Schizophrenia shows a unique metabolomics signature in plasma

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Schizophrenia is a severe complex mental disorder affecting 0.5–1% of the world population. To date, diagnosis of the disease is mainly based on personal and thus subjective interviews. The underlying molecular mechanism of schizophrenia is poorly understood. Using targeted metabolomics we quantified and compared 103 metabolites in plasma samples from 216 healthy controls and 265 schizophrenic patients, including 52 cases that do not take antipsychotic medication. Compared with healthy controls, levels of five metabolites were found significantly altered in schizophrenic patients (P-values ranged from $2.9 \times 10^{-8}$ to $2.5 \times 10^{-11}$) and in neuroleptics-free probands (P-values ranging between 0.006 and 0.03), respectively. These metabolites include four amino acids (arginine, glutamine, histidine and ornithine) and one lipid (PC ae C38:6) and are suggested as candidate biomarkers for schizophrenia. To explore the genetic susceptibility on the associated metabolic pathways, we constructed a molecular network connecting these five aberrant metabolites with 13 schizophrenia risk genes. Our result implicated aberrations in biosynthetic pathways linked to glutamine and arginine metabolism and associated signaling pathways as genetic risk factors, which may contribute to patho-mechanisms and memory deficits associated with schizophrenia. This study illustrated that the metabolic deviations detected in plasma may serve as potential biomarkers to aid diagnosis of schizophrenia.

Translational Psychiatry (2012) 2, e149; doi:10.1038/tp.2012.76; published online 14 August 2012

Introduction

Schizophrenia is a severe chronic brain disorder with symptomatic onset in early adulthood and persists throughout life. It affects approximately 0.5–1% of the world population with high heritability. Currently, the diagnosis of schizophrenia is still merely based on interview of the person and family members as there are no reliable medical diagnostic tests available. Therefore, the identification of blood markers for schizophrenia has become a crucial issue, which would not only improve our understanding of the disease, but also help diagnose the disease in the early-onset phase and lead to efficient and specific therapies.

Genetic factors and gene–environment interactions contribute largely the liability to the disease. Family, twin and adoption studies have demonstrated that the rising risk of schizophrenia in relatives relate to strong genetic factors. A number of genome-wide association studies have confirmed that common variations and individually rare structural variants would increase the disease risk. Environmental factors such as living environment, drug abuse and childhood trauma also have important roles in the pathogenesis of this disease.

Recently, several metabolomic studies also demonstrated metabolic abnormalities in schizophrenia. Holmes used $^1$H NMR spectroscopy to measure the metabolic profiles of 152 cerebrospinal fluid samples from drug-naïve patients compared with matched controls and found elevated glucose concentrations in patients, whereas serum glucose level showed no difference. Kaddurah–Daoku et al. compared lipid levels in 27 schizophrenic patients with 16 healthy controls and revealed significantly decreased levels in patients' plasma. Another Finland population study performed serum metabolome of 45 cases and controls each on a lipidomics platform and a small polar metabolites platform. They observed that six lipid clusters mainly including saturated triglycerides and two small-molecular clusters containing glutamic acid and proline were elevated in schizophrenia, which suggested the relevance between glucoregulatory, proline metabolism and this disease. A recent research reported a global metabolic profiling study on 112 schizophrenia and 110 controls using serum and urine samples. The study detected the potential metabolite markers consisting of several fatty acids and ketone bodies, and they presumed the upregulated fatty acid catabolism may result...
from an insufficiency of glucose supply in the brains of schizophrenia. These studies discovered various metabolite signatures and provided valuable clues to the understanding of the disease mechanism; however, they contained relative small numbers of individuals and thus limit their abilities to identify the diagnostic signatures in the disorder.

To identify the metabolite alteration in schizophrenia, which may serve as potential biomarker signatures, we targeted 103 metabolites, including amino acids, acylcarnitines, glycerophospholipids, sphingolipids and hexose, to analyse the plasma metabolite-concentration changes between 265 patients with schizophrenia and 216 healthy individuals.

**Materials and methods**

**Study population**

**Schizophrenic patients.** We recruited patients ascertained with schizophrenia from the Munich area in Germany. Case participants had a DSM-IV and ICD-10 diagnosis of schizophrenia with the following subtypes: paranoid 75.6% disorganized 17.3%, catatonic 2.3% and undifferentiated 4.9%. Detailed medical and psychiatric histories were collected, including the Structured Clinical Interview for DSM-IV, to evaluate lifetime Axis I and II diagnoses. Detailed depictions of all subjects were conducted according to the ethical standards laid down in the Declarations of Helsinki. All subjects gave their written informed consent. Age at onset was 24.3 (8.6) years old, 18.8% of the patients had a schizophrenia diagnosis in first-degree relatives, 31.2% in relatives in general.

**Healthy individuals.** Healthy volunteers were randomly selected from the general population of Munich, Germany and contacted by mails. We excluded subjects with central neurological diseases and psychotic disorders or subjects who had first-degree relatives with psychotic disorders by several screenings (for detailed exclusion criterion, see Braunewell et al. 21). Finally, 224 control subjects, 20–67 years of age, were included in this study. Eighty percentage of them are under- and sixty-six patients, aged from 19 to 67 years old, were included in this study. Eighty percentage of them are under- and sixty-six patients, aged from 19 to 67 years old, were included in this study. Eighty percentage of them are under- and sixty-six patients, aged from 19 to 67 years old, were included in this study.

**Plasma samples.** Blood (approximately 7.5 ml) was obtained from forearm vein and drawn in EDTA-containing tubes for biomarker studies. The samples were centrifuged for 10 min at 4 °C and the resulting plasma aliquoted into Eppendorf tubes, which were frozen immediately at – 80 °C.

**Metabolite measurements.** All plasma samples were measured by the commercially available Absolute/OQ p150 kit (BIOCRATES Life Sciences AG, Innsbruck, Austria), which allows simultaneous quantification of 163 metabolites using flow injection analysis mass spectrometry. This method has been proven to be in conformance with FDA-Guidelines ‘Guidance for Industry—Bioanalytical Method Validation (May 2001)’, which implies proof of reproducibility within a given error range. The procedures of this assay were described in more details elsewhere.

Metabolite measurements for all 490 samples were performed in three batches. All sample analyses underwent quality assurance and control procedures. Nine samples were excluded for further analysis as they had missing values. Ten replicate samples were measured across the three batches, and concentration of each metabolite was adjusted based on that to minimize potential batch effects.

To ensure data quality, each metabolite had to go through following stringent criteria, which we applied previously: (1) average value of the coefficient of variance for the metabolite in three quality control samples, which represents mixed human plasma samples provided by the manufacturer, should be <25%. (2) 90% of all measured sample concentrations for the metabolite should be above the limit of detection. (3) The correlation coefficient of the metabolite between measurements in ten duplicate samples should exceed 0.5. In total, 103 metabolites passed the three criteria, and the final metabolite panel contained 13 amino acids, 8 acylcarnitines, 60 phosphatidylcholines (PCs), 8 lysoPCs, 13 sphingomyelins and total hexoses (H1). Concentrations of all analyzed metabolites were reported in μM. Characteristics and average values of all measured metabolites were summarized in Supplementary Table 1.

**Statistics analysis.** The design and analysis strategy were depicted in Supplementary Figure S1. Multivariate logistic regression analysis was used to identify the metabolites with significant concentration difference between schizophrenic patients (cases) and healthy controls. Odds ratios for single metabolites were calculated between two groups. We normalized concentrations of each metabolite to have a mean of zero and a s.d. of one for the working data set. Age, sex and body mass index (BMI) were added to the logistic regression analysis as covariates. The significance level was adjusted using the Bonferroni correction to 4.8 × 10⁻⁴ (0.05/103).

To exclude the metabolites with significant concentration changes influenced by neuroleptics-taken (NT), one-way ANOVA test coupled with Tukey’s honestly significant difference test were conducted on the metabolites, which

| Table 1 Demographic data of schizophrenic cases and healthy controls |
|------------------------|------------------------|------------------------|
|                        | **Cases**              | **Controls**           |
|                        | **NF**                 | **NT**                 |
| Total number (N)       | 52                     | 213                    |
| Sex (male)             | 55.8%                  | 62.0%                  |
| Age (mean ± s.d.)      | 39.3 ± 11.2            | 36.9 ± 11.7            | 38.9 ± 10.6 |
| BMI (mean ± s.d.)      | 24.9 ± 4.8             | 26.4 ± 4.9             | 25.3 ± 4.4  |

Abbreviations: BMI, body mass index; NF, neuroleptics-free; NT, neuroleptics-taken.
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Metabolomics discriminate schizophrenic patients from healthy individuals. (a) Partial least square (PLS) analysis demonstrate a clear separation of metabolite concentration between patients with schizophrenia (red) and healthy individuals (black). (b) The concentrations of five metabolite signatures were shown in beanplots, which provide information on the mean level (solid black line), individual data points (short gray lines) and the density of the distribution between controls and cases. (c) The receiver-operating characteristic (ROC) curves indicated the effect of discrimination based on the combination of five metabolite signatures in the model (red) and the combination of age, sex and BMI (green) with 95% confidence intervals s.e. of the true-positive rate.

were significant in logistic regression. Multiple comparisons were performed between NT cases and neuroleptics-free (NF) cases, as well as between NF versus controls. The metabolites were categorized into three following groups: (1) No significant neuroleptics influence (that is metabolites showed no significant difference in NT–NF comparison, but they differed in NF versus controls.) (2) Potential neuroleptics influence (that is metabolites showed significant difference in NT versus NF comparison, whereas they showed no difference in NF versus controls.) (3) Unknown (that is the metabolites did not show significant difference in either comparison.). We kept the metabolites with no significant neuroleptics influence for further analysis, and excluded those in (2) and (3) groups.

To evaluate the performance of the five metabolites, we randomly split the participants into detection and validation groups for one hundred times to cross validate our classification model. Each time, we randomly chose 90% participants for model estimation, and used the remaining 10% samples for validation. The receiver-operating characteristic curve indicated the power of discrimination between the cases and controls with 95% confidence intervals s.d. of the true-positive rate. All calculations were performed under the R statistical environment (http://www.r-project.org/).
Network analysis. We collected 152 schizophrenia risk genes from the latest genome-wide association studies (Supplementary Table 2). Schizophrenia molecular network was constructed by connecting candidate metabolites, enzymes, intermediate proteins and schizophrenia risk genes. A background network was structured based on the metabolite–enzyme interaction from the Human Metabolome Database (HMDB) and the protein functional association from the Search Tool for the Retrieval of Interacting Genes/Proteins (STRING) database (STRING score at least 0.7). Using the metaP-Server we mapped the candidate metabolites to HMDB IDs, and derived their associated enzymes according to the annotations. We first mapped the candidate metabolites and their associated enzymes to the background network, and then connected the enzymes with the schizophrenia risk genes allowing for one intermediate protein in the paths. The paths from enzymes to risk genes were searched by the shortest path algorithm implemented in the igraph library (http://igraph.sourceforge.net/index.html), which finds the appropriate paths with the highest scores. The network consist of nodes (metabolites, enzymes, intermediate proteins and risk genes) and edges (associations and connections), and visualized by Cytoscape (http://www.cytoscape.org/). We used Database for Annotation, Visualization and Integrated Discovery (DAVID) to analyze the enriched biological function and involved pathway of the genes and proteins in the network based on Gene Ontology and Kyoto Encyclopedia of Genes and Genomes annotations under Bonferroni-corrected P-value < 0.05.

Results

Schizophrenic patients separated from healthy individuals. After the quality controls, a total of 481 individuals and 103 metabolites were covered in the final data set. Among them, 216 subjects were healthy controls and 265 subjects were cases diagnosed with schizophrenia, including 52 patients without taking any neuroleptic (Table 1). The cases and controls showed no significant difference in sex, age and BMI demographics.

To explore the global differences between the cases and controls in metabolite concentration levels, a partial least square analysis was carried out on the metabolite profiles. The partial least square result (Figure 1a) demonstrated a separation of metabolite profiles between the two groups, indicating some schizophrenic signatures were captured among the metabolic difference.

Metabolites with significant alteration in schizophrenia. To identify significant metabolite concentration changes between schizophrenic patients and healthy controls, multivariate logistic regression analysis were carried out on each of the 103 metabolites, with concentration as dependent and schizophrenia as explanatory variable besides the covariates sex, age and BMI. Twelve metabolites, including five amino acids, one acetylcarnitine and six PCs (Table 2), displayed significantly different concentrations in schizophrenic subjects compared with controls.
Schizophrenia risk gene under the receiver-operating characteristic curve value for the influence of genetic risks on the metabolic pathways. The Figure 3 in metabolite concentrations or gene/protein level. In previous publications. Bigger/smaller arrows indicate reported/observed changes and green arrows next to them. Blue arrows suggest the possible alteration reported (purple) and physical interaction (gray). The observed metabolite concentration visualized as following: activation (red), inhibition (blue), neurotrophin signaling (schizophrenia risk genes (marked by red stars) are supposed to be involved in the regulation of these metabolic pathways (black). Types of regulations are visualized as following: activation (red), inhibition (blue), neurotrophin signaling (purple) and physical interaction (gray). The observed metabolite concentration changes between schizophrenics and healthy individuals are denoted by the red and green arrows next to them. Blue arrows suggest the possible alteration reported in previous publications. Bigger/smaller arrows indicate reported/observed changes in metabolite concentrations or gene/protein level.

Furthermore, multiple comparisons were conducted between NT and NF, as well as NF and controls, considering the potential neuroleptics influence on these 12 metabolites. We further applied ANOVA test coupled with Turkey’s honestly significant difference test and finally identified five metabolites (that is, higher concentration of ornithine and lower concentration of arginine, glutamine, histidine and PC ae C38:6 in schizophrenic patients) showing no significant difference in NT versus NF comparison, but differed in NF versus controls. To visualize these five metabolites, the concentration distributions in cases and controls were displayed in Figure 1b by beanplots. The average area under the receiver-operating characteristic curve value for the five metabolites reached 80.5%, indicating a relatively high sensitivity and specificity for detection of schizophrenia compared with the value of the combination of sex, age and BMI (56.3%).

Pathways in schizophrenia molecular network. To explore the biological functions and pathways that are related to the significantly changed metabolites, we applied a metabolite-gene/protein bioinformatic approach to construct a schizophrenia-specific molecular network, starting from 152 schizophrenia genetic risk genes to connect with the founded five aberrant metabolites through one intermediate protein and enzyme in the path. Thirty schizophrenia risk genes (that is nitric oxide synthase 1 (NOS1), transcription factor 4 (TCF4), neurotrophin receptor kinase (NTRK3), catechol-O-methyltransferase (COMT), mitochondrial proline dehydrogenase (PRODH), leucine carboxyl methyltransferase 1 (LCMT1), cytosolic purine 5’-nucleotidase (NT5C2), DNA-directed RNA polymerase III subunit RPC3 (POLR3C), integrin beta-1 (ITGB1), integrin alpha-10 (ITGA10), clathrin, heavy chain-like 1 (CLTCL1), cell division control protein 42 homolog (CDC42) and cut-like homebox 1 (CUX1)) were connected to the five metabolites (Figure 2).

Function annotation analysis was performed on the proteins/genes in the network. Sixty biological functions such as glutamine metabolic process, learning or memory, and so on were found significantly involved in the constructed network (Figure 2 and Supplementary Table 3). Arginine, ornithine, glutamine and histidine and their associated proteins/genes were mainly involved in nitrogen compound biosynthetic processes as well as glutamine and arginine metabolism process. PC ae C38:6 and its connecting proteins/genes dominantly participated in several signaling pathways, such as immune-related signaling (for example, Fc epsilon RI signaling and Fc gamma R-mediated phagocytosis, Supplementary Table 3) and neurotrophin signaling.

Discussion
In our study, we quantified 103 metabolites concentration with 265 schizophrenic patients and 216 healthy individuals. We identified five metabolites (ornithine, arginine, glutamine, histidine and PC ae C38:6), which differed significantly between schizophrenic patients and controls, but are not influenced by antipsychotic medication. Furthermore, using a network approach, the altered five metabolites were associated with 13 schizophrenia risk genes.

Arginine and ornithine. Higher blood concentration of ornithine in schizophrenia is consistent with our observation in plasma; however, arginine concentration conflicts with ours. A key risk gene NOS1 is involved in the arginine–nitric oxide pathway, in which cases NOS compete with arginase (ARG1) for the same substrate L-arginine and produce nitric oxide (NO) (Figure 3). Alteration of NO metabolism has been reported in schizophrenia, which may impact on processes such as synapse formation, neurotransmission and so on. Increased NOS activity and decreased ARG1 activity were both reported in plasma samples of schizophrenics. According to previous findings,
we propose that substrate arginine consumption and downstream ornithine accumulation in the urea cycle may be induced by the aberrant NO metabolism in schizophrenia, although it is not clear whether over- or under-production of NO occurs.42,43 Another high-risk gene, TCF4 8,31 that has key roles in neurodevelopment may also have impact on these two amino acids. Differential expression studies of TCF4 demonstrated its downregulation in red blood cells and upregulation in cerebellar cortex of schizophrenics.46 We assume that the dysfunction of TCF4 in blood may suppress its activation of ornithine decarboxylase 1 (ODC1). Decreased expression of ODC in the ornithine-polyamine metabolism,47 which supports our assumption of the suppressed ornithine decarboxylation process, might be responsible for the higher concentration of ornithine in the disease.

**Histidine.** Histidine is the precursor of the ubiquitous neurohormone neurotransmitter histamine. Opposite to our observation, higher level of histidine was reported in schizophrenic plasma samples.41,46,48 As of histamine, lower concentration in whole blood50 and higher level in cerebrospinal fluid51 occurs in schizophrenic patients. The downregulation of TCF4, as we mentioned above, could probably release its inhibition on histidine decarboxylase (Figure 3), which implies an accelerated histidine decarboxylation process in schizophrenia.

**Glutamine.** Dysfunction of glutamine–glutamate cycle between neurons and glia was proposed having roles in the pathophysiology of schizophrenia.52–54 Lower concentration of glutamine was detected in our schizophrenic plasma samples, whereas elevated glutamate was reported both in serum and cortex of patients.54 Nevertheless, the dependency of glutamate concentration between brain and blood seems weak due to the restricted controlled passage across the blood–brain barrier.55 A glutamine-dependent carbamoylphosphate synthase enzyme involved in glutamine hydrolysis (Figure 3) was found with a single-nucleotide polymorphism-associated with schizophrenia,56 but the detailed mechanism is yet unclear.

Although discussed controversially, it appears that amino-acid alteration occurs in the schizophrenia pathology, and the plasma amino-acid concentration changes, to some extent, could reflect the susceptibility to the brain disorder.

**Phosphatidylycerolines.** Cell membrane abnormalities associated with disordered phospholipids composition and metabolism are indicated in schizophrenia pathogenesis.56,57 Significantly lower concentrations of PC in schizophrenic postmortem brain tissues58 are in agreement with our observation of lower PC levels in schizophrenia. However, several similar studies15,16 did not show the significant difference in phospholipids. In Oresić’s15 study, significantly higher metabolite levels in saturated triglycerides were detected, whereas in Yang’s recent report16 fatty acids and ketone bodies were found significantly elevated in both the serum and urine sample of the patients. The detected metabolite signatures varied in different studies probably due to sample size, varied samples and different platforms. Oresić et al measured serum samples by a global lipidomics platform and a small polar metabolites platform based on mass spectrometry. Yang et al quantified serum and urine samples using a GC-TOF spectral-based global metabolic profiling. Both studies cover a global metabolic profiling in schizophrenia, while our study focused on a targeted metabolomics profiling mainly covering acylcarnitines, amino acids, hexoses, PCs and sphingomyelins in schizophrenic plasma samples. In addition, deficiencies of plasmalogens (that is, phospholipid with vinyl-ether double bonds in the sn-1 position) were also found in Alzheimer’s disease and depression, which could be implicated in abnormal signal transduction associated with learning disability and cognitive deficit.59 We found a schizophrenia high-risk gene NTRK3 involved in neurotrophin signaling pathway may disturb the lipid metabolism in schizophrenia (Figure 3). NTRK3 coupled with its ligands BDNF, NGF and partner protein PLCG1 are all involved in the neurotrophin signaling, which promotes the survival and differentiation of peripheral and central neuron. Variation in NTRK3 may disturb the signaling cascade, affecting the binding between protein PLCG1 and PIPNA, then disturbing the regulation on PC transfer, which leads ultimately to a lower PC concentration as we observed. An alternative explanation could be expected that the increased phospholipase A2 activity in schizophrenia implicates an accelerated breakdown of membrane phospholipids in the disease.60 In our network analysis, PC ae C38:6 and its connecting proteins/genes mainly participated in several signaling pathways (for example immune-related signaling and neurotrophin signaling). The neurotrophins have important roles in neural development and are transmitted to intracellular through signaling cascades such as MAPK pathway, in turn leading to regulation of cell differentiation and survival, ultimately impacting higher-order activities such as learning and memory.

We also found that PC ae C34:3 showed no significant difference in NF versus controls, but significantly differed in NT versus NF, indicating a potential antipsychotic effect on its concentration change. An association study in Korean inpatients63 was to investigate possible associations between single-nucleotide polymorphisms within ten genes with schizophrenia and response to antipsychotics. ABCB4, one of the enzymes of PC ae C34:3 included in the ten candidate genes, which might involve in the metabolism to neuroleptics and need further investigation.

**Limitations.** The metabolite panels we used in our study contain mainly lipids, amino acids and carnitines and were unable to cover the whole metabolome and some of the metabolites we measured were not specific enough (for example the detailed structure of the PCs are not clear), which leaves us more area to be explored regarding metabolite concentration alterations in schizophrenia. The participants in our study were only Caucasian, which limits our findings to a more general population. However, even with these limits, we shed some lights on the mechanism of schizophrenia. With a broad coverage of metabolites and population, more signatures might be discovered to help better understand schizophrenia.
We identified the underlying molecular signatures in plasma of schizophrenic patients compared with healthy individuals. The aberrant concentrations of metabolites in plasma decipher possible disturbed pathways, including glutamine and arginine metabolism, nitrogen compound biosynthetic process, learning memory behavior, several signaling pathways, may reflect the genetic susceptibility of the disease. Although we encapsulated merely a small picture of metabolites, we believe that metabolomics analyses have huge potential to investigate candidate disease markers that will benefit the diagnostics and therapeutics of schizophrenia in the future.

Conflict of Interest

The authors declare no conflict of interest.

Acknowledgements

We thank all the participants in the study. We also thank all the staff from Genome Analysis Center, Metabolomics Core Facility, who helped in the sample logistics, data collection and metabolite quantification assays, especially Dr Werner Römisch-Margl, Julia Scarpa, Katharina Sckell and Arsin Sabunchi. The research was kindly funded by EU FP7 grant HEALTH-2009-2.2.1-3/242114 (Project OPTiMiSE) and in part supported by grants from the German Federal Ministry of Education and Research (BMBF) to Project-FKZ: 01DK012030 and to the German Center Diabetes Research (DZD e.V.).

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Supplementary Information accompanies the paper on the Translational Psychiatry website (http://www.nature.com/tp)