

**Dissertation**

**Drug-Induced Liver Injury in Psychiatric Disorders**

submitted by

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## **Statutory declaration**

I hereby declare that this thesis is my own original work and that I have fully acknowledged by name all of those individuals and organizations that have contributed to the research for this thesis. Due acknowledgement has been made in the text to all other material used. Throughout this thesis and in all related publications I followed the “Standards of Good Scientific Practice and Ombuds Committee at the Medical University of Graz“.

Graz, August 2020

Anna Maria Birkl-Töglhofer

## Disclosures

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### **Hepatic gene expression explains primary drug toxicity in bipolar disorder**

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*'Life need not be easy, provided only that it is not empty.'*

*(Lise Meitner)*

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## Abbreviations and definitions

AADAC	arylacetamide deacetylase
ACBD5	acyl-CoA binding domain containing 5
ACOT13	acyl-CoA thioesterase 13
ADAM10	ADAM metallopeptidase domain 10
AGL	amylo-alpha-1, 6-glucosidase, 4-alpha-glucanotransferase
AGTR1	angiotensin II receptor type 1
ALDH2	aldehyde dehydrogenase 2 family member
ANAPC13	anaphase promoting complex subunit 13
APOB	apolipoprotein B
A-SAA	acute-phase serum amyloid A
ATP	adenosine triphosphate
ATP5O	ATP synthase peripheral stalk subunit OSCP
ATP5PB	ATP synthase peripheral stalk-membrane subunit b
BMI	body mass index
BPD	bipolar disorder
C3	complement C3
CAT	catalase
CD	cluster of differentiation
CDH1	cadherin 1
cDNA	complementary desoxyribonucleic acid
CEP57	centrosomal protein 57
CMBL	carboxymethylenebutenolidase homolog
Ct	cycle treshold
CTRL	control
CYP	cytochrome P450
CYP3A5	cytochrome P450 family 3 subfamily A member 5
DECR1	2,4-dienoyl-CoA reductase 1

DEGs	differentially expressed genes
DHTKD1	dehydrogenase E1 and transketolase domain containing 1
DNA	desoxyribonucleic acid
DSM	Diagnostic and Statistical Manual of Mental Disorders
ERK	extracellular signal-regulated kinase
FDR	false discovery rate
FMO3	flavin containing dimethylaniline monooxygenase 3
FN1	fibronectin 1
GAPDH	glyceraldehyde-3-phosphate dehydrogenase
GATM	glycine amidinotransferase
GO	Gene ontology
HAO1	hydroxyacid oxidase 1
HCC	hepatocellular carcinoma
HGNC	Human Genome Organization (HUGO) Gene Nomenclature Committee
HIBADH	3-hydroxyisobutyrate dehydrogenase
HMGB1	high mobility group box 1
IDH1	isocitrate dehydrogenase (NADP(+)) 1
IL1RN	interleukin-1 receptor antagonist
IL-1R $\alpha$	interleukin-1 receptor antagonist protein
IPO8	importin 8
LRPPRC	leucine rich pentatricopeptide repeat containing
LSM5	LSM5 homolog, U6 small nuclear RNA and mRNA degradation associated
MAOI	Monoamine oxidase inhibitor
MAPK	mitogen-activated protein kinase
MCODE	Molecular Complex Detection
mRNA	messenger ribonucleic acid
NAFL	non-alcoholic fatty liver
NAFLD	non-alcoholic fatty liver disease
NASH	non-alcoholic steatohepatitis

NDUFA9	NADH:ubiquinone oxidoreductase subunit A9
PCA	principal component analysis
PGM1	phosphoglucomutase 1
PMI	<i>post mortem</i> intervall
PON1	paraoxonase 1
PPP2CA	protein phosphatase 2 catalytic subunit alpha
PRKAB2	protein kinase AMP-activated non-catalytic subunit beta 2
PSMA2	proteasome subunit alpha 2
PTEN	phosphatase and tensin homolog
RAP1	Ras-related protein
RAP1A	RAP1A, member of RAS oncogene family
RETC	respiratory electron transport chain
RIN	RNA integrity number
RMA	Robust Mult-array Average algorithm
RNA	ribonucleic acid
RPL41	ribosomal protein L41
S100A12	S100 calcium binding protein A12
S100A8	S100 calcium binding protein A8
S100A9	S100 calcium binding protein A9
SAA2	serum amyloid A2
SDHB	succinate dehydrogenase complex iron sulfur subunit B
SLC38A4	solute carrier family 38 member 4
SNAP23	synaptosome associated protein 23
SNRI	Serotonin norepinephrine reuptake inhibitors
SOD1	superoxide dismutase 1
SSRI	selective serotonin reuptake inhibitors
STRING	Search Tool for the Retrieval of Interacting Genes/Proteins
SULT1E1	sulfotransferase family 1E member 1
TARDBP	TAR DNA binding protein pseudogene 1

TCA	tricyclic antidepressants
TCGA LIHC	The Cancer Genome Atlas Liver Hepatocellular Carcinoma
TDO2	tryptophan 2,3-dioxygenase
TNF	tumor necrosis factor
TNFRSF1B	TNF receptor superfamily member 1B
TNFSF10	TNF superfamily member 10
TRAIL	tumor necrosis factor related apoptosis inducing ligand
TUBE1	tubulin epsilon 1
UBQLN2	ubiquilin 2
UDP	uridine diphosphate
UGT	UDP-glucuronosyltransferase
UGT2B15	UDP glucuronosyltransferase family 2 member B15
VAMP8	vesicle associated membrane protein 8
WebGestalt	WEB-based GEne SeT AnaLysis Toolkit

## **Zusammenfassung**

Bei schweren psychischen Erkrankungen, wie der bipolaren Störung, ist häufig eine akute aber auch langfristige medikamentöse Behandlung mit Psychopharmaka notwendig um auftretende Symptome zu lindern beziehungsweise zu verhindern. Neben deren positiven Effekten, sind einige Nebenwirkungen bekannt zu denen Störungen des Stoffwechsels zählen. Da die Leber die Funktion als zentrales Stoffwechselorgans für Nährstoffe aber auch pharmakologische Substanzen erfüllt, sind psychopharmakologische Nebenwirkungen durch eine eingeschränkte Entgiftung in Bezug auf die hepatische Funktion anzunehmen. Obwohl die nicht-alkoholische Fettlebererkrankung (NAFLD) häufig in der westlichen Bevölkerung auftritt, zeigen Patienten mit einer bipolaren Störung ein erhöhtes Risiko NAFLD zu entwickeln. Diese reicht von einer einfachen Steatosis bis hin zu einer nicht-alkoholischen Stetaohepatitis. Inwiefern Psychopharmaka die Entstehung von Leberschäden beeinflussen, insbesondere der zugrundeliegende molekulare Mechanismus, ist bisher noch unbekannt. Um die möglichen Psychopharmaka-induzierten Störungen auf molekularer Ebene zu erfassen, nimmt die Anwendung von ‚-omics‘ Technologien und der Einsatz von bioinformatischen Tools in Kombination mit histopathologischen Erkenntnissen eine zentrale Rolle ein. Die Erforschung von psychopharmakologischen Nebenwirkungen auf molekularer Ebene in der Leber trägt somit zur Erhöhung der Medikamentensicherheit und im weiteren Sinne zur Verbesserung des therapeutischen Effekts von Psychopharmaka für Patienten mit bipolarer Störung bei.

Der Fokus dieser Dissertation lag auf der Bestimmung der globalen Genexpression in postmortalem Lebergewebe von bipolaren Patienten mit psychopharmakologischer Behandlung. Die globale Genexpression wurde mit einer Kontrollgruppe, die eine ähnliche hepatische Histopathologie aufwies, verglichen. Der Großteil der differenziell exprimierten Gene war in der bipolaren Gruppe vorwiegend schwächer exprimiert als in den Kontrollen. Daraus wurde geschlossen, dass Psychopharmaka allgemein die Genexpression eher reduzieren, als antreiben. Für die biologische Funktionszuordnung der differenziell exprimierten Gene wurden umfassende bioinformatische Analysen durchgeführt. Besonders Gene, die in der Atmungskette involviert sind, waren von einer reduzierten Genexpression betroffen. Es wird angenommen, dass dies eine erniedrigte zelluläre Energieversorgung nach sich zieht. Dies ist wahrscheinlich mit einem seneszenten Phänotypen assoziiert und könnte auch histopathologisch erkennbar sein. Des Weiteren wurde eine verminderte Genexpression

für Gene, die im Lipid-, Aminosäuremetabolismus, und im Metabolismus von Xenobiotika beteiligt sind, festgestellt. Es wurde angenommen, dass als Konsequenz der zelluläre oxidative Stress ansteigt und somit schwere Schädigungen in der Zelle entstehen. Zusätzlich wurde *in silico* die Funktion der codierenden Gene auf Proteinebene untersucht. Hiermit wurden Einblicke in Protein-Protein-Interaktionen gewonnen und die Funktion des jeweiligen Proteins in diesen Netzwerken bestimmt. Um Muster der Genexpression aufzudecken, die eine Psychopharmaka-induzierte histopathologische Schädigung aufweisen, wurden mögliche Assoziationen zwischen der Genexpression und histopathologischen Veränderung der Leber ermittelt. Vor allem konnte eine Assoziation von degenerativen Veränderungen der Leber mit der Genexpression gezeigt werden.

Zusammenfassend zeigte diese Studie, dass Patienten mit bipolarer Störung und der Behandlung von Psychopharmaka eine veränderte Genexpression in der Leber aufweisen. Vor allem ist die zelluläre Energieversorgung beeinträchtigt und erhöhter zellulärer Stress vorliegend. All dies lässt auf einen gesonderten pathophysiologischen Mechanismus schließen, der einer Leberschädigung durch Psychopharmaka in bipolarer Störung zugrunde liegt.

## Abstract

Acute and long-term psychotropic drug treatment is often necessary to improve symptoms and prevent relapses in severe mental disorders, such as bipolar disorder. Beside the beneficial effects of the pharmacological treatment, several adverse effects are known like the induction of metabolic disturbances. As the liver is the main site for metabolism of nutrients and pharmacological agents, an affected hepatic function by improper drug detoxification can be assumed. Although non-alcoholic fatty liver disease (NAFLD), ranging from simple steatosis to non-alcoholic steatohepatitis, is common in the Western population, patients with severe mental disorders have an increased risk for the development of NAFLD. To which extent psychotropic drug influence the underlying molecular mechanism leading to hepatic injury remains unknown. Therefore, to reveal the central physiology behind the emerging biological disturbances due to psychotropic drug treatment, the application of ‘-omics’ technologies and the integration of advanced bioinformatical analyses in combination with basic histopathology is crucial. The assessment of possible interferences of psychotropic drugs on the molecular level of hepatic function are needed to subsequently improve the safety and therapeutic effect of pharmacological treatment in bipolar disorder.

This exploratory study addressed the global gene expression in hepatic *post mortem* tissue of patients with bipolar disorder and psychotropic drug treatment compared to control samples with similar hepatic histopathology. Differentially expressed genes were predominantly downregulated assuming that psychotropic drugs rather reduce gene expression than shifting to an increased state of gene expression. To allocate the differentially expressed genes to their biological function, comprehensive bioinformatical analyses were conducted including pathway and enrichment analyses. Particularly, a decreased expression was observed in the biological pathway of the respiratory energy transport chain. The anticipated resulting diminished cellular energy supply is probably associated with a senescent phenotype which might be also evident in hepatic histopathology. Moreover, gene expression was reduced for genes involved in the metabolism of lipids, amino acids and xenobiotics, where a consequent increase in oxidative stress might entail serious cell damage. Furthermore, protein-protein interactions of the encoding respective differentially expressed genes were assessed *in silico* to provide insights in terms of their function in an intricate network of proteins. For the examination of gene expression patterns defining or triggering a harmed histopathological

phenotype due to psychotropic drugs, gene expression data and histopathological features were integrated. Particularly, histopathological degenerative changes in the liver were associated with the enriched pathways indicating adverse effects on molecular level for psychotropic drugs.

In summary, this study highlights an altered hepatic gene expression in patients with bipolar disorder and psychotropic drug treatment. The presence of a diminished cellular energy supply and an increased cellular stress level are defining the molecular pattern in the liver evoked by psychotropic drugs. All of which indicates a differential underlying mechanism initiating or triggering hepatic injury in bipolar disorder with psychotropic drug treatment.

# **1 Introduction**

## **1.1 Severe mental disorders**

Severe mental disorders are characterized by problems in thinking, behavior, cognition, perception, mood, emotion and moving functions, all of which lead to an impaired functioning in live activities [1]. According to the World Mental Health surveys on behalf of the World Health Organization, the global prevalence of mental disorders amounted to 15.4% in all assessed countries, whereby higher income countries showed increased prevalence compared to lower income countries [2]. In Europe, a lifetime prevalence of 25% has been demonstrated for mental disorders in adults, whereas the experience of any mental disorder in the past 12-month was reported to be 10% [3]. The diagnostic subgroup of mood disorders revealed a lifetime and 12-month prevalence of 14% and 4.2%, respectively, whereas higher odds exists for females and young adults. In addition to the high prevalence, disturbing reduced life-expectancy have been observed for individuals with severe mental illnesses, such as bipolar disorder, major depression or schizophrenia [4–7]. Compared to the general population the life expectancy difference accounts for more than 10 years for all three mentioned mental disorders in the United Kingdom [4], yet lower life expectancy was observed for men in Denmark, Finland and Sweden [5]. Interestingly, death due to physical illnesses was the most frequent cause for the decreased life expectancy, whereas suicide rates were decreasing from adolescence to adulthood and older age in bipolar disorder [7]. The underlying reason for the observed premature deaths as consequence of physical health conditions in these severe mental disorders remains unknown. A possible interaction between mental and physical health and disease may exist.

## **1.2 Causes of severe mental disorders**

Decades of extensive research chased the answer behind the underlying causes of mental disorders and the possibility of an effective therapeutic treatment. So far, it has been proposed that severe mental disorders are complex traits with environmental and multiple genetic contributing factors which may be shared between mental disorders.

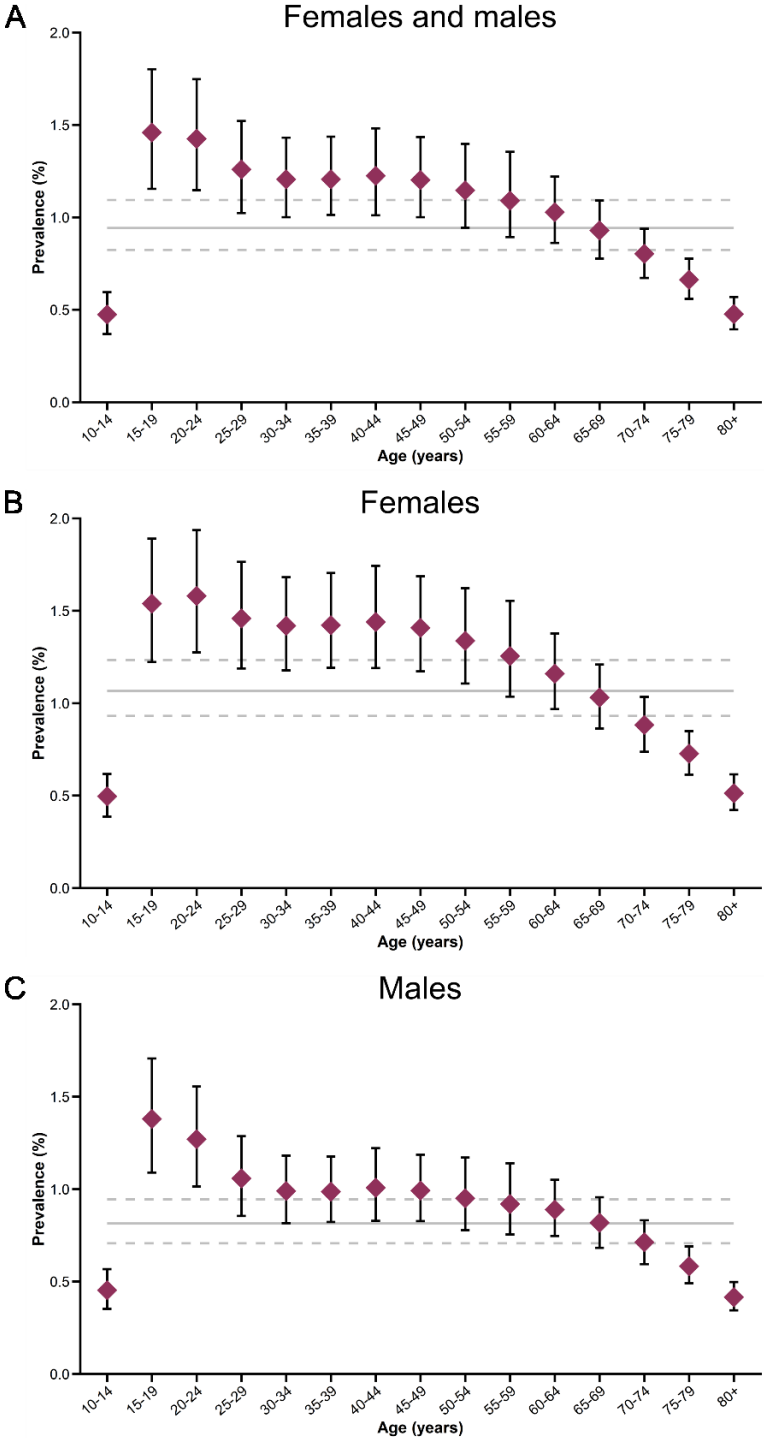
One of these supposed environmental factors is nutrition, where an unhealthy diet is linked to mental problems even during the prenatal phase [8]. An association between nutrition and gene

expression due to epigenetic modifications probably exists. Concerning the genetic factors, high heritability estimates were revealed for schizophrenia [9] and bipolar disorder [10,11], whereas on the contrary major depressive disorder exposed lower heritability [12]. Several genome-wide studies revealed genetic loci of interest considered to contribute to those mental disorders and indicating for a highly polygenic disorder [13–15]. A possible shared genetic background among mental disorders and mood disorders was evaluated [16,17]. A strong correlation of single nucleotide polymorphisms was observed for bipolar disorder and schizophrenia, whereas a moderate correlation was present for bipolar disorder and major depressive disorder as well as schizophrenia and major depressive disorder [16]. So, an underlying genetic contribution to the susceptibility of mental disorder can be assumed. Beside genetic predispositions affecting mental health, their influence on pharmacological treatment by gene-drug interactions is getting more and more attention. Recently, a minimum pharmacogenetics testing panel of 16 genetic variants has been proposed to be integrated in the standard protocol of the therapeutic pharmacological treatment [18]. These variants are associated with the drug effectiveness by modifying the available drug dose in the body and the drug tolerability. In general, pharmacological treatment in severe mental disorders is often accompanied with various side effects. However, probable altered underlying molecular effects due to pharmacotherapy such as an altered gene expression are not well understood. Therefore, the main focus of this thesis lies on bipolar disorder and the effect of psychotropic drug treatment on hepatic gene expression.

### **1.3 Bipolar disorder, a mood disorder**

Bipolar disorder, former known as manic depressive illness, manifests as chronic relapsing severe mental disorder typically characterized by unpredictable dramatic mood swings resulting in an impaired daily life functioning [11,19]. According to the World Health Organization report, the prevalence of bipolar disorder in the general population is 0.8%, whereas high income countries showed higher rates with 1.1% compared to low income countries with 0.4% [2]. These findings are consistent with data of the Global Burden of Disease Study 2017 for countries in Western Europe [20]. The age-standardized prevalence accounted for 1.0% for both sexes, whereas females have a slightly higher prevalence of 1.1% compared to males with a prevalence of 0.9%. With regard to the age at diagnosis, the highest prevalence of bipolar disorder is present in late adolescence and early adulthood for both sexes. Data sets of the

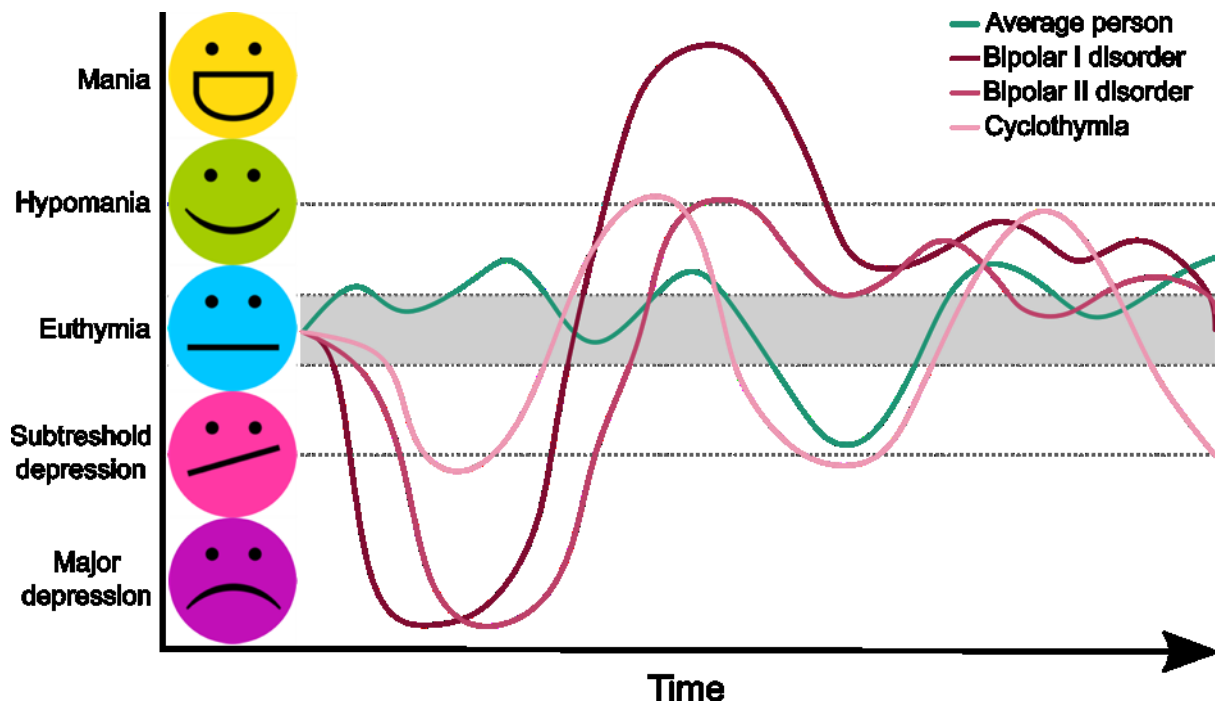
Global Burden of Disease Study 2017 has been used to outline the prevalence according to age and sex (*Figure 1*).



**Figure 1. Prevalence of bipolar disorder in Western Europe.** The prevalence according to the age categories are plotted for both sexes (A), and exclusively for females (B) and for males (C). The used data are provided by the Global Burden of Disease Study 2017 licensed under CC BY-NC 4.0 [20].

## 1.4 Subtypes of bipolar disorder

The classification and definition of severe mental disorders follows the widely used diagnostic criteria system of the Diagnostic and Statistical Manual of Mental Disorders (DSM), currently described in the fifth edition [21]. In general, bipolar disorder is characterized by the occurrence of interspersed mood swings comprising manic or hypomanic episodes, as well as depressive episodes alternating with phases of euthymia [19,22]. Based on the presence of those mood symptoms and their longitudinal appearance, different types of bipolar disorders are distinguishable (*Figure 2*).



*Figure 2. Main subtypes of bipolar disorder.* The different mood states over time for typical courses of bipolar disorder subtypes are illustrated in a simplified form.

In bipolar I disorder, the occurrence of at least one manic episode characterizes this subtype, whereas the presence of major depressive episodes is not necessarily needed for the diagnosis. In contrast, patients with bipolar II disorder have at least one milder hypomanic episode often followed by a major depressive episode. A special feature of bipolar I and II disorder is the occurrence of mixed states, where manic or hypomanic symptoms occur simultaneously with features of depression [22]. Cyclothymic disorder is characterized by episodes of hypomanic and depression over a duration of two years, where the diagnostic criteria for mania or major

depression are not met [11,19]. A special subtype represents the bipolar disorder with the feature of rapid cycling, where at least four mood episodes per year appear [11,22].

#### **1.4.1 Mood symptoms of bipolar disorder**

According to the DSM criteria, mania describes states of marked elevated mood and increased motor drive [19]. Severe manic symptoms might cause severe social or occupational impairment or psychotic symptom manifestation that frequently results in the requirement of hospital admission to prevent self-harm or harm to others [19,22]. The minimum duration of a manic episode is defined as one week [22], but may last for months with an observed average of 3.25 months [23]. Numerous symptoms occur, such as increased activity, less need for sleep, excessive talkativeness, heightened self-esteem, grandiosity, megalomania and increased risk-taking [19,22]. In contrast, hypomania defines mood elevations milder than during manic episodes and with absent triggering of social or occupational impairment and the necessity for hospitalization [11,19,22]. The symptoms usually remain present for at least four consecutive days to be defined as hypomanic episode [19] and may last for months with an observed average of 3.63 months [23]. For major depressive episodes in bipolar disorder, the same DSM-V criteria are valid as for major depressive disorder and last for at least two weeks [22]. Depressive episodes are commonly longer lasting than manic or hypomanic episodes with a duration of 5.18 months averaged for all bipolar disorder types [23]. Common depressive symptoms are reflected by reduced motivation and loss of interest or pleasure, as well as weight loss or gain, changes in sleeping behavior including insomnia and hypersomnia, agitation, fatigue, difficulties with concentrating, feeling of worthlessness and guilt, as well as thoughts of death [22]. Severe depressive states might entail suicidal ideation or suicide attempt.

Although most of the patients with bipolar disorder recover from the first episode, a high recurrence probability exists [24–26]. A recurrence probability of more than 80% within seven years is shown [25]. The highest relapse risk persists during the first year after recovery, where 37% [26] to 50% [25] of patients relapse during this period. Despite a continuing prophylactic pharmacotherapy, a recurrence probability of more than 70% is present in patients within 5 years [25].

## **1.5 Therapeutic approaches in bipolar disorder**

Those alternating and recurrent episodes of mania and depression considerably influence the functional and cognitive performance, as well as social and economic life of patients with bipolar disorder [25,27]. Due to fundamental differences of the therapeutic approaches in different mood states, a specified diagnosis is pivotal. Besides psychotherapy and psychoeducation with a wide range of therapy options including individual and interpersonal therapy settings, an effective drug treatment aiming to reduce symptoms and preventing mood episode recurrence by stabilizing the mood is essential [11,27–29]. In addition, electroconvulsive therapy might be applied to improve symptoms [29–31]. Therefore, to effectively manage and increase quality of life in patients with bipolar disorder, acute and long-term therapy and treatment is crucial.

## **1.6 Psychotropic drugs in the pharmacotherapy of bipolar disorder**

Pharmacotherapy of bipolar disorder encompasses the usage of mood stabilizers, antipsychotics and antidepressants [19,27–29,32,33]. The use of specific pharmacological compounds in bipolar disorder relies predominately on observations of therapeutic effects in randomized control trials and the arising response to mood symptoms. The definite mode of action on molecular or biochemical levels remains mainly unknown. In general, these psychotropic drugs are supposed to target the regulation of neurotransmitters and neuronal hormones, the intracellular signaling affecting neuroprotection, neuronal growth and ion channels, as well as sleep and circadian rhythm [27,34]. The pharmacotherapy with psychotropic drugs may vary, a combined or adjunctive usage or the usage as monotherapy are applied.

### **1.6.1 Mood stabilizers**

Pharmacological agents defined as mood stabilizers are drugs alleviating mood and psychotic symptoms predominantly during acute mania, and preventing recurrent mood episodes and prolong the inter-episode interval during prophylactic long-term treatment [35]. In addition, a sustained mood stabilizer treatment decreases the suicide rate in patients with bipolar disorder [36]. Despite their frequent clinical use, their underlying molecular mechanism of action remain unclear. Further, distinct drug characteristics, such as compound structure, aggravates the determination of overlapping molecular effects or targets [37]. Assumed downstream targets are supposed to be involved in neuronal and synaptic plasticity and

intracellular signaling [38]. Potentially affected intracellular signaling cascades are G-protein coupled signaling, protein kinase C signaling, glycogen synthase kinase-3 and Wnt signaling, and extracellular signal-regulated kinase (ERK) signaling cascade [37]. Moreover, possible epigenetic modifications based on mood stabilizers regulating gene expression are assumed [39].

### **1.6.2 Antipsychotics**

Atypical antipsychotics are used during acute episodes and for long-term treatment as monotherapy or concomitant therapy [40]. A potential mechanism of antipsychotic treatment includes the reduction of manic behavior during an acute mania episode [41]. In addition, antipsychotics improve depressive mood symptoms during bipolar depression [42,43]. Antipsychotics potentially contribute to depressive symptom relief via their mediated effects for neurotransmitters including serotonin, dopamine and noradrenalin and the corresponding receptors in the brain [44].

### **1.6.3 Antidepressants**

Since decades, acute depressive episodes are treated with antidepressants, though, a possible mood switch to a manic state, psychosis or aggressive behavior exist [33]. Despite their effectiveness for symptom relief, it is advised to use antidepressants in combination with mood stabilizers or antipsychotics, especially for bipolar I disorder, rather than as monotherapy [45,46]. Particularly the occurrence of mixed states represents a challenging setting to appropriate drug treatment [33,47]. Here, pharmacological treatment with antidepressants is discouraged due to their proposed potential to trigger a destabilized mood by provoking repeated induction of mixed states. Furthermore, long-term treatment with antidepressants is discouraged by most guidelines [29,48,49]. Sustained antidepressant monotherapy is associated with an increased risk of manic episode occurrence compared to mood stabilizer monotherapy or in combination with mood stabilizers [50]. In context of molecular effects, antidepressants are supposed to affect neuroplasticity probably by inducing gene expression and adaption of neurotransmitter function [51,52].

An overview of frequently used psychotropic drugs, the recommended scope of use and the targeted pharmacological profiles are listed (*Table 1*).

**Table 1. Recommended psychotropic drugs for bipolar disorder treatment.** Frequently used psychotropic drugs with their recommended application for the respective mood episode and during long-term treatment is indicated based on [19,29,47,48]. The pharmacological domains are listed for each drug and are based on the Neuroscience based Nomenclature devised by international psychopharmacological organizations [34,53].

<b>Psychotropic drug</b>	<b>Acute mania</b>	<b>Acute bipolar depression</b>	<b>Acute mixed state</b>	<b>Long-term treatment</b>	<b>Modified neurotransmitter/molecule/system</b>
<b>Mood stabilizers</b>					
Carbamazepine	+	+	-	+	glutamate
Lamotrigine	--	+	+	++	glutamate
Lithium	++	+	-	++	glutamate
Valproate	+	+	+	+	glutamate
<b>Antipsychotics</b>					
Clozapine	+	+	-	+	dopamine, serotonin, norepinephrine
Olanzapine	++	++	++	+	dopamine, serotonin
Quetiapine	++	++	-	++	dopamine, serotonin, norepinephrine
Risperidone	+	-	-	+	dopamine, serotonin, norepinephrine
Ziprasidone	+	-	+	+	dopamine, serotonin
<b>Antidepressants</b>					
Monoamine oxidase inhibitor (MAOI)	--	+	-	-	serotonin, norepinephrine, dopamine
Serotonin norepinephrine reuptake inhibitors (SNRI)	--	+	-	-	serotonin, norepinephrine
Selective serotonin reuptake inhibitors (SSRI)	--	+	-	-	serotonin
Tricyclic antidepressants (TCA)	--	+	-	-	serotonin, norepinephrine, dopamine

++ = highly recommended; + = recommended; - = non-recommended; -- = definitely non-recommended

## **1.7 Biotransformation of psychotropic drugs**

The kinetic of psychotropic drugs relies on drug absorption, distribution, metabolism and elimination [54]. The majority of psychotropic drugs are highly fat-soluble compounds and require biotransformation for the generation of a metabolic compound or prior to drug elimination, whereby the liver is the primary site for drug metabolism [55]. The biotransformation of xenobiotics such as psychotropic drugs consist of phase I and II. For a parent drug compound, biotransformation in phase I results in less lipophilic and more polar compound by oxidation, reduction or hydrolysis. Enzymes common in phase I biotransformation of psychotropic drugs are cytochrome P450 (CYP) enzymes [56–60]. In phase II, the conjugation of hydrophilic groups increase the water solubility of these compounds and, hence, their elimination from the body [55,56]. Enzymes mainly involved in the phase II biotransformation of psychotropic drugs are transferases for glucuronic acid, sulfate, acetate, glutathione, and methyl group [56,61]. Phase I and/or II biotransformation of the parent compound into metabolic compounds is sometimes necessary to achieve the therapeutic psychotropic drug effect [62]. The majority of drugs reveal reduced pharmacological activity after biotransformation, but some antipsychotics and antidepressants are metabolized into active metabolites [55]. However, biotransformation of the parent compound could also result in toxic metabolites, which are rather derived from phase I than phase II [57]. Due to the toxic potential of psychotropic drugs, it is hardly surprising that adverse drug effects are common during pharmacological treatment in bipolar disorder and, in general, in severe mental disorders.

## **1.8 Adverse side effects of psychotropic drugs**

The first line treatment in bipolar disorder is pharmacotherapy, yet frequent and infrequent adverse side effects are known in association with acute and long-term psychotropic drug treatment. The side effects range from adverse cognitive and neurological events [58,63,64] to adverse hormonal [58,64,65], gastrointestinal [58,64], and immunological events [64]. In addition, teratogenic effects [64] and adverse effects on the renal function [64,65] are observed, as well as weight gain [58,63,65,66] and the development of the metabolic syndrome [63,64], and hepatotoxicity [67,68].

## **1.9 Hepatotoxicity and drug-induced liver injury**

The liver, as the main site of systemic glucose and lipid biosynthesis [69], is frequently affected in severe mental disorder including bipolar disorder. Especially, patients with long-term psychotropic drug treatment are at risk of non-alcoholic liver disease (NAFLD) [70]. According to a recent meta-analysis, the general NAFLD prevalence in the population of Europe and North America is estimated to be 23.71% and 24.13%, respectively, which is slightly lower than the estimated global average prevalence of 25.24% [71]. The highest rates were observed in South America and Middle East with a prevalence more than 30%. In addition, increased NAFLD prevalence is associated with factors such as the presence of type 2 diabetes or an increased body mass index (BMI) [72]. Both, diabetes [73] and an increased BMI [74], have a higher prevalence in bipolar disorder compared to a reference population. Patients with NAFLD are faced with an elevated risk for the overall mortality, the development of liver cirrhosis and hepatocellular carcinoma (HCC) [75]. The determined hazard ratios were, 1.29, 3.20 and 6.55, respectively. With regard to patients with bipolar disorder, a 2.28 higher likelihood was determined for any chronic liver disease [76] and a 2.57 higher likelihood for the presence of NAFLD compared to controls without bipolar disorder [77]. Consequently, patients with bipolar disorder treated with psychotropic drugs represent a highly vulnerable group for the development NAFLD and its progression.

## **1.10 Spectrum of NAFLD**

NAFLD covers a wide spectrum of liver diseases ranging from simple hepatic steatosis to non-alcoholic steatohepatitis (NASH), progressive fibrosis and cirrhosis, and a possibly increased risk for the development of HCC, but with the absence of significant alcohol consumption [78]. The underlying mechanism for the development of lipid accumulation, hepatocyte injury and inflammation in NAFLD is complex with multiple supposed intrinsic and extrinsic factors. Factors such as genetic predisposition and epigenetic modifications, environmental factors and to a great extent dietary habits influence hepatocellular fat and inflammatory conditions. NAFLD has been linked to insulin resistance, obesity and type 2 diabetes, all of which are signs of the metabolic syndrome [79–81]. Today, the assumed origin of NAFLD development underlies a multiple hit hypothesis [78,82]. Lipotoxicity and glucotoxicity seem to play a central role [81]. Accumulation of lipids is a key event of NAFLD, whereas the presence of excessive fatty acids relies on dietary intake, lipolysis of adipose tissue and substantially on *de novo*

lipogenesis [80]. The grading and staging of NAFLD comprises the evaluation of histopathological features for steatosis, inflammation and cellular injury [79,80].

### **1.11 Histopathological features of NAFLD**

The lipid accumulation present during simple hepatic steatosis is reversible and defined as non-alcoholic fatty liver (NAFL) if at least 5% of hepatocytes present steatosis [79,83,84]. On the histological view, the intracellular lipid accumulation pushes the nucleus to the margins of the hepatocyte visually resembling an adipocyte [79]. The stored lipids visually manifest as lipid-containing vacuoles usually in macrovesicular or medio-macrovesicular form. However, the cellular excess of lipids due to a constant uptake, *de novo* synthesis and an insufficient lipid metabolism results in toxicity due to cellular injury by endoplasmic reticulum stress, mitochondrial damage and oxidative stress [80,85]. These intracellular stress signals based on the inability to resolve excessive lipids can induce apoptosis.

Advanced hepatocellular injury can be present as degenerative ballooned hepatocytes, a key feature for NASH [84]. These are hepatocytes without steatotic vacuoles but with a clear to flocculent cytoplasm and possible intracytoplasmic inclusion bodies primarily composed of misfolded cytokeratin 8 and 18 filaments, chaperone and heatshock proteins, which are defined as Mallory-Denk bodies [84,86]. Along with the occurrence of hepatocellular ballooning, an inflammatory response characterizes NASH [79,86]. Often mild chronic lobular inflammation is present at small foci with infiltrated lymphocytes, macrophages and neutrophils [79]. Usually, portal fibrosis is present in NASH [79,84,86], however, the degree of fibrosis highly varies and may already be present in NAFL [79]. During fibrosis, the architecture of the parenchymal liver tissue is being rearranged. Hepatocytes are separated and replaced by strands of collagen deposits along with extracellular matrix proteins [79,84]. Bridging fibrosis is present in advanced disease stages. With the progression of liver injury, functional liver tissue is largely replaced by scar tissue which severely interferes with the physiological liver function [84]. Cirrhosis is defined as end stage liver disease for chronic liver diseases, whereas the cirrhosis incidence based on NASH is higher than in NAFL [87]. In addition, NAFLD progressing to cirrhosis is assumed as primary indication for liver transplantation in future.

## **1.12 Complex underlying mechanisms of NAFLD**

The molecular mechanisms contributing to the development and progression of NAFLD are not fully elucidated, however, assumed affected pathway candidates are the metabolism of lipids and glucose and inflammatory pathways. These affected pathways are probably based on dysregulated genes arising from the loss of regular gene expression maintenance and of precise transcriptional control. A gene expression study investigating the possible associations between simple steatosis and hepatic gene expression revealed various involved genes of diverse molecular pathways [88]. Particularly pathways, such as the metabolism of glucose, lipids and amino acids, insulin signaling, inflammatory response and cell adhesion, are supposed to be implicated in NAFLD. In addition, a gene expression signature comprising 218 genes has been identified in a meta-analysis of publicly available datasets investigating their association with histopathological NAFLD features [89]. This emphasizes the possible application of biomarkers based on gene expression level for the identification and stratification of NAFLD and its disease progression which in turn could be beneficial for disease monitoring and therapeutic interventions. Even though patients with bipolar disorder and psychotropic drug treatment are at higher risk to develop NAFLD, the possible effect of pharmacological treatment altering the gene expression and adversely influencing hepatic function remains unclear.

## **1.13 Rationale of this thesis**

Up to now, less is known about the molecular effect of psychotropic drugs on hepatic gene expression and in the presence of NAFLD. Analysis of transcriptomics and their integration into comprehensive bioinformatical analyses, such as pathway, enrichment and interaction analyses, provides the basis for understanding complex disorders like NAFLD and in the special setting during psychotropic drug treatment. Therefore, the determination of deregulated genes in a disease and their respective protein function in an intricate network of protein interactions could provide insights into altered molecular processes and reactions associated with specific biochemical pathways. Further, the integration of gene expression data and histopathological features of NAFLD could be used to define molecular settings triggering an altered or a harmed phenotype.

This thesis addresses the fundamental effects of psychotropic drug treatment on gene expression level in the liver of patients with bipolar disorder compared to a histopathological

comparable control sample. Therefore, comprehensive bioinformatical analyses were established and the validation of the transcriptomics data was performed. Additionally, thorough association studies between gene expression and hepatic histopathological features were assessed.

## **2 Materials and Methods**

The Materials and Methods section is in part identical to the information that was published in Birkl-Toeglhofer et al. [90].

### **2.1 Human liver tissue samples**

Liver tissue of deceased patients with bipolar disorder and controls were collected as part of the Stanley Neuropathology Consortium Collection of the Stanley Medical Research Institute (Chevy Chase, MD, USA). Information regarding the patient's medical and psychiatric history was ascertained by the patient's medical records and by an interview with the next of kin. The clinical diagnosis was made according to the DSM-IV criteria. Liver tissue from patients was collected during autopsy after the donation permission was obtained from the patient's next of kin. For the sample collection, a technically standardized procedure was applied by trained medical examiners. Due to the presence of de-identifiable specimens of deceased human subjects the Institutional Review Board of the Uniformed Services University of the Health Sciences, Bethesda, decided that the Institutional Review Board ethical approval was not required for the collected samples of the Stanley Neuropathology Consortium Collection. A public statement of the Stanley Medical Research Institute is available at <http://www.stanleyresearch.org>. The study was approved by the local ethics committee of the Medical University of Graz (31-41 ex 18/19). Differences of demographic and clinical characteristics between the groups were tested for significance using the unpaired t-test for normally distributed data. The Fisher's exact test was used for binary data. All statistical analyses were performed using IBM SPSS Statistics, v23 (IBM Corp., Armonk, NY, USA).

### **2.2 Global gene expression analysis**

#### **2.2.1 Microarray processing**

*Post mortem* liver tissue of patients with bipolar disorder and control samples from the Stanley Medical Research Institute (Chevy Chase, MD, USA) were included in the global gene expression analysis. All sample preparation, including RNA extraction, RNA quality and quantity assessment and array processing steps of the microarray analysis, were performed at the Microarray Core Facility of the Johns Hopkins University (Baltimore, MD, USA). In brief, frozen liver tissue was used for the RNA extraction. The liver tissue was extracted using TRIzol

(Invitrogen, Carlsbad, CA, USA) and chloroform method. Total RNA was precipitated with isopropyl alcohol and washed with 70% ethanol. The precipitated and washed RNA was dissolved in diethyl dicarbonate water. An additional RNA purification step was added using the RNeasy column purification (Qiagen, Valencia, CA, USA). The RNA quantity and quality was assessed using the Agilent Bioanalyzer (Agilent Technologies, Palo Alto, CA, USA). The quality was represented by the RNA integrity number (RIN). For the investigation of global gene expression, the oligonucleotide array GeneChip Human Genome U133 Plus 2.0 array (Affymetrix, Santa Clara, CA, USA) was used. The CEL files were kindly provided by Prof. Serge Weis (Division of Neuropathology, Department of Pathology and Neuropathology, Neuromed Campus, Kepler University Hospital, Medical School, Johannes Kepler University, Linz, Austria) for further processing.

### **2.2.2 Data pre-processing and assessment of quality metrics**

Several steps were implemented prior to the assessment of differentially expressed genes (DEGs), including quality control steps, background adjustment, normalization and summarization. All of the bioinformatical analysis steps for the pre-processing and the quality assessment were performed using the open source software R 3.4.1 [91]. The Bioconductor package ‘affy’ [92] was used to read the generated CEL file of each chip.

Quality metrics were assessed for the raw data including average background and percentage of present calls. Visual exploration was performed on the probe intensity distribution. The relationship between probe intensity and difference to the median probe intensity was assessed. Quality control on the normalized data included a non-specific filtering step and the assessment of the overall sample variability. The average background and the percentage of genes called present were obtained using the Bioconductor package ‘simpleaffy’ [93]. To examine potential issues on RNA quality, the averaged probe intensity of all probes was plotted against the position of the probe. The focus was set on the appearance of the slope and the shape of the plotted curves from the 5’ to the 3’ end of the probesets. The Bioconductor package ‘affy’ [92] was used. To identify possible outliers deviating from the average intensity distribution, boxplots were generated prior and after normalization. Slight deviations were equalized by ensuing normalization, whereas extreme deviations would let to the exclusion of the sample. For the examination of the probe intensity and the difference to the median probe intensities, MA-plots were generated for each sample prior and after normalization. The M represents the

probe intensity ratio of one array and the median intensity of this probe over the entire arrays. The A denotes the average probe intensity of a single array and the median intensity of this probe over the entire arrays. The Bioconductor package 'affy' [92] was applied for the plotting.

An example R code is as follows:

```
# Load necessary packages
library('affy')
library('simpleaffy')
library('graphics')

# Read in .CEL files
data <- ReadAffy(ceIfiles)

# Perform quality control on the raw data
data.qc <- qc(data)

# Assess the average background for each sample
avbg <- avbg(data.qc)

# Obtain the percentage of present calls
pp <- percent.present(data.qc)

# Obtain RNA degradation
deg_data <- AffyRNAdeg(data)

# Plot the slope from the 5' to the 3' end
plotAffyRNAdeg(deg_data)

# Plot boxplot for perfect match probes for raw intensity data
boxplot(data,
         transfo = log2,
         which = 'pm')

# Plot boxplot for perfect match probes for normalized intensity data
boxplot(norm_data,
         transfo = identity,
         which = 'pm')

# Plot MA plot for raw intensity data
MAplot(object=data,
        which = 1)
```

```
# Plot MA plot for RMA normalized intensity data
MAplot(object=norm_data,
        which = 1)
```

### 2.2.3 Array-to-array normalization

To normalize the samples and, hence, control for array-to-array variation, the Robust Multi-array Average (RMA) algorithm [94] was applied. The RMA algorithm includes background correction, quantile normalization and the expression calculation. To generate RMA normalized expression values of the oligonucleotide arrays, the Bioconductor package ‘affy’ [92] was applied using the open source software R 3.4.1 [91].

Next, non-specific filtering was applied to the normalized expression data. Probes were removed when the low overall intensity was less than the  $\log_2(100)$  in a minimum of 10% of the samples and when the low overall variability of the  $\log_2$  intensities interquartile range was below 0.5. The Bioconductor package ‘genefilter’ [95] was applied to filter for low overall intensity values. To filter for low overall variability of expression values, an anonymous function was implemented.

An example R code for the non-specific filtering is as follows:

```
# Normalization of the raw expression data with RMA
norm_data = rma(data)

# Filtering of probes with low overall intensity values
filter1 <- poverA(0.10,
                 log2(100))

# Filtering of probes with low overall variability across samples
filter2 <- function(x)(IQR(x)>0.5)

# Combining filter functions into one function
filter <- filterfun(f1, f2)

# Implementation of non-specific filtering to normalized data
data_filtered <- genefilter(norm_data,
                             filter)
```

#### 2.2.4 Substructure analysis based on gene expression data

For the assessment of the overall sample variability, a principal component analysis (PCA) was performed with the normalized and filtered expression data using the open source software R 3.4.1 [91]. The PCA was used to acquire information on possible sample cluster formation and similarities between samples in an unsupervised manner.

An example R code for the substructure analysis is as follows:

```
# Computation of principal components of gene expression data
pca <- prcomp(t(exprs(norm_data)),
              center = TRUE,
              scale. = TRUE)

# Plotting of a screeplot of first ten principal components
fviz_screepLOT(pca,
               ncp = 10)

# Plotting first principal component against second principal component
autoplot(pca)
```

#### 2.2.5 Assessment of differential gene expression

To assess DEGs between bipolar disorder and the controls, a linear model with a moderate t-statistic was fitted to each probe. Several factors were used to adjust for probable confounding effects in the model. The factors included age, sex, BMI and *post mortem* interval (PMI). All probes were corrected for multiple testing by Benjamini and Hochberg adjustment. To define DEGs, the expression and the significance threshold of probes was set to a fold change of 1.5 and a false discovery rate (FDR) of less than 5%. All steps for the differential gene expression assessment were performed using R 3.4.1 [91]. The used Bioconductor package was ‘limma’ [96].

All probesets were annotated to the corresponding Human Genome Organization (HUGO) Gene Nomenclature Committee (HGNC) gene symbol, name and unique identifier using the Bioconductor package ‘biomaRt’ [97].

An example R code for the annotation is as follows:

```

# Create an appropriate design matrix
design <- model.matrix(~ 0 + factor)

# Fit the linear model
fit <- lmFit(data_filtered ~ design)

# Create the contrast matrix and compare groups
cont_matrix <- makeContrast(BPD-CTRL,
                           levels = design)
cont_fit <- eBayes(contrasts.fit(fit,
                                cont_matrix))

# Extract a list of differentially expressed genes in bipolar disorder
topTable(cont_fit,
         coef = 1,
         adjust.method = 'BH')

# Query databases for entries of the Affymetrix probeset IDs
getBM(attributes = affy_ensembl,
       mart = ensembl,
       values = "*",
       uniqueRows = TRUE)

```

## 2.3 Protein-protein interaction network analysis

Gene names of the DEGs were used as input for the stringApp, version 1.5.0 [98], a built-in app of Cytoscape, version 3.7.2 [99], an open source software for visualizing complex networks. The stringApp extracts data from the Search Tool for the Retrieval of Interacting Genes/Proteins (STRING) database [100,101]. The STRING protein query was used to find reliable protein-protein interactions with a minimum confidence score of 0.7. Using Cytoscape, the protein-protein interaction network was generated, whereas the  $\log_2$  fold change was used for the node color.

### 2.3.1 Analysis of key regulators of the protein-protein interaction network

A screening for topologically highly connected nodes, so called hubs, in the protein-protein interactome was performed using the Cytoscape built-in app cytoHubba, version 0.1 [102]. Global central nodes were measured based on the betweenness centralities providing normalized values which are independent to the network size.

### **2.3.2 Module analysis of the protein-protein interaction network**

The modules of closely-connected subnetworks of the generated protein-protein interactome were investigated using the Cytoscape built-in app ClusterViz, version 1.0.3 [103]. The Molecular Complex Detection (MCODE) clustering algorithm [104] based on density weighted nodes was used for module identification. The following cluster identification criteria were chosen: a minimum number of 2 edges per node (degree threshold), less than 20 % node score deviation from the seed node for new cluster members (node score threshold), at least 2 or more connections per node (k-core threshold), and an virtually unlimited distance from seed node to cluster members of 100 (maximum depth). A module cut-off with a score of >4 was applied.

## **2.4 Analysis of functional pathways**

For the identification of over-represented functional pathways in bipolar disorder compared to controls, the open-source tool WEB-based GENE SeT AnaLysis Toolkit (WebGestalt) [105,106] was used. The DEGs were examined for enriched functional pathways included in the Reactome Pathway Database. Pathways categories involving 5 to 1000 genes and a FDR of less than 0.05 for the multiple testing correction were set as analysis parameters. In addition, Gene Ontology (GO) terms were queried for cellular components, biological processes and molecular function using WebGestalt.

## **2.5 Validation of microarray data using quantitative real-time PCR**

### **2.5.1 RNA isolation**

For total RNA isolation, frozen post-mortem liver tissue from 16 bipolar disorder samples and 14 controls was used. Tissue homogenization was performed in 1 ml TRIzol Reagent (Thermo Fisher Scientific, Waltham, MA, USA) using a MagNA Lyser Instrument (Roche Diagnostics, Risch-Rotkreuz, Switzerland). The homogenization was performed in two rounds of 6500 rpm for 40 sec. After an incubation of 10 min on ice, the tissue homogenate was centrifuged at 10,000 x g for 10 min at 4 °C. For phase separation, 200 µl chloroform was added and vigorously mixed, followed by an incubation of 10 min at room temperature and a centrifugation at full speed for 15 min at 4 C. The upper aqueous phase containing RNA was precipitated with isopropyl alcohol and a high salt solution with 1.2 M sodium chloride and 0.8 M sodium citrate for 10 min at room temperature. After a centrifugation step at full speed

for 20 min at 4 °C and the removal of the supernatant, the pellet was washed twice with 80 % ethanol and dried for 5 min at 37 °C. The isolated RNA was dissolved in RNase-free water. RNA quantity and quality were measured using the NanoDROP 1000 (Thermo Fisher Scientific Inc., Waltham, MA, USA). The samples were stored at -20 °C until further processing.

### **2.5.2 Reverse transcription and quantitative real-time PCR**

In total, 2 µg RNA was reverse transcribed using the High Capacity cDNA Reverse Transcription Kit (Applied Biosystems, Foster City, CA, USA) containing an RNase inhibitor (Applied Biosystems, Foster City, CA, USA). A reaction volume of 20 µl was cycled according to the manufacturers instruction in the GeneAmp 9700 Thermocycler (Applied Biosystems, Foster City, CA, USA). The gene expression validation analysis was performed for 14 genes and 3 endogenous control genes. Gene specific primers were designed using Primer-BLAST [107] (**Table 2**). For quantitative real-time PCR, the Luna Universal qPCR Master Mix (New England Biolabs, Ipswich, MA, USA) was used according to the manufactures protocol. The reaction volume of 10 µl contained a final concentration of 10 ng cDNA, 0.2 µM of the gene specific forward and reverse primer and 5 µl of the Luna Universal qPCR Master Mix. To exclude cDNA cross contamination and contaminating DNA, a no template control with absent cDNA and a no amplification control with lacking reverse transcription reaction was included for each tested gene. Gene expression values were assessed as triplicates for each sample in 384-well plates using the QuantStudio 7 Flex Real-Time PCR System (Applied Biosystems, Foster City, CA, USA). The cycling conditions were as follows: 50 °C for 2 min and 95 °C for 10 min for the initial denaturation, followed by 45 cycles of denaturation at 95 °C for 15 sec and extension at 60 °C for 60 sec, and a melting curve stage at the end at 95 °C for 15 sec, 60 °C for 60 sec and 95 °C for 15 sec. A run quality control was performed using the QuantStudio Real-Time PCR Software, version 1.3 (Applied Biosystems, Foster City, CA, USA) including melting curve examination. A random mixture of amplified samples were used to validate the amplicon size for each gene. The amplicon sizes were examined by loading 10 µl of amplification product in each slot on a 2 % agarose gel stained with HDGreen Plus DNA Stain (Intas Science Imaging, Goettingen, Germany). Gel electrophoresis was performed for 30 min at 100 V. The amplicons were visualized using the Gel iX20 Imager (Intas Science Imaging, Goettingen, Germany). Only samples with a proper melting curve indicating a specific amplicon were included for further analysis.

**Table 2. Primers used for gene expression validation by quantitative real-time PCR.** The RefSeq accession numbers of targeted transcripts, the forward (F) and reverse (R) primer sequences and the expected amplicon size of each gene and the endogenous controls are listed. [Published in [90]]

Gene	RefSeq accession number	Primer sequence (5'-3')	Amplicon size (bp)
<i>LRPPRC</i>	NM_133259	F: 5'-GCCGGAGGACTACTGAGC-3' R: 5'-AGCAAGGCATGACTACCACC-3'	230
<i>ATP5PB</i>	NM_001688	F: 5'-TGCTACCTGGACTTTTCGTTGA-3' R: 5'-CTGCAATACCCCTGGACCTA-3'	110
<i>SDHB</i>	NM_003000	F: 5'-CACTCTAGCTTGCACCCGAA-3' R: 5'-CCGTCCAGTTTCTCACGCTCT-3'	219
<i>PONI</i>	NM_000446	F: 5'-ATCGAAACTGGCTCTGAAGAC-3' R: 5'-ACGACCACGCTAAACCCAAA-3'	469
<i>ACOT13</i>	NM_018473	F: 5'-GGACTTTCAGCTCTTCCGA-3' R: 5'-GAGTAATCTTTCCAAAACCTCTCT-3'	152
<i>AADAC</i>	NM_001086	F: 5'-GCGTGGGAAGTGCTGCTCTAA-3' R: 5'-CTGGGTCATCAAGGAGCTGTT-3'	273
<i>CMBL</i>	NM_138809	F: 5'-GTTCAAGTCGAGCACATCAAGG-3' R: 5'-ACAATGCCATAGACGGACACC-3'	401
<i>HIBADH</i>	NM_152740	F: 5'-TGCAGGTGAACAGGTAGTATCTT-3' R: 5'-CCAGAAACAGGGGCATCCAT-3'	243
<i>TDO2</i>	NM_005651	F: 5'-TATCTCCAGCATCAGGCTTCC-3' R: 5'-GCCATGCCTCCACTAATTCC-3'	186
<i>HAOI</i>	NM_017545	F: 5'-ACAAGGACCGAGAAGTCACC-3' R: 5'-CCTGGCATCATCACCTCTCAA-3'	335
<i>CAT</i>	NM_001752	F: 5'-CCTGTGAACTGTCCCTACCG-3' R: 5'-ATAGAATGCCCGCACCTGAG-3'	222
<i>PSMA2</i>	NM_002787	F: 5'-CTACATTCAGCCCGTCTGGT-3' R: 5'-GGTCGTCCCTCATTCCAAC-3'	400
<i>S100A9</i>	NM_002965	F: 5'-CAAGAAGGAGAATAAGAATGAAAAG-3' R: 5'-AGCATGATGAACTCCTCGAA-3'	144
<i>S100A12</i>	NM_005621	F: 5'-TGCAAACACCATCAAGAATATCAAA-3' R: 5'-GTGTGGTAATGGGCAGCCTT-3'	99
<i>GAPDH</i>	NM_002046, NM_001256799, NM_001289745, NM_001289746, NM_001357943	F: 5'-AAATCAAGTGGGGCGATGCT-3' R: 5'-CAAATGAGCCCCAGCCTTCT-3'	86
<i>RPL41</i>	NM_021104, NM_001035267	F: 5'-AAGATGAGGCAGAGGTCCAA-3' R: 5'-TCCAGAATGTCACAGGTCCA-3'	248
<i>IPO8</i>	NM_001190995, NM_006390	Hs_IPO8_1_SG (QuantiTect Primer Assay; Qiagen, Hilden, Germany)	124

*LRPPRC* - leucine rich pentatricopeptide repeat containing; *ATP5PB* - ATP synthase peripheral stalk-membrane subunit b; *SDHB* - succinate dehydrogenase complex iron sulfur subunit; *PONI* - paraoxonase 1; *ACOT13* - acyl-CoA thioesterase 13; *AADAC* - arylacetamide deacetylase; *CMBL* - carboxymethylglutamate lyase-like protein homolog; *HIBADH* - 3-hydroxyisobutyrate dehydrogenase; *TDO2* - tryptophan 2,3-dioxygenase; *HAOI* - hydroxyacid oxidase 1; *CAT* - catalase; *PSMA2* - proteasome subunit alpha 2; *S100A9* - S100 calcium binding protein A9; *S100A12* - S100 calcium binding protein A12; *GAPDH* - glyceraldehyde-3-phosphate dehydrogenase; *RPL41* - ribosomal protein L41; *IPO8* - importin 8;

Obtained raw expression values were used to evaluate the relative gene expression by the  $2^{-\Delta\Delta Ct}$  method [108]. The mean of the endogenous control genes was used for normalization. Differences in gene expression between the groups were assessed using the unpaired t-test or the Mann Whitney U test depending on data distribution. The differences were expressed as x-fold changes. The statistical analyses were performed using GraphPad Prism 6 (GraphPad Software, San Diego, CA, USA). The plots were generated using R 3.4.1 [91].

## **2.6 Histopathological assessment of liver tissue**

For the histological assessment, 5  $\mu\text{m}$  sections of formalin-fixed and paraffin-embedded liver tissue of patients with bipolar disorder and controls were stained with various histological and immunohistochemical staining methods for the features of interest. The histological parameters were assessed and scored by two experienced pathologists, both blinded to the diagnosis. The histological assessment and the scoring of the histopathological features were performed prior to this thesis and were provided for this study. The histopathological parameters were statistically analyzed using the chi-square test for binary data and the Mann Whitney U test for categorical data. The frequency distribution of the histopathological features were plotted as stacked bar plots. The statistical analyses and the graphical visualization were performed using R 3.4.1 [91].

## **2.7 Association analyses between gene expression and histopathological features**

The association of histologically assessed characteristics (*feature*) and the  $\log_2$  intensity values as measure of gene expression (*expression*) were visualized as feature-expression heatmaps based on a previously described approach [109]. All features were categorized into six groups: (1) degenerative changes, (2) hepatic activity, (3) endogenous hepatic activity, (4) steatosis, (5) fibrosis and (6) inflammation. Applying ordinal logistic regression and binary logistic regression with regard to the histological data type, association analyses with the  $\log_2$  intensity values were performed for the bipolar disorder samples. The analyzed genes for the association analyses included the ten most altered DEGs discovered by the over-represented functional pathways analysis. The feature-expression heatmaps display the effect size, the standard error, the *p*-value and the FDR adjusted *p*-value from the Benjamini and Hochberg correction. An FDR adjusted *p*-value of  $<0.2$  was considered as statistically significant. Statistical analyses

and plotting were performed using R 3.4.1 [91] and MATLAB 2012b (The MathWorks, Inc., Natick, Massachusetts, USA).

### 3 Results

#### 3.1 Subject characteristics

The subject characteristics were examined (**Table 3**). The mean age of the bipolar disorder group and the control group was comparable. For the control group, a lower number of females was included in the analysis. The BMI was comparable, the mean BMI for the bipolar disorder group was classified as obesity class I (BMI range: 30.0-34.9) and for the controls as pre-obesity (BMI range: 25.0-29.9), according to the World Health Organization BMI classification [110]. A statistically significant difference was observed for PMI between the groups. The bipolar disorder group revealed longer PMI values than the controls.

**Table 3. Characteristics of subjects with bipolar disorder and controls.** [Published in [90]]

	<b>Bipolar disorder (n=29)</b>		<b>Controls (n=20)</b>		<b>p-value</b>
<b>Age, years, mean (<math>\pm</math>SD)</b>	41.8	( $\pm$ 14.93)	45.9	( $\pm$ 11.13)	0.276
<b>Gender, n (%)</b>					
Female	15	(51.7)	4	(20.0)	0.037
Male	14	(48.3)	16	(80.0)	
<b>BMI<sup>1</sup>, kg/m<sup>2</sup>, mean (<math>\pm</math>SD)</b>	30.8	( $\pm$ 6.82)	27.8	( $\pm$ 6.34)	0.123
<b>PMI<sup>2</sup>, hours, mean (<math>\pm</math>SD)</b>	39.1	( $\pm$ 20.36)	23.1	( $\pm$ 10.48)	0.002
<b>Medication, n (%)</b>					
antidepressants	15	(51.7)			
antipsychotics	16	(55.2)			
mood stabilizers	21	(72.4)			
anticholinergic	1	(3.5)			

<sup>1</sup>BMI, body mass index; <sup>2</sup>PMI, *post mortem* interval;

#### 3.2 Pre-processing and normalization of microarray data

The microarray data were pre-processed and normalized in order to remove noise. In total, each array included 54675 probes in the analyses.

### 3.2.1 Assessment of average background and percent present calls

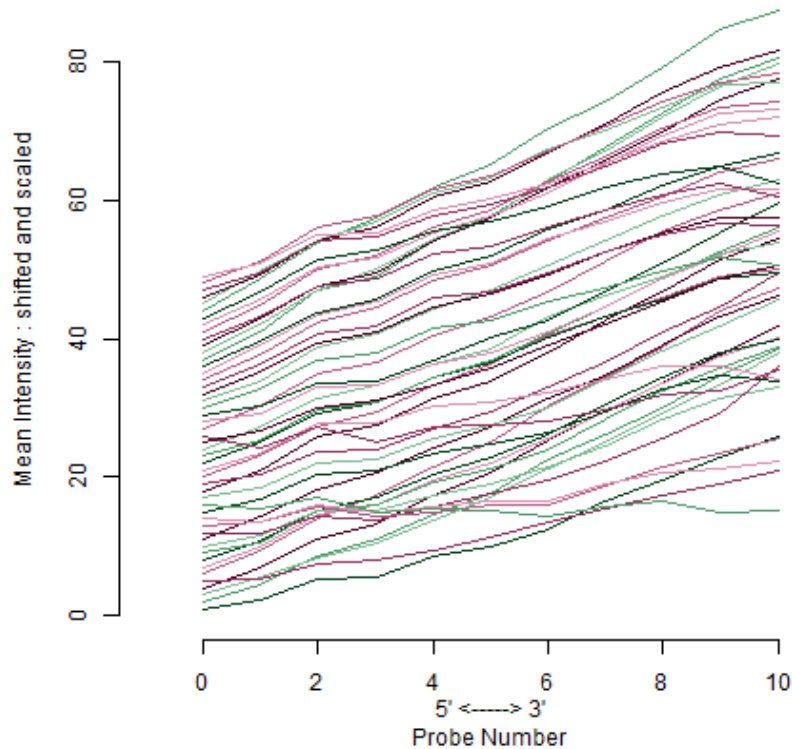
The individual average background values and percentage of genes called present were listed (*Table 4*). The average background and standard deviation of bipolar disorder and control samples was  $38.5 \pm 8.28$  and  $43.8 \pm 8.36$ , respectively, whereas the mean average background and the standard deviation of both groups was  $40.9 \pm 8.63$ . All of the samples fell into an average background range of 2.5 standard deviations. The mean and standard deviation of percent present calls in the groups was  $29.1\% \pm 3.06\%$  and  $19.7\% \pm 6.53\%$ , respectively. The overall mean and standard deviation of both groups was  $23.5 \pm 7.08$ , of which all but one bipolar disorder sample fall into a range of 2.5 standard deviations.

**Table 4. Assessed averaged background and present calls.** Individual average background and percentage present calls (% PP) of bipolar disorder (BPD) and control (CTRL) samples are listed.

Sample	Average background	% PP	Sample	Average background	% PP
CTRL01	54.3	28.0	BPD01	30.9	18.3
CTRL02	41.6	26.7	BPD02	32.9	25.3
CTRL03	47.5	28.7	BPD03	30.5	17.2
CTRL04	43.5	27.7	BPD04	32.0	23.2
CTRL05	35.4	33.7	BPD05	33.1	14.3
CTRL06	33.2	34.0	BPD06	38.8	28.5
CTRL07	35.3	27.8	BPD07	42.6	24.7
CTRL08	34.1	30.5	BPD08	52.9	23.1
CTRL09	37.7	32.4	BPD09	46.3	18.0
CTRL10	29.3	28.4	BPD10	34.0	16.3
CTRL11	42.1	29.2	BPD11	30.6	28.6
CTRL12	45.5	31.3	BPD12	37.4	13.3
CTRL13	59.4	27.1	BPD13	46.8	9.9
CTRL14	57.3	29.8	BPD14	42.2	8.9
CTRL15	45.4	30.6	BPD15	47.5	20.5
CTRL16	44.6	34.3	BPD16	36.0	5.5
CTRL17	55.4	27.8	BPD17	39.2	19.0
CTRL18	52.9	25.1	BPD18	46.0	23.3
CTRL19	42.5	27.0	BPD19	31.0	18.1
CTRL20	39.8	21.5	BPD20	29.8	27.6
			BPD21	56.0	22.6
			BPD22	28.5	25.9
			BPD23	40.1	21.8
			BPD24	29.7	32.6
			BPD25	34.8	19.0
			BPD26	29.2	7.0
			BPD27	59.1	24.6
			BPD28	35.6	16.8
			BPD29	42.5	18.4

### 3.2.2 Assessment of RNA degradation

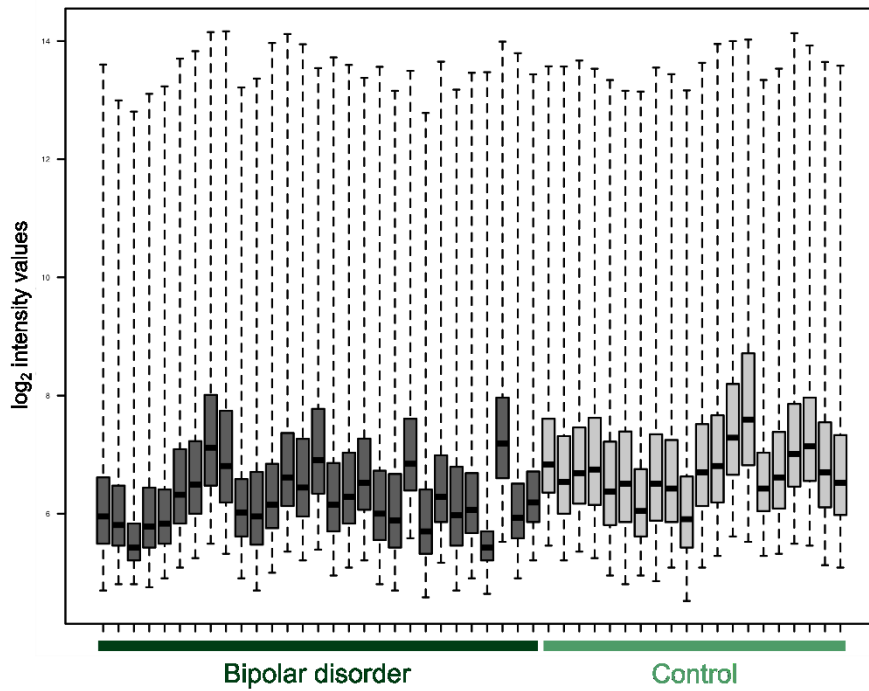
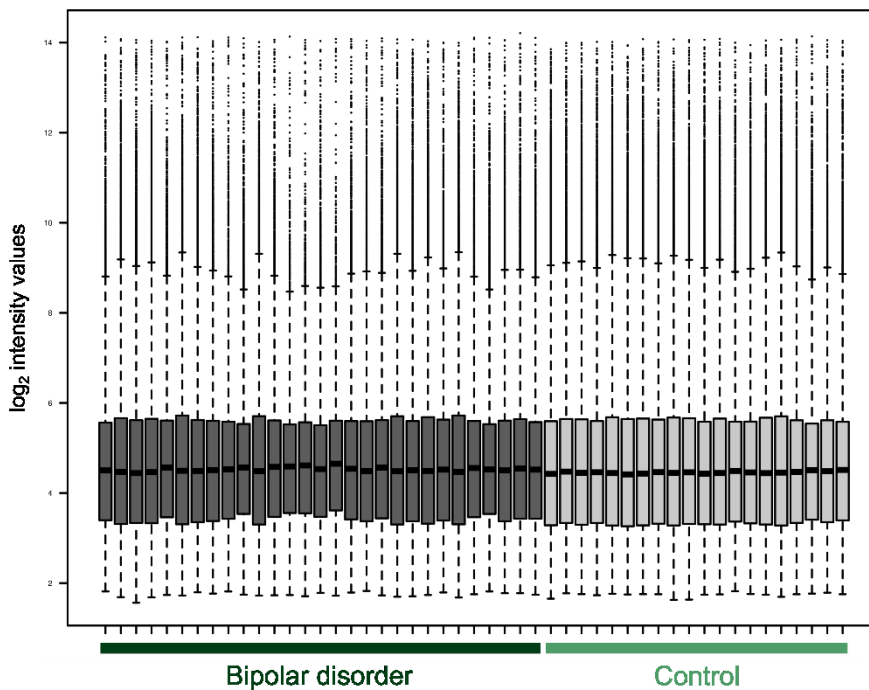
The RNA quality was compared between all samples and plotted as RNA degradation plot (**Figure 3**). An overall comparable RNA quality degradation was indicating similar RNA quality in all samples. The consistent ascending slopes of all samples represented an absence of intense RNA degradation differences between the samples.



**Figure 3. RNA degradation analysis of bipolar disorder and control samples.** The mean intensity of each probe (y-axis) is given across all probe sets from the 5' to the 3' end (x-axis). Each line corresponds to one sample.

### 3.2.3 Assessment of inter-sample intensity distributions

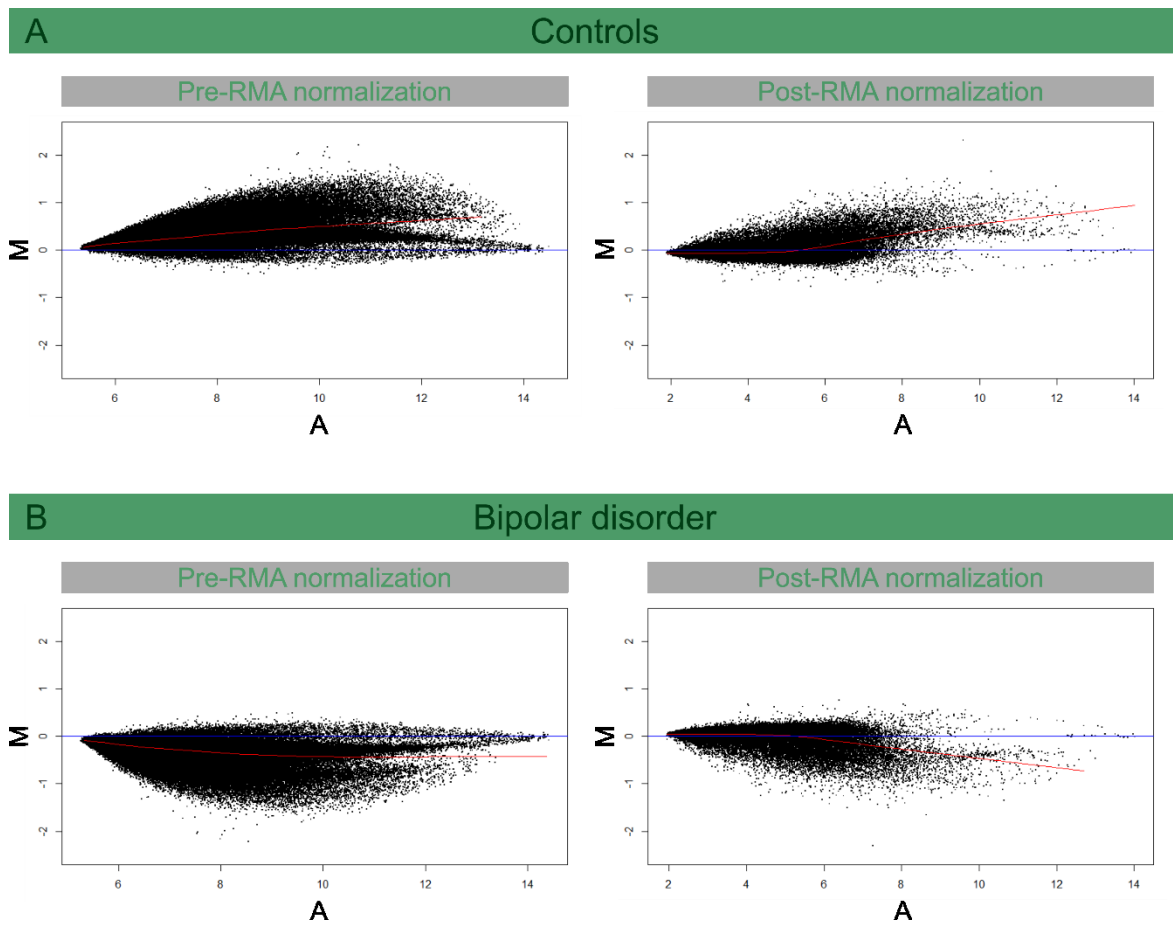
The between-array intensity boxplots of the samples were plotted to evaluate the inter-sample distribution of the gene expression data. A noise removal was observed after RMA normalization represented by the comparable distribution of  $\log_2$  intensity values across samples after normalization compared to the strong deviating distribution of the raw data (**Figure 4**).

**A****Pre-RMA normalization****B****Post-RMA normalization**

**Figure 4. Expression intensity distribution prior and after RMA normalization.** Each boxplot represents the chip-wise log<sub>2</sub> intensity of bipolar disorder (dark gray) and control samples (light gray). Bars represent interquartile range. (A) The samples prior to normalization deviate strongly from each other. (B) After RMA normalization, all intensity distributions appear similar.

### 3.2.4 Assessment of the averaged spread of gene expression

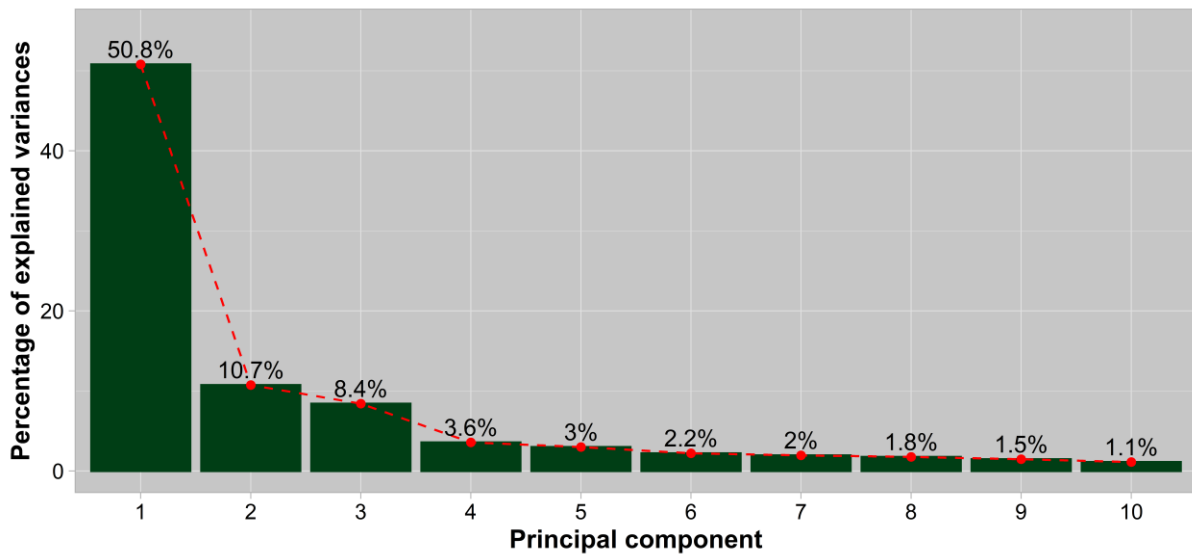
The MA-plots of the controls and bipolar disorder samples were generated to investigate the averaged spread of the gene expression data. By comparing the MA-plots prior and after normalization, the spread of the  $\log_2$  probe intensities and thus the variability is reduced in both groups by normalization (**Figure 5**). The centered distribution around  $M = 0$  in the normalized data refers to a higher number of genes that were non-differentially expressed than differentially expressed.



**Figure 5. MA-plots of raw data prior to RMA normalization and after RMA normalization for bipolar disorder and controls referenced to all samples.** Each dot represents one probe. In panel A, the averaged  $\log_2$  intensity ratio of the controls (M) were compared to the averaged  $\log_2$  intensity of all samples (A). In panel B, the averaged  $\log_2$  intensity ratio (M) of the bipolar disorder samples are compared to the averaged  $\log_2$  intensity of all samples (A). Non-differentially expressed probes are distributed tightly to  $M = 0$  (blue line). With increasing average intensity, the spread of probe intensities increases.

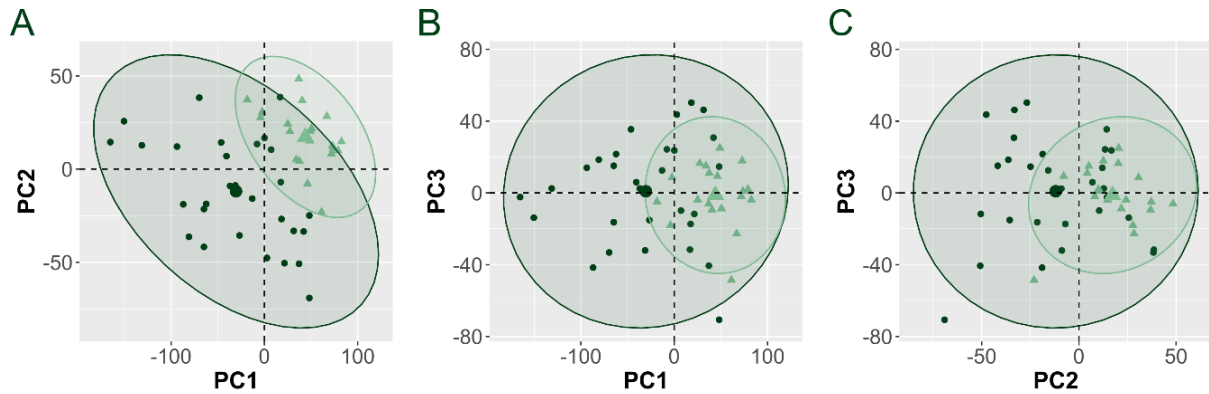
### 3.2.5 Assessment of sample substructures based on gene expression

The principal component analysis of the preprocessed, normalized and pre-filtered gene expression data uncovered biological substructures enabling the differentiation of liver tissue of bipolar disorder samples from control samples. The percentage of explained variance of the top 10 principal components based on the gene expression was evaluated and plotted as scree plot (*Figure 6*).



*Figure 6. Scree plot of principal components.* The percentage of the explained variance (y-axis) based on gene expression data is given for the principal components 1 to 10 (x-axis). Principal components 1, 2 and 3 explain the main variance in the groups amounting to 69.9%.

The principal components 1, 2 and 3 are contrasted (*Figure 7*). A clear separation was seen particularly for principal component 1 and 2 representing in total 61.5% of variance in the dataset.



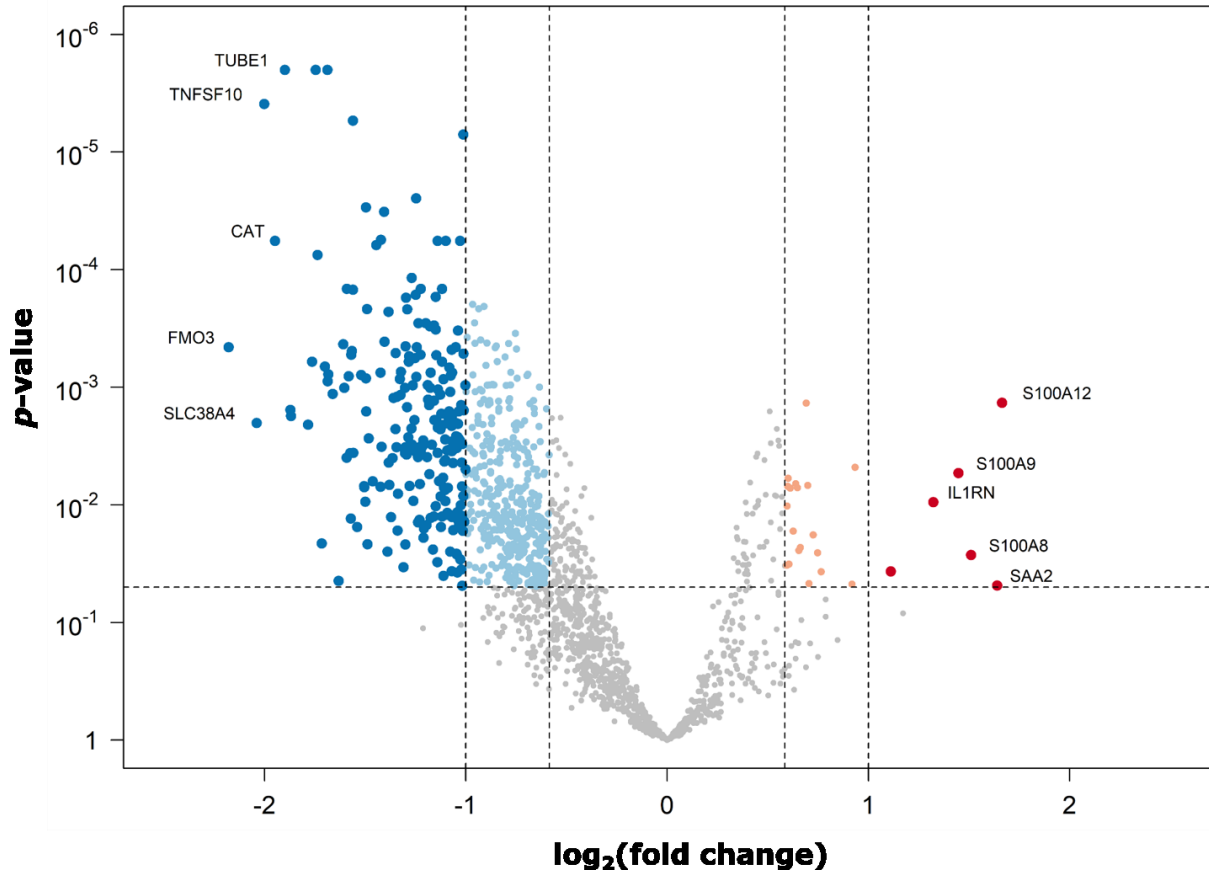
**Figure 7. Differences in gene expression profiles based on the principal component analysis.** The differences in gene expression profiles were indicated by the distance between the samples. Each dot represents one sample of the bipolar disorder (dark green) and control (light green) group representing the principal components (PC) of normalized and filtered  $\log_2$  gene expression values. In (A-C), two of first three principal components are plotted against each other. [Based on [90]]

Of the 54675 probes included in the preprocessing and normalization, 7280 probes remained after non-specific filtering and were used for the differential gene expression analysis.

### 3.3 Differential gene expression in bipolar disorder

Differential gene expression profiles were investigated by comparing non-specific filtered probes of the bipolar disorder group with the control group. Based on an adjusted  $p$ -value of  $<0.05$  and a fold change of  $>1.5$ , a total of 648 DEGs were obtained between samples with bipolar disorder and controls. Among those DEGs, only 25 genes were overexpressed, whereas 623 DEGs were found to be underexpressed. An extensive list of all DEGs with their annotation, as well as the corresponding fold change,  $\log_2$  fold change and the adjusted significance value is given in the Appendix (**Table A1**). The top five upregulated genes included three genes belonging to the S100 calcium binding protein family, namely S100 calcium binding protein A12 (*S100A12*), S100 calcium binding protein A9 (*S100A9*) and S100 calcium binding protein (*S100A8*), as well as the genes coding for the serum amyloid A2 (*SAA2*) and the interleukin-1 receptor antagonist (*IL1RN*). The top five downregulated genes identified were the flavin containing dimethylaniline monooxygenase 3 (*FMO3*), the solute carrier family 38 member 4 (*SLC38A4*), the TNF superfamily member 10 (*TNFSF10*), the catalase (*CAT*) and

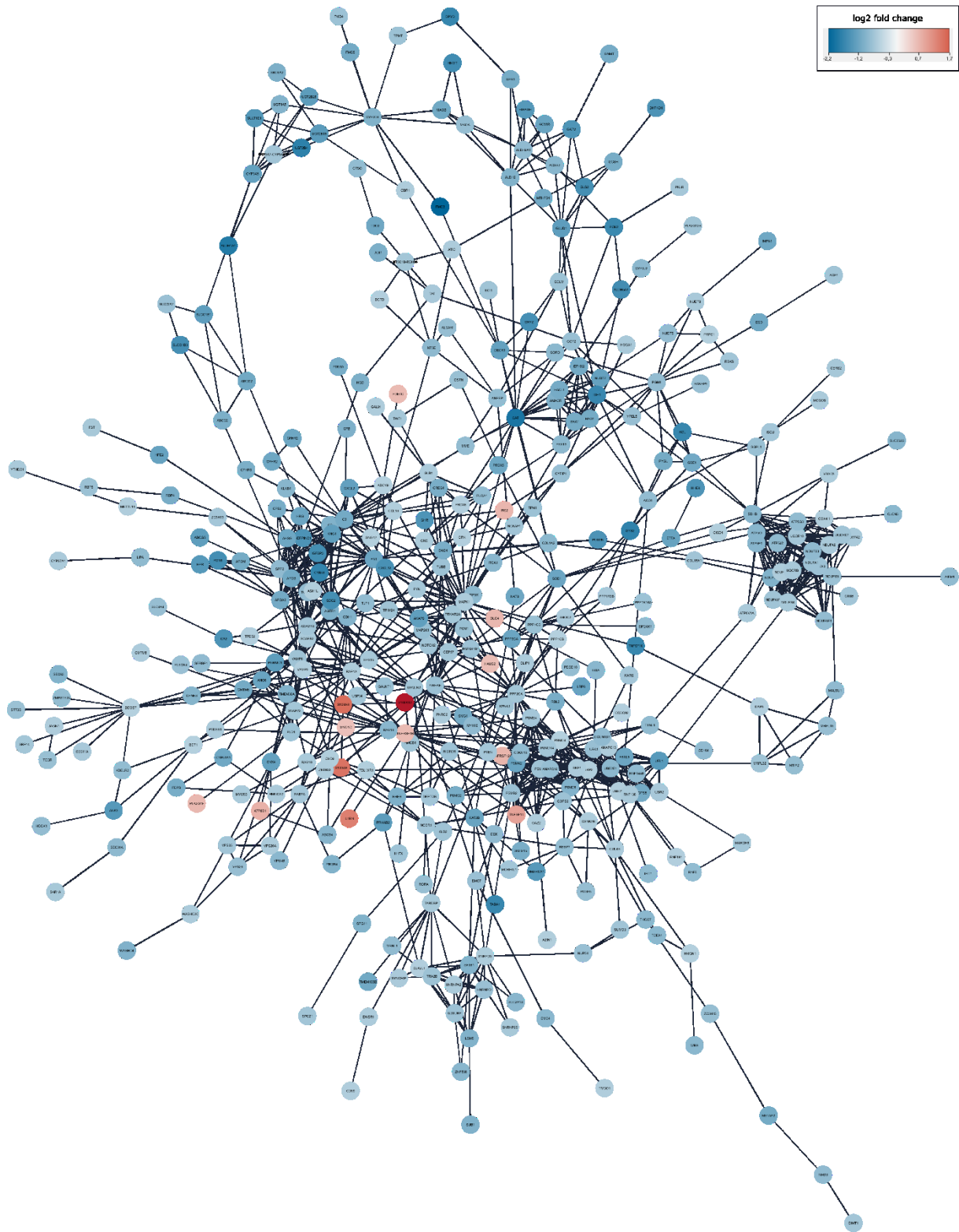
the tubulin epsilon 1 (*TUBE1*). The volcano plot depicts the fold change and the adjusted significance level of the DEGs (**Figure 8**).



**Figure 8. Volcano plot highlighting underexpressed and overexpressed DEGs in bipolar disorder.** For each DEG, the  $\log_2$  fold change (x-axis) is plotted against the corresponding adjusted  $p$ -value (y-axis). Thresholds for significance (adjusted  $p$ -value = 0.05; horizontal dashed line) and for expression (fold change >1.5 or >2; vertical dashed lines) are shown. Significant underexpressed genes with a fold change >1.5 or >2 are displayed as dots in light blue or dark blue, respectively. Significant overexpressed genes with a fold change of >1.5 or >2 are displayed as dots in light red or dark red, respectively. The five most underexpressed and overexpressed genes are labeled with their corresponding gene names. Gray dots represent genes with no significant difference. [Based on [90]]

### 3.4 Identified protein-protein interaction network

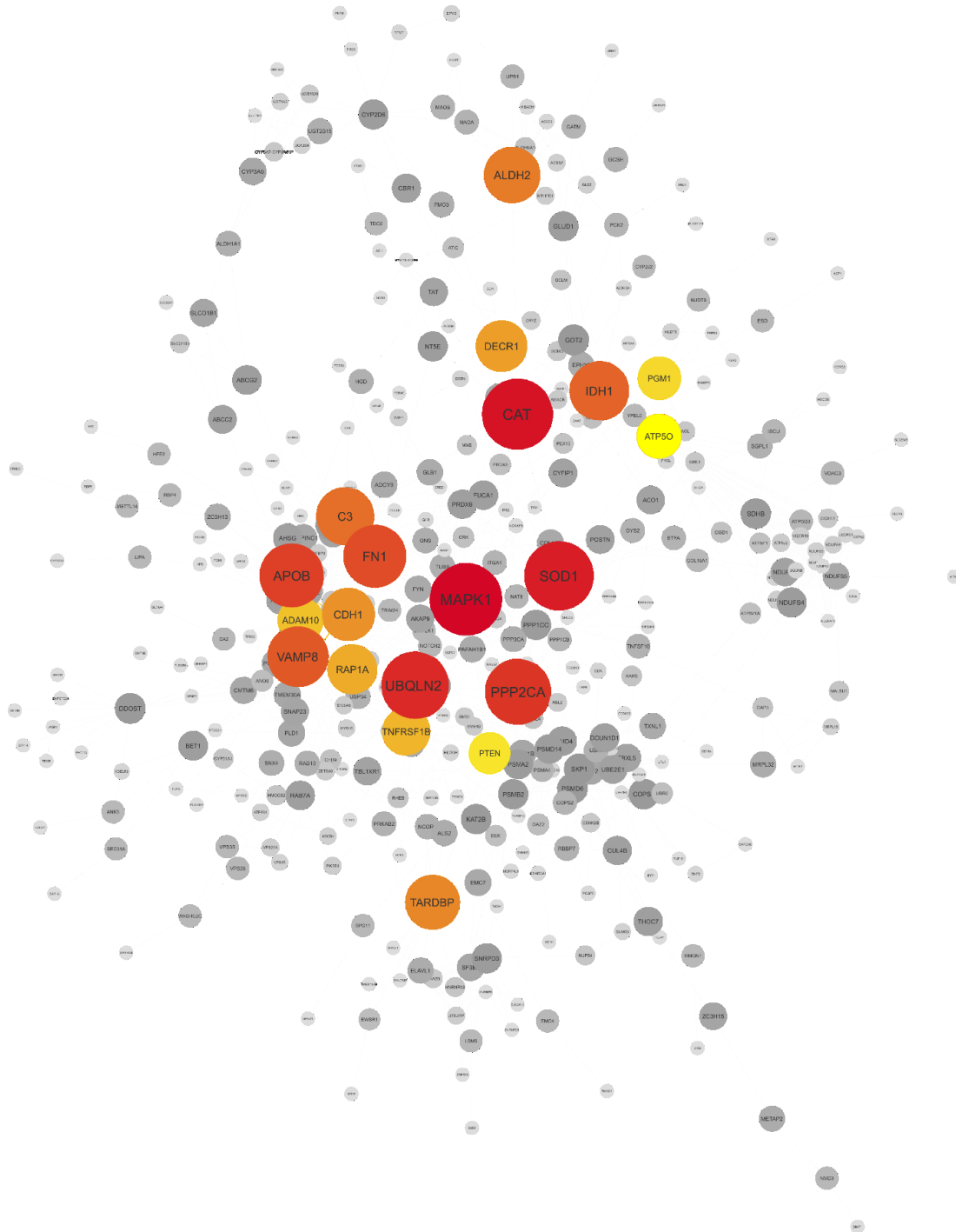
Underexpressed and overexpressed DEGs with reliable protein data were investigated using STRING. The protein-protein interaction network generated with a confidence level score of 0.7 included 610 nodes representing the proteins and 1229 edges representing interactions between proteins (**Figure 9**).



**Figure 9. Protein-protein interaction network of corresponding DEGs.** The overall protein-protein network generated by Cytoscape using the underexpressed and overexpressed DEGs with reliable protein data of the STRING database is displayed. The network included 610 connected nodes (proteins) and 1229 edges (interactions) shown as circles and lines, respectively. The coloring of the nodes corresponds to the  $\log_2$  fold change of the gene expression microarray analysis.

### 3.5 Determined key regulators of protein-protein interaction network

The top 20 hub proteins or key regulators of the overall protein-protein interaction network generated from the corresponding DEGs were identified and mapped (*Figure 10*).



**Figure 10. Hub proteins of the protein-protein interaction network.** The nodes layout including circle size and color correspond to the betweenness centrality values. The 20 hub proteins with the highest betweenness scores are shown in nuances from red to yellow and all other nodes in nuances of gray. Dark red indicates the highest betweenness score. Full names are listed in *Table A1* in the Appendix.

Those hub proteins indicating a pivotal role for the network's structure and function and their corresponding betweenness score were listed (*Table 5*). The three hubs with the highest degrees were identified as mitogen-activated protein kinase 1 (MAPK1), a crucial part of the MAP kinase signal transduction pathway involved in numerous biological cell functions, followed by CAT, a crucial antioxidant enzyme mitigating the harmful effect of hydrogen peroxide, and superoxide dismutase 1 (SOD1), an isoenzyme abolishing detrimental superoxide radicals.

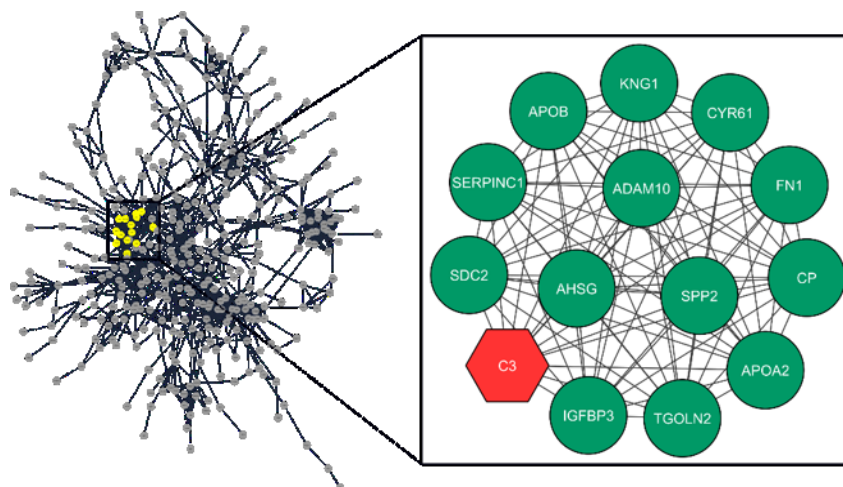
**Table 5. Hub proteins of the protein-protein interaction network.** The top 20 hub proteins with the corresponding betweenness scores are given.

Protein	Betweenness score	Protein	Betweenness score
MAPK1	26129.1	ALDH2	7611.9
CAT	21152.4	TARDBP	7560.8
SOD1	15946.9	CDH1	7523.5
UBQLN2	12262.7	DECRI	6808.1
PPP2CA	12249.0	RAP1A	6714.6
APOB	11510.8	TNFRSF1B	6657.2
FN1	10986.4	ADAM10	6581.8
VAMP8	9714.5	ATP5O	6523.2
IDH1	7953.7	PGM1	6433.7
C3	7821.5	PTEN	6227.6

*MAPK1* - mitogen-activated protein kinase 1; *CAT* - catalase; *SOD1* - superoxide dismutase 1; *UBQLN2* - ubiquilin 2; *PPP2CA* - protein phosphatase 2 catalytic subunit alpha; *APOB* - apolipoprotein B; *FN1* - fibronectin 1; *VAMP8* - vesicle associated membrane protein 8; *IDH1* - isocitrate dehydrogenase (NADP(+)) 1; *C3* - complement 3; *ALDH2* - aldehyde dehydrogenase 2 family member; *TARDBP* - TAR DNA binding protein; *CDH1* - cadherin 1; *DECRI* - 2,4-dienoyl-CoA reductase 1; *RAP1A* - RAP1A, member of RAS oncogene family; *TNFRSF1B* - TNF receptor superfamily member 1B; *ADAM10* - ADAM metalloproteinase domain 10; *ATP5O* - ATP synthase peripheral stalk subunit OSCP; *PGM1* - phosphoglucomutase 1; *PTEN* - phosphatase and tensin homolog;

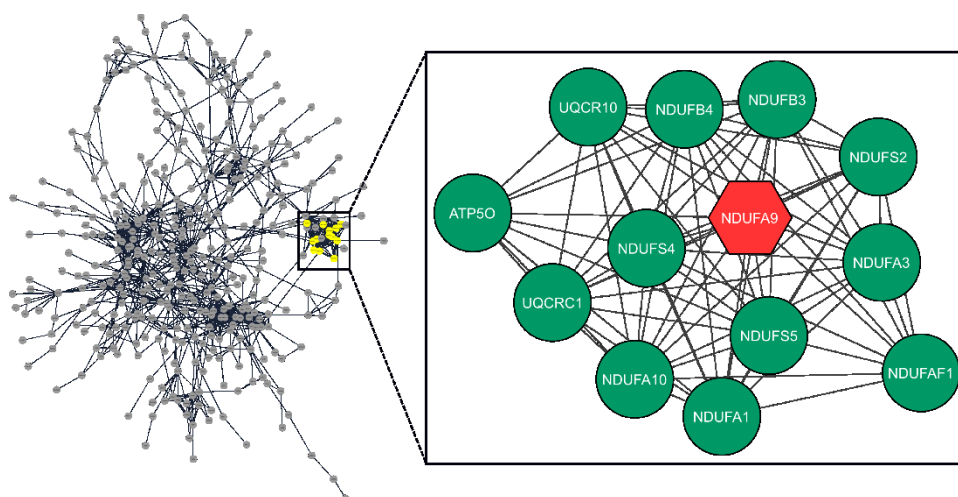
### 3.6 Identified modules of protein-protein interaction network

Functional module analysis of the protein-protein interaction network using ClusterViz and the MCODE algorithm was performed and identified 8 modules with a MCODE score higher than 4. Module 1 with a MCODE score of 14.00 consisted of 14 nodes and 91 edges, whereas complement 3 (C3) was identified as the seed, classified as the highest-weighted node (*Figure 11*).



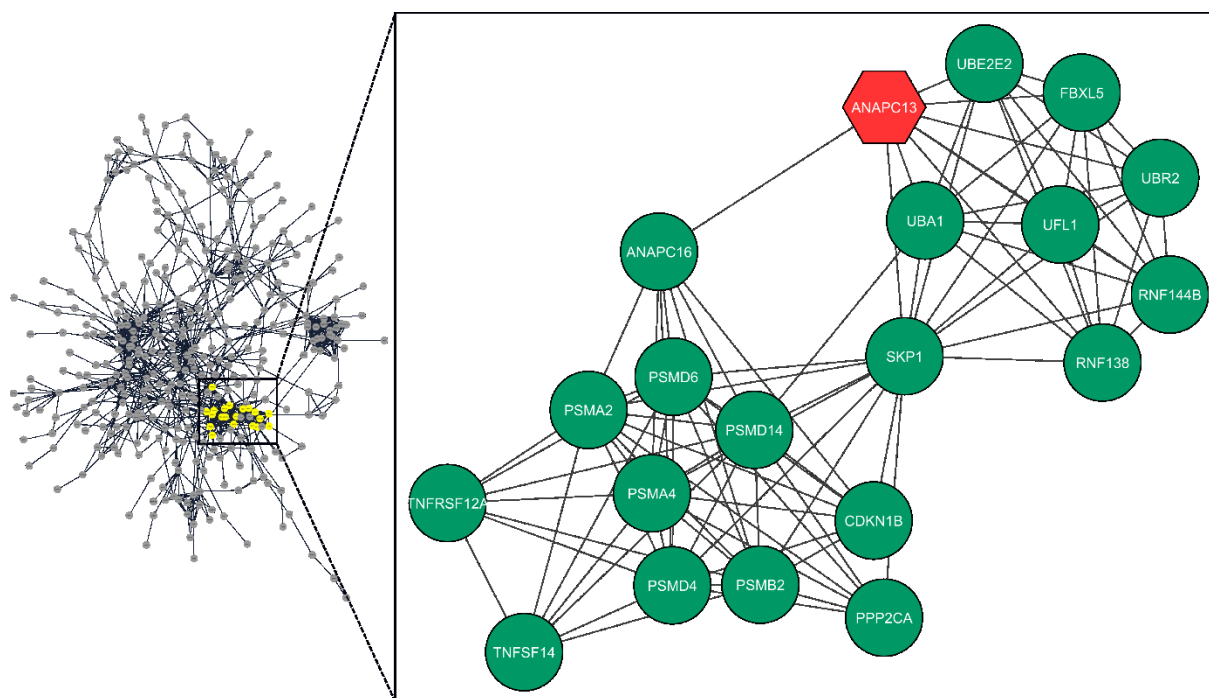
**Figure 11. Module 1 of the protein-protein interaction network.** Module 1 included 14 nodes and 91 edges and reach a score of 14. The seed and the other proteins included in the module are shown as red hexagon and green ellipses, respectively. The connecting lines indicate the edges representing protein interactions between the nodes. Full names are given in **Table A1** in the Appendix.

Module 2 with a MCODE score of 12.17 consisted of 13 nodes and 73 edges, where NADH:ubiquinone oxidoreductase subunit A9 (NDUFA9), a complex I subunit contributing to the respiratory chain, was determined as seed (**Figure 12**).



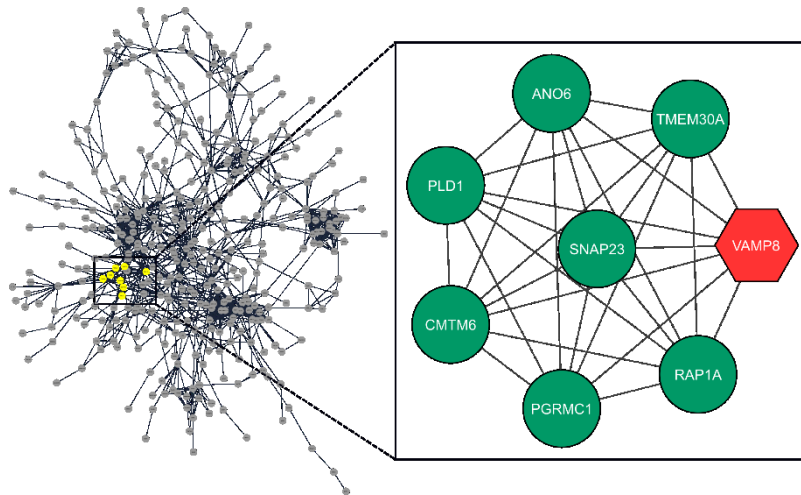
**Figure 12. Module 2 of the protein-protein interaction network.** Module 2 included 13 nodes and 73 edges and reach a score of 12.2. The seed protein and the other proteins included in the module are shown as red hexagon and green ellipses, respectively. The connecting lines indicate the edges representing protein interactions between the nodes. Full names are given in **Table A1** in the Appendix.

Module 3 with a MCODE score of 9.79 comprised 20 nodes and 93 edges (**Figure 13**). The seed in module 3 was identified as anaphase promoting complex subunit 13 (ANAPC13), a component of an ubiquitin ligase in the anaphase promoting complex regulating cell cycle progression.



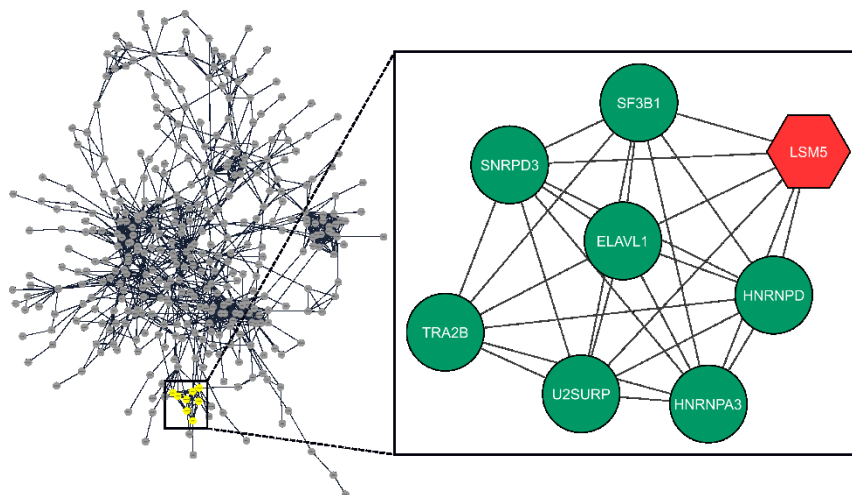
**Figure 13. Module 3 of the protein-protein interaction network.** Module 3 included 20 nodes and 93 edges and reach a score of 9.8. The seed protein and the other proteins included in the module are shown as red hexagon and green ellipses, respectively. The connecting lines indicate the edges representing protein interactions between the nodes. Full names are given in **Table A1** in the Appendix.

Module 4 with a MCODE score of 8.00 consisting of 8 nodes and 28 edges, whereas vesicle associated membrane protein 8 (VAMP8) was determined as seed, a membrane protein involved in the fusion process of cellular membranes such as synaptic vesicles (**Figure 14**).



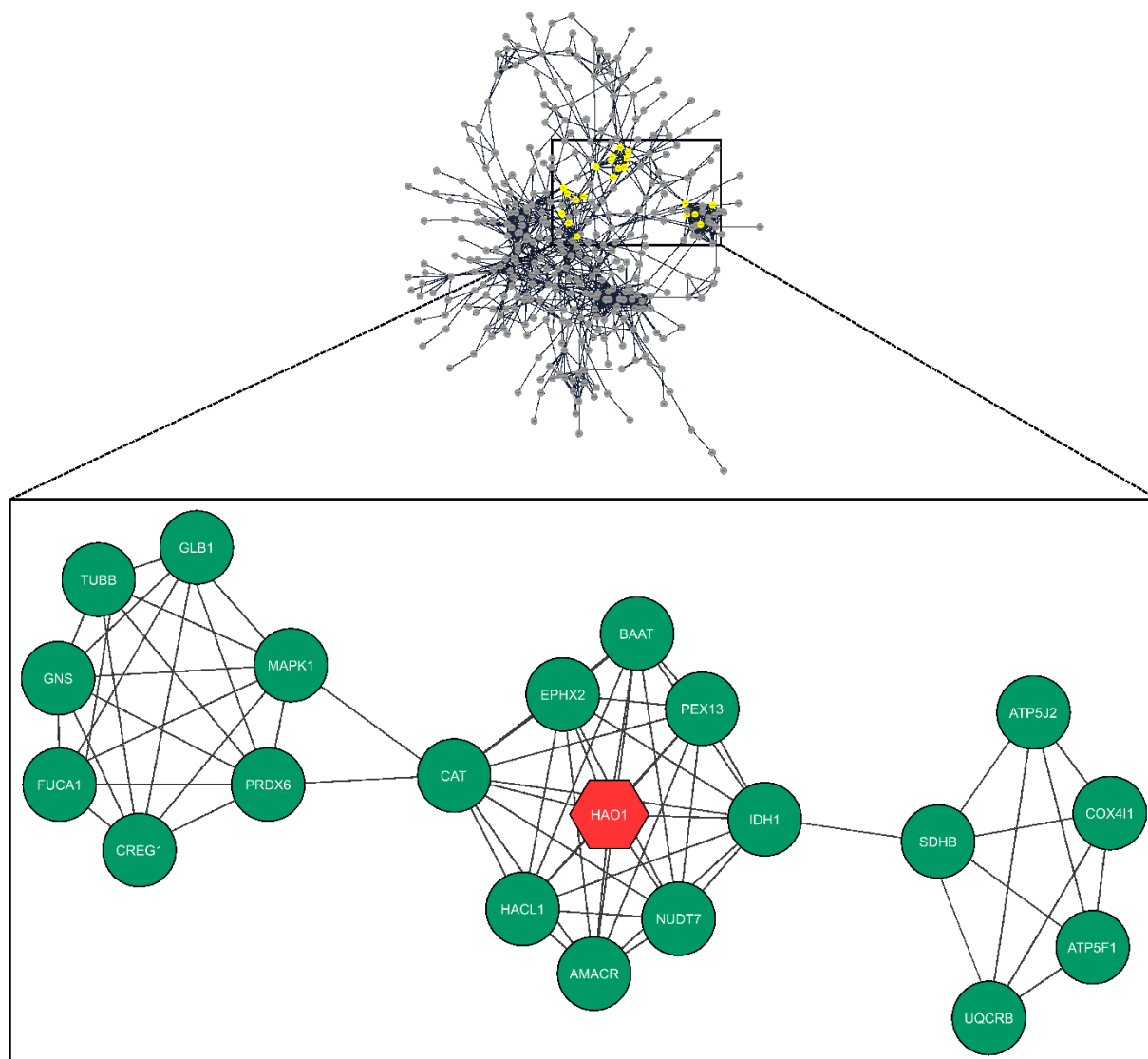
**Figure 14. Module 4 of the protein-protein interaction network.** Module 4 included 8 nodes and 28 edges and reach a score of 8.0. The seed protein and the other proteins included in the module are shown as red hexagon and green ellipses, respectively. The connecting lines indicate the edges representing protein interactions between the nodes. Full names are given in *Table A1* in the Appendix.

Module 5 with a MCODE score of 7.71 consisted of 8 nodes and 27 edges, and LSM5 homolog, U6 small nuclear RNA and mRNA degradation associated (LSM5), a component of a snRNP complex involved in spliceosome, was representing the seed (**Figure 15**).



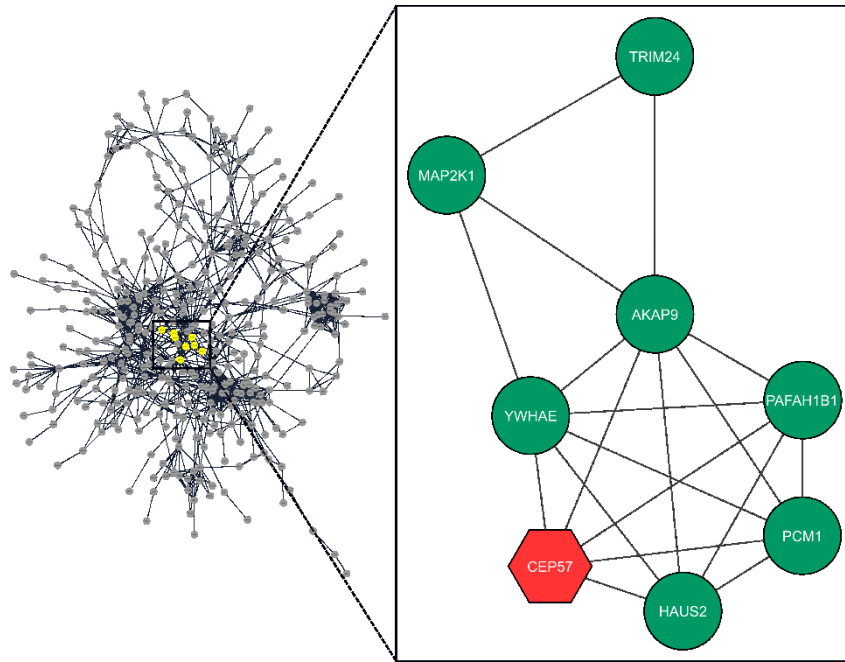
**Figure 15. Module 5 of the protein-protein interaction network.** Module 5 included 8 nodes and 27 edges and reach a score of 7.7. The seed protein and the other proteins included in the module are shown as red hexagon and green ellipses, respectively. The connecting lines indicate the edges representing protein interactions between the nodes. Full names are given in *Table A1* in the Appendix.

Module 6 with a MCODE score of 7.00 comprised 21 nodes and 70 edges, whereas hydroxyacid oxidase 1 (HAO1), an oxidase associated with glycolate and fatty acid oxidase activity, was revealed as seed (*Figure 16*).



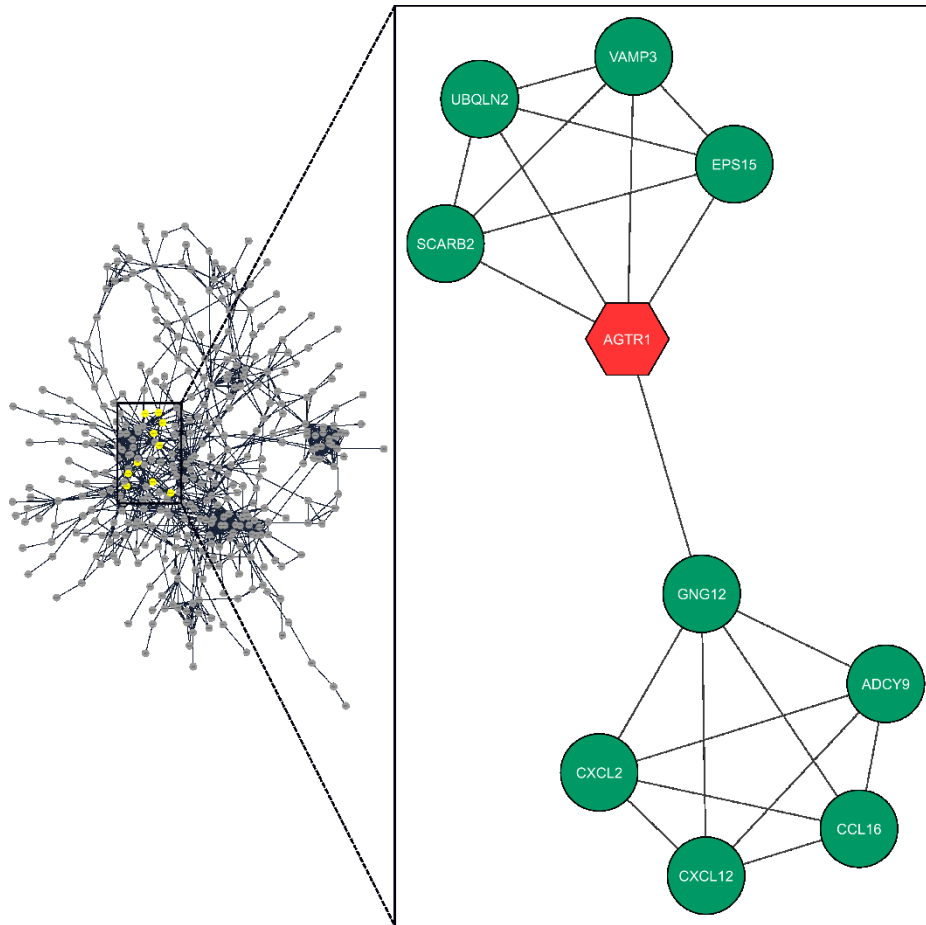
**Figure 16. Module 6 of the protein-protein interaction network.** Module 6 included 21 nodes and 70 edges and reach a score of 7.0. The seed protein and the other proteins included in the module are shown as red hexagon and green ellipses, respectively. The connecting lines indicate the edges representing protein interactions between the nodes. Full names are given in *Table A1* in the Appendix.

Module 7 with a MCODE score of 5.43 contained 8 nodes and 19 edges, and as seed centrosomal protein 57 (CEP57), which is supposed to interact with microtubules for their attachment to the centrosomes, was found (*Figure 17*).



**Figure 17. Module 7 of the protein-protein interaction network.** Module 7 included 8 nodes and 19 edges and reach a score of 5.4. The seed protein and the other proteins included in the module are shown as red hexagon and green ellipses, respectively. The connecting lines indicate the edges representing protein interactions between the nodes. Full names are given in **Table A1** in the Appendix.

Module 8 with a MCODE score of 4.67 consisted of 10 nodes and 21 edges, whereas angiotensin II receptor type 1 (AGTR1), which is involved in a G-protein mediated phosphatidylinositol-calcium second messenger system, was revealed as seed (**Figure 18**).



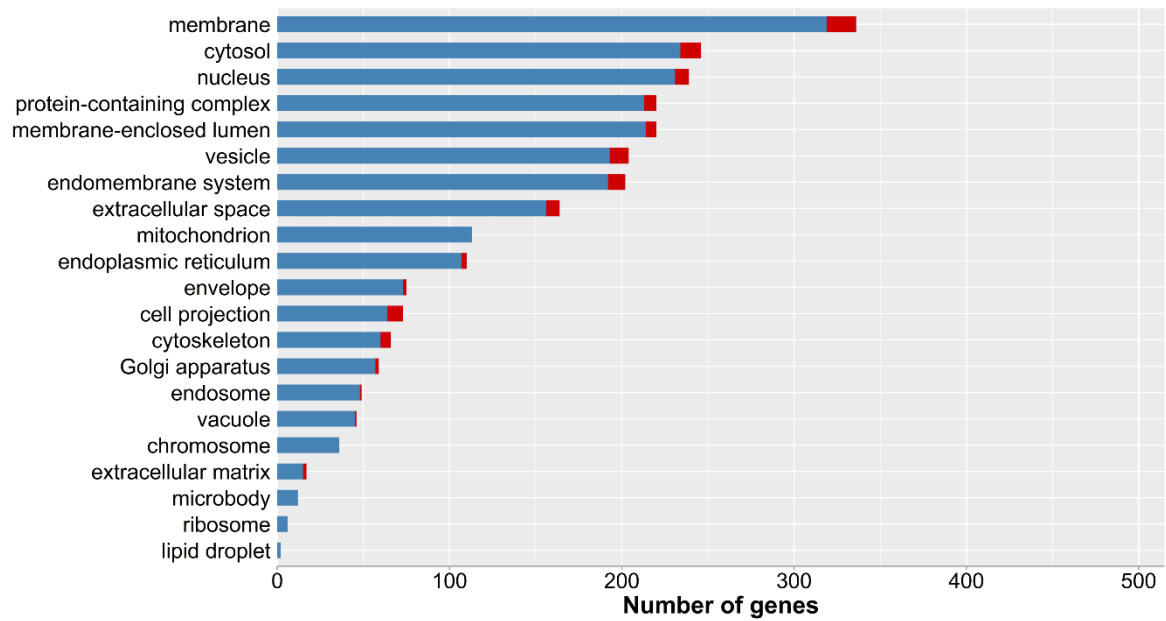
**Figure 18. Module 8 of the protein-protein interaction network.** Module 8 included 10 nodes and 21 edges and reach a score of 4.7. The seed protein and the other proteins included in the module are shown as red hexagon and green ellipses, respectively. The connecting lines indicate the edges representing protein interactions between the nodes. Full names are given in *Table A1* in the Appendix.

## 3.7 Functional enrichment analysis

### 3.7.1 Determination of gene ontology terms

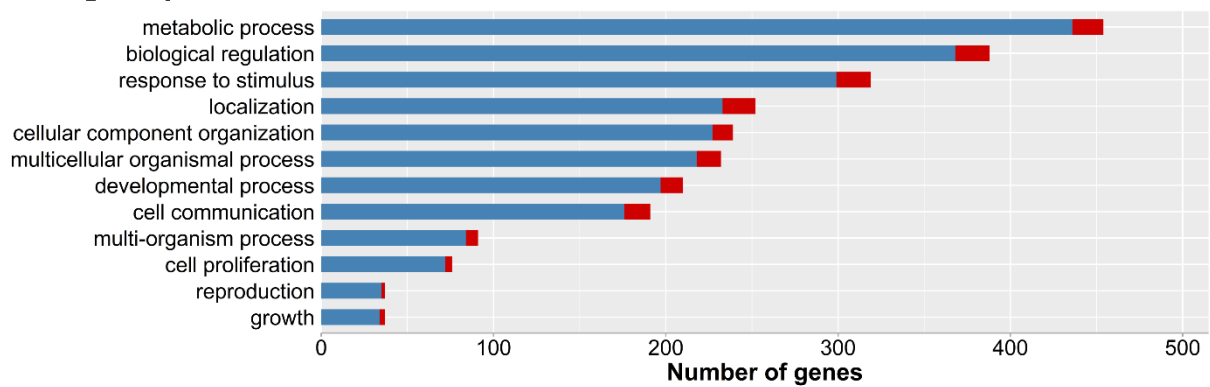
The GO terms for cellular components (*Figure 19*), biological processes (*Figure 20*) and molecular functions (*Figure 21*) were assessed for the 648 DEGs. With regard to the cellular component, the majority of the DEGs was present at the membrane followed by the cytosol and the nucleus. Most of the underexpressed DEGs were involved in metabolic process, biological regulation and stimulus response with regard to biological processes. In relation to the molecular function, upregulated DEGs and a large number of downregulated DEGs were predominantly involved in protein and ion binding.

## Cellular component



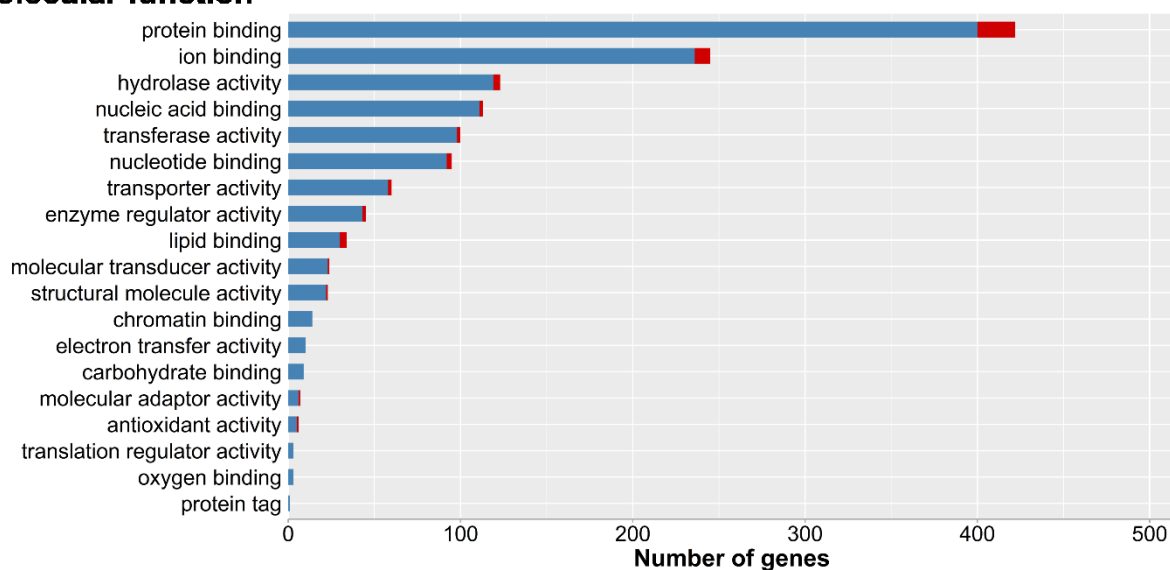
**Figure 19. Ranking of the cellular component terms based on the GO analysis.** The portion of under- and overexpressed genes based on the microarray analysis are shown in blue and red, respectively.

## Biological process



**Figure 20. Ranking of biological process terms based on the GO analysis.** The portion of under- and overexpressed genes based on the microarray analysis are shown in blue and red, respectively.

## Molecular function



**Figure 21. Ranking of molecular function terms based on the GO analysis.** The portion of under- and overexpressed genes based on the microarray analysis are shown in blue and red, respectively.

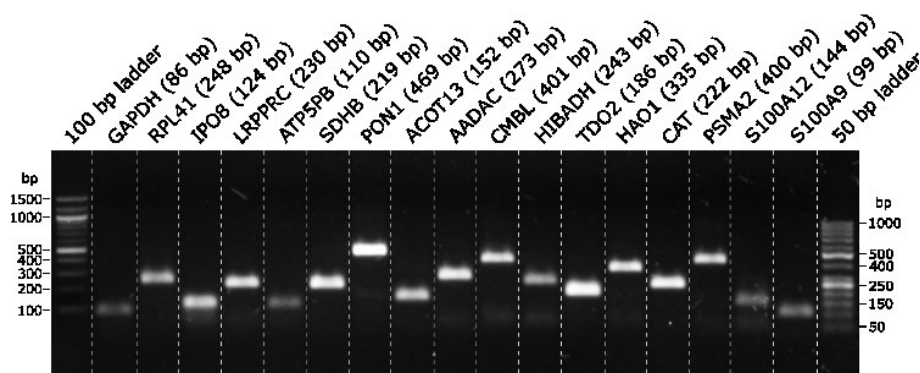
### 3.7.2 Identification of over-represented functional pathways

The over-represented functional pathway analysis revealed nine functional pathways included in the Reactome database with statistical significance. A marked over-representation of DEGs was present for the respiratory electron transport chain (RETC) involving four pathways. The four pathways were as follows: Respiratory electron transport, ATP synthesis by chemiosmotic coupling, and heat production by uncoupling proteins (R-HSA-163200, number of DEGs in pathway: 21, FDR = <0.001); Respiratory electron transport (R-HSA-611105, number of DEGs in pathway: 17, FDR = <0.001); The citric acid cycle and respiratory electron transport (R-HSA-1428517, number of DEGs in pathway: 22, FDR = <0.001); and Complex I biogenesis (R-HSA-6799198, number of DEGs in pathway: 10, FDR = <0.014). Other significantly over-represented pathways were found to be involved in the innate immune system with the pathway Neutrophil degranulation (R-HSA-6798695, number of DEGs in pathway: 42, FDR = <0.001), in the lipid metabolism with the pathway Metabolism of lipids and lipoproteins (R-HSA-556833, number of DEGs in pathway: 21, FDR = <0.001), in the biotransformation of endogenous compounds and xenobiotics with the pathway Biological oxidations (R-HSA-211859, number of DEGs in pathway: 21, FDR = <0.015), and the cellular amino acid and polyamine metabolism with the pathways Metabolism of amino acids and derivatives (R-HSA-

71291, number of DEGs in pathway: 21, FDR = <0.020) and Regulation of ornithine decarboxylase (R-HSA-350562, number of DEGs in pathway: 21, FDR = <0.027). The pathway allocation of the corresponding DEGs in the pathways is given in the Appendix (*Table A1*).

### 3.8 Gene expression validation of microarray data

The expression of 14 genes were evaluated in bipolar disorder and the controls to validate the microarray results. The amplicon size of the endogenous controls and each validation gene were verified by agarose gel electrophoresis (*Figure 22*). The gene specific amplicons showed the expected size and the absence of unspecific products.

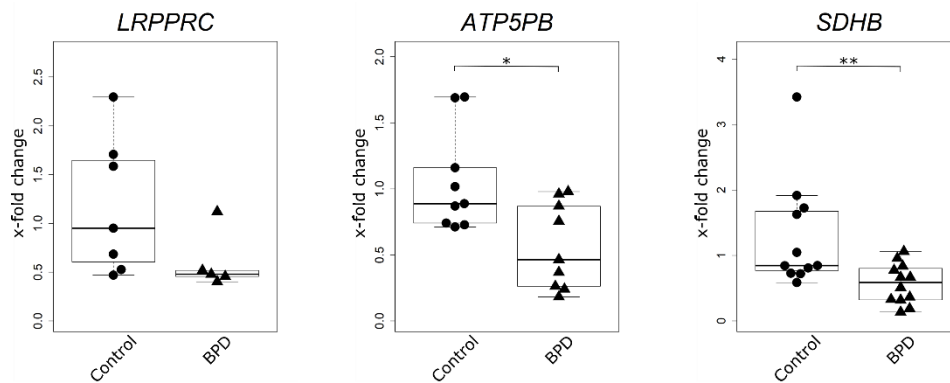


**Figure 22. Verification of gene specific products by amplicon size determination.** Gene specific amplicons were separated by gel electrophoresis using a 2% agarose gel. Each lane corresponds to one gene of interest. Gene name and amplicon size are given above the image. Molecular sizes of the 100 and 50 bp ladder are indicated on the left and right side of the image, respectively. Full names are given in *Table A1* in the Appendix. [Published in [90]]

As endogenous controls, *RPL41* and *IPO8* were used. Gene expression of *GAPDH* was not properly detected in all tested samples and therefore excluded as endogenous control gene. Despite the variation in the differential expression between bipolar disorder and controls, both, microarray and quantitative real-time PCR analyses showed comparable gene expression in most of the examined genes. A statistical significant difference in gene expression between bipolar disorder and controls was confirmed for 11 of the tested genes.

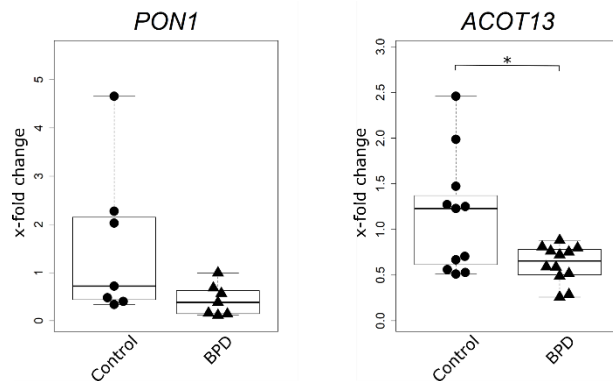
For the enriched RETC pathway, the genes *LRPPRC*, *ATP5PB* and *SDHB* were validated (*Figure 23*). No difference was revealed for *LRPPRC* between the groups. The genes *ATP5PB*

( $p = 0.032$ ) and *SDHB* ( $p = 0.007$ ) were significantly underexpressed in bipolar disorder. A reduction of 46.5% and 56.3% was observed for *ATP5PB* and *SDHB*, respectively.



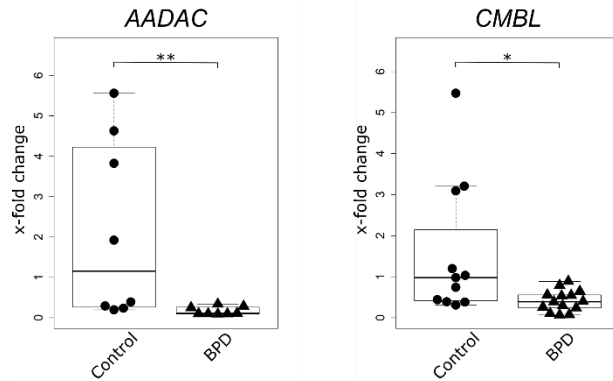
**Figure 23. Validation of selected DEGs allocated to the RETC pathway.** Relative gene expression is given as x-fold change assessed by sample triplicates and normalized using *RPL41* and *IPO8* as endogenous controls for the bipolar disorder (BPD) and control (Control) samples. Statistical analyses: Mann Whitney U test; \*  $p < 0.05$ , \*\*  $p < 0.01$  [Published in [90]]

The genes *PON1* and *ACOT13* allocated to the metabolism of lipids and lipoproteins were examined (**Figure 24**). For *PON1*, a difference was absent between the groups. The gene *ACOT13* ( $p = 0.013$ ) was significantly downregulated by 45.9% in bipolar disorder.



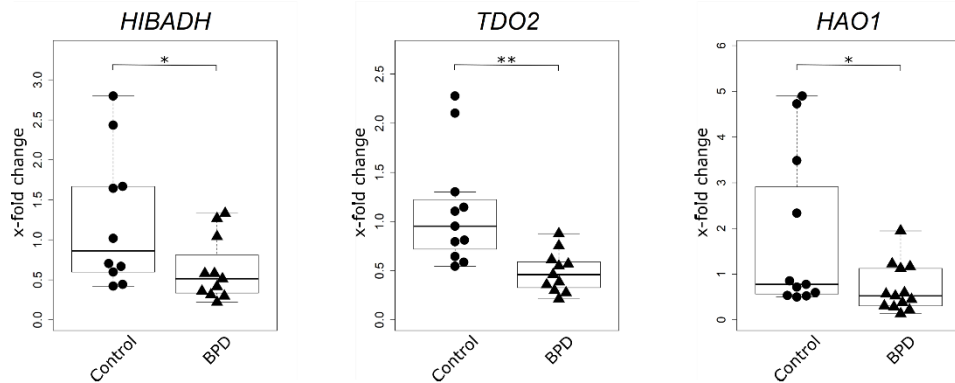
**Figure 24. Validation of selected DEGs allocated to the metabolism of lipids and lipoproteins.** Relative gene expression is given as x-fold change assessed by sample triplicates and normalized using *RPL41* and *IPO8* as endogenous controls for the bipolar disorder (BPD) and control (Control) samples. Statistical analyses: Mann Whitney U test (*PON1*) and unpaired t-test (*ACOT13*); \*  $p < 0.05$  [Published in [90]]

For the enriched pathway of biological oxidations, the gene expression of *AADAC* and carboxymethylenebutenolidase homolog (*CMBL*) were investigated (**Figure 25**). A statistically significant lower expression for both genes, *AADAC* ( $p = 0.007$ ) and *CMBL* ( $p = 0.013$ ), was shown. The reduction accounted for 92.0% and 73.6%, respectively.



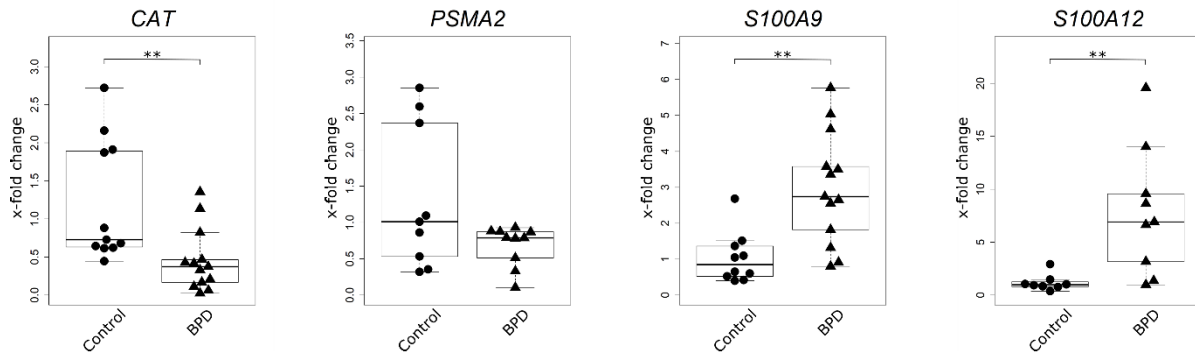
**Figure 25. Validation of selected DEGs allocated to biological oxidations.** Relative gene expression is given as x-fold change assessed by sample triplicates and normalized using *RPL41* and *IPO8* as endogenous controls for the bipolar disorder (BPD) and control (Control) samples. Statistical analyses: Mann Whitney U test; \*  $p < 0.05$ , \*\*  $p < 0.01$  [Published in [90]]

Diminished gene expression of the three tested genes related to the amino acid metabolism, *HIBADH* ( $p = 0.032$ ), *TDO2* ( $p = 0.002$ ) and *HAO1* ( $p = 0.037$ ) was observed with a decrease of 49.3%, 56.4% and 62.5%, respectively (**Figure 26**).



**Figure 26. Validation of selected DEGs allocated to the metabolism of amino acids.** Relative gene expression is given as x-fold change assessed by sample triplicates and normalized using *RPL41* and *IPO8* as endogenous controls for the bipolar disorder (BPD) and control (Control) samples. Statistical analyses: Mann Whitney U test; \*  $p < 0.05$ , \*\*  $p < 0.01$  [Published in [90]]

For the enriched pathway of neutrophil degranulation, the genes *CAT*, *PSMA2*, *S100A9* and *S100A12* were examined (**Figure 27**). The gene expression of *CAT* decreased by 62.5% ( $p = 0.004$ ) and increased for *S100A9* ( $p = 0.003$ ) and *S100A12* ( $p = 0.003$ ) by 189.1% and 574.1%, respectively, in the bipolar disorder samples compared to the controls. *PSMA2* gene expression did not differ between the groups.

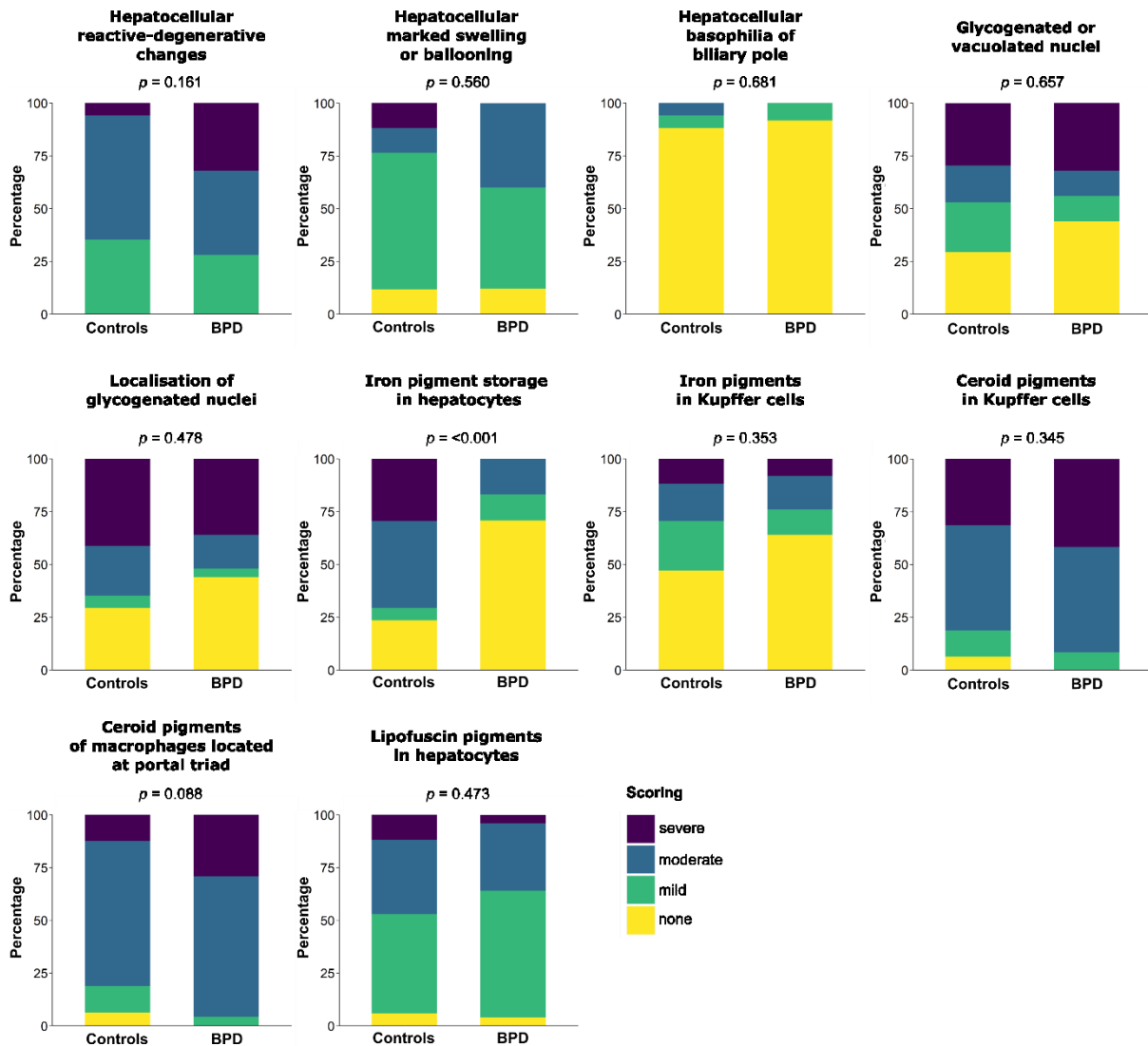


**Figure 27. Validation of selected DEGs allocated to neutrophil degranulation.** Relative gene expression is given as x-fold change assessed by sample triplicates and normalized using *RPL41* and *IPO8* as endogenous controls for the bipolar disorder (BPD) and control (Control) samples. Statistical analyses: Mann Whitney U test; \*\*  $p = <0.01$  [Published in [90]]

### 3.9 Frequency distributions of histopathological features in the liver

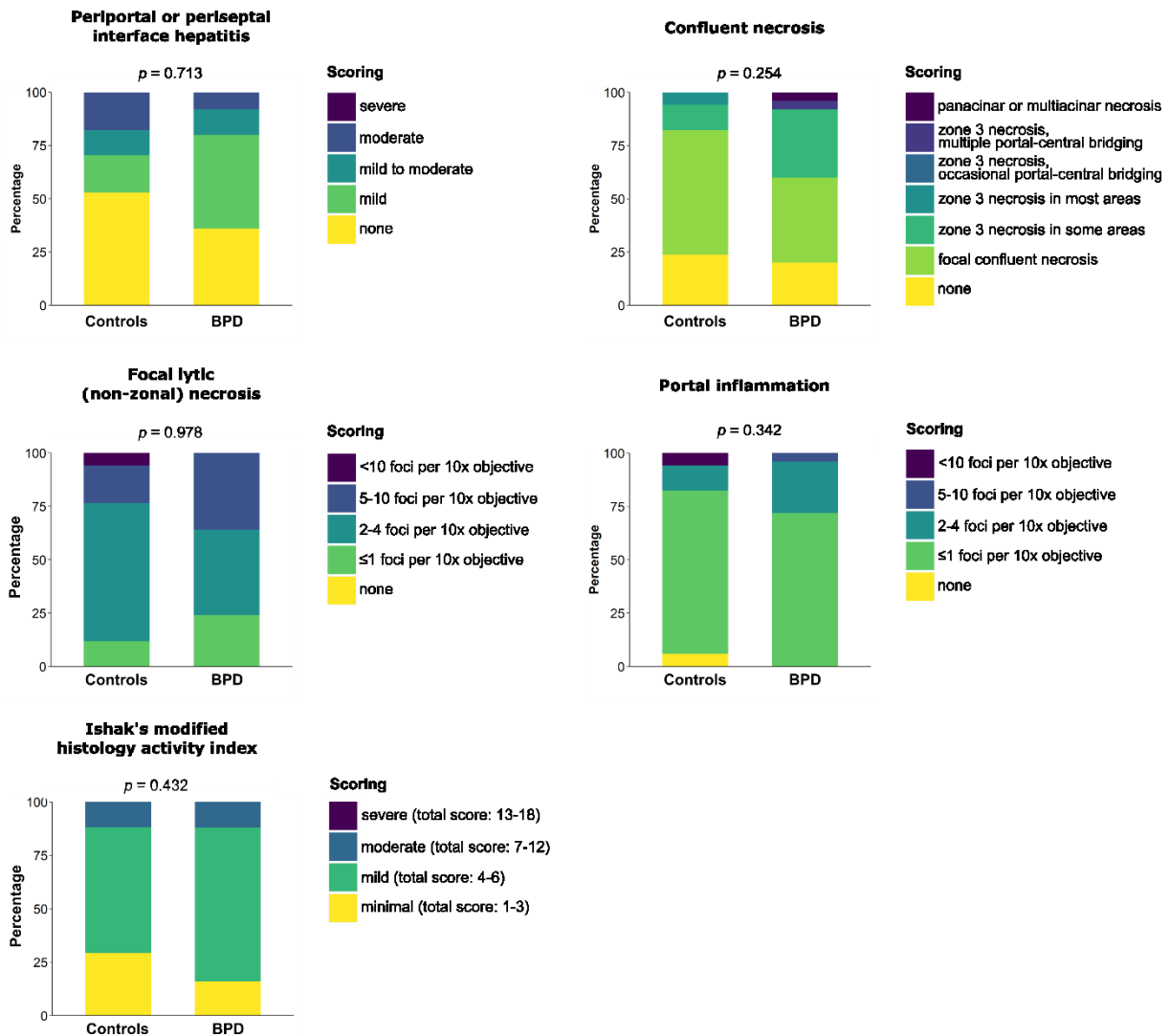
The presence of steatosis in liver tissue did not differ between patients with bipolar disorder and controls, whereby all bipolar disorder samples and 94.1% of the controls showed steatosis ( $p = 0.405$ ). The bipolar disorder group revealed a considerably higher percentage of steatohepatitis compared to the controls (68.0% and 35.3%, respectively), however, statistical significance was absent ( $p = 0.059$ ).

The assessed histopathological liver features were assigned to six categories. The first category termed as degenerative changes comprised features of reactive-degenerative changes, hepatocellular ballooning, hepatocellular accumulation of rough endoplasmic reticulum at the apical pole also known as basophilia of the biliary pole, as well as glycogen inclusions of the nucleus, and cytoplasmic inclusion and deposits of metals, such as iron, and endogenous pigments, such as ceroid and lipofuscin both formed by the degradation of lipids. The corresponding frequency distributions of histopathological features in this category were assessed (**Figure 28**). Comparable frequency distributions were observed for most of the features. The bipolar disorder group revealed a less severe iron pigment storage compared to the controls, where the majority of the control samples had moderate (41.2%) to severe (29.4%) stored iron pigments in the parenchymal cells.



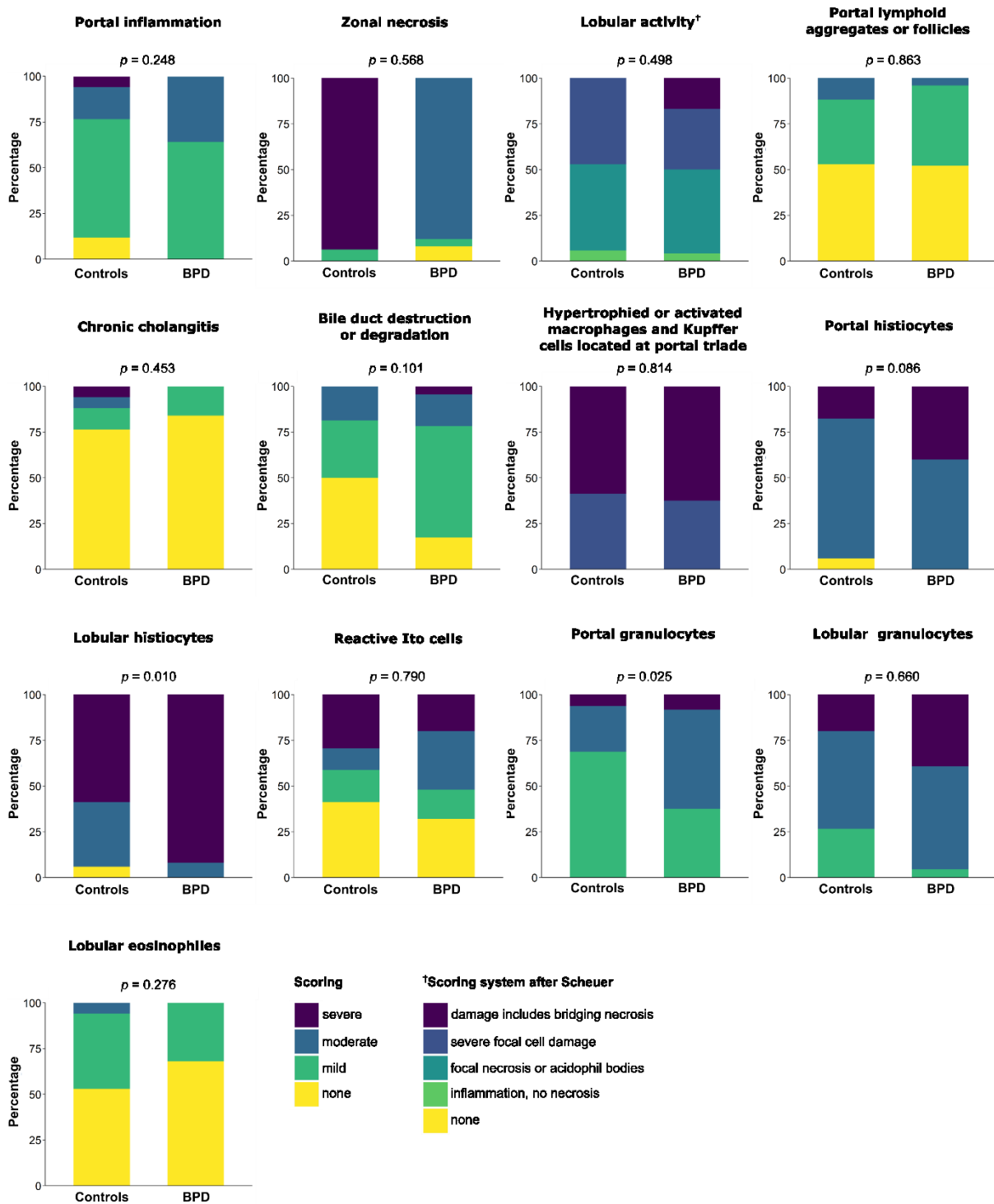
**Figure 28.** Frequency distribution of assessed histopathological features categorized to degenerative changes. The percentage of the respective histopathological feature score for the patients with bipolar disorder (BPD) and the control group (Controls) are given. Statistical analyses: Mann Whitney U test; [Based on [90]]

The second category comprised histopathological features assigned to hepatic reactivity. Frequency distributions describing the pathology of necrosis and inflammation according to the Ishak’s modified histological activity index [111] including the presence of interface hepatitis, confluent and focal necroses and inflammation at portal areas were assessed (**Figure 29**). No statistically significant frequency distribution differences for all features were observed between the groups.



**Figure 29. Frequency distribution of assessed histopathological features categorized to hepatic reactivity.** The percentage of the respective histopathological feature score for the patients with bipolar disorder (BPD) and the control group (Controls) are given. Statistical analyses: Mann Whitney U test; [Based on [90]]

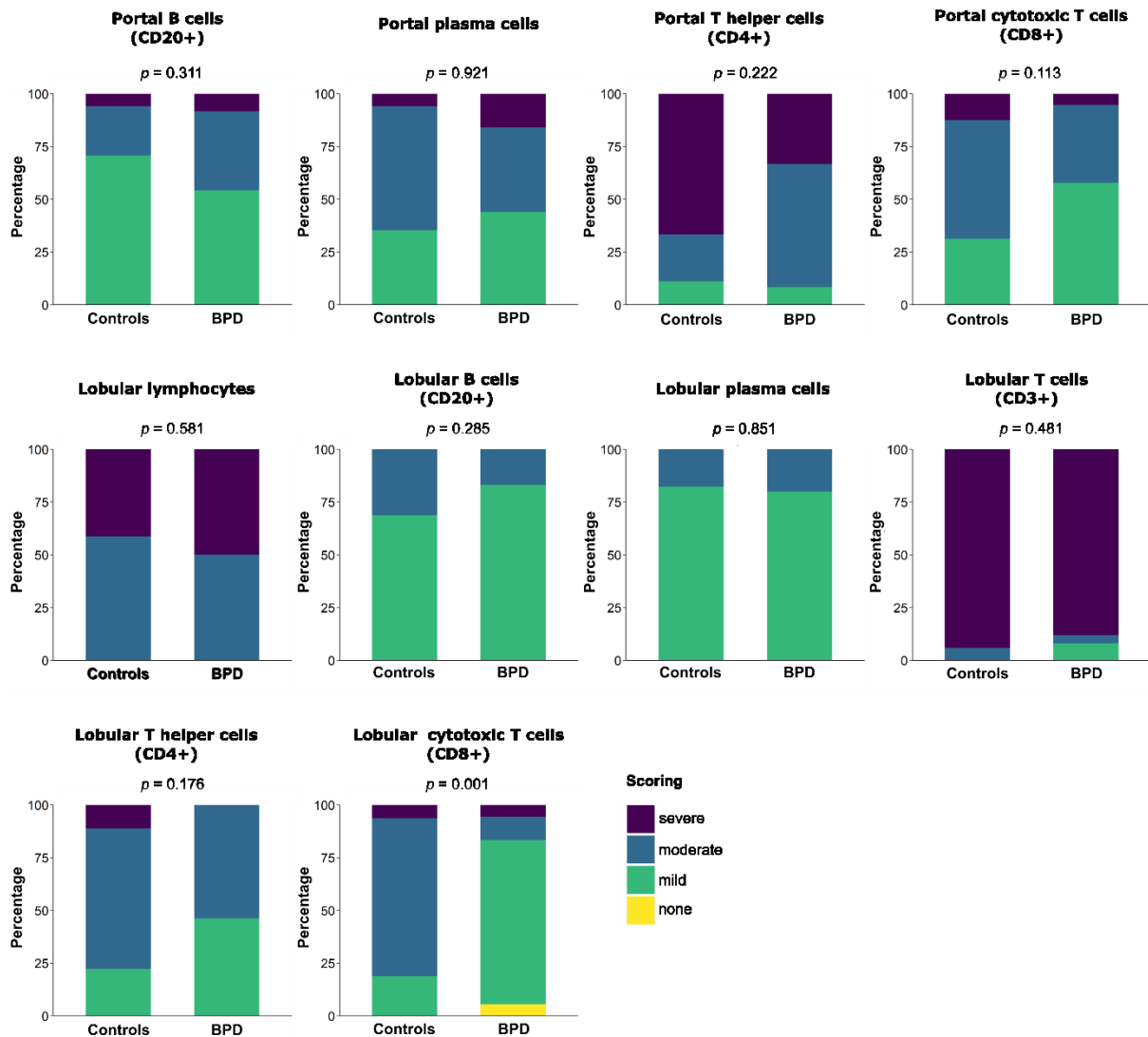
Frequency distributions of histopathological features of inflammation were assessed including features such as the presence of portal inflammation, zonal necrosis, lobular activity according to Scheurer [112], and lymph follicles, as well as the presence of chronic cholangitis and bile duct injury, the accumulation of immune cells such as macrophages including Kupffer cells and histiocytes, as well as hepatic stellate cells termed as Ito cells and granulocytes (**Figure 30**). The majority of bipolar disorder samples, 92.0%, revealed severe histiocytes presence in lobular areas, compared to 58.8% of the controls ( $p = 0.10$ ). The presence of granulocytes in portal areas was more dominant in liver samples of patients with bipolar disorder ( $p = 0.025$ ).



**Figure 30.** Frequency distribution of assessed histopathological features categorized to inflammation, part I. The percentage of the respective histopathological feature score for the patients with bipolar disorder (BPD) and the control group (Controls) are given. Statistical analyses: Mann Whitney U test; [Based on [90]]

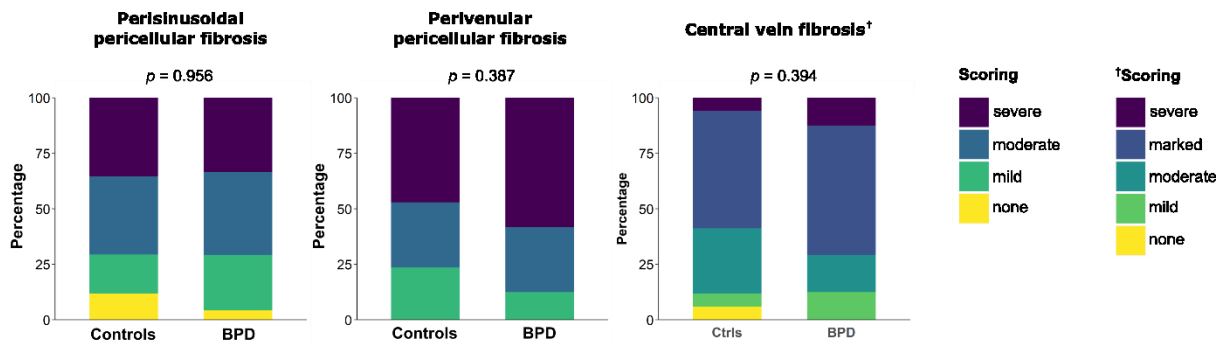
Further, the portal and lobular appearance of lymphocytes was assessed (**Figure 31**). Mainly, none or mild appearance of cytotoxic T-lymphocytes ( $CD8^+$ ) in lobular areas was observed in

the group of bipolar disorder (83.4%) as compared to a moderate to severe appearance in 81.3% of the controls ( $p = 0.001$ ). None of the other assessed inflammation features revealed statistical significant difference of the frequency distributions.



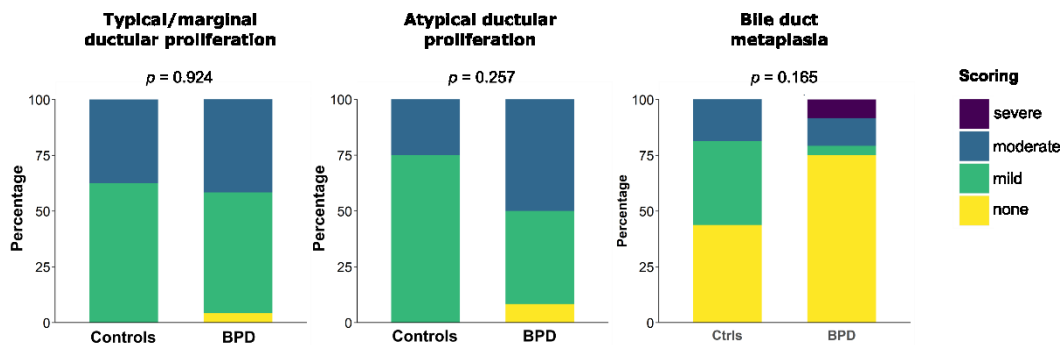
**Figure 31. Frequency distribution of assessed histopathological features categorized to inflammation, part II.** The percentage of the respective histopathological feature score for the patients with bipolar disorder (BPD) and the control group (Controls) are given. Statistical analyses: Mann Whitney U test; [Based on [90]]

The histopathological features of fibrosis were assessed with respect to their frequency distribution for the presence of fibrosis in perisinoidal and perivenular areas, as well as the appearance of fibrosis at areas surrounding the central vein (*Figure 32*). A statistical significant difference in these features was absent.



**Figure 32.** Frequency distribution of assessed histopathological features categorized to fibrosis. The percentage of the respective histopathological feature score for the patients with bipolar disorder (BPD) and the control group (Controls) are given. Statistical analyses: Mann Whitney U test; [Based on [90]]

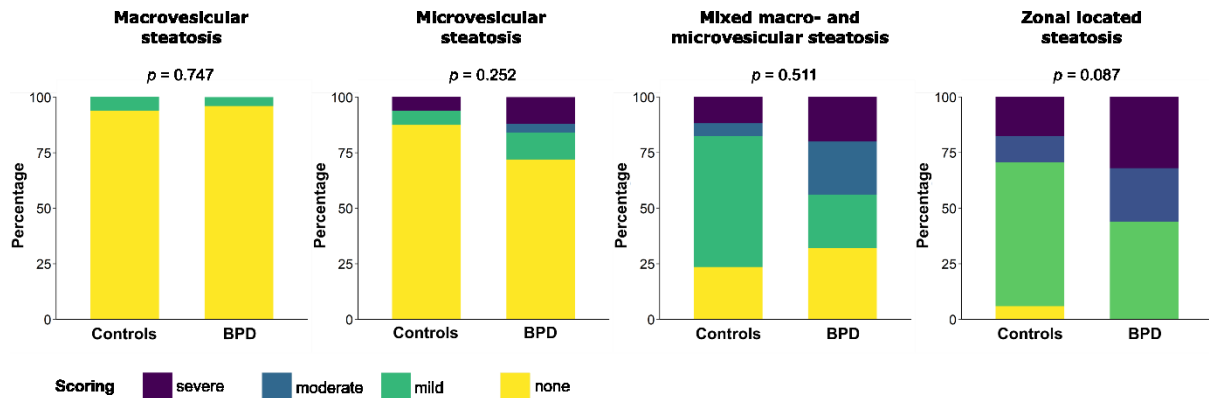
Frequency distributions of the category of endogenous reaction describing a possible cholestatic reaction including features for typical and atypical ductular proliferation, as well as bile duct metaplasia were assessed (**Figure 33**). Both groups revealed comparable frequencies of the respective scores for all features. For none of the features, statistical significant difference was observed between the bipolar disorder and the controls.



**Figure 33.** Frequency distribution of assessed histopathological features categorized to endogenous reaction. The percentage of the respective histopathological feature score for the patients with bipolar disorder (BPD) and the control group (Controls) are given. Statistical analyses: Mann Whitney U test; [Based on [90]]

For the category of steatosis, frequency distributions were assessed for histopathological features of intracytoplasmic accumulation of lipids, namely macrovesicular and microvesicular steatosis and the zonal distribution of the lipid

accumulation (*Figure 34*). No statistical significant differences were observable for all features of this category between the groups.



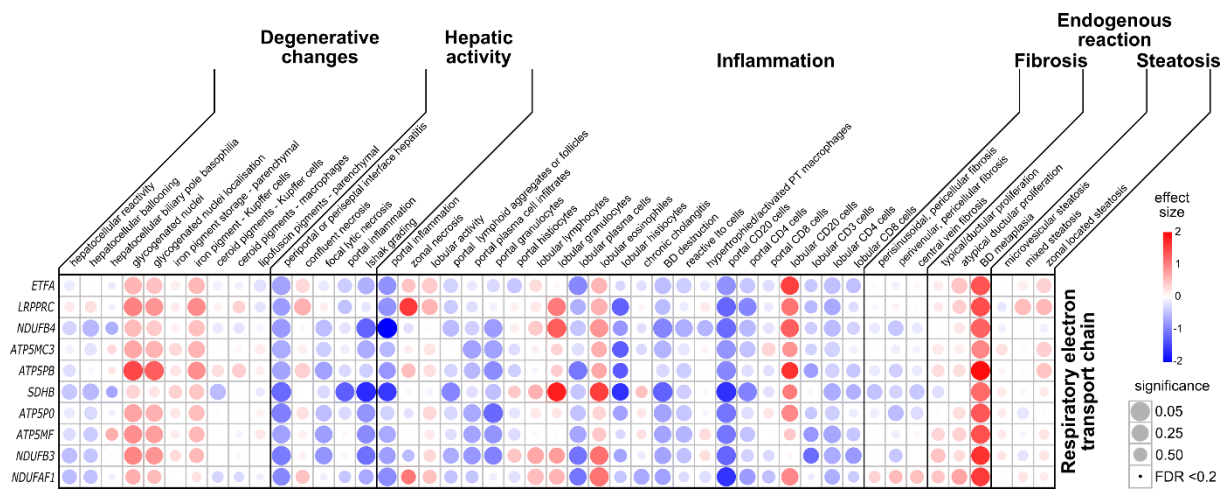
**Figure 34.** Frequency distribution of assessed histopathological features categorized to steatosis. The percentage of the respective histopathological feature score for the patients with bipolar disorder (BPD) and the control group (Controls) are given. Statistical analyses: Mann Whitney U test; [Based on [90]]

### 3.10 Associations between gene expression and histopathological features

Feature-expression heatmaps were generated in order to visualize the association between assessed histopathological characteristics and gene expression.

The associations between genes belonging to the RETC and histological features of the liver were depicted in the RETC feature-expression heatmap (*Figure 35*). Although an overall clustering of associations was absent between RETC genes and degenerative changes, the positive associations between the presence of glycogenated or vacuolated nuclei, located glycogenated nuclei and the presence of iron pigments in Kupffer cells were notable. A clustering of negative associations between RETC gene expression and hepatic activity was present. Prominent negative associations were observed for periportal or periseptal interface hepatitis and the Ishak grading. The heatmap showed a cluster of strong negative association between RETC genes and histological features of inflammation including the presence of portal inflammation, portal granulocytes, CD20<sup>+</sup> B-cells and CD4<sup>+</sup> T-cells, lobular plasma cells and histiocytes and biliary duct destruction. A cluster of weaker negative associations were found between RETC genes and the presence of portal lymphoid aggregates or follicles and plasma cell infiltrates, reactive Ito cells and lobular CD3<sup>+</sup>, CD4<sup>+</sup> and CD8<sup>+</sup> T-cells. Positive association

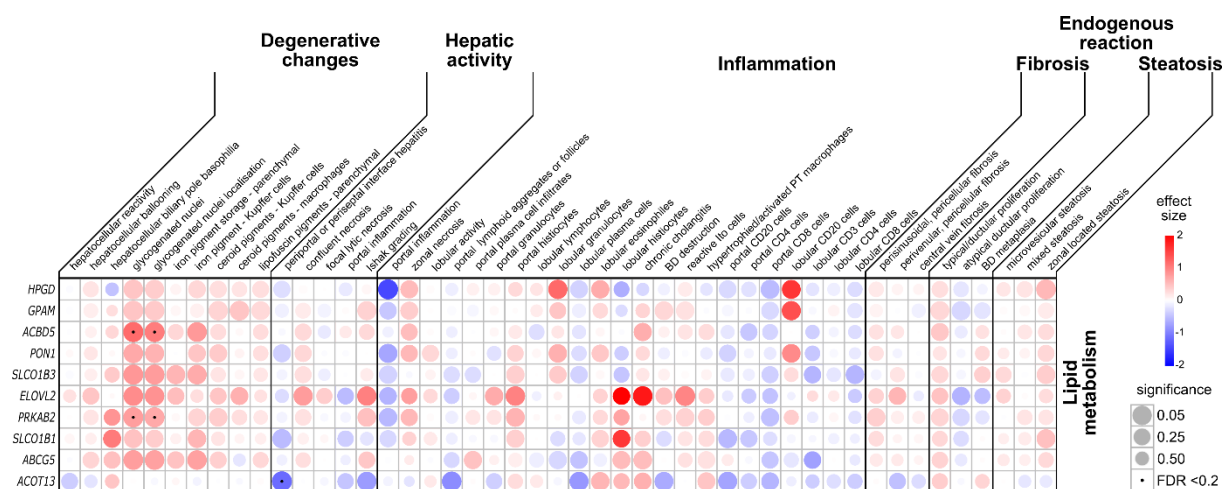
cluster were observed between RETC genes and the presence of lobular granulocytes, eosinophils and CD20<sup>+</sup> B-cells. Association clustering between genes of the RETC and fibrosis was absent. Strong positive associations were present between RETC genes and biliary duct metaplasia in the category of endogenous reaction. Clustering of associations were absent between RETC genes and the histological features of steatosis. After FDR correction, none of the individual analyses were statistically significant.



**Figure 35. Feature-expression heatmap depicting the association between genes related to respiratory electron transport chain (RETC) and histologically assessed characteristics of the liver.** Histological characteristics were ordinal or binary measured. Gene expression was expressed as log<sub>2</sub> intensity values. The color code (red = positive, blue = negative) and color intensity represent the regression coefficient. The circle radius represents the statistical significance of the regression analysis, whereas dotted circles represent a false discovery rate (FDR) below 0.2 for multiple testing. Blank squares correspond to a small regression coefficient and statistical probability or an analysis with insufficient observations. Full names are listed in **Table A1** in the Appendix. [Published in [90]]

The lipid metabolism feature-expression heatmap was created from associations between genes belonging to the lipid metabolism and histological features of the liver (**Figure 36**). A clustering of positive associations was observed between lipid metabolism genes and degenerative changes. Especially the association analyses between the presence of glycogenated or vacuolated nuclei, located glycogenated nuclei, and the presence of iron pigments in Kupffer cells revealed strong positive associations. Four associations were significant after FDR correction, corresponding to the acyl-CoA binding domain containing 5 (*ACBD5*) gene and the protein kinase AMP-activated non-catalytic subunit beta 2 (*PRKAB2*) gene. An overall clustering of associations between lipid metabolism genes and hepatic activity could not be

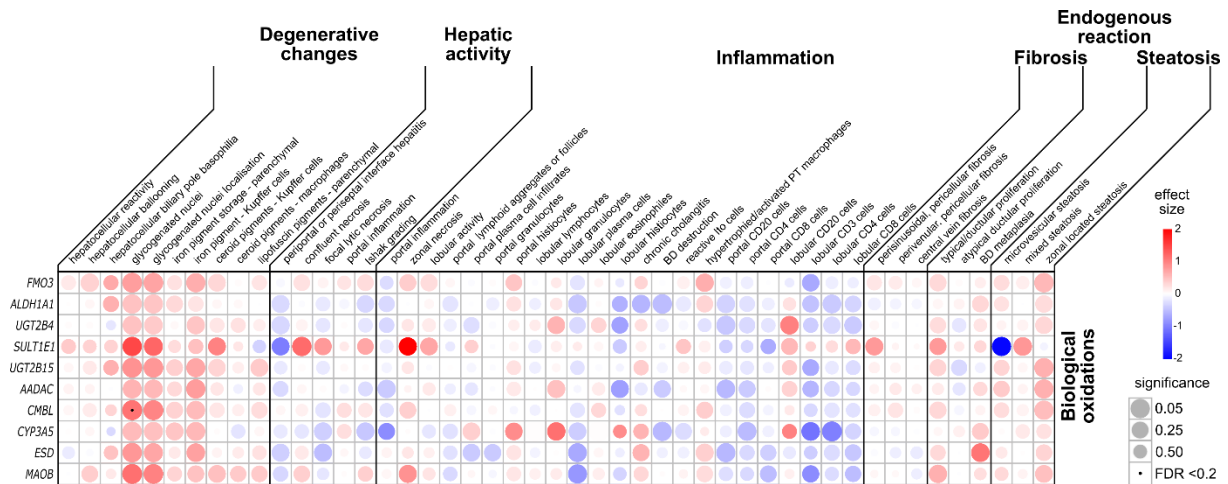
identified, however, the negative association between the *ACOT13* gene and periportal or periseptal interface hepatitis was significant after FDR correction. Although an overall clustering of associations was absent between lipid metabolism genes and inflammation, negative associations were revealed for portal inflammation and positive associations for lobular histiocytes and CD20<sup>+</sup> B-cells, and chronic cholangitis. A clustering of associations was absent between lipid metabolism genes and fibrosis, endogenous reaction and steatosis and none of the individual analyses was significant after FDR correction.



**Figure 36. Feature-expression heatmap depicting the association between the expression of genes related to lipid metabolism and histologically assessed characteristics of the liver.** Histological characteristics were ordinally or binary measured. Gene expression was expressed as log<sub>2</sub> intensity values. The color code (red = positive, blue = negative) and color intensity represent the regression coefficient. The circle radius represents the statistical significance of the regression analysis, whereas dotted circles depict a false discovery rate (FDR) below 0.2 for multiple testing. Blank squares correspond to a small regression coefficient and statistical probability or an analysis with insufficient observations. Full names are listed in **Table A1** in the Appendix. [Published in [90]]

The biological oxidations feature-expression heatmap was generated from associations between genes belonging to biological oxidations and histological features of the liver (**Figure 37**). A clustering of positive associations was observed between biological oxidations genes and degenerative changes. A strong positive association was observed between present glycogenated or vacuolated nuclei, located glycogenated nuclei and the presence of iron pigments in Kupffer cells. The association between the expression of gene coding for the *CMBL* and the presence of glycogenated or vacuolated nuclei was significant after FDR correction. A pronounced clustering of associations was absent for hepatic activity, inflammation, fibrosis, endogenous reaction and steatosis. However, a strong positive association was observed

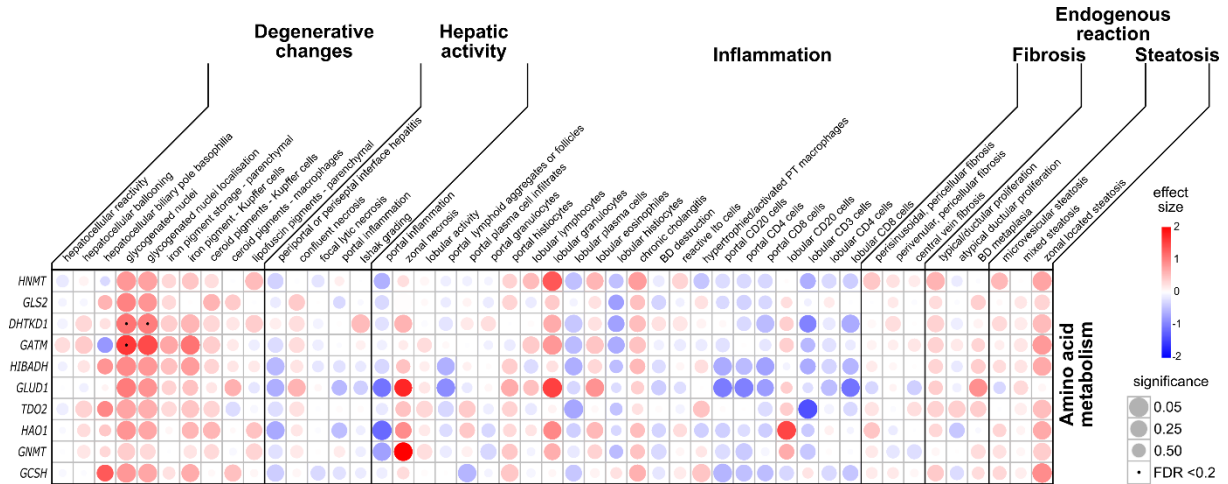
between sulfotransferase family 1E member 1 (*SULT1E1*) and portal zonal necrosis in the inflammation category. A strong negative association was revealed between *SULT1E1* expression and the presence of microvesicular steatosis in the steatosis category. After FDR correction, none of the other individual analyses were statistically significant.



**Figure 37. Feature-expression heatmap depicting the association between the expression of genes related to biological oxidations and histologically assessed characteristics of the liver.** Histological characteristics were ordinal or binary measured. Gene expression was expressed as log<sub>2</sub> intensity values. The color code (red = positive, blue = negative) and color intensity represent the regression coefficient. The circle radius represents the statistical significance of the regression analysis, whereas dotted circles depict a false discovery rate (FDR) below 0.2 for multiple testing. Blank squares correspond to a small regression coefficient and statistical probability or an analysis with insufficient observations. Full names are listed in **Table A1** in the Appendix. [Published in [90]]

The associations between genes belonging to the metabolism of amino acids and histological features of the liver were depicted in the amino acid metabolism feature-expression heatmap (**Figure 38**). Association analyses revealed a clustering of associations between amino acid metabolism genes and degenerative changes. Particularly strong positive associations were discovered between the presence of glycogenated or vacuolated nuclei, located glycogenated nuclei, and the presence of iron pigments in Kupffer cells. Three associations were significant after FDR correction, corresponding to the genes coding for dehydrogenase E1 and transketolase domain containing 1 (*DHTKD1*) and glycine amidinotransferase (*GATM*). An overall clustering of associations between amino acid metabolism genes and hepatic activity could not be identified. Although an overall clustering of associations was absent between amino acid metabolism genes and inflammation, negative associations were revealed for portal inflammation and portal CD3<sup>+</sup> T-cells, and positive associations for zonal necrosis and lobular histiocytes. A clustering of associations was absent between amino acid metabolism genes and

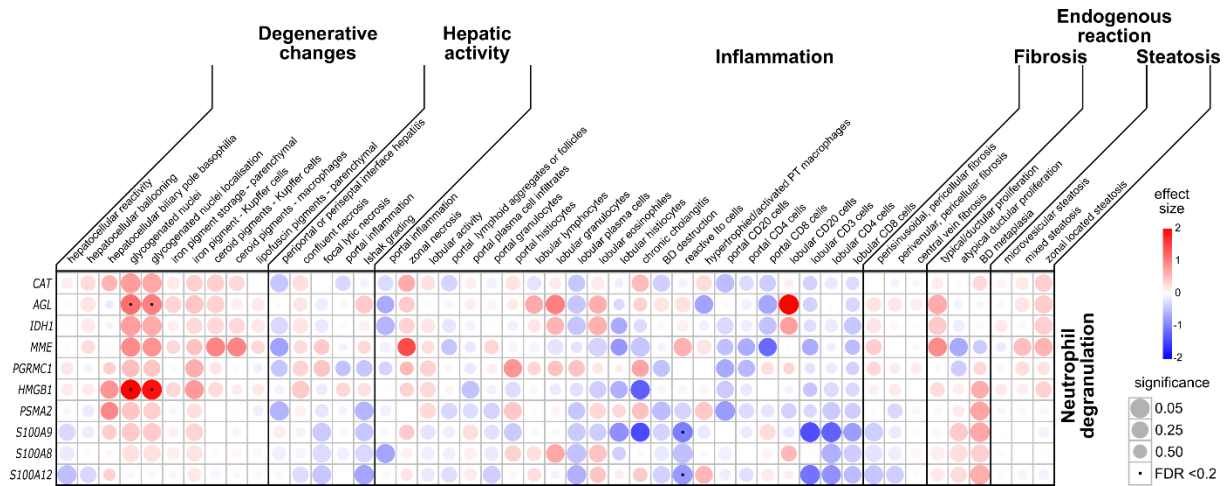
fibrosis, as well as endogenous reaction. Weak associations were found between neutrophil degranulation genes and zonal located steatosis in the steatosis category.



**Figure 38. Feature-expression heatmap depicting the association between the expression of genes related to amino acid metabolism and histologically assessed characteristics of the liver.** Histological characteristics were ordinal or binary measured. Gene expression was expressed as  $\log_2$  intensity values. The color code (red = positive, blue = negative) and color intensity represent the regression coefficient. The circle radius represents the statistical significance of the regression analysis, whereas dotted circles depict a false discovery rate (FDR) below 0.2 for multiple testing. Blank squares correspond to a small regression coefficient and statistical probability or an analysis with insufficient observations. Full names are listed in **Table A1** in the Appendix. [Published in [90]]

The associations between genes belonging to the neutrophil degranulation and histological features of the liver were depicted in the neutrophil degranulation feature-expression heatmap (**Figure 39**). A clustering of positive associations was observed between neutrophil degranulation genes and degenerative changes. Strong positive associations were observed between present glycogenated or vacuolated nuclei and located glycogenated nuclei, weaker positive associations for the presence of iron pigments in Kupffer cells. The associations between the expression of the genes coding for amylo-alpha-1, 6-glucosidase, 4-alpha-glucanotransferase (*AGL*) and high mobility group box 1 (*HMGB1*) and the presence of glycogenated or vacuolated nuclei were significant after FDR correction. An overall clustering of associations between amino acid metabolism genes and hepatic activity could not be identified. A weak clustering of negative associations was observed between neutrophil degranulation genes and inflammation, whereas the negative associations between the genes coding for *S100A9* and *S100A12* and the presence of reactive Ito cells were significant after FDR correction. Clustering of associations could not be identified for fibrosis and steatosis

features. A weak clustering of associations was present between neutrophil degranulation genes and biliary duct metaplasia.



**Figure 39. Feature-expression heatmap depicting the association between the expression of genes related to neutrophil degranulation and histologically assessed characteristics of the liver.** Histological characteristics were ordinal or binary measured. Gene expression was expressed as  $\log_2$  intensity values. The color code (red = positive, blue = negative) and color intensity represent the regression coefficient. The circle radius represents the statistical significance of the regression analysis, whereas dotted circles depict a false discovery rate (FDR) below 0.2 for multiple testing. Blank squares correspond to a small regression coefficient and statistical probability or an analysis with insufficient observations. Full names are listed in **Table A1** in the Appendix. [Published in [90]]

## **4 Discussion**

Parts of the discussion resemble the discussion section in Birkl-Toegelhofer et al. 2019 [90].

Psychotropic drugs, such as antipsychotics, antidepressants and mood stabilizers, represent a substantial treatment option in bipolar disorder. Despite their positive treatment effects and improved drug safety, potential adverse effects of the required acute and long-term drug treatment in the liver have to be considered. However, consequences on hepatic gene expression upon psychotropic drug treatment in bipolar disorder remain to be fully elucidated. Here, a genome-wide approach of gene expression with advanced comprehensive bioinformatical analyses was applied to address the influence of psychotropic drugs in bipolar disorder on the hepatic function. In addition, possible associations between gene expression and the presence of histopathological features of liver disease were examined.

### **4.1 Presence of an altered hepatic expression in bipolar disorder**

Differential gene expression was present in hepatic tissue of patients with bipolar disorder and psychotropic drug treatment. The differential expression analysis revealed only a few upregulated genes, whereas the larger portion were identified as downregulated genes. This suggests that psychotropic drug treatment rather turns off genes than shifting to an increased expression state of various genes. After annotation of the DEGs, assumptions on the individual gene level were feasible. The majority of downregulated genes were found to be associated with membranes in context to their subcellular localization and are involved in processes related to the metabolism. Resulting disturbances in energy metabolism, as well as lipid and glucose metabolism have already been shown for psychotropic drugs [113,114]. Psychotropic drugs for treating depression and mood disorders were shown to affect enzyme activity and respiration in the mitochondria [113]. Both, lipid and glucose metabolism were supposed to be affected by antipsychotics, such as clozapine, olanzapine, risperidone and quetiapine [114]. A higher estimated prevalence was shown for obesity and diabetes in patients with bipolar disorder and schizophrenia. These observations supported the assumption that psychotropic drug treatment possibly adversely affects metabolic pathways in the liver tissue of patients with bipolar disorder.

Beside the possible impairment of metabolic processes, the individual gene analysis revealed that the five respective major over- and underexpressed genes were found to play a role in immune response, oxidative stress response, apoptosis, and maintenance of the cell structure. Those genes are discussed and interpreted in accordance with recent literature.

#### **4.1.1 Upregulated genes involved in inflammatory response**

Revealing of the top five upregulated genes suggested the presence of increased inflammatory features in hepatic tissue of patients with bipolar disorder treated with psychotropic drugs. Those upregulated genes included the calgranulin genes *S100A8*, *S100A9* and *S100A12*, which are known to be abundantly expressed in neutrophils and associated with acute and chronic inflammatory disorders and various cancer types [115–118]. Altered levels of S100 genes and proteins were found in various tissue types of psychiatric disorders, such as bipolar disorder and schizophrenia [119–122]. It has been shown that increased *S100A8* and *S100A9* expression was present in the hippocampus of patients with bipolar disorder and schizophrenia [119]. Additionally, altered *S100A9* expression in blood were reported in a microarray approach of patients with bipolar disorder and schizophrenia [120]. Increased *S100A9* mRNA levels were found in both psychiatric disorder groups when compared to control subjects. With regard to *S100A12*, altered protein levels and mRNA expression have been observed in patients with bipolar disorder and patients with schizophrenia [121,122]. In both disorders, non-invasively measured S100A12 levels were found to be increased in saliva samples in comparison to healthy controls [121]. With regard to hepatic pathology, higher serum levels of S100A8 and S100A9 were present in patients with NASH and fibrotic features [123]. In HCC, malignancy progression was supposed to be promoted by an increase of S100A8 and S100A9 on gene and protein level [124]. A recent study investigating the knockout of *S100A9* in a mouse model of HCC observed reduced tumor size and a lower number of high-graded tumors in comparison to wild type mice [125]. The overexpression of S100 genes indicated a possible prominent inflammatory component in hepatic tissue of bipolar disorder. Probably, overexpression of calgranulins is intrinsically prevalent in severe mental disorders and might not be restricted to brain tissue. The overlapping findings in bipolar disorder and schizophrenia would indicate a relationship between these psychiatric disorders, but to which extent patients with schizophrenia also show hepatic overexpression of these S100 genes has to be examined.

Another highly overexpressed gene was identified as *SAA2*. The gene encodes for the serum amyloid A2 protein and belongs with the serum amyloid A1 protein to the acute-phase serum amyloid protein (A-SAA) primarily synthesized in the liver [126]. The SAA2 protein level was found to be induced by proinflammatory cytokines, such as interleukin-1-beta, interleukin-6 and tumor necrosis factor- $\alpha$  [127]. SAA2 has been shown to serve as chemoattractant for monocytes [128] and neutrophils [129]. With regard to psychiatric disorders, increased A-SAA plasma levels were associated with symptoms of depression and depressive disorder in a population-based cohort study (36). Another study showed that higher A-SAA levels were found in individuals with moderate to severe depressive symptoms [131]. The increased A-SAA levels correlated with several psychiatric rating scales items pointing to fatigue. In context to hepatic function, it has been shown that elevated A-SAA levels were present in primary hepatic stellate cells of healthy human livers and in an animal model investigating hepatic fibrosis [132]. It was proposed that A-SAA modulates fibrogenesis by its pro-proliferative effect on hepatic stellate cell at the sites of injury. These findings suggest that the presence of generally higher A-SAA levels in patients with bipolar disorder might entails a higher susceptibility to fibrogenesis during liver injury.

The *IL1RN* gene encoding for the protein interleukin-1 receptor antagonist (IL-1R $\alpha$ ) was found to be upregulated in bipolar disorder. This protein represents an anti-inflammatory cytokine which binds specifically to the interleukin-1 receptors leading to the inhibition of interleukin-1 downstream signaling and thereby functioning immunosuppressive [133]. In non-medicated patients with bipolar disorder, a state-dependent cytokine pattern was observed in manic and depressive phases [134]. A meta-analysis showed that patients with bipolar disorder revealed higher circulating IL-1R $\alpha$  levels during a manic episode and in euthymic patients compared to healthy controls [135]. Additionally, associations between increased IL-1R $\alpha$  levels and psychotic symptoms [136], as well as cognitive function in psychiatric disorders [137] were reported. Furthermore, a recent study revealed increased plasma IL-1R $\alpha$  levels in patients with NASH and correlate with markers for hepatic inflammation and damage [138]. Another study showed an association between increased serum IL-1R $\alpha$  levels and histologically assessed features, as well as a correlation between hepatic *IL1RN* expression and NASH [139]. In summary, these findings would suggest that patients with psychiatric disorders, such as bipolar disorder, have higher levels of IL-1R $\alpha$  per se. This could probably lead to a more detrimental effect in the context of hepatic injury.

#### 4.1.2 Downregulated genes involved in oxidative stress response, apoptosis and regeneration

*FMO3* was identified as the most underexpressed gene in the bipolar disorder group. Aside from CYPs, the *FMO3* enzyme reduces oxidative stress by metabolizing a broad variety of endogenous compounds and xenobiotics including plant alkaloids and synthetic therapeutic drugs [140]. Decreased *FMO3* expression has been observed in a mouse model of hepatotoxicity using carbon tetrachloride for oxidative stress induction [141]. Recently, a study analyzed the effect of epigenetically silenced *FMO3* expression in HCC-derived cell lines and further analyzed the effect on the HCC survival probability in the publicly available liver hepatocellular carcinoma data set of The Cancer Genome Atlas (TCGA LIHC) [142]. Cells of the HCC cell lines showed increased cell viability and invasion capability, as well as decreased apoptosis when *FMO3* was epigenetically silenced. The TCGA LIHC revealed that patients with HCC and a low *FMO3* expression indicated a poorer survival probability than those with higher expression. Another study showed detrimental effects on the hepatic antioxidative system by chronic administration of antidepressants, such as the SSRI fluoxetine, in an animal model for depression [143]. This leads to the suggestion that patients with bipolar disorder under psychotropic drug treatment probably have a higher rate of hepatic oxidative stress and would be more prone to develop liver disease, possibly even HCC. None of the analyzed patients with bipolar disorder were identified to suffer from HCC, but several histopathological features pointed to NAFLD or NASH, which are known to be precursors stages and can progress in unfavorable cases to HCC.

The amino acid transporter *SLC38A4* was found to be underexpressed in bipolar disorder upon psychotropic drug treatment. So far, less research has been performed on the effect of altered *SLC38A4* expression in the context of hepatic function, especially in context of psychiatric disorders. A genome-wide association study reported an association between the *SLC38A4* locus and cases of severe alcoholic hepatitis [144]. Genetic predispositions in *SLC38A4* were identified as potential risk to develop alcoholic hepatitis. However, as genetic variants were not explored in the bipolar disorder and control samples and a difference in alcohol intake between bipolar disorder and the control group was absent, the impact of such predisposition on the hepatic function under psychotropic drug treatment remains to be determined.

One of the top five underexpressed genes in bipolar disorder was *TNFSF10*. This gene encodes for the TNF-related apoptosis-inducing ligand super family member 10 or TRAIL, which is

mainly known in the context of apoptotic signaling [145]. It triggers the proapoptotic signaling cascade by binding to the death-signaling receptors followed by the activation of the caspase pathway [146]. A recent study reported that *TNFSF10* mRNA and protein levels were associated with NAFLD or NASH in humans, mice and *in vitro* [147]. Lower TNFSF10 plasma levels were shown in patients with NASH. *TNFSF10* knock-out mice developed more severe histopathological features of NAFLD and NASH than their wildtype littermates with high fat diet. Palmitate-treated HepG2 cells resembling an *in vitro* model of NAFLD showed increased lipid accumulation, inflammation and fibrosis, and about 50% reduced *TNFSF10* mRNA expression. Conversely, the palmitate-induced lipid uptake was reduced when treating the HepG2 cells with exogenous TNFSF10. These findings suggest a protective effect of *TNFSF10* expression in liver tissue. Therefore, it can be concluded that a detrimental effect of psychotropic drugs in bipolar disorder probably exists and leads to lowered *TNFSF10* expression and, hence, contributes to hepatic injury. In context of psychiatric disorders, decreased *TNFSF10* expression was observed in prefrontal brain tissue of patients with bipolar disorder and patients with schizophrenia [148]. An effect of antipsychotic treatment on *TNFSF10* expression in both groups was excluded by regression analysis. All of which indicates that decreased *TNFSF10* expression might rather be generally linked to bipolar disorder than affected by drug treatment.

Another top downregulated gene in bipolar disorder was found to be *CAT*. The protein is mainly located in peroxisomes and acts as antioxidative enzyme essential for decomposing hydrogen peroxide to water and oxygen in the setting of intracellular oxidative stress [149]. An altered CAT activity was reported in psychiatric disorders [150–152]. Lower levels of CAT were observed in blood samples of male patients with bipolar disorder and schizophrenia treated with psychotropic drugs suggesting increased oxidative stress [152]. A case report of identical twins with bipolar disorder showed lower CAT activity levels in serum samples in comparison to a matched control [150]. Decreased CAT activity remained even after psychotropic drug treatment in one of the twins. Additionally, another study showed reduced CAT activity in plasma red blood cells of patients with bipolar disorder and schizophrenia, whereas lowest activity was present in the schizophrenia samples [151]. The function of antioxidant enzymes including *CAT* has been extensively reviewed in the context of NAFLD pathogenesis [153]. It has been proposed that increased oxidative stress triggers the development of NAFLD and NASH, whereas *CAT* was shown to suppress hydrogen peroxide accumulation and, hence,

plays a prominent role in the cellular antioxidant defense system. In an early study, *CAT* overexpression was shown to have a beneficial effect by protecting cells from cytotoxicity arising by oxidative stress *in vitro* [154]. Another study investigated the hepatic gene expression of reactive oxygen species scavenging enzymes including *CAT* in NASH [155]. A significant downregulation of *CAT* and other genes encoding those scavenging enzymes was identified in NASH. Besides a decreased *CAT* expression, the activity of *CAT* was also reduced in human liver tissue samples [156]. The progression from steatosis to steatohepatitis showed a continuous decrease of the antioxidant capacity by *CAT*. All of this would suggest that patients with bipolar disorder and treated with psychotropic drugs are challenged with a higher level of oxidative stress and would be more prone to undergo the progress from simple fatty liver to NAFLD or NASH.

The fifth most downregulated gene was identified as *TUBE1*, encoding the  $\epsilon$ -tubulin protein a member of the tubulin superfamily. The protein is specifically localized at the centrosomes and is thought to be an integral centrosomal component [157]. The  $\epsilon$ -tubulin presented distinct spatial and temporal cell cycle-dependent localizations compared to the other tubulin family members. A pivotal role for *TUBE1* has been identified in the organization of the microtubule cytoskeleton [158]. An *in vitro* study has shown that after silencing *TUBE1* in *Paramecium* a progressive worsening of the cell structure was observable which was induced by an impaired basal body duplication [159]. This impaired basal body duplication revealed a phenotype presenting a diminished cell size and a more rounded cell shape, whereas cell death occurred after three to four cell divisions. With regard to hepatic cells in patients with bipolar disorder upon drug treatment this might suggest a lower regeneration rate of injured hepatic sites which remains to be determined.

## **4.2 Affected biological pathways upon psychotropic drug treatment**

As biological processes rely rather on intricate networks of interacting genes and proteins than individual gene or protein function, the application of extensive computational biological approaches was indispensable to capture and investigate the underlying mechanism in liver tissue of psychotropic drug treated patients with bipolar disorder. An intricate network of interacting proteins has been confirmed by the protein-protein network analyses. The

investigation of affected functional processes and networks included the comprehensive analyses of probable key regulators and underlying molecular complexes, as well as the determination of enriched biological pathways.

#### **4.2.1 Mitochondrial dysfunction upon psychotropic drug treatment**

The most affected biological process in bipolar disorder was found to be the cellular energy supply. Several energy-related genes and their corresponding proteins were found to be decreased. Based on betweenness centrality, the hub protein analysis revealed the ATP synthase subunit ATP5O as considerable key regulator. ATP5O was also found in module 2 alongside with various NADH dehydrogenases, where NDUFA9 was determined as the module's seed protein. Some of the module 6 proteins referred to mitochondrial proteins related to the energy supply. Furthermore, the enrichment analysis strengthened the possible functional deterioration of cellular energy supply. Several highly significant enriched pathways were related to the RETC. All proteins of module 2 overlapped with the queried and RETC-allocated DEGs. In addition, all RETC-allocating DEGs were underexpressed in the liver tissue of the investigated psychotropic drug-treated bipolar disorder group. Recent studies showed impairing effects on the energy metabolism and mitochondrial function upon psychotropic drug treatment [160–162]. One of those studies revealed reduced expression levels of mitochondria related genes in brain tissue of patients with bipolar disorder and antipsychotic treatment [160]. This observation was concluded as effect of the drug treatment, since untreated patients with bipolar disorder revealed rather an upregulation of those genes. Another study revealed decreased expression of mitochondria-related genes in blood samples of patients with bipolar disorder and known response to the mood stabilizer lithium [163]. Downregulated genes included genes encoding for the RETC and the oxidative phosphorylation pathway. Another study showed that patients with schizophrenia treated with antipsychotics revealed lower gene expression of mitochondria enzymes in peripheral blood mononuclear cells [161]. The antipsychotics, clozapine and olanzapine, showed the strongest decreasing effect on those genes, followed by quetiapine and risperidone. A recent study showed that bupropion, an antidepressant, decreased cellular ATP content at non-cytotoxic drug concentrations in two human hepatocyte cell lines [162]. The bupropion-treated cells presented the inhibition of complex I and II of the RETC, as well as increased reactive oxygen production. Among other antipsychotics and antidepressants,

all of the aforementioned antipsychotics and bupropion were under those psychotropic drugs recorded in the investigated patients with bipolar disorder. The previous findings have now been confirmed also in human liver tissue of patients with bipolar disorder and psychotropic drug treatment, since the decreased expression of RETC-genes was present in the bipolar disorder group. An animal study showed NAFLD features upon antipsychotic treatment [164]. Wistar albino rats treated with olanzapine or aripiprazole revealed increased hepatic steatosis, ballooned hepatocytes and inflammatory infiltrates. Therefore a diminished energy supply upon psychotropic drug treatment in liver cells, but obviously also in the periphery, and the resulting consequences, such as liver injury, have to be taken into consideration.

#### **4.2.2 Altered lipid metabolism upon psychotropic drug treatment**

In addition, diminished expression levels of all genes allocated to the enriched pathway metabolism of lipids and lipoproteins have been observed in the bipolar disorder group. As the catabolism of lipids produces substrates, such as pyruvate, for the citric acid cycle and thus for the RETC, it further suggest that psychotropic drugs could lead to a severe interference of the cell's energy supply. Several proteins of module 6 overlapped with the genes enriched in the lipid and lipoprotein metabolism. Further overlaps were present for module 1 and 4. In total, four supposed hub proteins were also present among the enriched pathway. Those included IDH1, DECR1, APOB and PTEN.

*IDH1*, coding for a peroxisomal and cytoplasmic enzyme, was one of the most underexpressed genes in this enriched pathway. A comprehensive gene expression profiling study searching for symptom-depending blood biomarkers in schizophrenia showed associations with *IDH1* [165]. Decreased *IDH1* expression was observed in blood samples of patients with schizophrenia and delusions and was confirmed in additional human tissue of schizophrenia patients, such as post-mortem brain tissue and fibroblasts, as well as in a schizophrenia mouse model treated with phencyclidine. Unfortunately, a possible association with psychotropic drug treatment in bipolar disorder still remains to be determined. In context of hepatic IDH1 function, *Idh1* knockout mice showed increased oxidative stress by accumulation of reactive oxygen species in the liver [166]. This led to the assumption that IDH1 might play a role in the antioxidant defense and probably suppress hepatic inflammation.

The enzyme encoded by *DECRI* is known to play a role in the mitochondrial  $\beta$ -oxidation of unsaturated fatty acids as auxiliary enzyme. In terms of psychiatric disorder and effects of psychotropic drug treatment, less is known about this gene. An *in silico* study of published data investigating a gene expression study using an animal model of NASH induced by high fat diet showed a possible fundamental role in NASH development [167]. It was suggested that altered *DECRI* levels possibly affect the development of NASH at later stage. To which extent decreased *DECRI* expression due to psychotropic drug treatment might affect liver injury in bipolar disorder remains to be solved.

The ApoB protein, where the isoform ApoB-100 is synthesized in the liver, is involved in transporting lipids in very low density lipoproteins [168]. A study investigating lipid metabolism upon antipsychotic treatment observed altered ApoB-100 secretion levels in several human hepatic cell lines [169]. The clozapine treatment did not alter ApoB-100 secretion, but after omitting clozapine increased ApoB-100 secretion was shown. It has been assumed that the antipsychotic-dependent effect entailing increased triglycerides levels regulates the ApoB secretion. If the decreased *APOB* gene expression in bipolar disorder, represent an adequate ApoB level after protein translation has not been investigated. So far, an effect of long-term psychotropic drug treatment on ApoB levels was absent in cell culture. It has to be taken into account that psychotropic drug treatment is generally a long-term treatment in patients meaning that effects of long-term drug exposure might be apparent only after a longer time period. In addition, possible compensating effects have to be considered where ApoB levels would be decreasing when fat accumulation in the cell exceeds a certain threshold. In context of *APOB* gene and protein expression in liver injury, an early study could show that patients with NASH have reduced ApoB-100 synthesis rates [170]. This has been shown when comparing the ApoB-100 synthesis rates in patients with NASH to lean and obese individuals. Due to reduced ApoB-100 availability, it has been suggested that fat accumulation in the liver might be more likely. Additionally, there might be an association between PTEN and APOB levels. The lipid phosphatase PTEN is primarily known as tumor suppressor mutated in various cancer entities, however, it plays an apparent pivotal role in liver metabolism too [171]. The investigation of hepatic *Pten* deficiency in mice demonstrated an increased fatty acid synthesis and hepatic steatosis [173]. It was supposed that hepatic *Pten* deficiency results in a constitutive insulin signaling. Another study showed that PTEN activity possibly regulates the availability of APOB in the liver [172]. Hepatic *Pten* deficiency in mice indicated a reduced lipid transport in

and from the cell and probably result in steatosis due to diminished APOB availability. Thus, all of this points to an increased probability for lipid accumulation in the liver due to underexpressed genes related to the lipid metabolism.

#### **4.2.3 Altered amino acid metabolism upon psychotropic drug treatment**

The impairment of adequate energy supply in cells of the liver in patients with bipolar disorder and psychotropic drug treatment was further emphasized by the reduced and enriched pathway of amino acid metabolism. Apart from glucose and lipids, amino acids can also serve as substrate for the citric acid cycle and further for energy generation. Every gene of this enriched pathway was underexpressed which implies an impaired amino acid metabolism. Further analyses revealed that several proteins of module 3 were overlapping with the enriched pathway. All of these were annotated as subunits of the proteolytic 26S proteasome complex or rather to the 19S and 20S proteasome. A gene expression study of brain tissue of patients with bipolar disorder showed decreased levels of genes related to protein degradation [174]. The GO results included processes associated to protein degradation including the 26S and 20S proteasome. Further analysis revealed that most genes of the proteasome were underexpressed in brain tissue of patients with bipolar disorder. In addition, gene expression related to the proteasome was also decreased in hippocampal tissue of patients with schizophrenia [175] and in blood samples of patients with psychosis [176], whereby both studies included patients under psychotropic drug treatment. These studies support the findings in the herein examined bipolar disorder group and indicated that an adequate protein degradation might be hampered by the underexpression of proteasomal subunits. Detrimental effects of an impaired protein degradation on hepatic cells have been observed in *in vitro* and *in vivo* studies [177–179]. Studies of hepatic cell lines showed that cytoplasmic inclusions were present upon proteasome inhibition [177,178]. One study showed that those cytoplasmic inclusion consisted of keratin 8 and 18, ubiquitin, heat-shock protein 70 and p62, which are typically found in Mallory-Denk bodies [178]. Similar results have been observed in an animal study investigating Mallory-Denk body formation in mice overexpressing keratin 8 upon proteasome inhibition [179]. It has been supposed that cytoplasmic inclusions have a higher occurrence in patients with NAFLD and obese individuals. Consequently, an increased susceptibility to develop Mallory-Denk bodies would exist for patients with psychotropic drug treatment. However, for none of the herein investigated individuals, Mallory-Denk bodies were observed. This does not exclude a possible detrimental effect of psychotropic drug treatment in bipolar disorder on proteasomal function.

The implication of accumulating ubiquitinated proteins in liver cells by proteasomal impairment in bipolar disorder remains to be elucidated.

#### **4.2.4 Reduced biotransformation upon psychotropic drug treatment**

The biotransformation of xenobiotics, such as drugs, and endogenous compounds was found to be disturbed in bipolar disorder. The pathway of biological oxidations was enriched based on the queried DEGs. This pathway includes the functionalization and conjugation of compounds, the so called phase I and phase II of biotransformation, respectively. The top downregulated gene *FMO3* and other enzymes of the phase I reaction were allocated to this pathway, such as enzymes belonging to the CYP450 family, such as cytochrome P450 family 3 subfamily A member 5 (*CYP3A5*) belonging to the subfamily of CYP3A. The subfamily CYP3A was determined to metabolize a high number of clinical drugs including psychotropic drugs [59]. The expression of *CYP3A5* seems to be largely dependent on genetic polymorphism. A study investigating hepatic CYP3A5 protein levels based on the three known *CYP3A5* genotypes revealed that CYP3A5 protein levels could represent up to 50% of total CYP3A protein levels [180]. With regard to psychotropic drugs, genetic polymorphism in *CYP3A5* were also shown to influence the clearance of olanzapine [181]. A higher olanzapine clearance was suggested for individuals carrying at least one of the *CYP3A5\*1* allele. Further, a study revealed a possible relationship between homozygous genotypes of *CYP3A5\*3* and another CYP450 gene, namely *CYP2D6\*10*, resulting in obviously high exposure and active moiety of risperidone [182]. It was supposed that the combined homozygous genotypes could be more prone to potential detrimental effects of psychotropic drug treatment. Unfortunately, genotype data was not available for the herein investigated samples. However, the genotype studies suggest that lower CYP3A5 levels probably lead to higher drug exposure due to an impaired first-pass metabolism. Furthermore, an increased exposure might also result in enhanced toxic effects on the liver in those individuals.

Beside enzymes of the phase I reaction, genes of phase II conjugation enzymes were also allocated to the biological oxidations pathway in bipolar disorder including transferases for glucuronidation and sulfonation, such as the UDP glucuronosyltransferase family 2 member B15 (*UGT2B15*) and *SULT1E1*, respectively. The conjugation of glucuronide is a major metabolizing pathway of psychotropic drugs, such as lamotrigine, olanzapine and others [183]. A study investigating inhibitory effects on uridine diphosphate (UDP)-glucuronosyltransferases

(UGT) by valproic acid treatment revealed impeded UGT2B15 activity in human liver microsomes and in a cell culture model [184]. The mode of inhibition could not be clarified, but an inhibition by valproic acid was supposed to be due to competitive inhibition. With regard to sulfate conjugation by sulfotransferases in liver disease, it has been shown that sulfotransferase activity decreased with increased liver disease severity [185]. Gene expression levels of *SULT1E1* were reduced in human diabetic liver cirrhosis tissue. A diminished phase II enzyme capacity could result in a reduced psychotropic drug metabolization, and, hence, a decreased clearing followed by a potential increase in their toxicity. Thus, impaired biotransformation of therapeutic drugs might induce detrimental effects on the liver and affect physiological liver function.

#### **4.2.5 Altered neutrophil degranulation upon psychotropic drug treatment**

The detected overexpressed S100 genes, *S100A8*, *S100A9* and *S100A12*, already indicated an involvement of the immune system in hepatic tissue of patients with bipolar disorder and psychotropic drug treatment. The pathway analysis revealed an enrichment of genes assigned to neutrophil degranulation, which is considered as response mechanism of the innate immune system. All genes in this pathway were underexpressed with the exception of the aforementioned overexpressed S100 genes. The *S100A8* and *S100A9* were determined to be involved in neutrophil chemotaxis [186]. This would lead to the suggestion that an accumulation of neutrophils, though with low degranulation potential, was supposed to be present in the hepatic tissue of patients with bipolar disorder and drug treatment. Indeed, patients with bipolar disorder revealed higher grades of portal and lobular granulocytes indicating an increased number of granulocytes. However, a differentiation in granulocytic subtypes was absent. Recent studies revealed an increased neutrophil to lymphocytes ratio in peripheral blood samples of patients with bipolar disorder [187] and in a meta-analysis of patients with non-affective psychosis [188]. The latter study considered that drug treatment could possibly influence the increased neutrophil amount, although not statistically significant. With regard to hepatic neutrophil presence upon psychotropic drug treatment, a study reported that atypical antipsychotic treatment caused hepatic injury defined by steatosis, fibrosis and features of inflammation [164]. Clozapine-treated animals showed mainly infiltrates of neutrophils and lymphocytes at sites of lobular inflammatory. This was also observed in the

herein studied bipolar disorder group. Patients with bipolar disorder revealed a higher percentage of portal and lobular granulocyte compared to the controls. But still, whether and how psychotropic drug treatment affects the neutrophil degranulation despite increased neutrophil presence, remains unclear. Early studies proposed that treatment with phenothiazines, which are used as antipsychotics, inhibit neutrophil degranulation [189,190]. Phenothiazine was supposed to prevent neutrophil degranulation by binding to calmodulin and, hence, altering the essential receptor binding property to fulfill degranulation [190]. It seems that psychotropic drug treatment probably result in an involvement of various molecular subclusters with regard to neutrophil degranulation. A large number of hub proteins overlapped with genes queried and allocated to this pathway, such as *CAT*, *MAPK1*, *C3* and *VAMP8*. Whereas the latter two were also determined as seeds in module 1 and module 4. Predominantly, proteins of module 4 and module 6 overlapped with this pathway. The gene encoding for the antioxidant enzyme *CAT* was revealed as one of the most underexpressed genes in the bipolar disorder group. With regard to its activity in inflammatory cells, it has been shown that *CAT* revealed high activity in neutrophils [191]. In neutrophils, *CAT* has been suggested to be responsible for the oxidant resistance upon the respiratory burst and its resulting oxidative stress due to the cell's activation. The underexpression of *CAT* in bipolar disorder with psychotropic drug treatment implies that neutrophils seem to be more vulnerable for oxidative stress and that the present liver injury in bipolar disorder may partly be based on the drug-treatment.

For the hub protein *MAPK1*, the highest betweenness score was revealed in the bipolar disorder group. As member of the *MAPK* family, it belongs to the subfamily of extracellular signal-regulated kinases (*ERKs*) and is also known as *ERK2* [192]. The *ERK* signaling has been determined as important response mechanism to extracellular signals. It has been shown that neutrophil activation was influenced by *MAPK* signaling in an *in vitro* study of isolated human neutrophils [193]. Inhibition of *MAPK1*/*MAPK3* signaling resulted in reduced respiratory burst of the neutrophils upon activation reflected by diminished generation of reactive oxygen species. Similar to the detected probable pivotal role in the bipolar disorder group with psychotropic drug treatment, *MAPK1* was supposed to have a central role in the molecular response of psychotropic drug treatment. In a study investigating biological interaction networks, *MAPK1* was reported to act as one of the key regulators of the cellular response upon lithium and valproate treatment [194]. In addition, altered *ERK* signaling was shown in several psychiatric disorders. In *post mortem* brain tissue of suicide subjects with major depression,

decreased *ERK1* and *ERK2* levels on gene and protein level have been shown [195]. Another study revealed altered levels of proteins involved in the ERK pathway in psychiatric disorders [196]. Reduced levels for the MAPK1-upstream signal RAP1, which is a member of the Ras family, were observed in major depression and schizophrenia. In the herein studied bipolar disorder group, the underexpressed gene *RAP1A*, encoding for an isoform of RAP1, was allocated to the neutrophil degranulation pathway and was determined as hub protein in the bipolar disorder group with psychotropic drug treatment. All of this indicates a reduced MAPK signaling in neutrophils and, thus, a diminished neutrophil activation rate in bipolar disorder with psychotropic drug treatment.

The determined hub protein and seed of module 1, C3, was underexpressed in the bipolar disorder group. The C3 protein was found to be the most abundant complement protein [197]. The cleavage of C3 was shown to be involved in inflammatory response, phagocytosis and pathogen lysis. In an early study, C3 plasma levels were reported to be different in medicated patients with psychiatric disorders [198]. The combined analysis of patients with schizophrenia, mania or depression revealed decreased levels upon psychotropic drug treatment compared to the non-medicated group of patients. It was proposed that psychotropic drug treatment might suppresses the active phase response of the immune system. In contrast, the cleaved C3 component C3a and other complement components were defined to be increased in a sample of patients with bipolar disorder [199]. However, at the time of investigation all patients were in the euthymic state. A study of patients with bipolar disorder and drug treatment in the acute and chronic phase presented decreased serum levels of complement components [200]. Therefore, a possible effect of psychotropic drugs on complement activation and a potential dependency on the mood state cannot be excluded. With regard to liver injury, an animal study has reported the involvement of C3 in liver regeneration and that xenobiotic-induced liver injury led to the cleavage of C3 and further to the activation of the complement cascade [201]. C3-deficient mice revealed impaired liver regeneration upon toxic injury. This leads to the suggestion that complement activation or rather C3 cleavage might be impaired in bipolar disorder with psychotropic drug treatment and, so, liver regeneration after hepatic injury might be diminished.

Another possibly central gene allocated to the neutrophil degranulation pathway was *VAMP8*. It was identified as hub protein and seed of module 6 and was underexpressed in the bipolar disorder group. The membrane fusion catalyzing protein VAMP8 located on granules or vesicles was proposed to be involved in exocytosis of neutrophil-derived mediators [202]. For

VAMP8, it has been shown that an interaction with synaptosome associated protein 23 (SNAP23) exist during exocytosis. Apart from the underexpression of *VAMP8*, a decreased mRNA expression was observed in the bipolar disorder group for *SNAP23*. Unfortunately, VAMP8 and SNAP23 were less investigated in context of psychiatric disorders, psychotropic drug treatment and liver injury, therefore, their role in developing potential detrimental effects remains inconclusive.

### **4.3 Associations between genes expression and features of liver injury**

Most of the histopathological features assigned to the six categories characterizing liver injury, revealed comparable frequency distributions between patients with bipolar disorder and controls. Based on these findings, a similar liver histopathology was assumed in the studied groups. However, the detected altered gene expression between the studied groups and the assignment to various overrepresented biological pathways indicated separate underlying mechanism initiating or triggering hepatic injury. Therefore, a drug-induced etiology could exist. Further, a more intense effect on gene expression might be observable when comparing liver tissue of patients with bipolar disorder and liver tissue of healthy individuals. The association studies between gene expression and histopathological features of the liver in patients with bipolar disorder revealed a potential contribution of all enriched pathways to degenerative changes. All other histopathological feature groups and enriched pathways revealed weak overall association clustering with the exception of the RETC and hepatic activity, inflammation and endogenous reaction.

In context of degenerative changes, particularly, the histopathological features of glycogenated nuclei were found to be positively associated with nearly all genes of each pathway. An abundant presence of glycogenated nuclei is usually found in NAFL and NASH, but is typically absent in alcoholic steatohepatitis [84,86,203]. However, this histopathological feature is thought to be irrelevant for the diagnosis of these diseases. Until now, the pathophysiology of glycogenated nuclei is unknown. Though, a recent study revealed a possible association between the presence of glycogenated nuclei and cellular senescence [204]. This was suggested by the observation of absent cytoplasmatic apoptosis marker and present DNA damage marker in hepatocytes with glycogenated nuclei. Another study observed an impaired lipid metabolism

in senescent hepatocytes [205]. An involvement of mitochondrial dysfunction and a resulting decreased lipid metabolism was supposed to be related to the development of the senescent phenotype. This leads to the suggestion that glycogenated nuclei might represent a possible liver-specific senescence marker. Further, senescent cells could be associated with a decreased ability of lipid metabolism which possibly result in a diminished cellular energy supply accompanied by increased reactive oxygen species production. This is in line with observed enriched pathways of the herein studied bipolar disorder group, such as lipid metabolism and respiratory electron transport chain, and indicates a possible involvement of psychotropic drug treatment in the development of senescence in hepatocytes. Up to now, specific studies determining the impact of psychotropic drug treatment on senescence in hepatocytes are absent. Since both, the bipolar disorder group and the controls, revealed no difference between their frequency distribution of the presence of glycogenated nuclei in the examined liver sections, an obvious effect of psychotropic drugs influencing the development of glycogen inclusions could not be verified. A transition to a senescent phenotype could also originate from injured liver tissue in general leading to a decreased regeneration capability. This has already been observed during progression of fibrosis during cirrhosis in hepatocytes, suggesting a persistent liver damage with lowered regeneration rate [206]. Hepatic stellate cells, triggering hepatic fibrosis when activated following parenchymal damage, affect the fibrogenic response upon cell-cycle arrest [207]. A limiting effect has been proposed for senescent activated hepatic stellate cells where a restrained secretion of extracellular matrix proteins could probably regulate liver fibrosis. Hence, senescence of hepatic cells could have beneficial or detrimental effects during liver injury. Whether and how psychotropic drugs influence senescence in hepatic cells has to be evaluated in future.

A decreased function of the mitochondrial respiration has been observed to influence various features of liver injury. In NAFLD, mitochondrial dysfunction has been raised to a central pathogenic event [208–210]. The mitochondrial respiratory chain has already been shown to be involved in NASH [211]. Its dysfunction, has been shown to trigger extensive reactive oxygen species production, whereas the presence of increased reactive oxygen species can lead to the occurrence of cell death, inflammation and fibrosis [209,212]. In addition, mitochondrial dysfunction in brain tissue has been linked to bipolar disorder [174]. Moreover, inhibiting effects on mitochondrial respiration induced by various psychotropic drugs used in the treatment of mood disorders and depression have been observed in an animal study [113].

Therefore, patients with bipolar disorder might innately have a mitochondrial dysfunction, which probably exists also in other tissue than the brain. Furthermore, the treatment with psychotropic drugs induce additional detrimental effects on mitochondria and, hence, the energy supply of cells. In conclusion, the reduced RETC function might have a devastating impact on the pathogenesis of liver injury in bipolar disorder. All of this highlights the crucial aspect of mitochondrial function, where a diminished energy supply and increased oxidative stress are the suspected main drivers of hepatic injury upon psychotropic drug treatment.

#### **4.4 Approaches and limitations of the study**

The usage of global gene expression data implied intense bioinformatical processing of the data. To ensure proper and valuable data, an adequate pre-processing preceded the post-analyses. Despite the extensive application of gene expression microarrays, a strict guideline on pre-processing is unavailable and is rather based on recommendations [213]. Human *post mortem* tissue serves as fundamental resource for biological investigations, though effects on RNA integrity have to be considered in subsequent experiments. In *post mortem* tissue, increased PMI typically effects tissue autolysis and is associated with lower RNA quality values [214]. Low RIN values of the herein studied samples were present. However, the low RIN values were comparable with other studies investigating *post mortem* tissue of the liver [214,215]. The pre-processing revealed that despite a difference in PMI comparable quality metrics were present in the investigated groups. In addition, RMA normalization resulted in a between array noise reduction resting on non-biological differences. This lead to the suggestion that the pre-processing and normalization of the raw gene expression data was successful and further analyses were feasible. However, to overcome adverse effects of prolonged PMI, it was included in the linear model for the analysis of DEGs.

Already the substructure analysis discovered differences in the groups based on pre-processed and normalized gene expression data excluding a pure clustering due to confounding effects. Especially, principal component 1 and 2 were enabling a distinction between bipolar disorder and control samples. The exposed group clustering strengthened the assumption that differential gene expression was present in bipolar disorder with psychotropic drug treatment compared to the controls. The subsequent differential expression analysis confirmed the presence of

differences in gene expression and was successfully validated by examining a subset of genes by quantitative real-time PCR. All of the determined pre-processing and normalization metrics, as well as the validation of the detected DEGs, emphasized further structured *in silico* analyses covering the evaluation of subclusters, as well as the functional analyses of the corresponding proteins influenced by psychotropic drug treatment in bipolar disorders.

In addition, it has to be considered that human tissue has a high degree of cellular heterogeneity. So, the herein examined transcriptomes were based on tissue containing an assembly of different liver cells and not on hepatocytes solely. This study, like other gene expression studies, was limited by providing key insights into gene expression, but not at the level of post-transcriptional regulation or translation *in vivo*. Unfortunately, the sample size and the present mixed medication history of the included patients with bipolar disorder limited explicit statements of the effect on hepatic gene expression by individual psychotropic drugs. To clarify the effects of single psychotropic drugs on liver tissue in psychiatric disorders, cell culture and animal studies besides human studies are inevitable. Residual biological and non-biological factors, such as genetic predispositions and non-registered clinical characteristics, influencing an individual's gene expression may not be negligible. In particular, genetic predispositions were shown to influence drug metabolism and response with a high degree of heterogeneity [18,216]. Therefore, a potential influence of pharmacogenomics on adverse effects of psychotropic drugs can be assumed and might be included in future studies.

## **4.5 Conclusion**

In summary, the findings of this study revealed altered gene expression upon psychotropic drug treatment in liver tissue of patients with bipolar disorder. Different biological pathways were found to be enriched, whereas predominantly an effect was observed for the cellular energy supply. It might be that the observed downregulation of RETC-related genes could result in cellular senescence, while senescence possibly could be detected in hepatic histopathology. Further affected biological pathways included the metabolism of lipids, amino acids and xenobiotics. All of this probably lead to increased oxidative stress in the cells entailing serious cell damage. In addition, the innate immune system response was found to be enriched. On the one hand, the accumulation of neutrophils was supposed to exist, but on the other hand, a

reduced neutrophil activation was assumed. To which extent a reduced neutrophil degranulation contributes to liver injury in bipolar disorder remains unclear.

Further investigations are necessary to clarify possible detrimental effects of psychotropic drugs on hepatic function in patients with bipolar disorder, whereas the inclusion of genetic factors should be considered in particular, as well as the impact of Western diet. An additional dietary intervention focusing on lipid, glucose and antioxidants intake might contribute to mitigate side effects of psychotropic drug treatment. In order to reveal the fundamental physiology behind complex biological function and dysfunction, the application of bioinformatical analyses of transcriptomics and other ‘-omics’ technologies in combination with wet lab experiments is crucial. Advanced computational analyses are necessary to extract and integrate essential information from the vast amount of such data. All of which could improve the understanding of hepatic function and dysfunction upon psychotropic drug treatment and promote the development of clinically innovative psychotropic drugs with enhanced safety for their therapeutic usage in psychiatric disorders.

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## 6 Appendix

**Table A1. Differentially expressed genes and enriched pathways assessed in liver tissue of patients with bipolar disorder compared to controls.** Annotated genes are listed with their gene symbol, the appropriate HGNC names and identifiers, as well as the corresponding probe ID of the Affymetrix microarray GeneChip Human Genome U133 Plus 2.0. Genes are listed in ascending order with regard to their expression level presented as fold change (FC) and log<sub>2</sub> fold change (log<sub>2</sub>FC). Blue fold changes indicating underexpression and red fold changes indicate overexpression. Adjusted significance levels are presented as false discovery rate (FDR). The respective allocated pathway (P) is denoted with ‘R’ for genes related to the respiratory electron transport chain, ‘L’ for the lipid metabolism, ‘B’ for the biological oxidations, ‘A’ for the amino acid metabolism and ‘I’ for the neutrophil degranulation. [Adapted from [90]]

Gene symbol	HGNC gene name	HGNC ID	Affymetrix probe ID	FC	log <sub>2</sub> FC	FDR	P
<i>FMO3</i>	flavin containing dimethylaniline monooxygenase 3	3771	40665_at	0,22	-2,18	4,56E-04	B
<i>SLC38A4</i>	solute carrier family 38 member 4	14679	220786_s_at	0,24	-2,04	2,01E-03	
<i>TNFSF10</i>	TNF superfamily member 10	11925	214329_x_at	0,25	-2,00	3,90E-06	
<i>CAT</i>	catalase	1516	211922_s_at	0,26	-1,95	5,68E-05	I
<i>TUBE1</i>	tubulin epsilon 1	20775	226181_at	0,27	-1,90	2,00E-06	
<i>ALDH1A1</i>	aldehyde dehydrogenase 1 family member A1	402	212224_at	0,27	-1,87	1,56E-03	B
<i>HPGD</i>	15-hydroxyprostaglandin dehydrogenase	5154	211548_s_at	0,27	-1,87	1,75E-03	L
<i>UGT2B4</i>	UDP glucuronosyltransferase family 2 member B4	12553	206505_at	0,29	-1,78	2,08E-03	B
<i>GYS2</i>	glycogen synthase 2	4707	214621_at	0,30	-1,76	6,07E-04	
<i>SLC41A2</i>	solute carrier family 41 member 2	31045	235299_at	0,30	-1,75	2,00E-06	
<i>NADK2</i>	NAD kinase 2, mitochondrial	26404	229299_at	0,30	-1,74	7,51E-05	
<i>CYR61</i>	cysteine rich angiogenic inducer 61	2654	201289_at	0,30	-1,72	2,12E-02	
<i>DPYD</i>	dihydropyrimidine dehydrogenase	3012	204646_at	0,31	-1,70	6,66E-04	
<i>TADA1</i>	transcriptional adaptor 1	30631	225455_at	0,31	-1,69	2,00E-06	
<i>TNFSF10</i>	TNF superfamily member 10	11925	202687_s_at	0,31	-1,69	4,63E-05	
<i>NADK2</i>	NAD kinase 2, mitochondrial	26404	226946_at	0,31	-1,69	1,70E-04	
<i>AGL</i>	amylase-1, 6-glucosidase, 4-alpha-glucanotransferase	321	203566_s_at	0,31	-1,69	8,90E-04	I
<i>IDH1</i>	isocitrate dehydrogenase (NADP(+)) 1, cytosolic	5382	1555037_a_at	0,31	-1,68	7,76E-04	L/I
<i>PCK2</i>	phosphoenolpyruvate carboxykinase 2, mitochondrial	8725	202847_at	0,32	-1,66	1,14E-03	
<i>HPGD</i>	15-hydroxyprostaglandin dehydrogenase	5154	203914_x_at	0,32	-1,66	2,60E-03	
<i>GPAM</i>	glycerol-3-phosphate acyltransferase, mitochondrial	24865	225424_at	0,32	-1,63	4,43E-02	L
<i>HNMT</i>	histamine N-methyltransferase	5028	204112_s_at	0,33	-1,61	4,31E-04	A
<i>ACBD5</i>	acyl-CoA binding domain containing 5	23338	225663_at	0,33	-1,60	1,01E-03	L
<i>LRRC40</i>	leucine rich repeat containing 40	26004	218577_at	0,33	-1,59	1,46E-04	
<i>ALDH5A1</i>	aldehyde dehydrogenase 5 family member A1	408	203608_at	0,33	-1,59	3,97E-03	
<i>PON1</i>	paraoxonase 1	9204	206345_s_at	0,33	-1,58	8,05E-04	L
<i>RNF128</i>	ring finger protein 128, E3 ubiquitin protein ligase	21153	219263_at	0,33	-1,58	3,63E-03	
<i>SDC2</i>	syndecan 2	10659	212154_at	0,34	-1,57	4,91E-04	
<i>GLS2</i>	glutaminase 2	29570	1564706_s_at	0,34	-1,57	5,29E-04	A
<i>IGFBP3</i>	insulin like growth factor binding protein 3	5472	212143_s_at	0,34	-1,57	1,31E-02	
<i>ART4</i>	ADP-ribosyltransferase 4 (Dombrock blood group)	726	242496_at	0,34	-1,56	5,40E-06	

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<i>CRYZ</i>	crystallin zeta	2419	202950_at	0,34	-1,56	1,48E-04	
<i>SLCO1B3</i>	solute carrier organic anion transporter family member 1B3	10961	206354_at	0,34	-1,56	3,61E-03	L
<i>CXCL12</i>	C-X-C motif chemokine ligand 12	10672	209687_at	0,34	-1,54	1,54E-02	
<i>FAM162A</i>	family with sequence similarity 162 member A	17865	223193_x_at	0,35	-1,52	7,85E-04	
<i>RTN4</i>	reticulon 4	14085	211509_s_at	0,35	-1,51	6,93E-03	
<i>TMEM27</i>	collectrin, amino acid transport regulator	29437	223784_at	0,35	-1,50	2,95E-05	
<i>POSTN</i>	periostin	16953	210809_s_at	0,35	-1,50	8,39E-04	
<i>CA2</i>	carbonic anhydrase 2	1373	209301_at	0,35	-1,50	1,60E-03	
<i>DHTKD1</i>	dehydrogenase E1 and transketolase domain containing 1	23537	227094_at	0,35	-1,50	7,08E-03	A
<i>SERPINC1</i>	serpin family C member 1	775	210049_at	0,35	-1,50	9,41E-03	
<i>HHEX</i>	hematopoietically expressed homeobox	4901	215933_s_at	0,36	-1,49	2,17E-04	
<i>GHR</i>	growth hormone receptor	4263	205498_at	0,36	-1,49	2,16E-02	
<i>GATM</i>	glycine amidinotransferase	4175	216733_s_at	0,36	-1,48	2,71E-03	A
<i>RTN4</i>	reticulon 4	14085	210968_s_at	0,36	-1,47	7,05E-03	
<i>KNG1</i>	kininogen 1	6383	206054_at	0,36	-1,46	6,31E-03	
<i>SULT1E1</i>	sulfotransferase family 1E member 1	11377	222940_at	0,37	-1,45	6,17E-05	B
<i>MME</i>	membrane metalloendopeptidase	7154	203434_s_at	0,37	-1,42	5,57E-05	I
<i>HIBADH</i>	3-hydroxyisobutyrate dehydrogenase	4907	224812_at	0,37	-1,42	7,50E-04	A
<i>ANK3</i>	ankyrin 3	494	206385_s_at	0,37	-1,42	3,20E-03	
<i>HPR</i>	haptoglobin-related protein	5156	208471_at	0,37	-1,42	7,02E-03	
<i>SMARCA1</i>	SWI/SNF related, matrix associated, actin dependent regulator of chromatin, subfamily a, member 1	11097	203874_s_at	0,38	-1,41	3,22E-05	
<i>UFL1</i>	UFM1 specific ligase 1	23039	212633_at	0,38	-1,40	4,09E-04	
<i>SLC22A7</i>	solute carrier family 22 member 7	10971	231398_at	0,38	-1,39	2,50E-02	
<i>ELOVL2</i>	ELOVL fatty acid elongase 2	14416	213712_at	0,38	-1,38	2,28E-04	L
<i>UGT2B15</i>	UDP glucuronosyltransferase family 2 member B15	12546	207392_x_at	0,38	-1,38	4,36E-03	B
<i>KLHDC10</i>	kelch domain containing 10	22194	210111_s_at	0,38	-1,38	6,75E-03	
<i>PRKAB2</i>	protein kinase AMP-activated non-catalytic subunit beta 2	9379	225278_at	0,39	-1,37	1,27E-02	L
<i>PGRMC1</i>	progesterone receptor membrane component 1	16090	201120_s_at	0,39	-1,36	1,23E-03	I
<i>UGT2B28</i>	UDP glucuronosyltransferase family 2 member B28	13479	211682_x_at	0,39	-1,36	4,00E-03	
<i>GLUD1</i>	glutamate dehydrogenase 1	4335	215794_x_at	0,39	-1,35	5,12E-04	A
<i>SLCO1B1</i>	solute carrier organic anion transporter family member 1B1	10959	210366_at	0,39	-1,35	2,26E-03	L
<i>ABCG5</i>	ATP binding cassette subfamily G member 5	13886	220383_at	0,40	-1,34	1,21E-03	L
<i>KNG1</i>	kininogen 1	6383	217512_at	0,40	-1,34	1,84E-03	
<i>AADAC</i>	arylacetamide deacetylase	17	205969_at	0,40	-1,34	3,23E-03	B
<i>CYP51A1</i>	cytochrome P450 family 51 subfamily A member 1	2649	202314_at	0,40	-1,34	8,04E-03	
<i>CMBL</i>	carboxymethylenebutenolidase homolog	25090	227522_at	0,40	-1,34	1,66E-02	B
<i>AKAP9</i>	A-kinase anchoring protein 9	379	210962_s_at	0,40	-1,33	8,49E-04	
<i>HMGB1</i>	high mobility group box 1	4983	224731_at	0,40	-1,32	7,39E-04	I
<i>GOLIM4</i>	golgi integral membrane protein 4	15448	238002_at	0,40	-1,32	1,16E-03	
<i>CXCL2</i>	C-X-C motif chemokine ligand 2	4603	209774_x_at	0,40	-1,31	3,39E-02	
<i>SNX4</i>	sorting nexin 4	11175	212652_s_at	0,41	-1,30	1,73E-04	

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<i>SLC35A3</i>	solute carrier family 35 member A3	11023	226894_at	0,41	-1,30	4,48E-04	
<i>ACOT13</i>	acyl-CoA thioesterase 13	20999	204565_at	0,41	-1,30	1,01E-03	L
<i>C6</i>	complement C6	1339	210168_at	0,41	-1,30	3,23E-03	
<i>AMACR</i>	alpha-methylacyl-CoA racemase	451	209425_at	0,41	-1,30	3,57E-03	L
<i>DECRI</i>	2,4-dienoyl-CoA reductase 1	2753	202447_at	0,41	-1,30	3,63E-03	L
<i>CYP3A5</i>	cytochrome P450 family 3 subfamily A member 5	2638	205765_at	0,41	-1,30	2,17E-02	B
<i>NUDT7</i>	nudix hydrolase 7	8054	228855_at	0,41	-1,29	2,17E-04	L
<i>C1RL</i>	complement C1r subcomponent like	21265	218983_at	0,41	-1,29	1,48E-03	
<i>ACSM2A</i>	acyl-CoA synthetase medium chain family member 2A	32017	214069_at	0,41	-1,29	2,66E-03	
<i>NR2F2</i>	nuclear receptor subfamily 2 group F member 2	7976	209120_at	0,41	-1,29	3,73E-03	
<i>HFE2</i>	hemojuvelin BMP co-receptor	4887	228621_at	0,41	-1,28	5,45E-04	
<i>CRLS1</i>	cardiolipin synthase 1	16148	223978_s_at	0,41	-1,28	6,07E-04	L
<i>PANK3</i>	pantothenate kinase 3	19365	221751_at	0,41	-1,28	6,90E-03	
<i>LGR4</i>	leucine rich repeat containing G protein-coupled receptor 4	13299	218326_s_at	0,41	-1,27	1,18E-04	
<i>APOM</i>	apolipoprotein M	13916	214910_s_at	0,41	-1,27	2,23E-03	
<i>ETFA</i>	electron transfer flavoprotein subunit alpha	3481	201931_at	0,41	-1,27	3,08E-03	R
<i>UBR3</i>	ubiquitin protein ligase E3 component n-recognin 3	30467	230029_x_at	0,42	-1,26	9,60E-04	
<i>LRP6</i>	LDL receptor related protein 6	6698	225745_at	0,42	-1,26	1,90E-03	
<i>FNI</i>	fibronectin 1	3778	216442_x_at	0,42	-1,26	9,25E-03	
<i>TSPAN12</i>	tetraspanin 12	21641	219274_at	0,42	-1,25	2,47E-05	
<i>ANO6</i>	anoctamin 6	25240	224906_at	0,42	-1,25	1,64E-04	I
<i>KAT2B</i>	lysine acetyltransferase 2B	8638	203845_at	0,42	-1,25	5,65E-04	
<i>ACSS3</i>	acyl-CoA synthetase short chain family member 3	24723	229222_at	0,42	-1,25	8,16E-04	L
<i>PSMA2</i>	proteasome subunit alpha 2	9531	201317_s_at	0,42	-1,25	3,48E-03	A/I
<i>FNI</i>	fibronectin 1	3778	210495_x_at	0,42	-1,25	9,32E-03	
<i>PPP3CA</i>	protein phosphatase 3 catalytic subunit alpha	9314	202429_s_at	0,42	-1,24	2,86E-04	
<i>TBC1D9</i>	TBC1 domain family member 9	21710	212956_at	0,42	-1,24	4,56E-04	
<i>CPEB2</i>	cytoplasmic polyadenylation element binding protein 2	21745	226939_at	0,42	-1,24	3,92E-03	
<i>FNI</i>	fibronectin 1	3778	212464_s_at	0,42	-1,24	1,32E-02	
<i>SOD1</i>	superoxide dismutase 1	11179	200642_at	0,42	-1,24	1,41E-02	
<i>FBXL5</i>	F-box and leucine rich repeat protein 5	13602	209004_s_at	0,43	-1,23	3,20E-03	
<i>HRG</i>	histidine rich glycoprotein	5181	206226_at	0,43	-1,23	6,66E-03	
<i>HMGCs2</i>	3-hydroxy-3-methylglutaryl-CoA synthase 2	5008	204607_at	0,43	-1,23	1,34E-02	L
<i>HLTF</i>	helicase like transcription factor	11099	202983_at	0,43	-1,22	1,46E-04	
<i>NAT8</i>	N-acetyltransferase 8 (putative)	18069	210289_at	0,43	-1,22	5,29E-04	
<i>ARHGAP29</i>	Rho GTPase activating protein 29	30207	203910_at	0,43	-1,22	3,40E-03	
<i>PRDX5</i>	peroxiredoxin 5	9355	1560587_s_at	0,43	-1,21	2,83E-03	
<i>LYPLA1</i>	lysophospholipase 1	6737	212449_s_at	0,43	-1,21	3,14E-03	
<i>CFHR2</i>	complement factor H related 2	4890	206910_x_at	0,43	-1,21	1,58E-02	
<i>AHSG</i>	alpha 2-HS glycoprotein	349	204551_s_at	0,43	-1,21	1,90E-02	I
<i>PAWR</i>	pro-apoptotic WT1 regulator	8614	226223_at	0,44	-1,20	2,86E-04	

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<i>TDO2</i>	tryptophan 2,3-dioxygenase	11708	205943_at	0,44	-1,20	1,48E-02	A
<i>LRPPRC</i>	leucine rich pentatricopeptide repeat containing	15714	211971_s_at	0,44	-1,19	9,59E-04	R
<i>IMPA1</i>	inositol monophosphatase 1	6050	203011_at	0,44	-1,19	1,27E-03	
<i>CYP51A1</i>	cytochrome P450 family 51 subfamily A member 1	2649	216607_s_at	0,44	-1,19	1,40E-03	
<i>ESD</i>	esterase D	3465	215096_s_at	0,44	-1,19	3,92E-03	B
<i>RALGPS2</i>	Ral GEF with PH domain and SH3 binding motif 2	30279	227533_at	0,44	-1,18	3,03E-04	
<i>ABCG2</i>	ATP binding cassette subfamily G member 2 (Junior blood group)	74	209735_at	0,44	-1,18	7,50E-04	
<i>SCOC</i>	short coiled-coil protein	20335	224786_at	0,44	-1,18	1,01E-03	
<i>TMEM106B</i>	transmembrane protein 106B	22407	226529_at	0,44	-1,18	1,40E-03	
<i>MAOB</i>	monoamine oxidase B	6834	204041_at	0,44	-1,18	1,42E-03	B
<i>APOB</i>	apolipoprotein B	603	223579_s_at	0,44	-1,18	5,47E-03	L
<i>MTHFD1</i>	methylenetetrahydrofolate dehydrogenase, cyclohydrolase and formyltetrahydrofolate synthetase 1	7432	202309_at	0,44	-1,17	3,08E-03	
<i>GBE1</i>	1,4-alpha-glucan branching enzyme 1	4180	203282_at	0,44	-1,17	1,29E-02	
<i>COP5</i>	COP9 signalosome subunit 5	2240	201652_at	0,45	-1,16	2,98E-04	
<i>FMO5</i>	flavin containing monooxygenase 5	3773	205776_at	0,45	-1,16	1,30E-03	
<i>EPHX2</i>	epoxide hydrolase 2	3402	209368_at	0,45	-1,16	1,90E-03	L
<i>C3</i>	complement C3	1318	217767_at	0,45	-1,16	1,25E-02	I
<i>APOA2</i>	apolipoprotein A2	601	219465_at	0,45	-1,16	2,40E-02	L
<i>PYGL</i>	glycogen phosphorylase L	9725	202990_at	0,45	-1,15	1,70E-04	I
<i>RGN</i>	regucalcin	9989	210751_s_at	0,45	-1,15	3,22E-04	
<i>METAP2</i>	methionyl aminopeptidase 2	16672	209861_s_at	0,45	-1,15	5,33E-04	
<i>HAOI</i>	hydroxyacid oxidase 1	4809	220224_at	0,45	-1,15	1,90E-03	A
<i>APOM</i>	apolipoprotein M	13916	205682_x_at	0,45	-1,15	3,24E-03	
<i>CMTM6</i>	CKLF like MARVEL transmembrane domain containing 6	19177	223047_at	0,45	-1,15	1,02E-02	I
<i>HACL1</i>	2-hydroxyacyl-CoA lyase 1	17856	223211_at	0,45	-1,14	5,68E-05	L
<i>SMG1</i>	SMG1, nonsense mediated mRNA decay associated PI3K related kinase	30045	224842_at	0,45	-1,14	1,04E-03	
<i>OSBPL9</i>	oxysterol binding protein like 9	16386	218047_at	0,45	-1,14	2,20E-03	L
<i>EID1</i>	EP300 interacting inhibitor of differentiation 1	1191	208669_s_at	0,45	-1,14	3,61E-03	
<i>GNMT</i>	glycine N-methyltransferase	4415	210328_at	0,45	-1,14	3,07E-02	A
<i>TMEM19</i>	transmembrane protein 19	25605	226860_at	0,46	-1,13	1,15E-03	
<i>UPB1</i>	beta-ureidopropionase 1	16297	224043_s_at	0,46	-1,13	2,28E-03	
<i>BAAT</i>	bile acid-CoA:amino acid N-acyltransferase	932	206913_at	0,46	-1,13	6,27E-03	L
<i>GCSH</i>	glycine cleavage system protein H	4208	213129_s_at	0,46	-1,13	8,48E-03	A
<i>RBL2</i>	RB transcriptional corepressor like 2	9894	212331_at	0,46	-1,12	1,46E-04	
<i>PDE8A</i>	phosphodiesterase 8A	8793	212522_at	0,46	-1,12	6,07E-04	
<i>ENOSF1</i>	enolase superfamily member 1	30365	204143_s_at	0,46	-1,12	1,68E-03	
<i>PSMG2</i>	proteasome assembly chaperone 2	24929	218467_at	0,46	-1,12	2,09E-03	
<i>CPB2</i>	carboxypeptidase B2	2300	206651_s_at	0,46	-1,12	1,25E-02	
<i>FDPS</i>	farnesyl diphosphate synthase	3631	201275_at	0,46	-1,12	1,54E-02	L
<i>FMCI</i>	formation of mitochondrial complex V assembly factor 1 homolog	26946	226780_s_at	0,46	-1,11	8,52E-04	

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<i>SMIM15</i>	small integral membrane protein 15	33861	224740_at	0,46	-1,11	4,28E-03	
<i>ABCB4</i>	ATP binding cassette subfamily B member 4	45	207819_s_at	0,46	-1,11	5,88E-03	L
<i>OCIAD1</i>	OCIA domain containing 1	16074	223011_s_at	0,46	-1,11	7,10E-03	
<i>RBP4</i>	retinol binding protein 4	9922	219140_s_at	0,46	-1,11	4,00E-02	
<i>RABGAP1L</i>	RAB GTPase activating protein 1 like	24663	203020_at	0,47	-1,10	5,68E-05	
<i>BAG4</i>	BCL2 associated athanogene 4	940	228189_at	0,47	-1,10	2,77E-03	
<i>CREG1</i>	cellular repressor of E1A stimulated genes 1	2351	201200_at	0,47	-1,10	4,15E-03	I
<i>TM4SF4</i>	transmembrane 4 L six family member 4	11856	209937_at	0,47	-1,10	7,45E-03	
<i>HSD17B11</i>	hydroxysteroid 17-beta dehydrogenase 11	22960	217989_at	0,47	-1,10	9,26E-03	L
<i>TMEM64</i>	transmembrane protein 64	25441	225974_at	0,47	-1,09	1,83E-03	
<i>ABCC2</i>	ATP binding cassette subfamily C member 2	53	206155_at	0,47	-1,09	2,03E-03	
<i>SMLR1</i>	small leucine rich protein 1	44670	230716_at	0,47	-1,09	3,43E-03	
<i>FAM96A</i>	cytosolic iron-sulfur assembly component 2A	26235	224779_s_at	0,47	-1,09	7,16E-03	
<i>MSRB2</i>	methionine sulfoxide reductase B2	17061	218773_s_at	0,47	-1,09	1,17E-02	
<i>CFHR5</i>	complement factor H related 5	24668	208088_s_at	0,47	-1,09	1,21E-02	
<i>SRD5A2</i>	steroid 5 alpha-reductase 2	11285	235945_at	0,47	-1,08	6,80E-04	L
<i>ENPEP</i>	glutamyl aminopeptidase	3355	204844_at	0,47	-1,08	7,95E-04	
<i>VPS13C</i>	vacuolar protein sorting 13 homolog C	23594	218396_at	0,47	-1,08	1,09E-03	
<i>PCMTD1</i>	protein-L-isoaspartate (D-aspartate) O-methyltransferase domain containing 1	30483	226119_at	0,47	-1,08	1,60E-03	
<i>TMEM14C</i>	transmembrane protein 14C	20952	223106_at	0,47	-1,08	1,83E-03	
<i>TSPYL1</i>	TSPY like 1	12382	221493_at	0,47	-1,08	3,39E-03	
<i>HGD</i>	homogentisate 1,2-dioxygenase	4892	205221_at	0,47	-1,08	1,21E-02	A
<i>ADII</i>	acireductone dioxygenase 1	30576	217761_at	0,47	-1,08	1,21E-02	A
<i>GNE</i>	glucosamine (UDP-N-acetyl)-2-epimerase/N-acetylmannosamine kinase	23657	205042_at	0,47	-1,08	2,49E-02	
<i>SKP2</i>	S-phase kinase associated protein 2	10901	203625_x_at	0,48	-1,07	4,80E-04	
<i>AGTR1</i>	angiotensin II receptor type 1	336	208016_s_at	0,48	-1,07	7,50E-04	
<i>ARSE</i>	arylsulfatase E	719	205894_at	0,48	-1,07	1,80E-03	L
<i>RMDN1</i>	regulator of microtubule dynamics 1	24285	218549_s_at	0,48	-1,07	2,13E-03	
<i>CHN2</i>	chimerin 2	1944	213385_at	0,48	-1,07	2,88E-03	
<i>RHEB</i>	Ras homolog, mTORC1 binding	10011	213404_s_at	0,48	-1,07	3,49E-03	
<i>SSBP1</i>	single stranded DNA binding protein 1	11317	202591_s_at	0,48	-1,07	1,27E-02	
<i>CFB</i>	complement factor B	1037	202357_s_at	0,48	-1,07	3,66E-02	
<i>AMACR</i>	alpha-methylacyl-CoA racemase	451	209426_s_at	0,48	-1,06	9,00E-04	
<i>FAM35A</i>	shieldin complex subunit 2	28773	220547_s_at	0,48	-1,06	4,41E-03	
<i>TMEM97</i>	transmembrane protein 97	28106	212282_at	0,48	-1,06	1,64E-02	
<i>SUB1</i>	SUB1 homolog, transcriptional regulator	19985	224587_at	0,48	-1,05	4,56E-04	
<i>REEP5</i>	receptor accessory protein 5	30077	208873_s_at	0,48	-1,05	2,01E-03	
<i>WASHC4</i>	WASH complex subunit 4	29174	212795_at	0,48	-1,05	2,61E-03	
<i>HHIP</i>	hedgehog interacting protein	14866	230135_at	0,48	-1,05	3,29E-03	
<i>SLC35D1</i>	solute carrier family 35 member D1	20800	209712_at	0,48	-1,05	2,60E-02	B
<i>SLC3A1</i>	solute carrier family 3 member 1	11025	205799_s_at	0,48	-1,05	2,63E-02	

Gene symbol	HGNC gene name	HGNC ID	Affymetrix probe ID	FC	log <sub>2</sub> FC	FDR	P
<i>SLC35B3</i>	solute carrier family 35 member B3	21601	222691_at	0,49	-1,04	3,30E-04	B
<i>SPG11</i>	SPG11, spatacsin vesicle trafficking associated	11226	203513_at	0,49	-1,04	1,60E-03	
<i>TMEM30A</i>	transmembrane protein 30A	16667	217743_s_at	0,49	-1,04	2,04E-03	I
<i>LSM5</i>	LSM5 homolog, U6 small nuclear RNA and mRNA degradation associated	17162	211747_s_at	0,49	-1,04	2,65E-03	
<i>PIK3R4</i>	phosphoinositide-3-kinase regulatory subunit 4	8982	212740_at	0,49	-1,04	2,72E-03	L
<i>CYP2D6</i>	cytochrome P450 family 2 subfamily D member 6	2625	207498_s_at	0,49	-1,04	1,15E-02	
<i>NDUFB4</i>	NADH:ubiquinone oxidoreductase subunit B4	7699	218226_s_at	0,49	-1,04	1,19E-02	R
<i>ALDH2</i>	aldehyde dehydrogenase 2 family member	404	201425_at	0,49	-1,04	1,35E-02	B
<i>CP</i>	ceruloplasmin	2295	204846_at	0,49	-1,04	3,75E-02	
<i>NT5DC1</i>	5'-nucleotidase domain containing 1	21556	223177_at	0,49	-1,03	5,68E-05	
<i>BDH1</i>	3-hydroxybutyrate dehydrogenase 1	1027	211715_s_at	0,49	-1,03	2,83E-03	L
<i>PSMD6</i>	proteasome 26S subunit, non-ATPase 6	9564	202753_at	0,49	-1,03	1,01E-02	A/I
<i>ARL6IP1</i>	ADP ribosylation factor like GTPase 6 interacting protein 1	697	211935_at	0,49	-1,03	1,48E-02	
<i>PLS3</i>	plastin 3	9091	201215_at	0,49	-1,03	2,91E-02	
<i>RTN4</i>	reticulon 4	14085	214629_x_at	0,49	-1,03	3,06E-02	
<i>GLOD4</i>	glyoxalase domain containing 4	14111	209092_s_at	0,49	-1,02	1,42E-03	
<i>PLEKHA5</i>	pleckstrin homology domain containing A5	30036	220952_s_at	0,49	-1,02	3,04E-03	L
<i>ALG5</i>	ALG5, dolichyl-phosphate beta-glucosyltransferase	20266	218203_at	0,49	-1,02	4,37E-03	
<i>SEC63</i>	SEC63 homolog, protein translocation regulator	21082	201916_s_at	0,49	-1,02	6,95E-03	
<i>SF3B1</i>	splicing factor 3b subunit 1	10768	211185_s_at	0,49	-1,02	1,24E-02	
<i>UGT1A5</i>	UDP glucuronosyltransferase family 1 member A5	12537	208596_s_at	0,49	-1,02	3,54E-02	
<i>CDH1</i>	cadherin 1	1748	201131_s_at	0,49	-1,02	4,87E-02	
<i>ZMYM4</i>	zinc finger MYM-type containing 4	13055	202051_s_at	0,50	-1,01	7,10E-06	
<i>CPNE3</i>	copine 3	2316	202118_s_at	0,50	-1,01	5,18E-04	L/I
<i>MARCH6</i>	membrane associated ring-CH-type finger 6	30550	212498_at	0,50	-1,01	1,57E-03	
<i>CHP1</i>	calcineurin like EF-hand protein 1	17433	214665_s_at	0,50	-1,01	8,41E-03	
<i>ANAPC16</i>	anaphase promoting complex subunit 16	26976	229145_at	0,50	-1,01	1,39E-02	
<i>ATP5G3</i>	ATP synthase membrane subunit c locus 3	843	207507_s_at	0,50	-1,01	1,66E-02	R
<i>N4BP2L2</i>	NEDD4 binding protein 2 like 2	26916	202258_s_at	0,50	-1,00	9,60E-04	
<i>ATXN10</i>	ataxin 10	10549	208833_s_at	0,50	-1,00	1,09E-03	
<i>PCBD1</i>	pterin-4 alpha-carbinolamine dehydratase 1	8646	203557_s_at	0,50	-1,00	4,98E-03	A
<i>TMEM14B</i>	transmembrane protein 14B	21384	221452_s_at	0,50	-1,00	5,53E-03	
<i>SLTM</i>	SAFB like transcription modulator	20709	217828_at	0,50	-1,00	7,33E-03	
<i>ABCB4</i>	ATP binding cassette subfamily B member 4	45	209994_s_at	0,50	-1,00	1,71E-02	
<i>ABCC6</i>	ATP binding cassette subfamily C member 6	57	214033_at	0,50	-0,99	3,78E-04	
<i>SLC25A5</i>	solute carrier family 25 member 5	10991	200657_at	0,50	-0,99	7,55E-03	
<i>ATP5F1</i>	ATP synthase peripheral stalk-membrane subunit b	840	211755_s_at	0,50	-0,99	8,41E-03	R
<i>CD302</i>	CD302 molecule	30843	203799_at	0,50	-0,99	1,22E-02	
<i>SDHB</i>	succinate dehydrogenase complex iron sulfur subunit B	10681	202675_at	0,50	-0,99	1,76E-02	R
<i>TMX4</i>	thioredoxin related transmembrane protein 4	25237	201581_at	0,51	-0,98	3,29E-03	
<i>ATP5O</i>	ATP synthase peripheral stalk subunit OSCP	850	200818_at	0,51	-0,98	9,93E-03	R

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<i>KLKB1</i>	kallikrein B1	6371	206541_at	0,51	-0,98	1,58E-02	
<i>ACSS2</i>	acyl-CoA synthetase short chain family member 2	15814	234312_s_at	0,51	-0,98	3,25E-02	B
<i>SLC2A13</i>	solute carrier family 2 member 13	15956	227176_at	0,51	-0,97	1,97E-04	
<i>ZNF638</i>	zinc finger protein 638	17894	211257_x_at	0,51	-0,97	2,20E-03	
<i>ITGA1</i>	integrin subunit alpha 1	6134	226731_at	0,51	-0,97	3,53E-03	
<i>HP1BP3</i>	heterochromatin protein 1 binding protein 3	24973	224591_at	0,51	-0,97	5,46E-03	
<i>RAB18</i>	RAB18, member RAS oncogene family	14244	224787_s_at	0,51	-0,97	5,51E-03	I
<i>SIGIRR</i>	single Ig and TIR domain containing	30575	52940_at	0,51	-0,97	1,14E-02	
<i>WRB</i>	tryptophan rich basic protein	12790	202749_at	0,51	-0,96	2,83E-04	
<i>VPS4B</i>	vacuolar protein sorting 4 homolog B	10895	218171_at	0,51	-0,96	7,85E-04	
<i>EMC4</i>	ER membrane protein complex subunit 4	28032	223043_at	0,51	-0,96	3,68E-03	
<i>GDA</i>	guanine deaminase	4212	224209_s_at	0,51	-0,96	7,62E-03	
<i>MFAP3L</i>	microfibril associated protein 3 like	29083	205442_at	0,51	-0,96	1,36E-02	
<i>PANK1</i>	pantothenate kinase 1	8598	226649_at	0,51	-0,96	4,36E-02	
<i>CLCN3</i>	chloride voltage-gated channel 3	2021	201734_at	0,52	-0,95	4,23E-04	
<i>THOC7</i>	THO complex 7	29874	218334_at	0,52	-0,95	1,09E-03	
<i>COL1A2</i>	collagen type 1 alpha 2 chain	2198	202403_s_at	0,52	-0,95	5,06E-03	
<i>TCEA1</i>	transcription elongation factor A1	11612	216241_s_at	0,52	-0,95	5,35E-03	
<i>SORD</i>	sorbitol dehydrogenase	11184	201562_s_at	0,52	-0,95	5,93E-03	
<i>PTPRK</i>	protein tyrosine phosphatase, receptor type K	9674	203038_at	0,52	-0,95	3,41E-02	
<i>CAAPI1</i>	caspase activity and apoptosis inhibitor 1	25834	219276_x_at	0,52	-0,94	2,15E-04	
<i>USP9X</i>	ubiquitin specific peptidase 9 X-linked	12632	201099_at	0,52	-0,94	1,76E-03	
<i>UBE2E1</i>	ubiquitin conjugating enzyme E2 E1	12477	212519_at	0,52	-0,94	1,92E-03	
<i>CYP2J2</i>	cytochrome P450 family 2 subfamily J member 2	2634	205073_at	0,52	-0,94	2,03E-03	B/L
<i>PEX13</i>	peroxisomal biogenesis factor 13	8855	235093_at	0,52	-0,94	2,58E-03	
<i>SS18L1</i>	SS18L1, nBAF chromatin remodeling complex subunit	15592	213140_s_at	0,52	-0,94	8,02E-03	
<i>PCSK6</i>	proprotein convertase subtilisin/kexin type 6	8569	207414_s_at	0,52	-0,94	1,20E-02	L
<i>NT5E</i>	5'-nucleotidase ecto	8021	203939_at	0,52	-0,93	3,97E-04	
<i>CGGBP1</i>	CGG triplet repeat binding protein 1	1888	224599_at	0,52	-0,93	1,04E-03	
<i>NR1I3</i>	nuclear receptor subfamily 1 group I member 3	7969	207007_at	0,52	-0,93	1,37E-03	
<i>MBTPS2</i>	membrane bound transcription factor peptidase, site 2	15455	226760_at	0,52	-0,93	4,73E-03	L
<i>SEC31A</i>	SEC31 homolog A, COPII coat complex component	17052	200945_s_at	0,52	-0,93	2,42E-02	
<i>TMEM176B</i>	transmembrane protein 176B	29596	220532_s_at	0,52	-0,93	4,53E-02	
<i>IFIT1</i>	interferon induced protein with tetratricopeptide repeats 1	5407	203153_at	0,53	-0,92	1,08E-03	
<i>RALGPS2</i>	Ral GEF with PH domain and SH3 binding motif 2	30279	227224_at	0,53	-0,92	1,12E-03	
<i>ATP5J2</i>	ATP synthase membrane subunit f	848	202961_s_at	0,53	-0,92	9,52E-03	R
<i>LIPA</i>	lipase A, lysosomal acid type	6617	201847_at	0,53	-0,92	1,60E-02	L
<i>FAM120A</i>	family with sequence similarity 120A	13247	200774_at	0,53	-0,92	3,60E-02	
<i>DCUN1D1</i>	defective in cullin neddylation 1 domain containing 1	18184	218583_s_at	0,53	-0,91	2,06E-04	
<i>MYH10</i>	myosin heavy chain 10	7568	212372_at	0,53	-0,91	1,27E-03	
<i>SERBP1</i>	SERPINE1 mRNA binding protein 1	17860	217724_at	0,53	-0,91	2,08E-03	

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<i>MEST</i>	mesoderm specific transcript	7028	202016_at	0,53	-0,91	3,15E-03	
<i>DEK</i>	DEK proto-oncogene	2768	200934_at	0,53	-0,91	3,65E-03	
<i>GGCX</i>	gamma-glutamyl carboxylase	4247	205351_at	0,53	-0,91	3,84E-03	
<i>NDUFB3</i>	NADH:ubiquinone oxidoreductase subunit B3	7698	203371_s_at	0,53	-0,91	4,64E-03	R
<i>CRLS1</i>	cardiolipin synthase 1	16148	225324_at	0,53	-0,91	5,60E-03	
<i>ALDH9A1</i>	aldehyde dehydrogenase 9 family member A1	412	201612_at	0,53	-0,91	6,38E-03	A
<i>RNF144B</i>	ring finger protein 144B	21578	228153_at	0,53	-0,91	6,61E-03	
<i>NMD3</i>	NMD3 ribosome export adaptor	24250	231870_s_at	0,53	-0,91	1,18E-02	
<i>CD164</i>	CD164 molecule	1632	208405_s_at	0,53	-0,91	1,49E-02	
<i>MAOA</i>	monoamine oxidase A	6833	212741_at	0,53	-0,91	1,74E-02	B
<i>SPP2</i>	secreted phosphoprotein 2	11256	214478_at	0,53	-0,91	1,83E-02	
<i>GHITM</i>	growth hormone inducible transmembrane protein	17281	209248_at	0,53	-0,91	1,93E-02	
<i>MALSU1</i>	mitochondrial assembly of ribosomal large subunit 1	21721	226385_s_at	0,53	-0,91	3,38E-02	
<i>AIFM1</i>	apoptosis inducing factor mitochondria associated 1	8768	205512_s_at	0,54	-0,90	6,52E-04	
<i>ZNF644</i>	zinc finger protein 644	29222	222580_at	0,54	-0,90	2,87E-03	
<i>LIFR</i>	LIF receptor alpha	6597	225575_at	0,54	-0,90	7,89E-03	
<i>CDKN1B</i>	cyclin dependent kinase inhibitor 1B	1785	209112_at	0,54	-0,90	8,48E-03	
<i>ATP5G3</i>	ATP synthase membrane subunit c locus 3	843	207508_at	0,54	-0,90	1,29E-02	
<i>MBNL1</i>	muscleblind like splicing regulator 1	6923	201153_s_at	0,54	-0,90	1,36E-02	
<i>TXNL1</i>	thioredoxin like 1	12436	201588_at	0,54	-0,90	1,94E-02	
<i>RNF5</i>	ring finger protein 5	10068	209111_at	0,54	-0,89	4,21E-04	
<i>RBBP7</i>	RB binding protein 7, chromatin remodeling factor	9890	201092_at	0,54	-0,89	9,60E-04	
<i>ACO1</i>	aconitase 1	117	207071_s_at	0,54	-0,89	1,41E-03	
<i>CTSO</i>	cathepsin O	2542	203758_at	0,54	-0,89	2,49E-03	
<i>KDELRL2</i>	KDEL endoplasmic reticulum protein retention receptor 2	6305	200698_at	0,54	-0,89	1,54E-02	
<i>TJPI1</i>	tight junction protein 1	11827	202011_at	0,54	-0,89	2,01E-02	
<i>NCOA4</i>	nuclear receptor coactivator 4	7671	210774_s_at	0,54	-0,89	2,23E-02	
<i>ANPEP</i>	alanyl aminopeptidase, membrane	500	202888_s_at	0,54	-0,89	4,50E-02	I
<i>MME</i>	membrane metalloendopeptidase	7154	203435_s_at	0,54	-0,88	2,47E-04	
<i>FUCA1</i>	alpha-L-fucosidase 1	4006	202838_at	0,54	-0,88	9,59E-04	I
<i>CAMTA1</i>	calmodulin binding transcription activator 1	18806	225693_s_at	0,54	-0,88	1,30E-03	
<i>SLC39A6</i>	solute carrier family 39 member 6	18607	202088_at	0,54	-0,88	1,37E-03	
<i>NUDT5</i>	nudix hydrolase 5	8052	223100_s_at	0,54	-0,88	3,25E-03	
<i>NDUFAF1</i>	NADH:ubiquinone oxidoreductase complex assembly factor 1	18828	204125_at	0,54	-0,88	3,36E-03	R
<i>YME1L1</i>	YME1 like 1 ATPase	12843	201352_at	0,54	-0,88	4,31E-03	
<i>PDGFC</i>	platelet derived growth factor C	8801	218718_at	0,54	-0,88	5,22E-03	
<i>HOOK1</i>	hook microtubule tethering protein 1	19884	225792_at	0,54	-0,88	5,36E-03	
<i>NFIA</i>	nuclear factor I A	7784	224976_at	0,54	-0,88	7,34E-03	
<i>EIF2AK1</i>	eukaryotic translation initiation factor 2 alpha kinase 1	24921	217736_s_at	0,54	-0,88	8,67E-03	
<i>TUBB</i>	tubulin beta class I	20778	211714_x_at	0,54	-0,88	1,23E-02	I
<i>DNAJC15</i>	DnaJ heat shock protein family (Hsp40) member C15	20325	218435_at	0,54	-0,88	1,79E-02	

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<i>PGM1</i>	phosphoglucomutase 1	8905	201968_s_at	0,54	-0,88	1,86E-02	I
<i>TGOLN2</i>	trans-golgi network protein 2	15450	212043_at	0,54	-0,88	3,98E-02	
<i>DIAPH2</i>	diaphanous related formin 2	2877	205726_at	0,55	-0,87	9,46E-04	
<i>EPB41L5</i>	erythrocyte membrane protein band 4.1 like 5	19819	229292_at	0,55	-0,87	1,82E-03	
<i>PDCD10</i>	programmed cell death 10	8761	210907_s_at	0,55	-0,87	2,04E-03	
<i>MTIF2</i>	mitochondrial translational initiation factor 2	7441	203095_at	0,55	-0,87	2,70E-03	
<i>HDHD2</i>	haloacid dehalogenase like hydrolase domain containing 2	25364	223155_at	0,55	-0,87	2,88E-03	
<i>MOCS2</i>	molybdenum cofactor synthesis 2	7193	218212_s_at	0,55	-0,87	3,22E-03	
<i>ZMPSTE24</i>	zinc metallopeptidase STE24	12877	202939_at	0,55	-0,87	8,43E-03	
<i>ATL2</i>	atlastin GTPase 2	24047	222700_at	0,55	-0,87	9,93E-03	
<i>KARS</i>	lysyl-tRNA synthetase	6215	200840_at	0,55	-0,87	2,03E-02	A
<i>TIMM23</i>	translocase of inner mitochondrial membrane 23	17312	225535_s_at	0,55	-0,87	3,22E-02	
<i>ALCAM</i>	activated leukocyte cell adhesion molecule	400	201952_at	0,55	-0,86	9,58E-04	
<i>MAOA</i>	monoamine oxidase A	6833	204388_s_at	0,55	-0,86	3,96E-03	
<i>PCGF5</i>	polycomb group ring finger 5	28264	226326_at	0,55	-0,86	8,42E-03	
<i>ANAPC16</i>	anaphase promoting complex subunit 16	26976	224664_at	0,55	-0,86	9,16E-03	
<i>MAP2K1</i>	mitogen-activated protein kinase kinase 1	6840	202670_at	0,55	-0,86	1,13E-02	
<i>ZC3H15</i>	zinc finger CCCH-type containing 15	29528	201595_s_at	0,55	-0,86	1,62E-02	
<i>SUN1</i>	Sad1 and UNC84 domain containing 1	18587	212074_at	0,55	-0,86	1,63E-02	
<i>MCCCI</i>	methylcrotonoyl-CoA carboxylase 1	6936	218440_at	0,55	-0,85	4,47E-04	A
<i>CREBL2</i>	cAMP responsive element binding protein like 2	2350	201989_s_at	0,55	-0,85	4,61E-04	
<i>MRPL15</i>	mitochondrial ribosomal protein L15	14054	218027_at	0,55	-0,85	4,31E-03	
<i>VPS29</i>	VPS29, retromer complex component	14340	223026_s_at	0,55	-0,85	1,21E-02	
<i>SLC4A4</i>	solute carrier family 4 member 4	11030	203908_at	0,55	-0,85	1,33E-02	
<i>TM9SF3</i>	transmembrane 9 superfamily member 3	21529	222399_s_at	0,55	-0,85	1,64E-02	
<i>YPEL5</i>	yippee like 5	18329	222408_s_at	0,55	-0,85	1,94E-02	I
<i>RORA</i>	RAR related orphan receptor A	10258	226682_at	0,55	-0,85	3,46E-02	L
<i>PIGP</i>	phosphatidylinositol glycan anchor biosynthesis class P	3046	221689_s_at	0,56	-0,84	6,07E-04	
<i>CYFIP1</i>	cytoplasmic FMR1 interacting protein 1	13759	208923_at	0,56	-0,84	1,31E-03	I
<i>NUP54</i>	nucleoporin 54	17359	218256_s_at	0,56	-0,84	1,60E-03	
<i>GOT2</i>	glutamic-oxaloacetic transaminase 2	4433	200708_at	0,56	-0,84	6,19E-03	A
<i>GALNT1</i>	polypeptide N-acetylgalactosaminyltransferase 1	4123	201724_s_at	0,56	-0,84	6,33E-03	
<i>RANBP9</i>	RAN binding protein 9	13727	202582_s_at	0,56	-0,84	7,24E-03	
<i>UBA1</i>	ubiquitin like modifier activating enzyme 1	12469	200964_at	0,56	-0,84	8,00E-03	
<i>NR0B2</i>	nuclear receptor subfamily 0 group B member 2	7961	206410_at	0,56	-0,84	8,03E-03	
<i>CLCN5</i>	chloride voltage-gated channel 5	2023	226274_at	0,56	-0,84	1,02E-02	
<i>SGPL1</i>	sphingosine-1-phosphate lyase 1	10817	212322_at	0,56	-0,83	1,51E-03	L
<i>FOPNL</i>	FGFR1OP N-terminal like	26435	225087_at	0,56	-0,83	3,50E-03	
<i>TTC3</i>	tetratricopeptide repeat domain 3	12393	210645_s_at	0,56	-0,83	3,97E-03	
<i>WNK3</i>	WNK lysine deficient protein kinase 3	14543	236236_at	0,56	-0,83	4,42E-03	
<i>PCMI</i>	pericentriolar material 1	8727	202174_s_at	0,56	-0,83	5,99E-03	

Gene symbol	HGNC gene name	HGNC ID	Affymetrix probe ID	FC	log <sub>2</sub> FC	FDR	P
<i>PPP1CC</i>	protein phosphatase 1 catalytic subunit gamma	9283	200726_at	0,56	-0,83	6,25E-03	L
<i>GLMP</i>	glycosylated lysosomal membrane protein	29436	225401_at	0,56	-0,83	6,26E-03	
<i>COL4A3BP</i>	collagen type IV alpha 3 binding protein	2205	226277_at	0,56	-0,83	7,14E-03	L
<i>SNTB1</i>	syntrophin beta 1	11168	226438_at	0,56	-0,83	1,45E-02	
<i>C4orf3</i>	chromosome 4 open reading frame 3	19225	224602_at	0,56	-0,83	1,61E-02	
<i>TECR</i>	trans-2,3-enoyl-CoA reductase	4551	208336_s_at	0,56	-0,83	1,64E-02	
<i>CUX2</i>	cut like homeobox 2	19347	213920_at	0,56	-0,83	2,07E-02	
<i>DSTN</i>	destrin, actin depolymerizing factor	15750	201021_s_at	0,56	-0,83	2,76E-02	
<i>CCDC6</i>	coiled-coil domain containing 6	18782	225010_at	0,57	-0,82	2,99E-03	
<i>PSMB2</i>	proteasome subunit beta 2	9539	200039_s_at	0,57	-0,82	7,16E-03	A
<i>SLC22A7</i>	solute carrier family 22 member 7	10971	220554_at	0,57	-0,82	7,89E-03	
<i>PLSCR4</i>	phospholipid scramblase 4	16497	218901_at	0,57	-0,82	1,29E-02	
<i>PLPP1</i>	phospholipid phosphatase 1	9228	210946_at	0,57	-0,82	1,41E-02	L
<i>EMC7</i>	ER membrane protein complex subunit 7	24301	217898_at	0,57	-0,82	1,43E-02	
<i>MXRA5</i>	matrix remodeling associated 5	7539	209596_at	0,57	-0,82	1,55E-02	
<i>CLIC4</i>	chloride intracellular channel 4	13518	201560_at	0,57	-0,82	1,58E-02	
<i>HNRNPD</i>	heterogeneous nuclear ribonucleoprotein D	5036	200073_s_at	0,57	-0,82	2,10E-02	
<i>CDHR5</i>	cadherin related family member 5	7521	220075_s_at	0,57	-0,82	3,30E-02	
<i>RBP5</i>	retinol binding protein 5	15847	223820_at	0,57	-0,81	6,87E-03	
<i>STT3B</i>	STT3B, catalytic subunit of the oligosaccharyltransferase complex	30611	224700_at	0,57	-0,81	9,25E-03	
<i>LIN7C</i>	lin-7 homolog C, crumbs cell polarity complex component	17789	221568_s_at	0,57	-0,81	1,14E-02	
<i>DPM3</i>	dolichyl-phosphate mannosyltransferase subunit 3	3007	219373_at	0,57	-0,81	1,35E-02	
<i>NDUFS5</i>	NADH:ubiquinone oxidoreductase subunit S5	7712	201757_at	0,57	-0,81	2,91E-02	R
<i>YWHAE</i>	tyrosine 3-monooxygenase/tryptophan 5-monooxygenase activation protein epsilon	12851	210996_s_at	0,57	-0,81	3,13E-02	
<i>TMEM14B</i>	transmembrane protein 14B	21384	223105_s_at	0,57	-0,81	3,59E-02	
<i>ARL5A</i>	ADP ribosylation factor like GTPase 5A	696	226617_at	0,57	-0,80	5,54E-04	
<i>TPMT</i>	thiopurine S-methyltransferase	12014	203672_x_at	0,57	-0,80	9,67E-04	B
<i>PFN2</i>	profilin 2	8882	204992_s_at	0,57	-0,80	2,91E-03	
<i>FMO4</i>	flavin containing monooxygenase 4	3772	206263_at	0,57	-0,80	3,32E-03	
<i>PPP1R16A</i>	protein phosphatase 1 regulatory subunit 16A	14941	225203_at	0,57	-0,80	1,29E-02	
<i>GCLM</i>	glutamate-cysteine ligase modifier subunit	4312	203925_at	0,57	-0,80	2,36E-02	B/A
<i>EIF1B</i>	eukaryotic translation initiation factor 1B	30792	201738_at	0,57	-0,80	2,41E-02	
<i>TPM1</i>	tropomyosin 1	12010	210986_s_at	0,57	-0,80	4,16E-02	
<i>PLD1</i>	phospholipase D1	9067	226636_at	0,58	-0,79	4,26E-04	L/I
<i>GGCX</i>	gamma-glutamyl carboxylase	4247	235413_at	0,58	-0,79	9,93E-04	
<i>RNF170</i>	ring finger protein 170	25358	226104_at	0,58	-0,79	1,31E-03	
<i>COPS2</i>	COP9 signalosome subunit 2	30747	202467_s_at	0,58	-0,79	1,02E-02	
<i>RICTOR</i>	RPTOR independent companion of MTOR complex 2	28611	226312_at	0,58	-0,79	1,21E-02	
<i>SECISBP2L</i>	SECIS binding protein 2 like	28997	212450_at	0,58	-0,79	1,41E-02	
<i>TUBB</i>	tubulin beta class I	20778	209026_x_at	0,58	-0,79	1,44E-02	
<i>PAFAH1B1</i>	platelet activating factor acetylhydrolase 1b regulatory subunit 1	8574	200816_s_at	0,58	-0,79	1,64E-02	

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<i>U2SURP</i>	U2 snRNP associated SURP domain containing	30855	212058_at	0,58	-0,79	1,66E-02	
<i>NDUFA3</i>	NADH:ubiquinone oxidoreductase subunit A3	7686	218563_at	0,58	-0,79	2,18E-02	R
<i>NDUFA1</i>	NADH:ubiquinone oxidoreductase subunit A1	7683	202298_at	0,58	-0,79	2,52E-02	R
<i>TRA2B</i>	transformer 2 beta homolog	10781	200893_at	0,58	-0,79	3,34E-02	
<i>MAF</i>	MAF bZIP transcription factor	6776	209348_s_at	0,58	-0,79	3,34E-02	
<i>H1FX</i>	H1 histone family member X	4722	204805_s_at	0,58	-0,79	3,81E-02	
<i>RNLS</i>	renalase, FAD dependent amine oxidase	25641	223824_at	0,58	-0,78	7,47E-04	
<i>ALS2</i>	ALS2, alsin Rho guanine nucleotide exchange factor	443	226291_at	0,58	-0,78	9,58E-04	
<i>PLA2G12B</i>	phospholipase A2 group XIIB	18555	231009_at	0,58	-0,78	1,25E-03	
<i>TRIM24</i>	tripartite motif containing 24	11812	204391_x_at	0,58	-0,78	1,51E-03	
<i>KPNA1</i>	karyopherin subunit alpha 1	6394	202059_s_at	0,58	-0,78	3,01E-03	
<i>HMGB1</i>	high mobility group box 1	4983	200679_x_at	0,58	-0,78	5,44E-03	
<i>USF3</i>	upstream transcription factor family member 3	30494	227433_at	0,58	-0,78	6,27E-03	
<i>MRPL32</i>	mitochondrial ribosomal protein L32	14035	225260_s_at	0,58	-0,78	6,80E-03	
<i>PPTC7</i>	PTC7 protein phosphatase homolog	30695	225213_at	0,58	-0,78	1,52E-02	
<i>SLC35E2B</i>	solute carrier family 35 member E2B	33941	217122_s_at	0,58	-0,78	1,58E-02	
<i>VAMP8</i>	vesicle associated membrane protein 8	12647	202546_at	0,58	-0,78	1,72E-02	I
<i>DST</i>	dystonin	1090	215016_x_at	0,58	-0,78	1,80E-02	
<i>PSMA4</i>	proteasome subunit alpha 4	9533	203396_at	0,58	-0,78	2,70E-02	A
<i>CHST13</i>	carbohydrate sulfotransferase 13	21755	239647_at	0,58	-0,78	2,74E-02	
<i>LIPG</i>	lipase G, endothelial type	6623	219181_at	0,58	-0,78	4,11E-02	L
<i>ACSM5</i>	acyl-CoA synthetase medium chain family member 5	26060	89977_at	0,58	-0,78	4,36E-02	B
<i>COLEC11</i>	collectin subfamily member 11	17213	219873_at	0,58	-0,78	4,47E-02	
<i>SYPL1</i>	synaptophysin like 1	11507	201259_s_at	0,59	-0,77	7,92E-03	
<i>ECII</i>	enoyl-CoA delta isomerase 1	2703	209759_s_at	0,59	-0,77	8,65E-03	L
<i>NDUFA10</i>	NADH:ubiquinone oxidoreductase subunit A10	7684	217860_at	0,59	-0,77	1,14E-02	R
<i>DENR</i>	density regulated re-initiation and release factor	2769	221509_at	0,59	-0,77	1,22E-02	
<i>MAGED1</i>	MAGE family member D1	6813	209014_at	0,59	-0,77	1,67E-02	
<i>MINPP1</i>	multiple inositol-polyphosphate phosphatase 1	7102	209585_s_at	0,59	-0,77	1,85E-02	
<i>SUMO3</i>	small ubiquitin-like modifier 3	11124	200740_s_at	0,59	-0,77	2,29E-02	
<i>LCLAT1</i>	lysocardiolipin acyltransferase 1	26756	226996_at	0,59	-0,76	2,10E-03	L
<i>THNSL1</i>	threonine synthase like 1	26160	222931_s_at	0,59	-0,76	2,69E-03	
<i>ACPI</i>	acid phosphatase 1	122	201630_s_at	0,59	-0,76	3,41E-03	
<i>NCOR1</i>	nuclear receptor corepressor 1	7672	200854_at	0,59	-0,76	3,82E-03	L
<i>NCKAP1</i>	NCK associated protein 1	7666	207738_s_at	0,59	-0,76	1,37E-02	
<i>NOTCH2</i>	notch 2	7882	202443_x_at	0,59	-0,76	1,71E-02	
<i>UQCRC1</i>	ubiquinol-cytochrome c reductase core protein 1	12585	201903_at	0,59	-0,76	2,15E-02	R
<i>CAMK2N1</i>	calcium/calmodulin dependent protein kinase II inhibitor 1	24190	218309_at	0,59	-0,76	2,43E-02	
<i>COL18A1</i>	collagen type XVIII alpha 1 chain	2195	209081_s_at	0,59	-0,76	3,59E-02	
<i>RNF11</i>	ring finger protein 11	10056	208924_at	0,59	-0,76	4,59E-02	
<i>CYP3A7-CYP3A51P</i>	CYP3A7-CYP3A51P readthrough	51504	211843_x_at	0,59	-0,76	4,88E-02	

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<i>VPS35</i>	VPS35, retromer complex component	13487	222388_s_at	0,59	-0,75	3,48E-04	
<i>OARD1</i>	O-acyl-ADP-ribose deacylase 1	21257	213322_at	0,59	-0,75	4,74E-04	
<i>SMC6</i>	structural maintenance of chromosomes 6	20466	218781_at	0,59	-0,75	8,39E-04	
<i>DIMT1</i>	DIM1 dimethyladenosine transferase 1 homolog	30217	204405_x_at	0,59	-0,75	2,01E-03	
<i>DEPTOR</i>	DEP domain containing MTOR interacting protein	22953	218858_at	0,59	-0,75	2,03E-03	
<i>RBKS</i>	ribokinase	30325	57540_at	0,59	-0,75	2,30E-03	
<i>MSRA</i>	methionine sulfoxide reductase A	7377	219281_at	0,59	-0,75	3,01E-03	
<i>SPG21</i>	SPG21, maspardin	20373	217827_s_at	0,59	-0,75	4,57E-03	
<i>C7orf55-LUC7L2</i>	C7orf55-LUC7L2 readthrough	44671	226781_at	0,59	-0,75	9,53E-03	
<i>USP34</i>	ubiquitin specific peptidase 34	20066	212066_s_at	0,59	-0,75	1,32E-02	
<i>CTSB</i>	cathepsin B	2527	200838_at	0,59	-0,75	1,67E-02	I
<i>MORF4L1</i>	mortality factor 4 like 1	16989	221381_s_at	0,59	-0,75	1,95E-02	
<i>ADAM10</i>	ADAM metallopeptidase domain 10	188	202603_at	0,59	-0,75	2,44E-02	I
<i>SON</i>	SON DNA binding protein	11183	214988_s_at	0,59	-0,75	3,50E-02	
<i>PSMD14</i>	proteasome 26S subunit, non-ATPase 14	16889	212296_at	0,59	-0,75	3,77E-02	A/I
<i>SLC38A3</i>	solute carrier family 38 member 3	18044	205972_at	0,59	-0,75	4,02E-02	
<i>OCIAD1</i>	OCIA domain containing 1	16074	223010_s_at	0,60	-0,74	2,11E-03	
<i>TCAIM</i>	T cell activation inhibitor, mitochondrial	25241	1555906_s_at	0,60	-0,74	3,07E-03	
<i>NT5C1B-RDH14</i>	NT5C1B-RDH14 readthrough	38831	222203_s_at	0,60	-0,74	4,00E-03	
<i>CEP57</i>	centrosomal protein 57	30794	203494_s_at	0,60	-0,74	7,17E-03	
<i>TMEM256</i>	transmembrane protein 256	28618	227063_at	0,60	-0,74	7,45E-03	
<i>MARCH6</i>	membrane associated ring-CH-type finger 6	30550	201736_s_at	0,60	-0,74	9,35E-03	
<i>SMDT1</i>	single-pass membrane protein with aspartate rich tail 1	25055	225795_at	0,60	-0,74	1,42E-02	
<i>SNAP23</i>	synaptosome associated protein 23	11131	209130_at	0,60	-0,74	1,79E-02	I
<i>SRP14</i>	signal recognition particle 14	11299	200007_at	0,60	-0,74	2,25E-02	I
<i>PPP2CA</i>	protein phosphatase 2 catalytic subunit alpha	9299	208652_at	0,60	-0,74	2,79E-02	
<i>UBQLN2</i>	ubiquilin 2	12509	215884_s_at	0,60	-0,74	3,36E-02	
<i>PRKAR2A</i>	protein kinase cAMP-dependent type II regulatory subunit alpha	9391	225011_at	0,60	-0,74	4,44E-02	
<i>ADGRL2</i>	adhesion G protein-coupled receptor L2	18582	206953_s_at	0,60	-0,73	2,09E-03	
<i>PTEN</i>	phosphatase and tensin homolog	9588	228006_at	0,60	-0,73	3,29E-03	L
<i>RAB7A</i>	RAB7A, member RAS oncogene family	9788	211961_s_at	0,60	-0,73	7,21E-03	I
<i>IL13RA1</i>	interleukin 13 receptor subunit alpha 1	5974	201887_at	0,60	-0,73	7,73E-03	
<i>CASC4</i>	cancer susceptibility 4	24892	224619_at	0,60	-0,73	8,95E-03	
<i>PUM2</i>	pumilio RNA binding family member 2	14958	201493_s_at	0,60	-0,73	9,93E-03	
<i>CPED1</i>	cadherin like and PC-esterase domain containing 1	26159	228728_at	0,60	-0,73	1,17E-02	
<i>KARS</i>	lysyl-tRNA synthetase	6215	200079_s_at	0,60	-0,73	1,18E-02	
<i>VDAC3</i>	voltage dependent anion channel 3	12674	208845_at	0,60	-0,73	1,93E-02	
<i>IER3IP1</i>	immediate early response 3 interacting protein 1	18550	223071_at	0,60	-0,73	2,24E-02	
<i>ZC3H13</i>	zinc finger CCCH-type containing 13	20368	212402_at	0,60	-0,73	2,54E-02	
<i>PSME4</i>	proteasome activator subunit 4	20635	212219_at	0,60	-0,73	2,57E-02	A
<i>UQCRI0</i>	ubiquinol-cytochrome c reductase, complex III subunit X	30863	218190_s_at	0,60	-0,73	2,70E-02	R

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<i>ISCU</i>	iron-sulfur cluster assembly enzyme	29882	209075_s_at	0,60	-0,73	3,39E-02	
<i>LRRC8D</i>	leucine rich repeat containing 8 VRAC subunit D	16992	218684_at	0,61	-0,72	7,29E-03	
<i>ZNF664</i>	zinc finger protein 664	25406	224593_at	0,61	-0,72	1,44E-02	
<i>PKLR</i>	pyruvate kinase L/R	9020	210451_at	0,61	-0,72	1,46E-02	
<i>UBXN4</i>	UBX domain protein 4	14860	212006_at	0,61	-0,72	2,16E-02	
<i>FYN</i>	FYN proto-oncogene, Src family tyrosine kinase	4037	210105_s_at	0,61	-0,72	2,25E-02	L
<i>TWF1</i>	twinfilin actin binding protein 1	9620	201745_at	0,61	-0,72	3,49E-02	
<i>NOL7</i>	nucleolar protein 7	21040	210097_s_at	0,61	-0,72	3,71E-02	
<i>ANXA9</i>	annexin A9	547	211712_s_at	0,61	-0,71	1,93E-03	
<i>SNRNP25</i>	small nuclear ribonucleoprotein U11/U12 subunit 25	14161	218493_at	0,61	-0,71	3,35E-03	
<i>MTFR1</i>	mitochondrial fission regulator 1	29510	203208_s_at	0,61	-0,71	3,41E-03	
<i>TTC3</i>	tetratricopeptide repeat domain 3	12393	208073_x_at	0,61	-0,71	3,63E-03	
<i>TARDBP</i>	TAR DNA binding protein	11571	200020_at	0,61	-0,71	9,66E-03	
<i>EPS15</i>	epidermal growth factor receptor pathway substrate 15	3419	217887_s_at	0,61	-0,71	1,02E-02	
<i>NDUFS4</i>	NADH:ubiquinone oxidoreductase subunit S4	7711	209303_at	0,61	-0,71	1,15E-02	R
<i>SARIA</i>	secretion associated Ras related GTPase 1A	10534	201542_at	0,61	-0,71	1,32E-02	
<i>MASP2</i>	mannan binding lectin serine peptidase 2	6902	210798_x_at	0,61	-0,71	1,93E-02	
<i>CERS2</i>	ceramide synthase 2	14076	222212_s_at	0,61	-0,71	2,47E-02	L
<i>FST</i>	folistatin	3971	207345_at	0,61	-0,71	2,57E-02	
<i>UBR2</i>	ubiquitin protein ligase E3 component n-recognin 2	21289	212760_at	0,62	-0,70	7,95E-04	
<i>SCAF11</i>	SR-related CTD associated factor 11	10784	225336_at	0,62	-0,70	2,93E-03	
<i>UBL3</i>	ubiquitin like 3	12504	201535_at	0,62	-0,70	4,45E-03	
<i>TMEM220</i>	transmembrane protein 220	33757	229693_at	0,62	-0,70	6,93E-03	
<i>SEC11A</i>	SEC11 homolog A, signal peptidase complex subunit	17718	216274_s_at	0,62	-0,70	1,87E-02	
<i>YME1L1</i>	YME1 like 1 ATPase	12843	201351_s_at	0,62	-0,70	1,93E-02	
<i>TAT</i>	tyrosine aminotransferase	11573	206916_x_at	0,62	-0,70	2,20E-02	A
<i>DIP2C</i>	disco interacting protein 2 homolog C	29150	212503_s_at	0,62	-0,70	3,78E-02	
<i>UQCRB</i>	ubiquinol-cytochrome c reductase binding protein	12582	205849_s_at	0,62	-0,70	3,88E-02	R
<i>PPP1R3B</i>	protein phosphatase 1 regulatory subunit 3B	14942	1552670_a_at	0,62	-0,70	4,42E-02	
<i>POLR3GL</i>	RNA polymerase III subunit G like	28466	223269_at	0,62	-0,69	7,76E-04	
<i>NDUFA9</i>	NADH:ubiquinone oxidoreductase subunit A9	7693	208969_at	0,62	-0,69	3,61E-03	R
<i>LACTB</i>	lactamase beta	16468	226354_at	0,62	-0,69	3,90E-03	
<i>NUDT9</i>	nudix hydrolase 9	8056	218375_at	0,62	-0,69	5,44E-03	
<i>EFHD1</i>	EF-hand domain family member D1	29556	209343_at	0,62	-0,69	6,08E-03	
<i>SYNPO2</i>	synaptopodin 2	17732	225720_at	0,62	-0,69	1,29E-02	
<i>TMEM41B</i>	transmembrane protein 41B	28948	212623_at	0,62	-0,69	1,58E-02	
<i>TKFC</i>	triokinase and FMN cyclase	24552	218688_at	0,62	-0,69	3,52E-02	
<i>RNF138</i>	ring finger protein 138	17765	218738_s_at	0,62	-0,69	4,59E-02	
<i>SEC14L2</i>	SEC14 like lipid binding 2	10699	204541_at	0,62	-0,69	4,75E-02	
<i>HOGA1</i>	4-hydroxy-2-oxoglutarate aldolase 1	25155	239093_at	0,62	-0,69	4,79E-02	
<i>CDK14</i>	cyclin dependent kinase 14	8883	204604_at	0,62	-0,68	1,38E-03	

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<i>ATP6V1A</i>	ATPase H+ transporting V1 subunit A	851	201972_at	0,62	-0,68	4,36E-03	
<i>RAB10</i>	RAB10, member RAS oncogene family	9759	222981_s_at	0,62	-0,68	9,10E-03	I
<i>MYO5B</i>	myosin VB	7603	225301_s_at	0,62	-0,68	9,83E-03	
<i>METTL5</i>	methyltransferase like 5	25006	221570_s_at	0,62	-0,68	1,41E-02	
<i>HNRNPA3</i>	heterogeneous nuclear ribonucleoprotein A3	24941	211932_at	0,62	-0,68	1,61E-02	
<i>MOCOS</i>	molybdenum cofactor sulfurase	18234	219959_at	0,62	-0,68	1,77E-02	
<i>YTHDC1</i>	YTH domain containing 1	30626	212455_at	0,62	-0,68	2,82E-02	
<i>SCARB2</i>	scavenger receptor class B member 2	1665	201646_at	0,62	-0,68	2,89E-02	
<i>COX7A2</i>	cytochrome c oxidase subunit 7A2	2288	201597_at	0,62	-0,68	3,28E-02	
<i>SULT1A3</i>	sulfotransferase family 1A member 3	11455	203615_x_at	0,62	-0,68	4,85E-02	
<i>PTDSS1</i>	phosphatidylserine synthase 1	9587	201433_s_at	0,63	-0,67	7,91E-03	L
<i>RETSAT</i>	retinol saturase	25991	218124_at	0,63	-0,67	1,14E-02	
<i>PNRC2</i>	proline rich nuclear receptor coactivator 2	23158	217779_s_at	0,63	-0,67	1,29E-02	
<i>SYNCRIP</i>	synaptotagmin binding cytoplasmic RNA interacting protein	16918	217833_at	0,63	-0,67	1,66E-02	
<i>SHOC2</i>	SHOC2, leucine rich repeat scaffold protein	15454	202777_at	0,63	-0,67	3,19E-02	
<i>PPM1B</i>	protein phosphatase, Mg <sup>2+</sup> /Mn <sup>2+</sup> dependent 1B	9276	209296_at	0,63	-0,67	3,52E-02	
<i>UBP1</i>	upstream binding protein 1	12507	218082_s_at	0,63	-0,67	4,32E-02	
<i>SKP1</i>	S-phase kinase associated protein 1	10899	200711_s_at	0,63	-0,67	4,38E-02	
<i>CYP27A1</i>	cytochrome P450 family 27 subfamily A member 1	2605	203979_at	0,63	-0,67	4,78E-02	B/L
<i>HMGS2</i>	3-hydroxy-3-methylglutaryl-CoA synthase 2	5008	240110_at	0,63	-0,67	4,93E-02	
<i>WASHC2C</i>	WASH complex subunit 2C	23414	212370_x_at	0,63	-0,66	1,30E-03	
<i>OXSM</i>	3-oxoacyl-ACP synthase, mitochondrial	26063	219133_at	0,63	-0,66	1,52E-03	
<i>CUL4B</i>	cullin 4B	2555	202214_s_at	0,63	-0,66	1,67E-03	
<i>ATIC</i>	5-aminoimidazole-4-carboxamide ribonucleotide formyltransferase/IMP cyclohydrolase	794	208758_at	0,63	-0,66	3,22E-03	
<i>GLB1</i>	galactosidase beta 1	4298	201576_s_at	0,63	-0,66	4,96E-03	L/I
<i>ARHGAP15</i>	Rho GTPase activating protein 15	21030	218870_at	0,63	-0,66	8,15E-03	
<i>CLIP1</i>	CAP-Gly domain containing linker protein 1	10461	201975_at	0,63	-0,66	8,33E-03	
<i>CMTM8</i>	CKLF like MARVEL transmembrane domain containing 8	19179	235099_at	0,63	-0,66	1,42E-02	
<i>VPS26A</i>	VPS26, retromer complex component A	12711	201807_at	0,63	-0,66	1,43E-02	
<i>EWSR1</i>	EWS RNA binding protein 1	3508	209214_s_at	0,63	-0,66	1,80E-02	
<i>MAP3K20</i>	mitogen-activated protein kinase kinase kinase 20	17797	225665_at	0,63	-0,66	2,57E-02	
<i>MAPK1</i>	mitogen-activated protein kinase 1	6871	224621_at	0,63	-0,66	3,40E-02	I
<i>KIAA0232</i>	KIAA0232	28992	212441_at	0,63	-0,66	3,50E-02	
<i>PPP1CB</i>	protein phosphatase 1 catalytic subunit beta	9282	201409_s_at	0,64	-0,65	4,99E-03	L
<i>KCNJ8</i>	potassium voltage-gated channel subfamily J member 8	6269	205304_s_at	0,64	-0,65	1,04E-02	
<i>DCTD</i>	dCMP deaminase	2710	210137_s_at	0,64	-0,65	1,25E-02	
<i>VDAC3</i>	voltage dependent anion channel 3	12674	208846_s_at	0,64	-0,65	1,93E-02	
<i>GNS</i>	glucosamine (N-acetyl)-6-sulfatase	4422	212334_at	0,64	-0,65	2,13E-02	I
<i>SNRPD3</i>	small nuclear ribonucleoprotein D3 polypeptide	11160	202567_at	0,64	-0,65	3,31E-02	
<i>ZFR</i>	zinc finger RNA binding protein	17277	201857_at	0,64	-0,65	3,70E-02	
<i>TM2D3</i>	TM2 domain containing 3	24128	221702_s_at	0,64	-0,65	3,83E-02	

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<i>MPC1</i>	mitochondrial pyruvate carrier 1	21606	218024_at	0,64	-0,65	4,92E-02	R
<i>GPR89B</i>	G protein-coupled receptor 89B	13840	220642_x_at	0,64	-0,64	2,16E-03	
<i>NDUFS2</i>	NADH:ubiquinone oxidoreductase core subunit S2	7708	201966_at	0,64	-0,64	5,66E-03	R
<i>ZNF22</i>	zinc finger protein 22	13012	218005_at	0,64	-0,64	8,45E-03	
<i>ZBTB33</i>	zinc finger and BTB domain containing 33	16682	226255_at	0,64	-0,64	9,41E-03	
<i>ANAPC13</i>	anaphase promoting complex subunit 13	24540	209001_s_at	0,64	-0,64	1,02E-02	
<i>JAGN1</i>	jagunal homolog 1	26926	223104_at	0,64	-0,64	1,25E-02	
<i>RAP1A</i>	RAP1A, member of RAS oncogene family	9855	202362_at	0,64	-0,64	1,42E-02	I
<i>OAZ2</i>	ornithine decarboxylase antizyme 2	8096	201364_s_at	0,64	-0,64	1,54E-02	A
<i>GTF3C6</i>	general transcription factor IIIC subunit 6	20872	225083_at	0,64	-0,64	1,97E-02	
<i>CSNK2B</i>	casein kinase 2 beta	2460	201390_s_at	0,64	-0,64	2,19E-02	L/I
<i>GALM</i>	galactose mutarotase	24063	235256_s_at	0,64	-0,64	2,65E-02	
<i>TTC19</i>	tetratricopeptide repeat domain 19	26006	217964_at	0,64	-0,64	3,29E-02	
<i>CUTA</i>	cutA divalent cation tolerance homolog	21101	221488_s_at	0,64	-0,64	4,11E-02	
<i>CBR1</i>	carbonyl reductase 1	1548	209213_at	0,64	-0,64	4,66E-02	L
<i>TTC39C</i>	tetratricopeptide repeat domain 39C	26595	1552307_a_at	0,64	-0,64	4,77E-02	
<i>MOGAT1</i>	monoacylglycerol O-acyltransferase 1	18210	1553062_at	0,64	-0,64	4,83E-02	L
<i>TMCO1</i>	transmembrane and coiled-coil domains 1	18188	208715_at	0,65	-0,63	9,96E-04	
<i>UBE2E2</i>	ubiquitin conjugating enzyme E2 E2	12478	225651_at	0,65	-0,63	2,91E-03	
<i>USP16</i>	ubiquitin specific peptidase 16	12614	218386_x_at	0,65	-0,63	1,29E-02	
<i>PSMD4</i>	proteasome 26S subunit, non-ATPase 4	9561	211609_x_at	0,65	-0,63	1,49E-02	A
<i>COX4I1</i>	cytochrome c oxidase subunit 4I1	2265	200086_s_at	0,65	-0,63	1,61E-02	R
<i>METTL14</i>	methyltransferase like 14	29330	227601_at	0,65	-0,63	2,22E-02	
<i>DAP3</i>	death associated protein 3	2673	208822_s_at	0,65	-0,63	2,24E-02	
<i>VAMP3</i>	vesicle associated membrane protein 3	12644	201336_at	0,65	-0,63	2,41E-02	
<i>AZIN1</i>	antizyme inhibitor 1	16432	212461_at	0,65	-0,63	2,70E-02	A
<i>OSBPL11</i>	oxysterol binding protein like 11	16397	218304_s_at	0,65	-0,63	3,64E-02	
<i>CD99</i>	CD99 molecule (Xg blood group)	7082	201029_s_at	0,65	-0,63	4,23E-02	
<i>SKP2</i>	S-phase kinase associated protein 2	10901	211042_x_at	0,65	-0,62	3,18E-03	
<i>CISD1</i>	CDGSH iron sulfur domain 1	30880	218597_s_at	0,65	-0,62	1,31E-02	
<i>CCL16</i>	C-C motif chemokine ligand 16	10614	207354_at	0,65	-0,62	1,37E-02	
<i>TROVE2</i>	TROVE domain family member 2	11313	210438_x_at	0,65	-0,62	1,54E-02	
<i>SMIM10L1</i>	small integral membrane protein 10 like 1	49847	221847_at	0,65	-0,62	1,87E-02	
<i>CHD9</i>	chromodomain helicase DNA binding protein 9	25701	212616_at	0,65	-0,62	2,04E-02	L
<i>AZIN1</i>	antizyme inhibitor 1	16432	201772_at	0,65	-0,62	2,97E-02	
<i>TBL1XR1</i>	transducin beta like 1 X-linked receptor 1	29529	223013_at	0,65	-0,62	3,47E-02	L
<i>ASH1L</i>	ASH1 like histone lysine methyltransferase	19088	222667_s_at	0,65	-0,62	3,50E-02	
<i>SEPT11</i>	septin 11	25589	201307_at	0,66	-0,61	5,32E-03	
<i>SPRYD4</i>	SPRY domain containing 4	27468	225616_at	0,66	-0,61	5,53E-03	
<i>DCTD</i>	dCMP deaminase	2710	201572_x_at	0,66	-0,61	1,13E-02	
<i>ELAVL1</i>	ELAV like RNA binding protein 1	3312	201726_at	0,66	-0,61	1,39E-02	

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<i>PPMIK</i>	protein phosphatase, Mg <sup>2+</sup> /Mn <sup>2+</sup> dependent 1K	25415	226773_at	0,66	-0,61	1,78E-02	A
<i>RNF181</i>	ring finger protein 181	28037	223064_at	0,66	-0,61	1,88E-02	
<i>TPD52</i>	tumor protein D52	12005	201690_s_at	0,66	-0,61	1,93E-02	
<i>SYNPO2</i>	synaptopodin 2	17732	225721_at	0,66	-0,61	3,28E-02	
<i>PRDX6</i>	peroxiredoxin 6	16753	200845_s_at	0,66	-0,61	3,68E-02	I
<i>CRK</i>	CRK proto-oncogene, adaptor protein	2362	202225_at	0,66	-0,61	3,98E-02	
<i>WEE1</i>	WEE1 G2 checkpoint kinase	12761	212533_at	0,66	-0,61	4,22E-02	
<i>CCDC92</i>	coiled-coil domain containing 92	29563	218175_at	0,66	-0,61	4,64E-02	
<i>GPR89A</i>	G protein-coupled receptor 89A	31984	225463_x_at	0,66	-0,60	1,53E-03	
<i>DNAJA2</i>	DnaJ heat shock protein family (Hsp40) member A2	14884	209157_at	0,66	-0,60	1,90E-02	
<i>GNG12</i>	G protein subunit gamma 12	19663	212294_at	0,66	-0,60	2,49E-02	
<i>HMGNI</i>	high mobility group nucleosome binding domain 1	4984	200944_s_at	0,66	-0,60	2,90E-02	
<i>RIOK3</i>	RIO kinase 3	11451	202130_at	0,66	-0,60	3,74E-02	
<i>PRPS1</i>	phosphoribosyl pyrophosphate synthetase 1	9462	209440_at	0,66	-0,59	3,77E-03	
<i>PPP1CB</i>	protein phosphatase 1 catalytic subunit beta	9282	201408_at	0,66	-0,59	3,92E-03	
<i>ADCY9</i>	adenylate cyclase 9	240	204497_at	0,66	-0,59	1,19E-02	
<i>BET1</i>	Bet1 golgi vesicular membrane trafficking protein	14562	202710_at	0,66	-0,59	1,52E-02	
<i>PSMD4</i>	proteasome 26S subunit, non-ATPase 4	9561	200882_s_at	0,66	-0,59	2,10E-02	
<i>DDOST</i>	dolichyl-diphosphooligosaccharide--protein glycosyltransferase non-catalytic subunit	2728	208675_s_at	0,66	-0,59	3,07E-02	I
<i>SLC6A2</i>	solute carrier family 6 member 2	11048	216611_s_at	1,52	0,60	5,94E-03	
<i>MYO1C</i>	myosin IC	7597	214656_x_at	1,52	0,60	6,96E-03	
<i>APOL2</i>	apolipoprotein L2	619	221013_s_at	1,52	0,60	1,02E-02	
<i>CCDC152</i>	coiled-coil domain containing 152	34438	237475_x_at	1,52	0,60	3,26E-02	
<i>DUX4</i>	double homeobox 4	50800	216473_x_at	1,53	0,61	7,21E-03	
<i>OCIAD1</i>	OCIA domain containing 1	16074	239748_x_at	1,53	0,61	3,20E-02	
<i>HAUS2</i>	HAUS augmin like complex subunit 2	25530	220071_x_at	1,55	0,63	1,67E-02	
<i>MAPK8IP2</i>	mitogen-activated protein kinase 8 interacting protein 2	6883	205050_s_at	1,56	0,64	6,62E-03	
<i>SSC4D</i>	scavenger receptor cysteine rich family member with 4 domains	14461	236529_at	1,57	0,65	7,15E-03	
<i>PDE4C</i>	phosphodiesterase 4C	8782	206792_x_at	1,58	0,66	2,31E-02	
<i>TNFRSF1B</i>	TNF receptor superfamily member 1B	11917	203508_at	1,58	0,66	2,45E-02	
<i>PLA2G4F</i>	phospholipase A2 group IVF	27396	235958_at	1,61	0,69	1,37E-03	L
<i>DLG4</i>	discs large MAGUK scaffold protein 4	2903	210684_s_at	1,62	0,70	6,84E-03	
<i>IRS2</i>	insulin receptor substrate 2	6126	209184_s_at	1,64	0,71	4,66E-02	
<i>TNFRSF12A</i>	TNF receptor superfamily member 12A	18152	218368_s_at	1,66	0,73	1,80E-02	
<i>PIM3</i>	Pim-3 proto-oncogene, serine/threonine kinase	19310	224739_at	1,68	0,75	2,56E-02	
<i>ATP8B1</i>	ATPase phospholipid transporting 8B1	3706	214594_x_at	1,71	0,77	3,70E-02	
<i>CSRNP1</i>	cysteine and serine rich nuclear protein 1	14300	225557_at	1,89	0,92	4,72E-02	
<i>TNFSF14</i>	TNF superfamily member 14	11930	207907_at	1,91	0,93	4,79E-03	
<i>DNAJB1</i>	DnaJ heat shock protein family (Hsp40) member B1	5270	200666_s_at	2,16	1,11	3,68E-02	
<i>IL1RN</i>	interleukin 1 receptor antagonist	6000	212657_s_at	2,50	1,32	9,48E-03	
<i>S100A9</i>	S100 calcium binding protein A9	10499	203535_at	2,73	1,45	5,37E-03	I

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<i>S100A8</i>	S100 calcium binding protein A8	10498	202917_s_at	2,85	1,51	2,67E-02	I
<i>SAI2</i>	serum amyloid A2	10514	208607_s_at	3,12	1,64	4,84E-02	
<i>S100A12</i>	S100 calcium binding protein A12	10489	205863_at	3,16	1,66	1,35E-03	I