

Proteomics Analysis of Serum from COVID-19 Patients

Xiaoling Liu,[#] Yinghao Cao,[#] Hongmei Fu, Jie Wei, Jianhong Chen, Jun Hu,^{*} and Bende Liu^{*}Cite This: *ACS Omega* 2021, 6, 7951–7958

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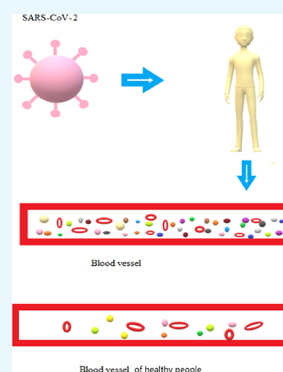


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ABSTRACT: Coronavirus disease 2019 (COVID-19) is a worldwide pandemic. To understand the changes in plasma proteomics upon SARS-CoV-2 infection, we analyzed the protein profiles of plasma samples from 10 COVID-19 patients and 10 healthy volunteers by using the DIA quantitative proteomics technology. We compared and identified differential proteins whose abundance changed upon SARS-CoV-2 infection. Bioinformatic analyses were then conducted for these identified differential proteins. The GO/KEEG database was used for functional annotation and enrichment analysis. The interaction relationship of differential proteins was evaluated with the STRING database, and Cytoscape software was used to conduct network analysis of the obtained data. A total of 323 proteins were detected in all samples. Difference between patients and healthy donors was found in 44 plasma proteins, among which 36 proteins were up-regulated and 8 proteins were down-regulated. GO functional annotation showed that these proteins mostly composed of cellular anatomical entities and proteins involved in biological regulation, cellular processes, transport, and other processes. KEGG functional annotation further showed that these proteins were mainly involved in complement system activation and infectious disease processes. Importantly, a KEGG pathway (natural killer cell-mediated cytotoxicity) was enriched, with three important activators of this pathway, ICAM1/2 and IgG, being up-regulated. Protein–protein interaction (PPI) statistics indicated that, among these 44 proteins, 6 were the most significantly up-regulated (DBH, SHGB, TF, ICAM2, THBS1, and C1RL) while 2 were the most significantly down-regulated (APCS and ORM1). Results from this study showed that a few proteins associated with immune activation were up-regulated in patient plasma. In addition, this study established a method for extraction and quantitative determination of plasma components in convalescent plasma from COVID-19 patients.



1. INTRODUCTION

Coronavirus disease 2019 (COVID-19) is an epidemic disease characterized by acute respiratory distress syndrome caused by SARS-CoV-2. Since its first discovery in Wuhan, it has rapidly spread to 212 countries and regions around the world. Being the first pandemic in the 21st century, by August 4, 2020, the total number of confirmed cases worldwide has exceeded 18 million, with a crude mortality rate of about 3.7%. Based on the severity of the disease, symptoms were divided into mild and severe. Eighty percent of patients presented with mild symptoms such as fever, cough, loss of appetite, dizziness, fatigue, etc. Conventional antiviral therapy and other supportive treatments could effectively alleviate the conditions. Patients with severe symptoms required life-sustaining treatment such as intensive care and mechanical ventilation. At present, a specific antiviral treatment for COVID-19 is very limited. To effectively control the spread of the epidemic, tremendous efforts are being made all over the world in vaccine and medicine research and development for prevention and treatment of this disease.

SARS-CoV-2 and SARS-CoV share 79.5% sequence homology, and the mechanisms of infection and pathogenesis of them are similar.¹ SARS-CoV-2 binds to host cells using the spike protein (S protein) on the surface of the virus envelope and mediates the fusion of the host cell and virus during the

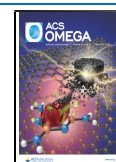
infection process.² SARS-CoV-2 has one polyprotein (Orf1ab), four structural proteins (S, E, M, and N), and six accessory proteins (Orf3a, Orf6, Orf7a, Orf7b, Orf8, and Orf10) involved in virus replication and transcription. The interaction between the S protein receptor binding domain (RBD) of SARS-CoV-2 and the ACE2 (angiotensin converting enzyme 2) molecule was a necessary process for SARS-CoV-2 to invade host cells.^{3,4}

Some studies have investigated the vulnerability of medical staff in the COVID-19 epidemic. Korth et al. detected SARS-CoV-2-IgG antibodies in 1.6% of 316 health care workers, suggesting that the medical staff were vulnerable to SARS-CoV-2 infection.⁵ Another study in the United States found that 1.5% of 3500 medical workers were positive for SARS-CoV-2 antibodies.⁶ In addition to conventional antiviral therapy and basic supportive therapy, convalescent plasma therapy was considered a promising treatment approach. Historically, this approach has been adopted for a variety of

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infectious diseases. The Spanish flu in the 1920s was the first viral infection to be treated effectively with convalescent plasma. Convalescent plasma therapy was found to be effective in treating SARS in 2003, the Middle East respiratory syndrome in 2014, Ebola in 2015, and some other viral infectious diseases.^{7–9} Convalescent plasma contains neutralizing antibodies, natural antibodies, and anti-inflammatory factors, clotting factors, and other proteins of uncertainty.¹⁰ As such, convalescent plasma not only directly neutralizes the virus but also reduces inflammatory response, which mediates inflammatory storm in COVID-19 patients.¹¹ Analyzing the plasma protein components of COVID-19 patients may provide insights into body response to SARS-CoV-2 infection and will have profound implication to vaccine and therapeutic research and development against COVID-19. In this study, we quantitatively analyzed the plasma protein profiles of COVID-19 patients and healthy subjects and identified protein components, which showed different abundance between the two groups. Results from this work demonstrated that multiple factors involved in humoral and cellular immune responses were increased in patient plasma, suggesting that both humoral and cellular immune responses may be required for fighting SARS-CoV-2 infection.

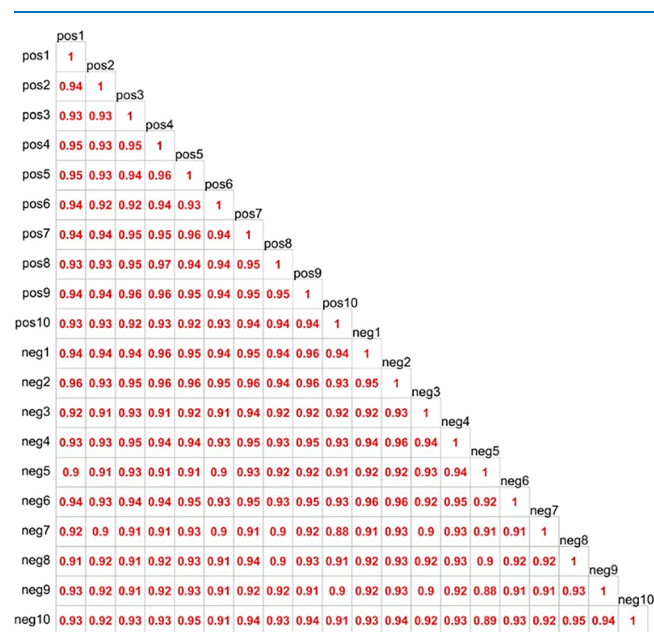


Figure 1. Quantitative correlation analysis of sample protein.

2. MATERIAL AND METHODS

2.1. Subjects. Inclusion criteria: (i) aged over 18 years; (ii) SARS-CoV-2 infection was confirmed as respiratory tract specimens by real-time reverse transcription polymerase chain reaction (RT-PCR); (iii) the severity of the disease is classified as mild according to the COVID-19 guidelines (Trial Version 7).¹²

Exclusion criteria: (i) under the age of 18; (ii) patients with severe cardiopulmonary disease, immunodeficiency syndrome, malignant tumor, or mental abnormalities who were unable to cooperate.

Ten patients diagnosed with COVID-19 in the Union Hospital and Liyuan Hospital of Tongji Medical College of Huazhong University of Science and Technology from January

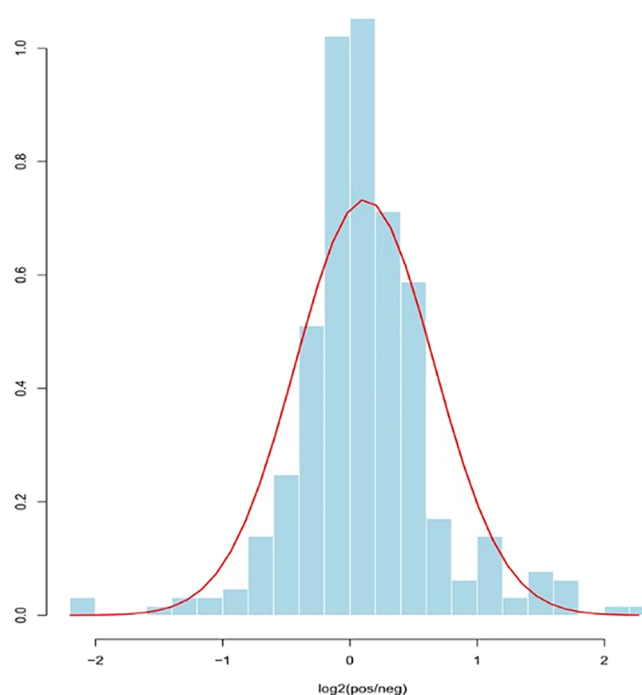


Figure 2. Histogram distribution of cases.

Table 1. Baseline Characteristics of Patients Included in the Study

subject/group	case group (mean \pm SD)	control group (mean \pm SD)	P value
average age	62.8 \pm 14.597	50.7 \pm 14.103	0.076
sex/men/n	4	5	0.395
complications/n	6		
severity (light/severe)	light		
WBC $\times 10^9$	6.44 \pm 1.810	6.53 \pm 1.443	0.909
Lym %	26.07 \pm 11.393	36.64 \pm 6.337	0.02
Neu %	61.08 \pm 13.376	60.05 \pm 8.268	0.838
RBC $\times 10^{12}$	3.79 \pm 0.907	4.42 \pm 0.546	0.078
PLT $\times 10^9$	187.7 \pm 77.424	227.6 \pm 41.078	0.167
ALT (u/L)	83.01 \pm 192.283	26.98 \pm 15.589	0.371
AST (μ /L)	58.84 \pm 105.445	26.1 \pm 7.790	0.34
TBIL (μ mol/L)	19.26 \pm 14.527	9.6 \pm 5.039	0.064
BUN (μ mol/L)	7.4 \pm 5.609	5.8 \pm 1.755	0.41
Crea (μ mol/L)	101.39 \pm 108.967	63.16 \pm 16.568	0.287
PCT (ng/mL)	0.49 \pm 1.206	0.03 \pm 0.026	0.258
LDH (μ /L)	245.6 \pm 91.504	201.3 \pm 46.070	0.188

to April 2020 were included in the case group, and 10 healthy subjects were included in the control group. All investigators were informed of the purpose of the study and precautions that should be taken. All participants in this study signed written informed consent forms. This study was approved by the ethics committee of the respective hospitals.

2.2. Preparation of Plasma Samples. Fasting venous blood of patients and healthy subjects was collected in the morning. Reaction solution (1% SDC, 10 mM TCEP, 40 mM CAA) was added to the plasma samples and incubated at 60 $^{\circ}$ C for 30 min for protein denaturation disulfide bond reduction as well as sulfhydryl alkylation. Protein concentration in samples was determined using the Bradford assay. To conduct enzymatic digestion for the samples, the concentration of SDC was diluted to less than 0.5%, trypsin was added to

Table 2. Differential Protein Information

protein names	protein description	FC (pos/ neg)	log2FC (pos/ neg)	P value	state
HV5X1_HUMAN	immunoglobulin heavy variable 5-10-1	4.82	2.26872	0.005737	up
GPV_HUMAN	platelet glycoprotein V	4.17	2.05876	0.001849	up
SHBG_HUMAN	sex hormone-binding globulin	3.32	1.7328	0.006005	up
TSP1_HUMAN	thrombospondin-1	3.11	1.63661	0.015421	up
DOPO_HUMAN	dopamine β -hydroxylase	2.82	1.49417	0.003681	up
ICAM2_HUMAN	intercellular adhesion molecule 2	2.66	1.40923	0.005953	up
KV621_HUMAN	immunoglobulin kappa variable 6-21	2.34	1.22735	0.02518	up
LV469_HUMAN	immunoglobulin lambda variable 4-69	2.26	1.17911	0.02029	up
ARAP1_HUMAN	Arf-GAP with Rho-GAP domain, ANK repeat and PH domain-containing protein 1	2.17	1.11569	0.041246	up
S10AC_HUMAN	protein S100-A12	2.16	1.11386	0.005249	up
HV205_HUMAN	immunoglobulin heavy variable 2-5	2.06	1.04224	0.024872	up
LV325_HUMAN	immunoglobulin lambda variable 3-25	2.02	1.01292	0.008503	up
HV43D_HUMAN	immunoglobulin heavy variable 3-43D	1.93	0.94639	0.012763	up
TRFE_HUMAN	serotransferrin	1.79	0.83607	0.046888	up
CRAC1_HUMAN	cartilage acidic protein 1	1.63	0.70302	0.010019	up
KVD15_HUMAN	immunoglobulin kappa variable 3D-15	1.61	0.69054	0.007783	up
KV320_HUMAN	immunoglobulin kappa variable 3-20	1.57	0.651	0.012032	up
HV323_HUMAN		1.57	0.64824	0.012064	up
C1RL_HUMAN	complement C1r subcomponent-like protein	1.56	0.64178	0.002045	up
HVD82_HUMAN	immunoglobulin heavy variable 4-38-2	1.54	0.62082	0.017265	up
HV118_HUMAN	immunoglobulin heavy variable 1-18	1.53	0.61393	0.046497	up
CETP_HUMAN	cholesteryl ester transfer protein	1.47	0.55667	0.029024	up
A2MG_HUMAN	alpha-2-macroglobulin	1.44	0.52321	0.015519	up
IGHG2_HUMAN	immunoglobulin heavy constant gamma 2	1.43	0.51655	0.022341	up
KV108_HUMAN	immunoglobulin kappa variable 1-8	1.40	0.48435	0.02486	up
AFAM_HUMAN	afamin	1.39	0.47767	0.041463	up
HV428_HUMAN	immunoglobulin heavy variable 4-28	1.38	0.46815	0.034601	up
PLTP_HUMAN	phospholipid transfer protein	1.38	0.46297	0.00479	up
TENX_HUMAN	tenascin-X	1.38	0.46195	0.031929	up
KV106_HUMAN		1.35	0.43012	0.037791	up
KV37_HUMAN	probable non-functional immunoglobulin kappa variable 3-7	1.33	0.4118	0.013726	up
IGG1_HUMAN		1.31	0.3898	0.008494	up
PGRP2_HUMAN	N-acetylmuramoyl-L-alanine amidase	1.30	0.37717	0.011929	up
APOA1_HUMAN	apolipoprotein A-I	1.29	0.36524	0.02888	up
IGKC_HUMAN	immunoglobulin kappa constant	1.28	0.35081	0.011717	up
C1QA_HUMAN	complement C1q subcomponent subunit A	1.27	0.34757	0.025015	up
C4BPA_HUMAN	C4b-binding protein alpha chain	0.77	-0.38011	0.00728	down
CERU_HUMAN	ceruloplasmin	0.76	-0.38873	0.009802	down
SAA4_HUMAN	serum amyloid A-4 protein	0.70	-0.51451	0.000425	down
IC1_HUMAN	plasma protease C1 inhibitor	0.68	-0.56084	0.007387	down
LG3BP_HUMAN	galectin-3-binding protein	0.66	-0.60309	0.018873	down
LBP_HUMAN	lipopolysaccharide-binding protein	0.59	-0.77189	0.029039	down
A1AG1_HUMAN	alpha-1-acid glycoprotein 1	0.58	-0.78207	0.018007	down
SAMP_HUMAN	serum amyloid P component	0.48	-1.06577	0.001006	down

reach a mass ratio of 1:50 between the enzyme and protein, and the samples were incubated at 37 °C overnight. On the following day, TFA was added to terminate the enzymatic digestion, and the pH value of the solution was adjusted to about 6.0. The solution was centrifuged at 12,000g for 15 min, and the supernatant was taken for desalting. After desalination, the peptide solution was filtered by using centrifugal concentrators and stored at -20 °C until being analyzed.

2.3. Data-Independent Acquisition (DIA) Analysis.

DIA technology was used for mass spectrometry analysis. The analysis was conducted on an Ultimate3000 (capillary flow) combined with a Q Exactive HF-X. While conducting a SWATH scan, the scan of each cycle contained one MS1 scan (scan range, 350–1250 m/z ; resolution, 60 K; AGC, 3×10^6 ;

max interval time (IT), 20 ms) and 40 variable window MS2 scans (resolution, 30 K; AGC, 1×10^6 ; max IT, auto). Secondary mass spectrometry files obtained from the DIA scan were processed by DIA-Umpire, which could be used for database retrieval. The MSFragger software was used for database retrieval of the secondary mass spectrogram, and the results obtained were used as the spectrogram library. Quantitative algorithm DIA-NN was used for subsequent DIA-targeted extraction. Quantitative information for components obtained by DIA analysis was screened by a 1% false discovery rate (FDR), the normalized data was converted to log2 data, which was then used for difference comparison and t test analysis, and the fold change ratio and P value were used to screen for protein components, which showed different

abundance between patients and the control group. Differential protein was screened by the ratio and P value. Screening criteria: ① if fold change ≥ 1.2 and $P \leq 0.05$, then the protein was over-represented; ② if fold change ≤ 0.833 and $P \leq 0.05$, then the protein was under-represented.

2.4. Bioinformatics Analysis. The GO/KEEG database was used for functional annotation and enrichment analysis. The interaction relationship of differential proteins was evaluated with the STRING database, and Cytoscape software was used to conduct network analysis of the obtained data.

2.5. Statistical Analysis. Continuous variables were represented by mean \pm standard deviation. The percentage of categorical variables was used to compare the difference between case and control groups. The chi-square test or Fisher test was used to compare categorical variables between the two groups. SPSS Version 24 (Chicago, IL, USA) was used for all statistical analyses. $P < 0.05$ was considered statistically significant.

3. RESULTS

3.1. Basic Patient Information. Basic data for all patients and healthy individuals included in the study are shown in Table 1. There were four males in the case group. The average age of this group was 62.8 ± 14.597 , and six patients in this group had basic diseases including diabetes, hypertension, and kidney stones. All 10 patients had mild symptoms. There were four males in the control group. The average age of the control group was 50.7 ± 14.103 . There was a statistically significant difference in the percentage of lymphocytes between the two groups ($P = 0.02$). No significant difference in age, gender, and other blood test results was found between the two groups ($P > 0.05$).

3.2. Proteomics Analysis. A total of 323 plasma proteins were detected in both groups, and the quantitative values for all proteins are shown in Table 1. Proteins with fold change ≥ 1.2 and $P \leq 0.05$ were classified as up-regulated proteins, and proteins with fold change ≤ 0.833 and $P \leq 0.05$ were classified as down-regulated proteins. The quantitative correlation for proteins in each sample was good, and the quantitative correlation coefficient R^2 values of different samples were all greater than 0.9 (Figure 1). The ratio distribution of the case group is roughly normal (Figure 2). A total of 44 differential proteins were detected between the two groups, of which 36 were up-regulated and 8 were down-regulated (Table 2 and Figures 3 and 4).

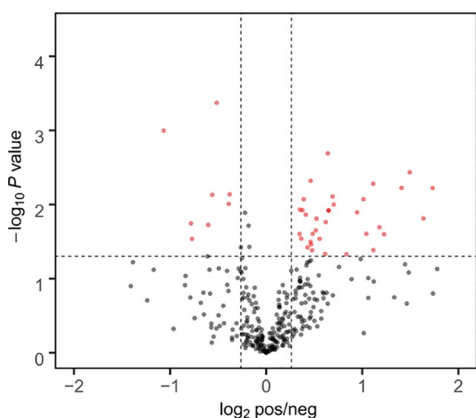


Figure 3. Volcano figure.

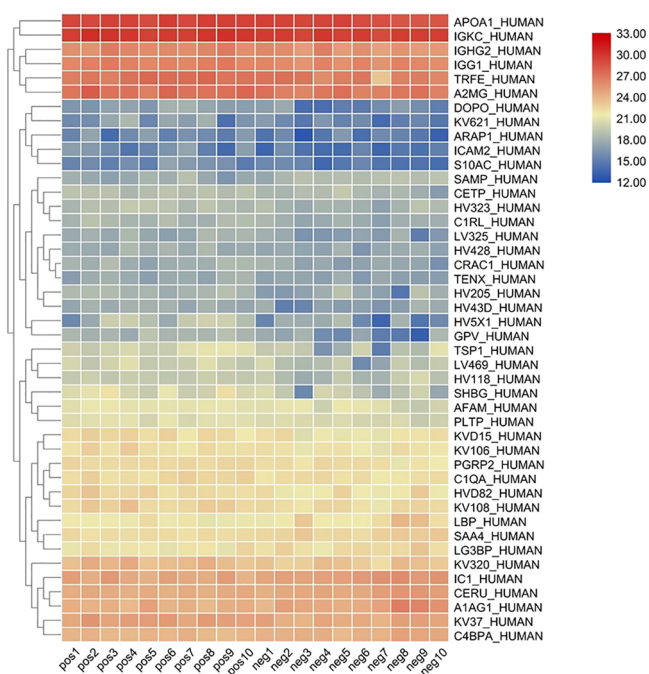


Figure 4. Clustering heat map of all proteins.

3.3. GO Functional Annotation and Enrichment Analysis. Differential proteins identified were annotated as three GO levels and three categories, namely, the biological process (BP), cellular component (CC), and molecular function (MF) (Figure 5A–C, respectively). In the BP category, differential proteins in the patient group were mainly involved in biological regulation (13.06%), cellular processes (12.16%), and the response to stimulation (10.36%). In the

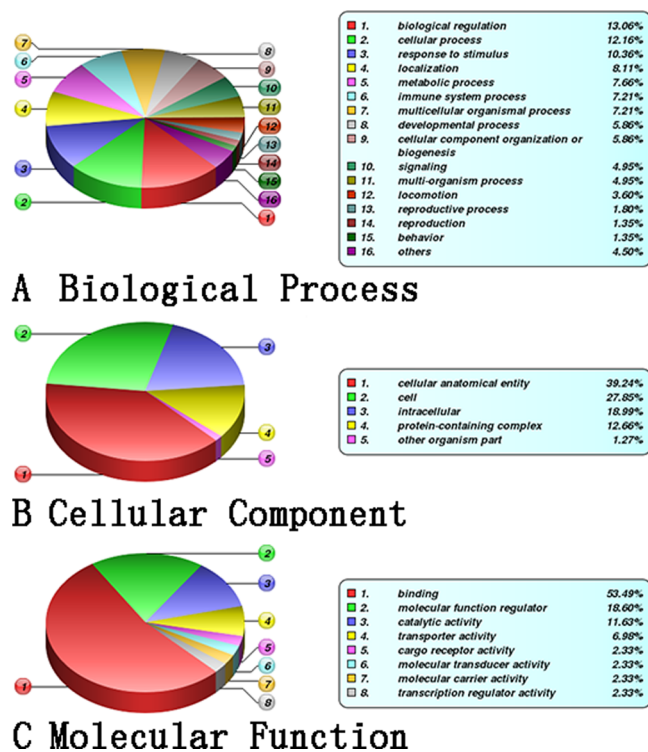


Figure 5. GO annotation and enrichment analysis. (A) Biological process, (B) cellular component, and (C) molecular function.

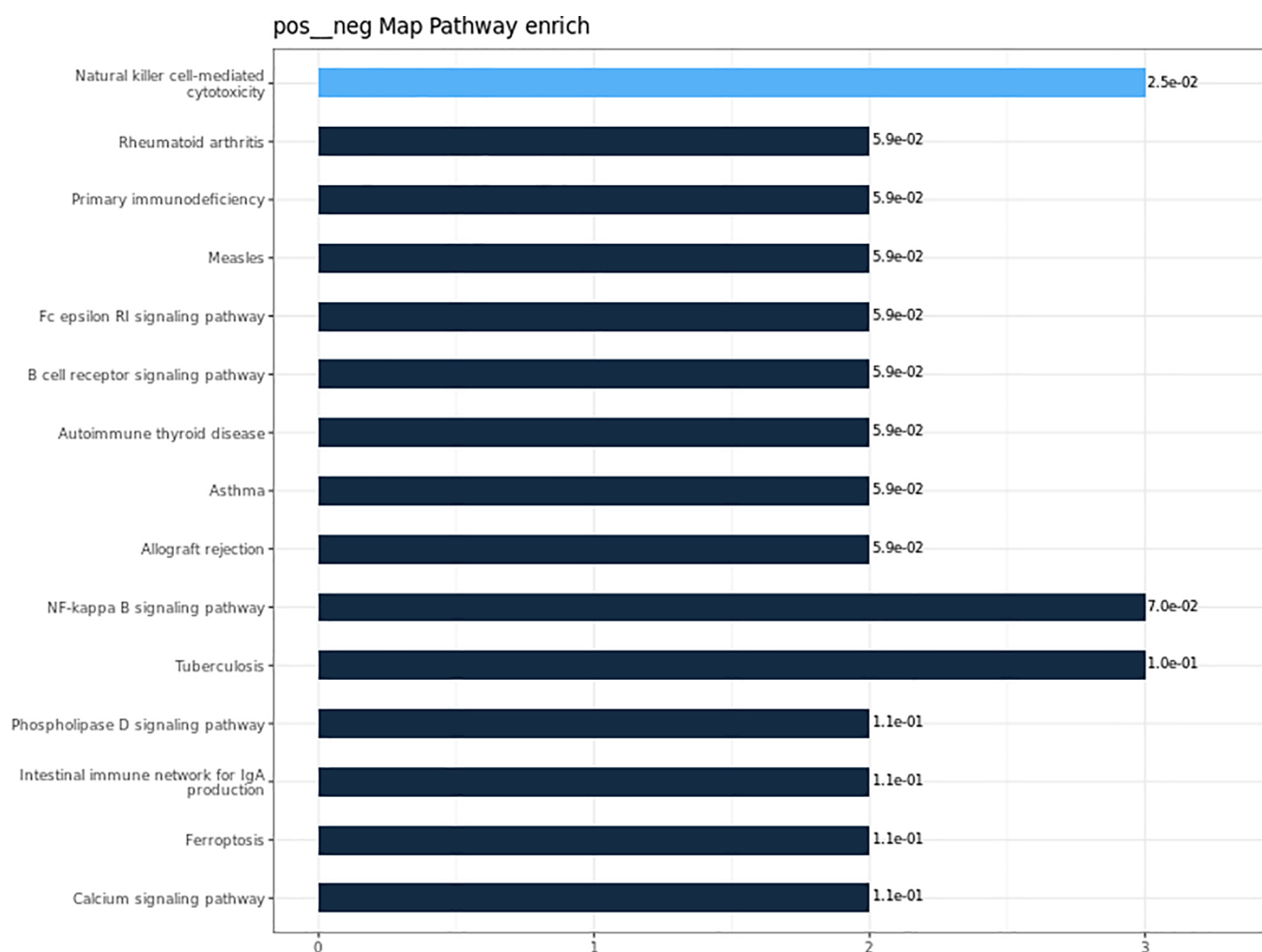


Figure 6. KEGG annotation and enrichment analysis.

CC category, 39.24% belonged to the anatomical entity of the cell, 27.85% belonged to the cell, and 18.3% belonged to the intracellular. In the MF category, 53.49% had a binding function, 18.6% were molecular function regulators, and 11.63% had catalytic activity. The hypergeometric test was used to analyze the functional classification or pathway in which differentially represented proteins were significantly enriched compared with the background proteins (the total number of proteins with quantitative and annotated information) ($P \leq 0.05$). This case group did not have the GO classification information enriched.

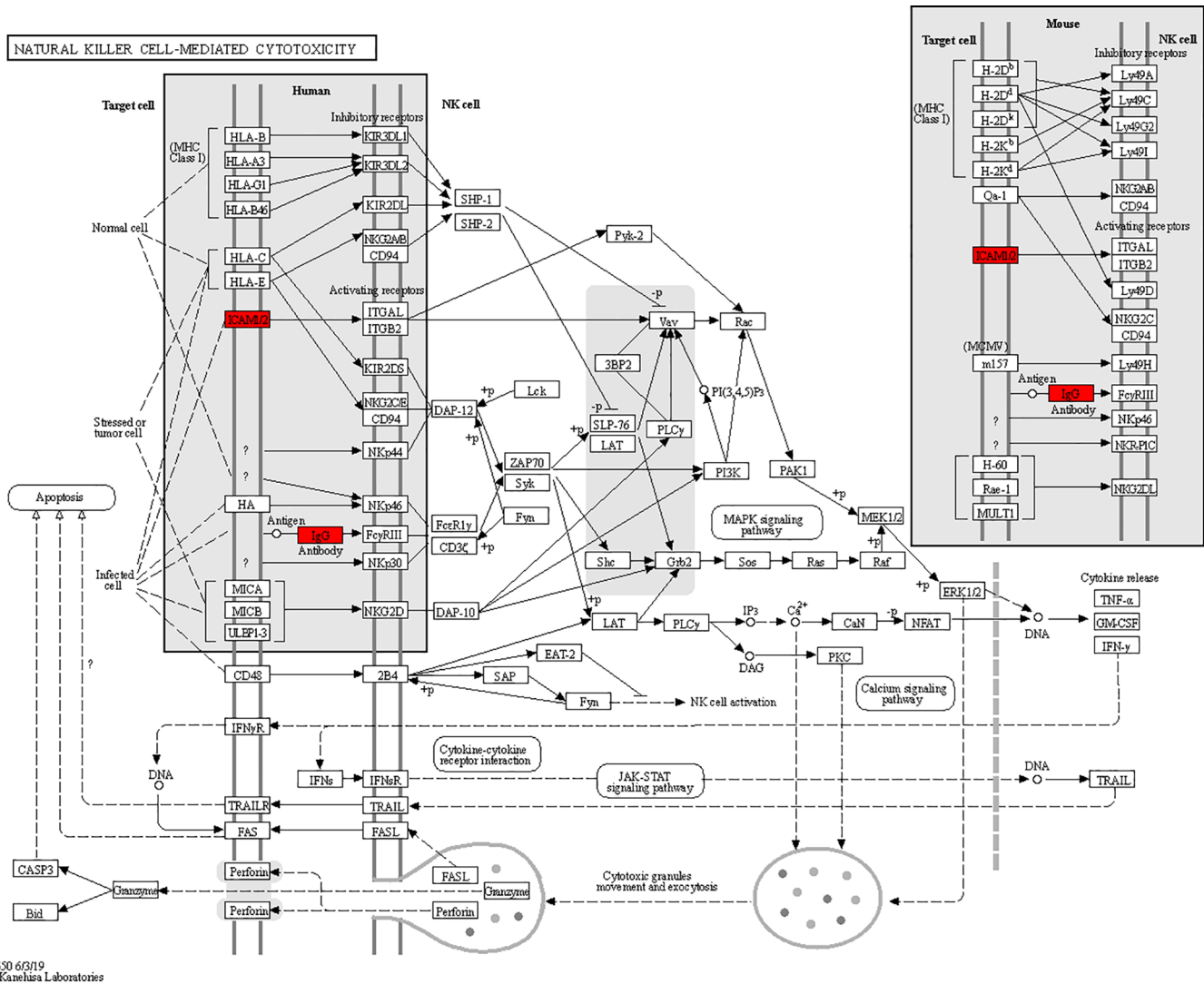
3.4. KEGG Functional Annotation and Enrichment Analysis. To better understand the functions of the identified differential proteins in physiological or pathological activities and to determine the metabolic and signaling pathways involved, we carried out KEGG pathway analysis for the differential proteins. Our analyses showed that the differential proteins were mainly involved in the following biological process: NK cell-mediated primary immunodeficiency NF- κ B signaling pathway tuberculosis in rheumatoid arthritis (Figure 6). The pathway with the highest KEGG enrichment was natural killer cell-mediated cytotoxicity (Figure 7).

3.5. PPI Network Construction. The interaction relationship of differential proteins was evaluated with the STRING database, and Cytoscape software was used to conduct network analysis of the obtained data (Figure 8). Each node in the

figure represented a protein, and the attachment between nodes represented the interaction between proteins. Red represented up-regulation, blue represented down-regulation, and the darker the color, the larger the variation multiple of the protein. As shown in the figure, over-represented proteins with significant changes included DBH (dopamine β -hydroxylase), SHBG (sex hormone-binding globulin), TF (serum transferrin), THBS1 (thrombospondin-1), ICAM2 (intercellular adhesion molecule 2), and C1RL (complement C1r sub-component-like protein). Two proteins, namely, APCS (serum amyloid P component) and ORM1 (alpha-1-acid glycoprotein 1) were significantly under-represented.

4. DISCUSSION

With the spread of COVID-19 and the urgent need for treatment and prevention options, a systematic study of virus–host interactions is critically important. In a recent study, Gordon et al. reported the SARS-CoV-2 protein interaction map and identified 66 human proteins or cytokines as potential therapeutic targets for COVID-19.¹³ As is well known, after viruses invade host cells, the body undergoes a series of antiviral reactions, including inflammatory cell aggregation, cytokine release, humoral immunity, and cellular immune activation. Among these factors, antiviral antibodies play a very critical role. By detecting IgM and IgG in the sera of 29 COVID-19 patients during convalescence, Jiang et al.¹⁴



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Figure 7. Enrichment KEGG pathway diagram.

constructed a proteomics microarray containing 18 predicted viral proteins and characterized the profile of SARS-CoV-2-specific serum antibodies, especially antibodies against S1 and N proteins.¹⁵ Long et al. analyzed the acute antibody response in 285 COVID-19 patients, and the results showed that 100% of patients tested positive for SARS-CoV-2 virus IgG within 19 days from the symptom onset.¹⁶ These studies characterized direct interactions between viral and host proteins and the antibody response to viral infection. However, a systematic investigation of change in protein components in patient serum as a response to SARS-CoV-2 infection was never reported yet. In this study, a total of 44 differential proteins were identified in plasma samples from COVID-19 patients with 36 being over-represented and 8 under-represented in patient samples in comparison with control samples.

Through GO functional annotation, it was found that these proteins were mostly cellular anatomical entities and factors involved in biological regulation, cell process, and transportation. KEGG functional annotation further revealed that these differential proteins were mainly involved in the complement system activation, pertussis, and other infectious diseases. Notably, members of the pathway of natural killer cell-mediated cytotoxicity were enriched, as evidenced by the

finding that ICAM1/2 and IgG were all over-represented in patient samples. ICAM1/2 participates in the NF- κ B signaling pathway, natural killer cell-mediated cytotoxicity, the TNF signaling pathway, the interaction between signaling molecules, leukocyte migration across the endothelium, etc.^{17,18} PPI interaction analysis suggested that SHBG participates in receptor-mediated processes; TF may play a role in stimulating cell proliferation; THBS1 mediates intracellular and inter-cellular interactions. APCS can interact with DNA and histones. ORM1 may play a role in regulating the activity of the immune system. These factors can directly or indirectly block viral invasion and replication and other processes, which are of great significance for virus defense.¹⁹

Plasma phospholipid transfer protein (PLTP), a member of the lipid transfer/lipoprotein binding protein gene family, is a major regulator of lipoprotein metabolism and also has the function of regulating inflammation and immune response. PLTP has a specific effect on Th1/Th2 polarization. Lack of PLTP will reduce the expression of IL-18, and antigen-presenting cells cannot induce Th1 cell differentiation due to lack of IL-18, thereby affecting the immune function of the body.²⁰ Ceruloplasmin is a protein that maintains iron and copper homeostasis in the body. It also has the functions of

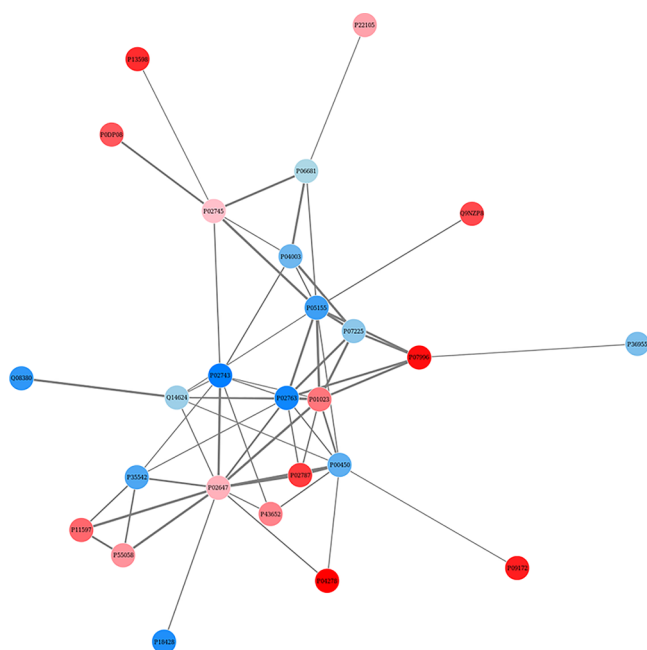


Figure 8. Protein interaction network analysis.

copper transport, blood coagulation, angiogenesis, and oxidative stress. Down-regulation of this protein has been observed in many viral infections.²¹ Cholesteryl ester transfer protein (CETP) can mediate the exchange of CE in HDL to TG in other lipoproteins.²² It is speculated that the up-regulation of CETP may be due to the important role of lipids in virus infection. Lipids participate in membrane fusion, envelope, and transformation as direct or indirect virus receptors, fusion cofactors, and entrance cofactors to help viruses invade host cells.^{23,24} Study has suggested that the lipid may play important roles in the virus life cycle and can be potential targets for therapeutic intervention of COVID-19.²⁵

In this study, 44 differential proteins were identified, among which 15 belonged to the immunoglobulin superfamily.

There are some limitations in this study. On the one hand, the sample size of the study was relatively small, which might lead to selection bias in the results. On the other hand, the differential proteins identified were not confirmed with alternative assays. Therefore, in the future, studies with larger sample sizes may be conducted if conditions allow, and differential proteins identified will need to be confirmed by other analytical methods.

AUTHOR INFORMATION

Corresponding Authors

Jun Hu – Department of endocrinology, Liyuan Hospital, Tongji Medical College, Huazhong University of Science and Technology, Wuhan, Hubei 430022, China; Email: junhu1108@163.com

Bende Liu – Department of Emergency Medicine, Union Hospital, Tongji Medical College, Huazhong University of Science and Technology, Wuhan 430022, China; Department of Nosocomial Infection Management, Union Hospital, Tongji Medical College, Huazhong University of Science and Technology, Wuhan 430022, China; orcid.org/0000-0002-4825-1402; Phone: +86-13907191851; Email: 13907191851@139.com

Authors

Xiaoling Liu – Department of endocrinology, Liyuan Hospital, Tongji Medical College, Huazhong University of Science and Technology, Wuhan, Hubei 430022, China

Yinghao Cao – Department of Gastrointestinal Surgery, Union Hospital, Tongji Medical College, Huazhong University of Science and Technology, Wuhan 430022, China; Department of Nosocomial Infection Management, Union Hospital, Tongji Medical College, Huazhong University of Science and Technology, Wuhan 430022, China

Hongmei Fu – Department of endocrinology, Liyuan Hospital, Tongji Medical College, Huazhong University of Science and Technology, Wuhan, Hubei 430022, China

Jie Wei – Department of endocrinology, Liyuan Hospital, Tongji Medical College, Huazhong University of Science and Technology, Wuhan, Hubei 430022, China

Jianhong Chen – Department of endocrinology, Liyuan Hospital, Tongji Medical College, Huazhong University of Science and Technology, Wuhan, Hubei 430022, China

Complete contact information is available at: <https://pubs.acs.org/10.1021/acsomega.1c00616>

Author Contributions

*X.L. and Y.C. contributed equally to this work.

Notes

The authors declare no competing financial interest.

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