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Proteomic blood profiling in mild, severe and critical COVID-19 patients

Hamel Patel^{1,2,16}✉, Nicholas J. Ashton^{3,4,5,6,16}, Richard J. B. Dobson^{1,2,7,8,9,10}, Lars-Magnus Andersson^{11,12}, Aylin Yilmaz^{11,12}, Kaj Blennow^{3,13}, Magnus Gisslen^{11,12} & Henrik Zetterberg^{3,13,14,15}

The recent SARS-CoV-2 pandemic manifests itself as a mild respiratory tract infection in most individuals, leading to COVID-19 disease. However, in some infected individuals, this can progress to severe pneumonia and acute respiratory distress syndrome (ARDS), leading to multi-organ failure and death. This study explores the proteomic differences between mild, severe, and critical COVID-19 positive patients to further understand the disease progression, identify proteins associated with disease severity, and identify potential therapeutic targets. Blood protein profiling was performed on 59 COVID-19 mild (n = 26), severe (n = 9) or critical (n = 24) cases and 28 controls using the OLINK inflammation, autoimmune, cardiovascular and neurology panels. Differential expression analysis was performed within and between disease groups to generate nine different analyses. From the 368 proteins measured per individual, more than 75% were observed to be significantly perturbed in COVID-19 cases. Six proteins (IL6, CKAP4, Gal-9, IL-1ra, LILRB4 and PD-L1) were identified to be associated with disease severity. The results have been made readily available through an interactive web-based application for instant data exploration and visualization, and can be accessed at <https://phidatalab-shiny.rosalind.kcl.ac.uk/COVID19/>. Our results demonstrate that dynamic changes in blood proteins associated with disease severity can potentially be used as early biomarkers to monitor disease severity in COVID-19 and serve as potential therapeutic targets.

The recent severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) pandemic manifests itself as a mild respiratory tract infection in most individuals leading to coronavirus disease 2019 (COVID-19) disease. However, in some infected individuals, this can progress to severe pneumonia and acute respiratory distress syndrome (ARDS), leading to multi-organ failure and death. The exact percentage of patients presenting with severe symptoms is currently impossible to calculate as the exact number of individuals who have contracted the virus is

¹Department of Biostatistics and Health Informatics, Institute of Psychiatry, Psychology and Neuroscience, King's College London, London, UK. ²NIHR BioResource Centre Maudsley, NIHR Maudsley Biomedical Research Centre (BRC) at South London and Maudsley NHS Foundation Trust (SLaM) & Institute of Psychiatry, Psychology and Neuroscience (IoPPN), King's College London, De Crespigny Park, London SE5 8AF, UK. ³Department of Psychiatry and Neurochemistry, Institute of Neuroscience and Physiology, The Sahlgrenska Academy, University of Gothenburg, Göteborg, Sweden. ⁴Department of Psychiatry and Neurochemistry, Wallenberg Centre for Molecular and Translational Medicine, Institute of Neuroscience and Physiology, the SAHLGRENKA Academy at the University of Gothenburg, Sahlgrenska University Hospital, MedTech West, Röda stråket 10B, 413 45 Göteborg, Sweden. ⁵Department of Old Age Psychiatry, Institute of Psychiatry, Psychology & Neuroscience, King's College London, London, UK. ⁶NIHR Biomedical Research Centre for Mental Health & Biomedical Research Unit for Dementia at South London & Maudsley NHS Foundation, London, UK. ⁷UK Dementia Research Institute at UCL, London, UK. ⁸Health Data Research UK London, University College London, 222 Euston Road, London, UK. ⁹Institute of Health Informatics, University College London, 222 Euston Road, London, UK. ¹⁰The National Institute for Health Research University College London Hospitals Biomedical Research Centre, University College London, 222 Euston Road, London, UK. ¹¹Department of Infectious Diseases, Institute of Biomedicine, Sahlgrenska Academy, University of Gothenburg, Gothenburg, Sweden. ¹²Department of Infectious Diseases, Region Västra Götaland, Sahlgrenska University Hospital, Gothenburg, Sweden. ¹³Clinical Neurochemistry Laboratory, Sahlgrenska University Hospital, Mölndal, Sweden. ¹⁴Department of Neurodegenerative Disease, UCL Institute of Neurology, London, UK. ¹⁵UK Dementia Research Institute at UCL, London, UK. ¹⁶These authors contributed equally: Hamel Patel and Nicholas J. Ashton. ✉email: hamel.patel@kcl.ac.uk

		Control	Mild	Severe	Critical	Total cases
Participants (n)		28	26	9	24	59
Follow-up (n)		0	6	0	5	11
Gender (M/F)		15/13	13/13	6/3	22/2	41/18
Age (95% CI)		63.1 (55.9–70.3)	51.3 (45.9–56.8)	64.8 (55.0–74.6)	61.0 (56.2–65.8)	57.8 (53.8 – 60.8)
Ethnicity (n)	African black	NA	0	0	1	1
	Caucasian	NA	20	8	23	51
	Persian	NA	6	1	0	7
“Days since symptom onset” (95% CI)		NA	20.9 (17.3–24.5)	12 (10.3–13.7)	11 (9.1–13)	15.5 (13.4–17.7)

Table 1. Cohort demographics. The table provides a summary of the available demographics in this study. Cases represent the mild, the severe, and the critical group merged into one group. The “Follow-up” represents the number of patients that had samples taken at two different time points. The mild group was identified to contain patients significantly younger when compared to the remaining groups (control p-value = 0.01, severe p-value = 0.02 and the critical p-value = 0.01). Gender was identified to be significantly different between the control and the critical group (p-value = 0.005) and between the mild and the critical group (p-value = 0.002) only. The number of “days since symptom onset” was found to be significantly higher in the mild group when compared to the severe (p-value = 5.41×10^{-5}) and the critical group (p-value = 1.60×10^{-5}). Ethnicity information was unavailable for the control group. *M* males, *F* females, *CI* confidence interval.

unknown and many who have, are unaware due to being asymptomatic. Nevertheless, according to the World Health Organization (WHO), it is estimated 80% of infections are asymptomatic or mild, 15% are severe infections requiring oxygen support, and 5% are critical and require ventilation¹.

SARS-CoV-2 shares an evolutionary relationship with SARS-CoV, the causative pathogen of the SARS outbreak in 2013². Currently, owing to the lack of reliable markers, it is challenging to monitor individuals that are progressing to severe COVID-19, which relies mainly on clinical manifestations³. Previous studies on highly pathogenic coronaviruses, e.g., SARS or Middle East Respiratory Syndrome (MERS), have highlighted an inflammatory cytokine storm and lymphocytopenia as common features relating to disease severity^{4–6}. It has also been suggested that the presenting cytokine storm is related to rapid disease progression or inadequate response to treatment⁶. Thus, it is vitally essential to tease out which peripheral markers are reliably related to disease severity to administer treatment at the earliest stage.

Furthermore, the central nervous system (CNS) involvement has also been reported in hospitalized patients infected with SARS⁷. It is plausible that patients suffering from COVID-19 might also exhibit CNS damage. Our previous results show neurochemical evidence of neuronal injury and glial response in patients with severe and critical COVID-19 and are associated with disease severity⁸. Therefore, it is plausible that other CNS injury markers can be detected in the blood to support the possible impact of COVID-19 on the CNS. This study examines an extensive collection of inflammatory, immune response, cardiovascular and neurological markers in the blood, from mild, severe and critical COVID-19 positive patients to further understand the role of CNS in the disease, identify other blood markers associated with disease severity and identify potential therapeutic targets.

Results

Cohort demographics. An overview of the patient demographics is provided in Table 1 and significance testing of age, gender and “days since symptom onset” between groups is provided in Supplementary Table S1. The mild group was identified to contain patients significantly younger when compared to the remaining groups (control p-value = 0.01, severe p-value = 0.02 and critical p-value = 0.01). No significant difference in age was identified between the remainder of the groups, including between the control and the case groups (p-value = 0.15).

Gender was identified to be significantly different between the control and the critical group (p-value = 0.005) and between the mild and the critical group (p-value = 0.002) only. There was no significant difference in gender distribution between the control group and the case group (p-value = 0.16). The number “days since symptom onset” was found to be significantly higher in the mild group when compared to the severe (p-value 5.41×10^{-5}) and critical group (p-value = 1.60×10^{-5}), with no difference between the severe and the critical group (p-value = 0.434).

As mentioned in the cohort originating publication⁸, four patients presented with symptoms of confusion before admission to the ICU, and one patient had a single episode of seizure before admission to the ICU, with no signs of epileptic activity when EEG was performed a day later. CT scans were normal in 2 of the 3 cases scanned; the third had signs of small vessel disease. MRI scans were not performed due to restrictions imposed by the protection of hospital workers and other patients in place at the time. No additional neurologic abnormalities were documented.

Summary of OLINK data processing. The OLINK cardiovascular, immune, inflammation and neurology panels consisted of 92 proteins each. Protein profiling the 87 samples resulted in the measurement of 368 proteins per sample. One sample, belonging to the mild symptom group, failed in two assays (immune and neurology) and was excluded from the failed panels, but were retained for analysis involving the cardiovascular and inflammation panels. Thirteen proteins had missing Normalized Protein eXpression (NPX) values or had NPX values below the protein-specific limit of detection (LOD) in more than 50% of samples in all four disease

Analysis	Total DE	Up	Down	Most sig	logFC	FDR p-value
Control vs mild	147	42	105	NF2	- 4.62E + 00	2.57E-42
Control vs severe	65	8	57	MANF	- 4.71E + 00	5.12E-16
Control vs critical	187	82	105	NF2	- 4.44E + 00	2.40E-36
Mild vs severe	146	97	49	EN-RAGE	2.16E + 00	5.33E-09
Mild vs critical	197	158	39	CTSL1	2.16E + 00	3.76E-13
Severe vs critical	64	63	1	IL-18R1	5.79E-01	0.00139
Control vs case	269	120	149	NF2	- 4.62E + 00	1.31E-86
Mild group longitudinal	13	10	3	BOC	1.57E + 00	0.001243
Critical group longitudinal	6	3	3	DECR1	0.6896971	0.023089

Table 2. Summary of differential expression analysis. The table summarises the number of significantly (FDR p -value ≤ 0.05) differentially expressed proteins identified in all nine analyses. “Total DE” is the total number of differentially expressed proteins identified in the analysis, “Up” represents the number of proteins that are up-regulated and “Down” represents the number of down-regulated proteins. “Most sig” represents the most significant perturbed protein in the analysis and is provided as the protein symbol. *logFC* log fold change, *FDR* false discovery rate.

groups. These 13 proteins were removed, leaving 355 proteins, of which 344 proteins were unique due to protein duplication across the four panels.

Summary of differential expression analysis. A summary of the number of proteins being significantly perturbed in each analysis is provided in Table 2. The complete differential expression (DE) results for all analyses are available in Supplementary Table S2 and can also be explored using the data explorer application developed in this study (<https://phidata-lab-shiny.rosalind.kcl.ac.uk/COVID19/>).

Proteins differentially expressed in COVID-19. The “control vs case” analysis identified 269 proteins as significantly differentially expressed in COVID-19 patients, with the NF2 protein identified as the most perturbed (FDR p -value = 1.31E-86, $\logFC = -4.62E+00$). The NF2 protein was also identified as the most perturbed protein in multiple analyses (“control vs mild”, “control vs critical” and “control vs case”) and was the single most perturbed protein from all nine analyses. The expression pattern of the NF2 protein per sample is illustrated in Fig. 1.

The 269 proteins perturbed in COVID-19 mapped to 265 unique Entrez gene IDs in the ConsensusPathDB database, which identified 285 significantly enriched biological pathways. The most significantly enriched pathway was the “Cytokine-cytokine receptor interaction” (FDR adjusted p -value = 1.72 e-41). The pathway, along with the enriched proteins, is illustrated in Fig. 2. The complete pathway enrichment analysis results are provided in Supplementary Table S3.

Proteins associated with COVID-19 severity. Eleven proteins were significantly differentially expressed between the control, the mild, the severe and the critical symptom groups. Eight of these proteins were consistently perturbed in the same direction as the infection symptoms increased (control \rightarrow mild \rightarrow severe \rightarrow critical). Due to the presence of the duplicate proteins across the different panels, these eight proteins correspond to six unique proteins. The six proteins are IL6, CKAP4, Gal-9, IL-1ra, LILRB4 and PD-L1 are associated with COVID-19 severity. None of these proteins originates from the neurology panel and their statistical significance across the nine analyses is included in Supplementary Table S2.

The IL6 protein was discovered to have been repeated on three different panels, and its expression pattern and significance level were found to be very similar between the disease groups. When comparing the overall magnitude of expression change from the control group to the critical group (“control vs critical” analysis), IL6 protein on the immune, inflammation and the cardiovascular panel has a \logFC of 4.23, 4.49, 4.79, and an FDR p -value of 4.01e-10, 2.74e-10 and 2.61e-10, respectively. The consistency in results provides validity across the different panels and demonstrates that IL6 protein is reliably detected as significantly increased in COVID-19.

The expression of all six proteins was observed to significantly increase from the control group to the mild group, further increasing in the severe group, and increasing even further in the critical group. Their expression patterns for these proteins is shown in Fig. 3. The proteins can be ranked by their magnitude of fold change from the control group to the critical group (as determined from the “control vs critical” analysis) as follows; IL6 ($\logFC = 4.79$), IL-1ra ($\logFC = 2.43$), CKAP4 ($\logFC = 1.98$), LILRB4 ($\logFC = 1.79$), Gal-9 ($\logFC = 1.60$), PD-L1 ($\logFC = 1.25$). Pathway analysis identified all six proteins as significantly enriched in the “Immune system” (FDR adjusted p -value = 1e-4).

Longitudinal data analysis. The mild symptom group (six patients) and the critical symptom group (five patients) consisted of patients with blood samples taken at two different time points after the onset of disease symptoms. A summary of the demographics of these longitudinal samples is provided in Table 3: Demographics of the longitudinal samples. A summary of the number of proteins perturbed in each longitudinal analysis

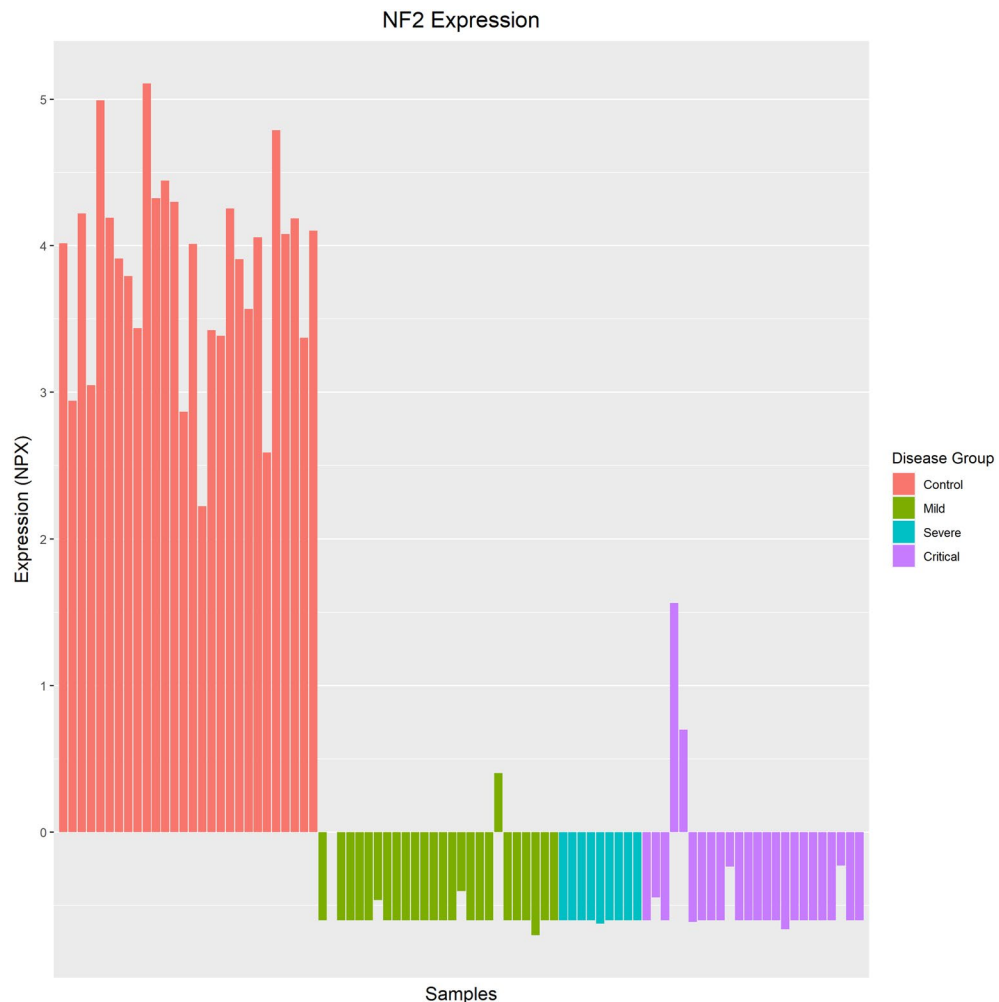


Figure 1. NF2 Expression across all disease groups. This protein was identified as the most perturbed protein in the “control vs case” analysis and is significantly down-regulated in all COVID-19 patients, regardless of infection severity.

is included in Table 2, and the complete DE analysis results from these analyses are included in Supplementary Table S2. No protein was identified to be significantly perturbed in both the “mild group longitudinal” and the “critical group longitudinal” analyses.

OLINK neuronal proteins correlated with markers of neural injury and astrogliosis. The three proteins measured on the Simoa platform were absent from the OLINK platform. The three proteins tau, NfL and GFAP were significantly ($p\text{-value} \leq 0.05$) correlated with 97, 233, 165 proteins from the OLINK platform, respectively (Supplementary Table S4), of which 20, 61 and 41 proteins belong to the Neurology panel, respectively.

NfL was identified to be most correlated with EDA2R ($r=0.66$, $p\text{-value}=4.01e-12$), which is not significantly perturbed in any of the nine DE analysis performed in this study (see Fig. 4a). Tau and GFAP were identified to be both most correlated with SCARB2 (tau: $r=0.39$ and $p\text{-value}=1.74e-4$, GFAP: $r=0.46$, $p\text{-value}=6.47e-6$). When compared to controls, SCARB2 is significantly up-regulated in the mild ($\log\text{FC}=0.59$, $p\text{-value}=4.7e-3$) and critical groups ($\log\text{FC}=0.72$, $p\text{-value}=0.01$) but not in the severe group ($\log\text{FC}=1.02$, $p\text{-value}=0.12$). Although SCARB2 is not statistically associated with disease severity, this protein’s expression is noted to increase inline with disease severity (Fig. 4b).

In addition, biomarkers of disease severity (Fig. 3) were vastly more associated with NfL (CKAP4, $r=0.642$; PD-L1, $r=0.623$; IL6, $r=0.528$; Gal-9, LILRB4, $r=0.508$; $r=0.486$; IL-1ra, $r=0.435$) than GFAP or tau indicating that disease severity includes increased CNS damage as marked by NfL and not GFAP and tau.

Differentially expressed OLINK neuronal proteins. From the “control vs case” analysis, 71 proteins on the neurology panel were identified to be perturbed in COVID-19, with LAT (see Fig. 4c) identified to be the most perturbed in this analysis ($\log\text{FC}=-3.9$, $p\text{-value}=4.46e-22$). LAT is also significantly differentially

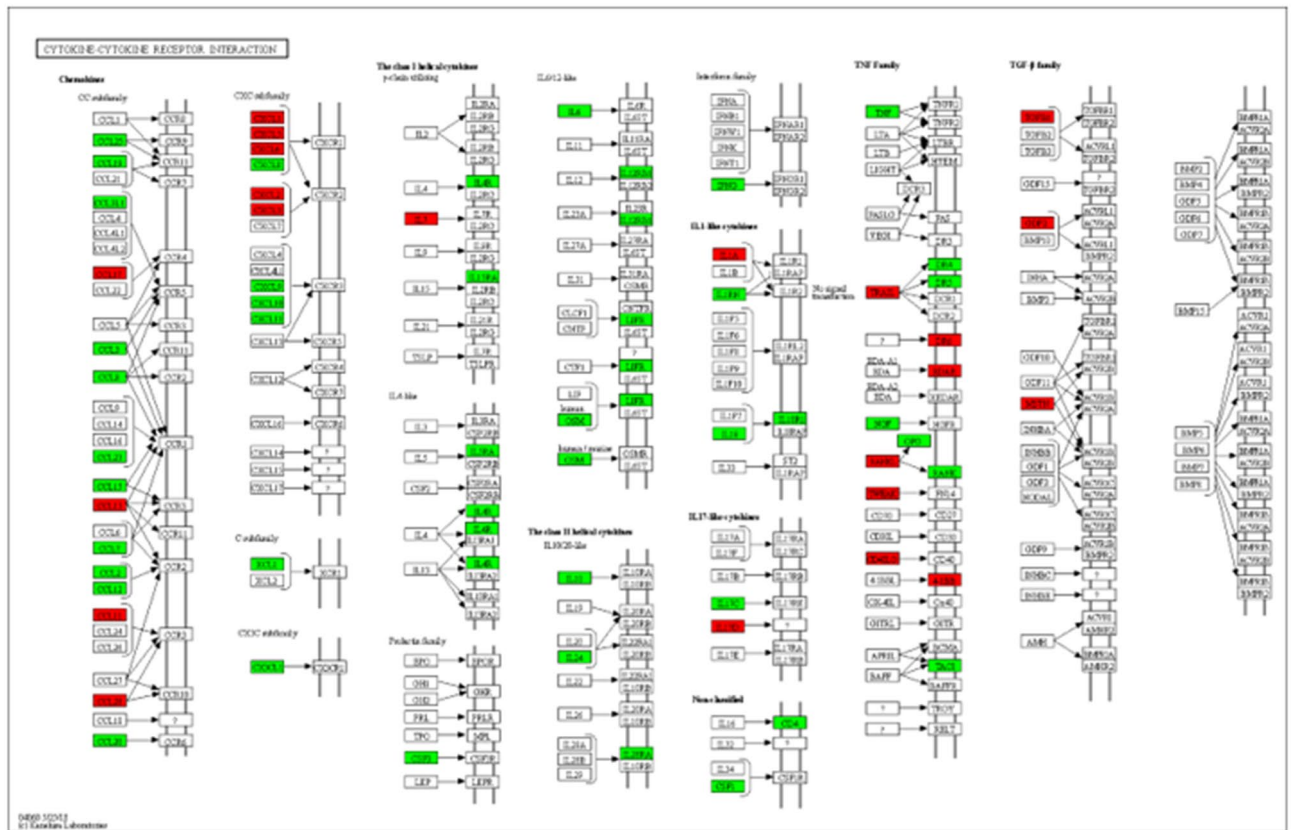


Figure 2. Cytokine-cytokine receptor interaction. The 269 proteins significantly differentially expressed in COVID-19 patients were significantly enriched in the “cytokine-cytokine receptor interaction” pathway, which is illustrated above. The proteins highlighted in green are up-regulated, and the proteins highlighted in red were down-regulated in COVID-19. The figure was generated using the “KEGG mapper—Search&Color Pathway” (Kanehisa & Sato, 2020).

expressed between the mild and severe group ($\log_{2}FC = 1.94$, $p\text{-value} = 7.7e-3$), but not between the severe and critical group ($\log_{2}FC = -0.08$, $p\text{-value} = 0.89$). This suggests LAT is down-regulated in COVID-19 but is not associated with disease severity. From all analyses involving the neuronal panel, MANF (see Fig. 4d) was the most perturbed protein, and is significantly down-regulated in the mild ($\log_{2}FC = -2.85$, $p\text{-value} = 2.53e-28$), severe ($\log_{2}FC = -4.71$, $p\text{-value} = 8.58e-17$) and critical symptom groups ($\log_{2}FC = -1.81$, $p\text{-value} = 5.98e-17$). However, MANF is not significantly perturbed between the mild and severe group ($\log_{2}FC = 0.8$, $p\text{-value} = 0.3$), or the severe and critical group ($\log_{2}FC = 0.3$, $p\text{-value} = 0.61$). This suggests MANF is down-regulated in COVID-19 and is not associated with disease severity.

Discussion

This study explored the changes in protein expression between four disease groups where patients were either controls (tested negative for COVID-19 using PCR) or had mild, severe or critical symptoms of COVID-19. In addition, longitudinal data were available where patients had blood drawn at two different time points after the onset of disease symptoms. In total nine DE analysis were performed (“control vs mild”, “control vs severe”, “control vs critical”, “mild vs severe”, “mild vs critical”, “severe vs critical”, “control vs case”, “mild group longitudinal” and “critical group longitudinal”) to identify proteins significantly perturbed between various disease groups and within groups across time. Furthermore, the results have been made readily available in an interactive web-based R Shiny application (<https://phidata-lab-shiny.rosalind.kcl.ac.uk/COVID19/>), allowing researchers to swiftly visualize and further investigate the expression changes of specific proteins in COVID-19 patients.

Biomarkers for COVID-19 infection. The three different symptom severity groups (mild, severe and critical) were merged to create a “case” group and were compared to the control group to identify common differentially expressed proteins in COVID-19. A total of 269 proteins were identified as significantly differentially expressed, of which 120 are up-regulated, and 149 are down-regulated in COVID-19 cases. Notably, over 75% of proteins measured in this study are significantly perturbed in COVID-19 cases compared to COVID-19 negative controls of similar age groups. Neurofibromin 2 (NF2) was identified as the most perturbed protein in this study and regardless of disease severity, was significantly down-regulated in all COVID-19 patients (Fig. 1). This protein was not perturbed according to the longitudinal analysis, and therefore, it is not regarded as associated with the duration of disease. The NF2 protein, or better known as the Merlin protein (moesin-ezrin-radixin-like

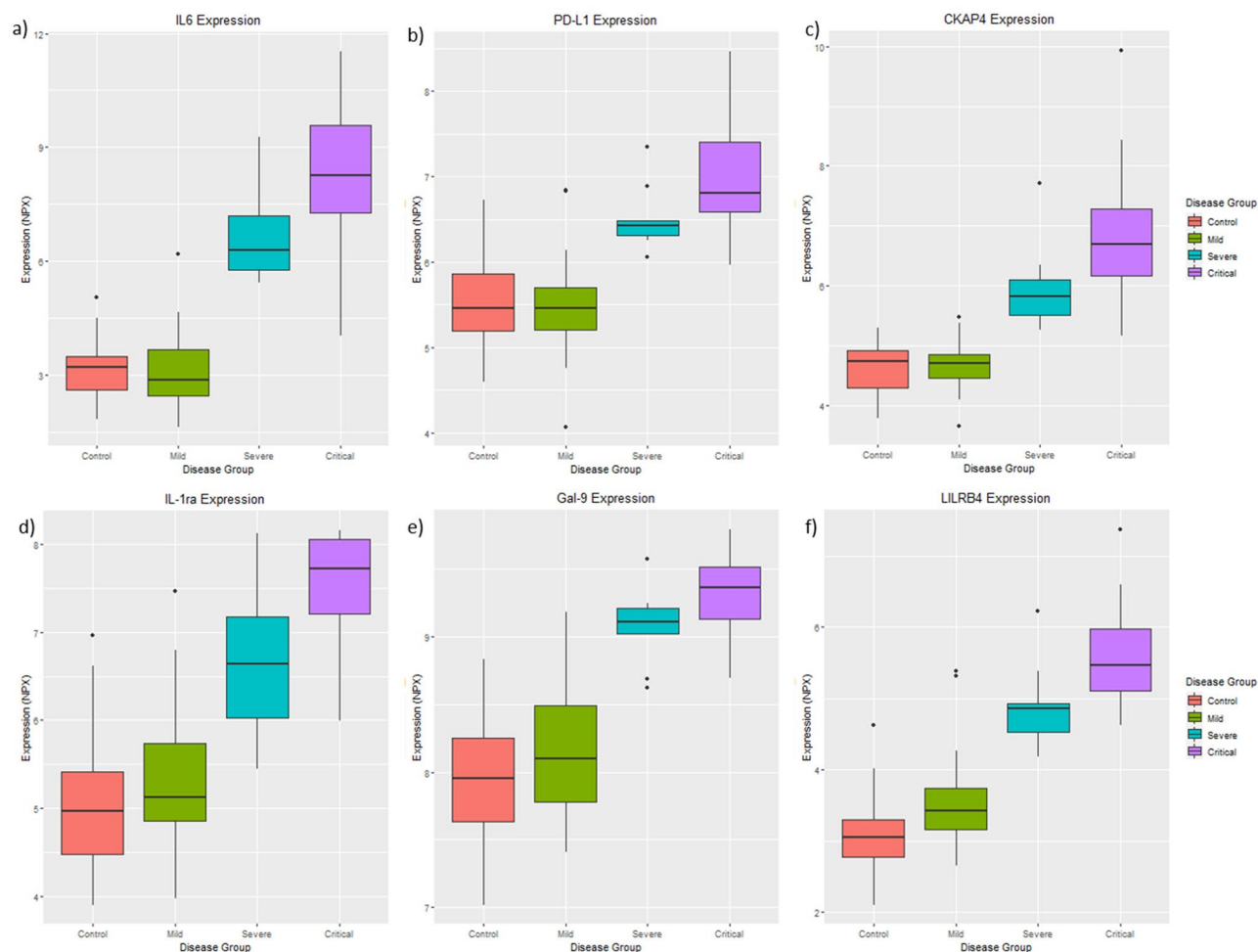


Figure 3. Six proteins (a) IL6, (b) PD-L1, (c) CKAP4, (d) IL-1ra (e) Gal-9 and (f) LILRB4 are consistently differentially expressed between the control, the mild, the severe and the critical symptom groups after controlling for age, gender, and “days since symptom onset”, suggesting these proteins may be associated with disease severity.

	Mild	Critical
No. patients	6	5
Gender (M/F)	3/3	5/0
Age (95% CI)	51.5 (36.8–66.2)	55.6 (45.8–73.4)
Ethnicity	Caucasian	4
	Persian	2
Baseline sample symptom duration (95% CI)	8.5 (3.4–13.6)	8.4 (5.5–11.3)
Repeat sample symptom duration (95% CI)	24.7 (22.2–27.1)	11.2 (7.33–15.1)
Average time between samples (95% CI)	16.2 (12.1–20.2)	2.8 (1.4–4.2)

Table 3. Demographics of the longitudinal samples. The table provides the demographics of the longitudinal samples in this study. *M* males, *F* females, *CI* confidence interval.

protein), functions as a tumour suppressor through impacting mechanisms related to proliferation, apoptosis, survival, motility, adhesion, and invasion^{9,10}. The Merlin protein also activates anti-mitogenic signalling at tight-junctions; hence, an inactivation of Merlin causes uncontrolled mitogenic signalling and tumorigenesis¹¹.

Biomarkers for infection severity. Six proteins (IL6, CKAP4, Gal-9, IL-1ra, LILRB4 and PD-L1) were consistently differentially expressed between the control, mild, severe and critical symptom groups. The expression of all six proteins increased as the severity of the symptoms increased, suggesting these proteins may be

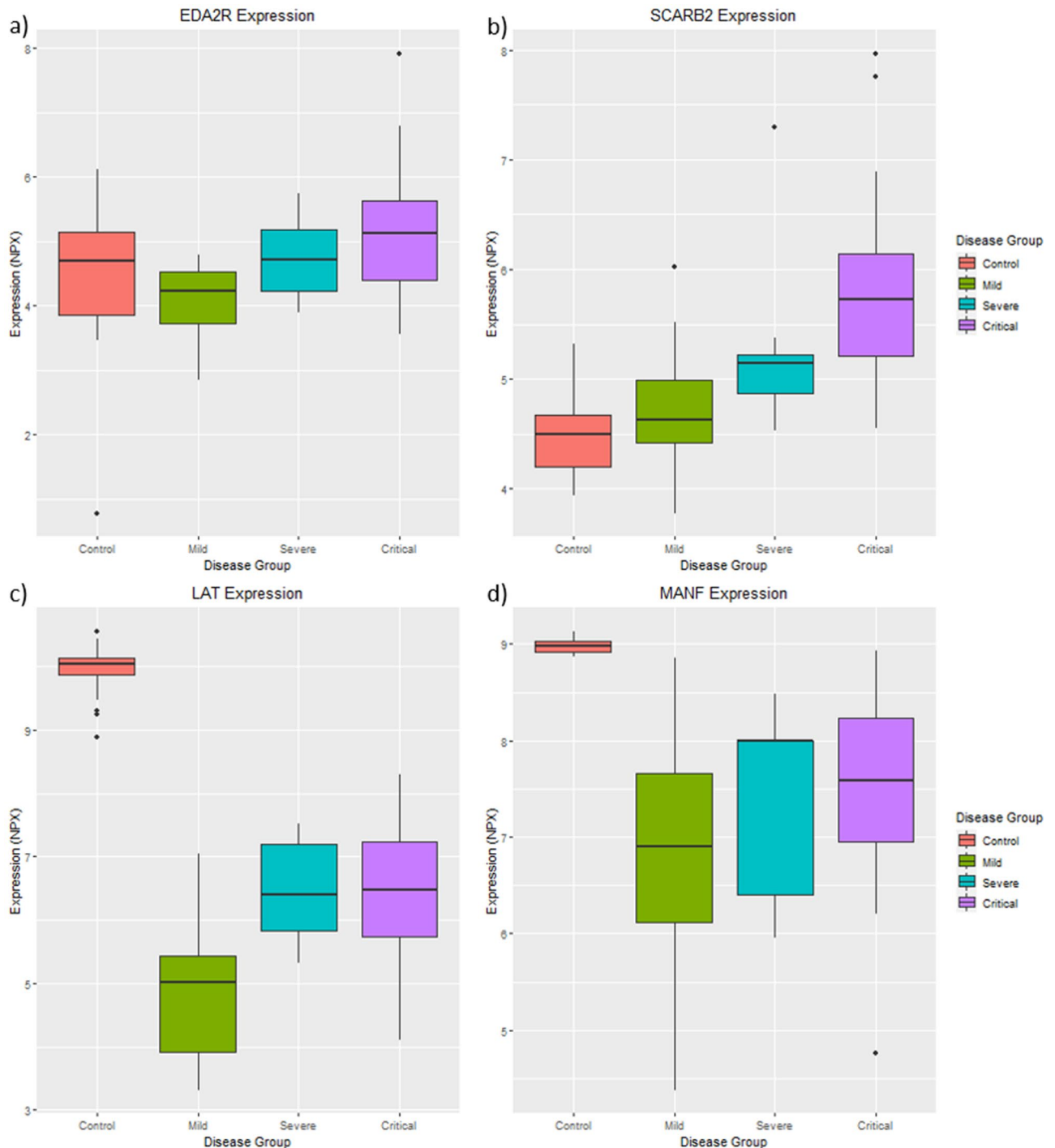


Figure 4. Expression patterns of the neuronal proteins perturbed in COVID-19. **(A)** EDA2R is correlated with the Simoa measured NfL. **(B)** SCARB2 is correlated with the Simoa measured tau and GFAP. **(C)** LAT is most differentially expressed protein in the “control vs case” analysis. **(D)** MANF is the most differentially expressed neuronal protein in this study from all analyses performed in this study. Statistical analysis was performed while controlling for age, gender, and “days since symptom onset”.

useful in monitoring the progression from mild to critical disease. Furthermore, these six proteins are not significantly perturbed within the longitudinal analysis, suggesting these proteins are not associated with infection duration.

Interleukin 6 (IL6) is an inflammatory cytokine that is an endogenous pyrogen of inducing fever in patients with autoimmune diseases or infections. IL6 is an acute phase inflammatory cytokine that has been suggested to reflect the inflammatory state of the lungs. Elevated IL6 levels have been discovered in ARDS and lung transplantation complications^{4,12}, and have already been shown to be elevated in COVID-19 patients^{12–14}. Furthermore, it

has also been suggested to be associated with increased COVID-19 mortality¹⁵. Our results add strong support that IL-6 expression increases with disease severity.

Furthermore, while not assessed in this study, IL6 shows some evidence of decreasing disease remission in COVID-19 patients^{12,16} and is, therefore, a viable target for treating the cytokine storm during disease progression. Tocilizumab is a monoclonal antibody targeted against IL6 and its receptor (IL6R) inhibitor, that is commonly used to treat inflammatory and autoimmune conditions¹⁷. It is currently being investigated in its effectiveness to treat COVID-19 patients^{18,19}.

In addition to IL6, we have highlighted novel protein markers associated with disease severity in COVID-19 patients and are all involved in the immune system, with many mediating cytokine productions, including IL6. Cytoskeleton Associated Protein 4 (CKAP4) is involved in the innate immune system and mediates the anchoring of endoplasmic reticulum to microtubules²⁰. A recent study identified that serum CKAP4 levels of lung cancer patients were significantly higher than those of healthy controls, suggesting CKAP4 as a potential early biomarker for lung cancer²¹. Galectin 9 (Gal-9) belongs to a family of beta-galactoside-binding proteins that are implicated in modulating cell–cell and cell–matrix interactions and has a diverse role in the innate and adaptive immune system [provided by RefSeq, Jul 2008]. Gal-9 has been demonstrated to activate ERL/2 phosphorylation, inducing chemokine and cytokine production, including IL-6^{22,23}. Serum Gal-9 concentrations have been observed to be significantly increased in patients with infections such as HIV²⁴, hepatitis C virus (HCV)²⁵ and malaria²⁶, suggesting increased gal-9 production is not specific to COVID-19 infection. Interleukin-1 receptor antagonist protein (IL-1ra) inhibits the activities of interleukin 1 alpha (IL1A) and interleukin 1 beta (IL1B) and modulates a variety of interleukin 1 related immune and inflammatory responses [provided by RefSeq, Jan 2016]. Essentially, IL-1ra is an inflammation-inhibitor protein that has also been identified to be significantly higher in COVID-19 patients with a severe symptom²⁷. Anakinra is a recombinant IL-1ra that has already been administered to COVID-19 patients with suggested improved clinical outcome in two small uncontrolled studies^{28,29}. Randomized controlled studies are still ongoing.

Leukocyte Immunoglobulin Like Receptor B4 (LILRB4) protein belongs to a family of cell surface receptors that have been suggested to down-regulate the immune response by Inhibiting monocyte activation and inhibiting the production of a critical pro-inflammatory cytokine (TNF α)³⁰. Increased expression of LILRB4 was associated with increased disease severity in this study, suggesting a possible decrease in monocyte activation leading to an immune-suppressive microenvironment. LILRB4 represents a compelling target to investigate COVID-19 treatment. Programmed cell death 1 ligand 1 (PD-L1) is a type 1 transmembrane protein that has immunoglobulin domains which bind to receptors commonly found on T-cells to inhibit T-cell activation and cytokine production. During infection or inflammation of healthy tissue, this interaction is essential for preventing autoimmunity by maintaining homeostasis of the immune response [provided by RefSeq, Sep 2015]. PD-L1 is found in higher concentrations on some types of cancer cells than healthy cells, which, when bound to PD-1 on T-cells, prevents the T cell from killing the PD-L1 containing cancer cell. To address this, immune checkpoint inhibitors (ICIs) are commonly used in various cancers to block PD-L1 on the cancer cell binding to the T cell, reinvigorating antitumor immune responses³¹. However, this immunodeficiency in cancer patients may be the primary cause of why they represent a vulnerable population in the COVID-19 pandemic³². As these novel proteins are not associated with the duration of infection in neither the mild nor the severe group, they may hold potential as biomarkers for disease severity. However, as members of the immune system, it is unknown if these proteins are markers of general infection rather than COVID-19, thus; they require further investigation for disease specificity, but together, may still be valuable as an additional biomarker for COVID-19 severity after disease confirmation.

Biomarkers for infection duration. In this study, six patients from the mild group and five patients from the critical group were sampled at two different time-points to identify biomarkers for disease duration. This identified thirteen (BOC, KYNU, SPRY2, KIM1, SCE, MANF, SLAMF1, CD84, SCE, PADI2, PAPP, CLEC4A, TANK) and six (DECRI, TPSAB1, TF, GDF-8, GZMA, BCAN) proteins in the mild and the critical group respectively, where proteins expression significantly changed from the baseline sample to the first repeat. No protein was discovered to be significantly perturbed in both groups over time; therefore, these biomarkers may be specific within their respective symptom severity groups. However, it is essential to note that sampling time between the baseline and repeat sample differed between the two severity groups, with the mild group averaging 16.2 days and the critical group patients averaging only 2.8 days. The smaller number of days between repeat sampling in the critical symptom group may have been inadequate to measure significant changes in protein expression that reflect infection duration.

CNS injury biomarkers. We have previously shown neurochemical evidence of neuronal injury and glial response in patients with severe and critical COVID-19. The results of this study indicate that astrocytic activation and/or injury (GFAP) may be a common feature in mild and severe stages of COVID-19, while neuronal injury (NfL) occurs later in the disease process and mainly in patients with critical symptoms⁸. In this current study, we expanded our proteomic profiling of CNS proteins, on the same patients, by using the OLINK neurology panel. Correlation analysis identified SCARB2 is most correlated with tau and GFAP, and EDA2R is most correlated with NfL, suggesting that these novel proteins are associated with COVID-19-related early or later CNS injury, respectively. Furthermore, MANF and LAT were significantly down-regulated in COVID-19 cases compared to controls, with expression patterns suggesting it is not associated with disease severity. Although MANF and LAT were present on the OLINK neuronal panel, both proteins are not specific to the brain. Mesencephalic Astrocyte Derived Neurotrophic Factor (MANF) is an endoplasmic reticulum (ER) stress protein and has been suggested to have neuroprotective effects against cerebral ischemia³³. Linker For Activation Of T Cells

(LAT) protein is primarily expressed in T-cells and is required for T-cell antigen receptor (TCR) and pre-TCR-mediated signalling. It is important to note that our putative biomarkers of disease progression in COVID-19 (IL6, CKAP4, Gal-9, IL-1ra, LILRB4 and PD-L1) were associated to NfL but to a much lesser extent to GFAP and tau.

The CNS involvement in COVID-19 is not known, and direct invasion of the virus may be unlikely. Support for a hypothesis of CNS infection through a nasopharyngeal route is provided by clinical observations of frequent and persistent anosmia dysgeusia¹⁶. Neurological symptoms are reported in severe cases, supporting the concept that CNS symptoms might be secondary to severe respiratory failure³⁴. CNS hypoxia from respiratory failure caused by COVID-19, thrombotic microangiopathy, or an indirect effect of extensive cytokine activation that is commonly found in severe COVID-19 is more probable explanations of these increases (Kanberg et al., 2020). Thus, this explorative study of the neurology-associated proteins in the blood may add further insight into hypoxia biomarkers in a broader sense, e.g., to monitor cardiac arrest. Furthermore, the evidence that specific CNS proteins are detectable in blood, as shown by this study, may open up avenues of investigation in CNS injury for neurodegenerative disorders, multiple sclerosis or HIV, where GFAP and NfL are regularly investigated.

Limitations. Due to blood samples being taken as part of a routine hospital procedure during an unprecedented time, this study was restricted to a small cohort, certain aspects of this study design were uncontrollable and valuable information was unobtainable. The number of “days since symptom onset” was significantly higher in the mild symptom group than the severe and the critical symptom group. Essentially, following the onset of symptoms, samples were drawn from the mild symptom group at a much later date compared to the severe and the critical symptom group. As a result, expression changes involving the mild group (“mild vs severe” and “mild vs critical”) may reflect the duration a patient has been infected with COVID-19 rather than being a reflection of symptom severity. However, as the longitudinal analysis in this study measures protein expression changes during infection, these results were used to differentiate between expression changes likely due to disease severity and duration of infection.

Furthermore, information on comorbidities, medical history and medications are unknown. Hospitalized patients with COVID-19 are known to be more likely to have an underlying health disorder such as hypertension, obesity and diabetes³⁵, and it is unknown if this cohort has the same characteristics. Moreover, it is unknown if any medication, in addition to oxygen supplementation, was administered to COVID-19 patients, therefore; protein expression changes in this study may reflect a combination of COVID-19, comorbidity and medication.

Conclusion

This extensive proteomic analysis unbiasedly identified IL6 and five novel proteins (CKAP4, Gal-9, IL-1ra, LILRB4 and PD-L1) to be associated with disease severity in COVID-19 cases. These proteins' expression significantly increased as the disease symptom severity deteriorated and highlight a shared mechanism of cytokine-mediated lung injury cause by viral injection. These proteins warrant further investigation but could provide potential as early biomarkers for disease severity and may serve as potential therapeutic targets, or as biomarkers to monitor the effect of treatments to modulate the immune system and/or suppress the infection. Overall, this study's results further increase the understanding of COVID-19, which includes CNS involvement, and have been made widely available to researchers as an interactive web-based tool accessible at <https://phidatalab-shiny.rosalind.kcl.ac.uk/COVID19/>.

Materials and methods

Cohort. The patients in this study originates from a previous published study, with this study including more patients and generating additional data⁸. Fifty-nine patients with confirmed COVID-19 and 28 healthy, age-matched controls were included. Samples were collected at diagnosis and repeated when possible. Patients were divided into three groups related to systemic disease severity: 26 patients with mild (i.e., not requiring hospitalization) 9 with severe (hospitalized and requiring oxygen supplementation), and 24 with critical disease (admitted to intensive care unit [ICU] and placed on mechanical ventilation [n = 23] or not considered a candidate for ICU treatment and with fatal outcome [n = 1]). The controls were initially recruited as cognitively unimpaired controls for an observational dementia study, and they were all neurologically and psychiatrically normal with any magnetic resonance imaging (MRI) abnormalities set as an exclusion criteria. Follow-up samples on patients with critical COVID-19 were collected when they were still in ICU.

COVID-19 confirmation. The diagnosis was confirmed using real-time polymerase chain reaction (rtPCR) analysis of nasal and throat swab specimens. Nucleic acid was extracted from clinical samples in a MagNA Pure 96 instrument using the Total Nucleic Acid isolation kit (Roche). rtPCR targeting the RdRP region was performed in a QuantStudio 6 instrument (Applied Biosystems, Foster City, CA) using the probe described by Corman et al. and the primers RdRP_Fi, GTCATGTGTGGCGGTTCACT and RdRP_Ri, CAACACTATTAG CATAAGCAGTTGT³⁶.

Plasma proteomics. Plasma glial fibrillary acidic protein (GFAP), neurofilament light (NfL), and tau were measured and reported in our recent COVID-19 related study⁸, which discovered neurochemical evidence to support the possible impact of COVID-19 on the CNS. The Simoa protein measurements are detailed in⁸. Plasma from the same participants was also used for the OLINK protein profiling for this study. Protein concentrations were measured on the Olink Multiplex platform (Olink Proteomics AB, Uppsala, Sweden) using the cardiovascular II (v.5006), immune response (v.3203), inflammation (v.3022) and neurology (v.8012) 96-plex panels. The OLINK immunoassays are based on the Proximity Extension Assay (PEA) technology³⁷, which uses

a pair of oligonucleotide-labelled antibodies to bind to their respective target protein. When the two antibodies are in close proximity, a new polymerase chain reaction target sequence is formed, which is then detected and quantified by quantitative real-time PCR.

Statistical analysis. The Olink-generated data was preprocessed and quality controlled using the platform-specific “Olink NPX manager” software, which background corrects, log₂ transforms and normalizes all samples to an arbitrary NPX scale. The NPX is a relative quantification unit where a difference of 1 NPX equates to a doubling of protein concentration.

Additional data processing was performed in RStudio (version 1.2.1335) using R (version 3.6.0). First, samples with a failure rate of more than 50% across all proteins were removed. Next, proteins were removed if the protein failed to quantify in more than 50% of samples in each disease group, or if the protein NPX value fell below the protein-specific LOD value in more than 50% of samples in each disease group. The remaining NPX values below the LOD were substituted by the proteins LOD/ $\sqrt{2}$.

The demographic variables available were ethnicity, age, gender and “days since symptom onset”. The “days since symptom onset” variable represents the number of days that a blood sample was taken after the first self-reported symptom date. The Welch Two Sample t-test and the Fisher’s exact test was performed where appropriate to identify any significant differences in age, gender and “days since symptom onset” between groups.

DE analysis was performed using the R package “limma” (version 3.42.2), which has been shown to be very powerful and stable at detecting significant changes in protein abundance³⁸. Multiple linear models using robust regression were fitted to each protein using gender, age and “days since symptom onset” as covariates where possible. A protein was determined to be significantly differentially expressed if the false discovery rate (FDR) adjusted p-value was ≤ 0.05 . The following self-explanatory comparisons were made: 1) “control vs mild”, 2) “control vs severe”, 3) “control vs critical”, 4) “mild vs severe”, 5) “mild vs critical”, 6) “severe vs critical”, 7) “control vs case” (where patients with mild, severe and critical symptoms are merged and treated as the cases).

Some patients within the mild and the severe symptom group had protein concentrations measured at two different time points; therefore, two additional DE analysis was performed independently in the mild and the severe symptom groups and are referred to as (1) “mild group longitudinal” and (2) “critical group longitudinal” analysis, respectively. The two longitudinal analyses were performed in their respective disease groups using a paired t-test approach in limma.

Pathway enrichment analysis was performed using an Over-Representation Analysis (ORA) implemented through the ConsensusPathDB (<http://cpdb.molgen.mpg.de>) web-based platform (version 34)³⁹. Significant results were then explored using the “KEGG mapper—Search&Color Pathway” to map and visualize proteins in a specific biological pathway⁴⁰.

This study further explores CNS injury-related biomarkers in COVID-19 by correlating the Simoa measured proteins with the OLINK measured proteins using Pearson’s correlation.

Ethics statement. This study has been approved by the Swedish Ethical Review Authority (2020–01771) and all experiments were performed in accordance with relevant guidelines and regulations. All participants provided written informed consent, in those with severe COVID-19, this was obtained before they were placed on mechanical ventilation and were deemed fully capable of understanding the nature of the study and their part in.

Data availability

The proteomic data is available in the BioStudies database (<http://www.ebi.ac.uk/biostudies>) under accession number S-BSST416. Additionally, an R shiny application was written in R using the “shiny” framework (version 1.4.0.2) to allow the quick and efficient visualization of the expression of specific proteins across the control, mild, severe and severe symptom groups. The application is hosted on the research computing facility at King’s College London (Rosalind), and allows researchers to quickly visualize and investigate the results across all nine DE analyses performed in this study. The application can be accessed at <https://phidatalab-shiny.rosalind.kcl.ac.uk/COVID19/>. All data analysis scripts used in this study have been deposited in zenodo under the <https://doi.org/10.5281/zenodo.3895886>.

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Author contributions

A.Y., L.M.A. and M.G. were involved in patient recruitment. HP analyzed the data with involvement from N.J.A., R.J.B.D., K.B., M.G. and H.Z. H.P. developed the interactive web-based tool to host the results from this study. H.P. and N.J.A. drafted the manuscript with all authors contributing to the result interpretation and critical review of the manuscript.

Competing interests

KB has served as a consultant, at advisory boards, or at data monitoring committees for Abcam, Axon, Biogen, Julius Clinical, Lilly, MagQu, Novartis, Roche Diagnostics, and Siemens Healthineers, and is a co-founder of Brain Biomarker Solutions in Gothenburg AB (BBS), which is a part of the GU Ventures Incubator Program (outside submitted work). HZ has served at scientific advisory boards for Denali, Roche Diagnostics, Wave, Samumed, Siemens Healthineers, Pinteon Therapeutics and CogRx, has given lectures in symposia sponsored by Fujirebio, Alzecure and Biogen, and is a co-founder of Brain Biomarker Solutions in Gothenburg AB (BBS), which is a part of the GU Ventures Incubator Program (outside submitted work).

Additional information

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Correspondence and requests for materials should be addressed to H.P.

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