Recapitulation of pathophysiological features of AD in SARS-CoV-2 infected subjects

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Abstract

Infection with the etiological agent of COVID-19, SARS-CoV-2, appears capable of impacting cognition, which some patients with Post-acute Sequelae of SARS-CoV-2 (PASC). To evaluate neuro-pathophysiological consequences of SARS-CoV-2 infection, we examine transcriptional and cellular signatures in the Broadman area 9 (BA9) of the frontal cortex and the hippocampal formation (HF) in SARS-CoV-2, Alzheimer’s disease (AD) and SARS-CoV-2 infected AD individuals, compared to age- and gender-matched neurological cases. Here we show similar alterations of neuroinflammation and blood-brain barrier integrity in SARS-CoV-2, AD, and SARS-CoV-2 infected AD individuals. Distribution of microglial changes reflected by the increase of Iba-1 reveal nodular morphological alterations in SARS-CoV-2 infected AD individuals. Similarly, HIF-1α is significantly upregulated in the context of SARS-CoV-2 infection in the same brain regions regardless of AD status. The finding may help to inform decision-making regarding therapeutic treatments in patients with neuro-PASC, especially those at increased risk of developing AD.

Teaser

SARS-CoV-2 and Alzheimer’s disease share similar neuroinflammatory processes, which may help explain neuro-PASC.
Introduction

The consequences of SARS-CoV-2 infection have been well-studied in the respiratory system; however, much less information is available regarding the neurological consequences of infection (1). Previous evidence tentatively suggests that SARS-CoV-2 may be neuroinvasive, leading to a vast array of neurological symptoms, including anosmia, ageusia, cognitive functions, and cerebrovascular disorders (2, 3). Neurological complications of SARS-CoV-2 infection manifest with increased severity related to age and shared medical history of metabolic disorders and other vascular risk factors (4–6). Some strains within the coronavirus family are implicated in neuronal degeneration, such as HCoV-OC43, which can lead to glutamate excitotoxicity and neuronal degradation in mice through cytokine production (7, 8). Because SARS-CoV-2 is a new addition to the coronavirus family, little information exists regarding potential long-term neurological complications that may impact COVID-19 survivors. This may be especially concerning for aging individuals with increased vulnerability to developing age-related neurodegenerative disease.

While it is currently unknown if neurological manifestations of SARS-CoV-2 infection arise solely from systemic inflammation in the periphery or brain infiltration, it is well established that secretion of cytokines and chemokines from the periphery allows the recruitment of leukocytes and other cells to specific tissues (9, 10). This type of immune response is speculated to accelerate blood-brain barrier (BBB) disruption and may potentially damage cells within the central nervous system (CNS) (11). Such effects are similar to neuropathology seen in Alzheimer’s disease (AD) and other neurodegenerative disorders, where leukocyte infiltration,
BBB dysregulation, and microglial activation are observed. Thus, it is possible that SARS-CoV-2 may accelerate the onset and severity of cognitive decline or AD in susceptible individuals (12, 13). Understanding the impact of SARS-CoV-2 infection on the CNS and subsequent mechanisms associated with cognition is particularly pressing, given the number of individuals presenting with neurological symptoms of post-acute sequelae of SARS-CoV-2 (neuro-PASC), colloquially termed ‘long-COVID’ (14, 15).

Here, we investigate potential brain mechanisms associated with SARS-CoV-2 infection in SARS-CoV-2, AD and SARS-CoV-2 infected AD individuals by comparing transcriptional and cellular responses in the cortical Broadman area 9 (cortical BA9) of the frontal cortex and the hippocampal formation (HF); two brain regions deeply involved in cognitive and emotional functions, to age- and gender-matched neurological cases. Here we report evidence suggesting that SARS-CoV-2 infection promotes similar pathophysiological features found in AD cortical BA9 and HF regions and possibly exacerbate pre-existing AD pathophysiology.
Results

Demographics of human postmortem cases

We examined postmortem tissue samples of cortical BA9 of the frontal cortex and the HF collected by the Neuropathology Brain Bank and Research CoRE at Mount Sinai from 4 different groups: SARS-CoV-2, AD, and AD individuals which became infected by SARS-CoV-2, compared to age- and gender-matched neurological cases (Table 1). Each group comprises an equal number of age-matched male and female patients, ensuring equal distribution with an average age of 79.6 among groups. Medical records indicate that all AD individuals have postmortem ABC scores indicative of AD (Table 2), where A is a measure of amyloid beta deposition, B is a measure of neurofibrillary degeneration based on the Braak and Brook score, and C is scored based on neuritic plaques outlined by the Consortium to Establish a Registry for Alzheimer’s Disease diagnosis (CERAD) (16). SARS-CoV-2 infection was confirmed by diagnostic polymerase chain reaction (PCR). Within the SARS-CoV-2 and SARS-CoV-2 infected AD groups, all patients were symptomatic, admitted to the hospital, and received oxygen supplementation by ventilator or cannula with disease onset to death occurring on average 27 days after diagnosis for patients with SARS-CoV-2 only and 32 days after diagnosis in infected AD patients (Table 1 and 2). Blood specimens were collected and indicate several laboratory features of severe COVID-19 (Table 2), such as increased C-reactive protein (CRP) and interleukin-6 (IL-6) (17).
The transcriptional profiles of AD and SARS-CoV-2 infected postmortem cases from the cortical BA9 were assessed by pairwise comparison to determine gene expression compared to neurological control cases (Figure 1A-1C). Volcano plot distributions of gene transcripts of AD and SARS-CoV-2 cases compared to neurological controls (Figure 1A and 1B) both reveal 39,901 genes which were used to determine differentially expressed genes based on a nominal \( p \)-value less than 0.01 and an absolute log2 fold-change (log2 FC) greater than 1. We find in the AD group, 287 genes are upregulated, and 131 genes are downregulated (Figure 1D). In the SARS-CoV-2 group, 179 genes are upregulated, and 426 genes are downregulated (Figure 1D).

The Volcano plot distribution of gene transcripts of SARS-CoV-2 infected AD cases compared to neurological controls (Figure 1C) revealed from 38,021 genes, and show 813 genes upregulated, 611 genes downregulated using the same cutoff for significance (Figure 1D).

Transcriptional profiles of AD and SARS-CoV-2 cases from the HF were assessed (\( p<0.01 \) and log2 FC>1) to determine gene expression, as compared to neurological control cases (Figure 1E-1G), using 39,901 genes for AD and SARS-CoV-2 cases (Figure 1E, and 1F), and 35,527 genes for the SARS-CoV-2 infected AD cases (Figure 1G). We find in the AD group, 304 genes upregulated, and 214 genes downregulated (Figure 1H). In the SARS-CoV-2 group, 230 genes are upregulated, and 464 genes are downregulated (Figure 1H). Additionally, in the SARS-CoV-2 infected AD group 466 genes upregulated, 582 genes downregulated (Figure 1H).
Similarity of gene expression in cortical BA9 and HF regions

In order to determine the relative similarity of gene expression changes induced by disease states, a collection of genes known to be differentially expressed due to either SARS-CoV-2 infection or AD were examined across three individual differential comparisons (Figure 2A and 2B). While all three comparisons (AD versus Control, SARS-CoV-2 versus Control, and SARS-CoV-2 infected AD versus Control) showed generally the same trends, the AD and SARS-CoV-2 infected AD comparisons were strikingly similar, in both BA9 (Figure 2A) and HF (Figure 2B).

Rank-rank hypergeometric overlap (RRHO) analysis comparing SARS-CoV-2 and AD cases also reveals a positive correlation between the AD and SARS-CoV-2 groups in the cortical BA9 (Figure 2- figure supplement 1) as well as in the HF (Figure 2- figure supplement 2), although to a lesser extent than in the cortical BA9. We then filtered genes using Ingenuity Pathway Analysis software (IPA) to determine the number of shared differentially regulated genes among the AD, SARS-CoV-2 and SARS-CoV-2 infected AD groups within each tissue region ($p<0.05$ and absolute $z > 0.0001$). Comparing these shared genes within the cortical BA9, we find that AD, SARS-CoV-2 and SARS-CoV-2 infected AD individuals shared 410 upregulated genes, 269 downregulated genes from 679 expressed genes (Figure 2C and Source data 1). In the HF, we found the three groups share 211 upregulated genes, 148 downregulated genes from 359 differentially expressed genes (Figure 2D and Source data 2).
Pathway activation of SARS-CoV-2 and AD groups

IPA was employed to further characterize canonical pathways in AD, SARS-CoV-2 and SARS-CoV-2 infected AD groups compared to the neurological controls. Within the cortical BA9 region (Figure 3A), for example, several pathways, including neuroinflammation, TREM1, and cell senescence show increased activation in the AD, SARS-CoV-2 and SARS-CoV-2 infected AD groups. TREM1 signaling, the most highly activated pathway in this dataset (Figure 3A and 3B), is expressed primarily on myeloid cells, such as macrophages and microglia, and is involved in pro-inflammatory immune responses (18). Although age-associated cellular senescence (Figure 3A) is a natural process where telomeres shorten over time, this process also occurs during cellular stress due to inflammation, including that caused by viral infections, leading to a senescence-associated secretory phenotype such as metalloproteinases (MMP’s), and inflammatory cytokines (19, 20). Another pathway of interest, SNARE, shows decreased activation in AD and SARS-CoV-2 groups (Figure 3A), with the most significant reduction seen in the SARS-CoV-2 infected AD cases group. SNARE proteins play an essential role in neurotransmitter release, and altered function is implicated in the pathophysiology of neurodegenerative diseases such as AD, where SNARE proteins affect β-amyloid (Aβ) accumulation and cytoplasmic transport of neurofibrillary tangles (21).

Within the HF (Figure 4A), for example interleukin-8 (IL-8) another neuroinflammatory pathway, and ciliary neurotrophic factor (CNTF) signaling is upregulated in AD, SARS-CoV-2
and SARS-CoV-2 infected AD groups, while cAMP response element-binding protein (CREB) is also upregulated in SARS-CoV-2 groups but downregulated in AD-only (Figure 4A). IL-8 has several important roles, including endothelial cell migration and chemoattraction of neutrophils (22). One way IL-8 aids cell migration is by enhancing the expression of molecules such as MMP-2, MMP-9, involved in BBB integrity and induction of neuronal apoptosis, and VEGF-A, involved in vascular permeability and angiogenesis, thus having a potential effect on vascular damage (23–25). CNTF signaling aids in the prevention of neuronal degeneration after injury and is neuroprotective in diseases such as multiple sclerosis (MS) and amyotrophic lateral sclerosis (ALS) (26). Interestingly, CREB signaling modulates processes in consolidating memory and information processing and is inhibited in AD (27). Consistent with these findings, we also see a reduction in CREB signaling in the AD individuals; however, this effect is reversed in SARS-CoV-2 infected AD individuals in the HF. It is important to note that TREM1 signaling is upregulated in two of the three sample groups (SARS-CoV-2 and SARS-CoV-2 infected AD cases), but no predictions for this pathway occurs in the AD group. As such, this pathway is not included in the analysis within the HF (Figure 4A and B).

Additionally, upregulation of integrin signaling and nerve growth factor (NGF) signaling occur in both cortical BA9 and HF (Figure 3B and 4B) in all SARS-CoV-2, AD and SARS-CoV-2 infected AD groups. Neuronal function shown by the upregulation of NGF signaling is critical for the survival of neurons, and alterations of NGF signaling has been implicated in neurodegenerative disorders such as AD (28). Integrin signaling is used for a diverse array of functions within the CNS via cell-to-cell and cell-to-extracellular matrix interactions (29).
Regulation of inflammatory microglia responses in cortical BA9 and HF

Region-matched formalin-fixed paraffin-embedded (FFPE) tissue from the contralateral hemisphere of cortical BA9 and HF used for transcriptional studies underwent immunohistochemical (IHC) analysis for evidence of microglial activation (Iba-1), the presence of SARS-CoV-2 (SARS-CoV-2 nucleocapsid), and vascular integrity, assessed via hypoxia-inducible factor-1α subunit (HIF-1α).

Neuroinflammation was assessed with a pan-microglia marker, Iba-1, which is upregulated by microglia in the context of inflammation and reveals morphological alterations associated with their activation state (Figure 5). Compared to neurological controls, microglia in AD, SARS-CoV-2, and SARS-CoV-2 infected AD cases are more numerous and display an activated phenotype, with retracted, thickened processes and enlarged somas (Figure 5A-5D). Microglia also appear to accumulate around blood vessels in the context of AD and SARS-CoV-2 groups but not in unaffected neurological controls brain (Figure 5E-5H), suggesting factors at the level of the BBB may participate in microglial activation. Microglial nodules, which are commonly observed in neuroinflammatory disease, were seen in all conditions but appeared larger and more frequently in infection, as compared to AD-only and controls (Figure 5I-5L). When detectable, SARS-CoV-2 appears to be restricted to the endothelium (Figure 5G, 5H, 5K, and 5L) and did not colocalize with Iba-1 or glial fibrillary acidic protein (GFAP; data not shown), suggesting neither microglia nor astrocytes harbor productive virus. Nonbiased quantitation revealed an overall increase in the number of microglia in the HF and cortical BA9 of patients with AD,
SARS-CoV-2, and SARS-CoV-2 infected AD cases. A statistically significant increase in the number of microglia occurs in AD alone in the HF and SARS-CoV-2 infected AD cases in both the HF and cortical BA9 (**Figure 5M and 5O**). An increase in the frequency of nodular lesions was seen with SARS-CoV-2 infection, with and without AD, in both brain regions assessed. However, a statistically significant increase in nodular lesions are present in SARS-CoV-2 infected AD cases (**Figure 5P-5R**).

**Regulation of vascular damage using hypoxia inducible factor- 1α (HIF-1α) detection**

Tissues were further assessed for possible hypoxia by IHC using an antibody against the α subunit of HIF-1, which is stabilized under hypoxic conditions. Upregulation and stabilization of HIF-1α is most pronounced in the context of SARS-CoV-2 infection, regardless of AD status in cortical BA9 (**Figure 6A-6D**) and HF (**Figure 6E-6H**). Nonbiased area positivity quantitation revealed HIF-1α is only slightly elevated in cortical BA9 of AD patients, as compared to age-matched neurological controls (**Figure 6J**). In contrast, a greater increase occurs in the HF from SARS-CoV-2 individuals, with and without AD. SARS-CoV-2 only samples showed a significant increase in HIF-1α of cortical BA9 when compared to controls (**Figure 6I-6K**).
Discussion

The finding from our study suggests that SARS-CoV-2 and AD infected individuals share similar alterations of regulatory patterns of immune-inflammatory pathways and pathways involved with cognition as suggested by recent meta-analyses showing shared neuroinflammation and microvascular injury, in particular AD and SARS-CoV-2 infected individuals (30). Additionally, Zhou and colleagues found a significant overlap in cerebrospinal fluid (CSF) monocytes and markers in AD and COVID-19, which also occurs in our dataset (30). The similarities of transcriptional profiles and cellular pathophysiology in SARS-CoV-2 and SARS-CoV-2 infected individuals support the potential role of SARS-CoV-2 infection on the CNS, leading to neuro-PASC symptoms such as brain fog and memory loss.

Notably, our study identifies similar neuroinflammatory profiles in SARS-CoV-2, AD and AD SARS-CoV-2 infected individuals at the transcriptional and cellular levels in both the cortical BA9 and HF brain regions. This suggests that SARS-CoV-2 generates a similar neuroinflammatory environment in neurodegenerative disorders like AD. This was highlighted by the regulation of TREM1, neuroinflammation, and cellular senescence/inflammatory pathways present in all groups and further established by the widespread microglial activation seen in AD and SARS-CoV-2 infected AD cases, and in particular, the nodular formation seen in SARS-CoV-2 infected AD cases. Microglia nodule formation is present in some neurodegenerative diseases, such as MS, and viral infections, such as herpes simplex virus (HSV) and human immunodeficiency virus (HIV) (31, 32). A potentially compounding finding is the nodular formation that is initially characterized by abundant presence of activated microglia.
and innate immune factors leading to toll-like receptor (TLR) signaling and upregulations of inflammasome genes, leading to T cell stimulation and ultimately the destruction of neurons (31). This may be of particular concern in the SARS-CoV-2 infected AD cases, where we primarily observe increased nodular lesions. This suggests SARS-CoV-2 further promotes neuroinflammation in AD, which likely advances the progression and severity of CNS disease in these individuals. Interestingly, a large retrospective study of patients 65 years or older revealed that patients with SARS-CoV-2 were at an increased risk for a new AD diagnosis within a year of their SARS-CoV-2 diagnosis, with the most significant risk seen among those 85 years and older (33). This may underscore the critical role of preexisting inflammation in the brain, which is seen in the context of ‘normal’ aging, in promoting or advancing AD progression.

Our findings also support the notion that SARS-CoV-2 may cause cognitive deficits via regulation of pathways associated with cognition and neuroinflammation. Here, we show changes in the transcriptional regulation of SNARE and NGF pathways, suggesting impaired neuronal health and function, presumably negatively impacting cognitive function. We also observed potential damage to the vasculature via increased regulation of HIF-1α, integrin signaling, and IL-8 signaling. It is important to note a possibility of vascular damage due to ventilation prior to death in SARS-CoV-2 infected individuals; however, vascular damage by SARS-CoV-2, as assessed by HIF-1α, is observed in non-human primates that were euthanized at designated end-of-life time points precluding breathing intervention (34). These findings may point to a possible route for lymphocyte transmigration following chemoattractant gradients such as IL-8 and enhancement of MMPs and VEGF, which are aided by integrins such as cellular...
adhesion molecules such as ICAM-1 and are implicated with inflammatory signal transduction (35, 36). This process leads to T cell activation, which may be responsible for the observed inflammatory environment. Our transcriptional data showed an abundant upregulation of TLRs (TLRs 1, 5, 7, and 8) related to the inflammasome. Further analysis of this dataset also indicated the potential role for T cells through IL-7 pathway activation and CD86 upregulation; however, we did not confirm changes in the number of T cells, and no evidence of leukocyte infiltration into the CNS compartment was observed in any disease state in this investigation. Future studies will be required to determine the role of leukocytes in the observed pathophysiology.

Although neuroinflammation and vascular injury were prominent features of SARS-CoV-2 and SARS-CoV-2 infected AD cases brain pathology, the direct role of the virus is unclear. We did detect SARS-CoV-2 nucleocapsid in some SARS-CoV-2 brain tissues that appear to be restricted to the vasculature. This finding is supported by other studies suggesting SARS-CoV-2 is sporadically present in brain tissue (37). Importantly, we only investigated CNS regions with the greatest significance in AD. These regions may be less prone to SARS-CoV-2 infection than others, such as the olfactory bulb and tract, where olfactory neurons are proposed to be infected through viral spread from olfactory epithelium (38). Our findings of the viral nucleocapsid limited to the endothelium suggest the hematological spread of SARS-CoV-2 to the CNS that may not extend to the neurons in the cortical BA9 and HF yet is still capable of inducing widespread inflammation in the brain. Even in the absence of a detectable virus in the neurons or neural cells, SARS-CoV-2 may impact cognitive dysfunction through TREM1 activation of the NLRP3 inflammasome and subsequent pyroptosis, where pro-caspase 1 cleavage and subsequent
cleavage and activation of IL-1β, IL-18 and gasmerdin D pore formation in cells ultimately leading to pyroptosis (39, 40). NLRP3 activation is reported in AD and COVID-19, and is suggested by formation of microglial nodules demonstrated in these cases (41–43).

SARS-CoV-2 is not the first virus to be implicated in cognitive dysfunction. This is a facet shown with other viruses such as HIV, HSV, and Epstein-Barr virus (EBV) (44, 45). While this study cannot predict the outcome of disease progression in COVID-19 survivors, the present findings that SARS-CoV-2 infection can recapitulate AD-type transcriptional and cellular neuroinflammatory patterns among other in a very short time frame makes it critical to understand how SARS-CoV-2 impacts long-term cognition. The increase in nodular formation present in SARS-CoV-2 infected AD cases tissue also demonstrates a critical need to functionally determine potentially synergistic effects of AD and SARS-CoV-2. This is underscored by the prevalence of cognitive dysfunction seen among neuro-PASC patients, making it imperative that the link between SARS-CoV-2 and cognition be intensively investigated to identify potential therapeutic strategies for halting cognitive decline in these individuals. Collectively, our results demonstrate several key areas of overlap between the neurological effects of SARS-CoV-2 infection and AD. These findings may help inform decision-making regarding therapeutic treatments in patients with COVID-19, especially those who may be at increased risk of developing AD.
Materials and Methods

Patients

Brain tissue was collected from the cortical BA9 and HF of four SARS-CoV-2 cases, four AD cases and four non-SARS-CoV-2 or AD autopsies. With each case, one hemisphere of the brain was formalin-fixed paraffin-embedded (FFPE) and the other hemisphere was frozen to generate fresh-frozen blocks. FFPE tissue was used for microscopy and region matched fresh-frozen tissue was used for sequencing.

Tissue was collected in accordance with IRB-approved guidelines and regulations by the Brain Bank at Mount Sinai, and clinical data including cardiovascular and neurological conditions were collected by the Department of Neurology.

RNA sequencing

Samples collected and homogenized in RNAzol RT (Molecular Research Center, Inc.) were then processed using the Zymo Clean and concentrator Kit (Zymo Research) to collect total RNA following the manufacturers protocols. cDNA library construction and sequencing were conducted by Genewiz (Azenta Life Sciences). Total RNA samples were quantified using Qubit 2.0 Fluorometer (Life Technologies, Carlsbad, CA, USA) and RNA integrity was checked with 4200 TapeStation (Agilent Technologies, Palo Alto, CA, USA).

Samples were treated with TURBO DNase (Thermo Fisher Scientific, Waltham, MA, USA) to remove DNA contaminants, followed by rRNA depletion using QIAseq® FastSelect™–rRNA HMR kit (Qiagen, Germantown, MD, USA); conducted following the manufacturer’s protocol. RNA sequencing libraries were constructed with the NEBNext Ultra II RNA Library Preparation
Kit for Illumina by following the manufacturer’s recommendations, where enriched RNAs were fragmented for 15 minutes at 94 °C. First strand and second strand cDNA are subsequently synthesized. cDNA fragments are end repaired and adenylated at 3’ends, and universal adapters are ligated to cDNA fragments, followed by index addition and library enrichment with limited cycle PCR. Sequencing libraries were validated using the Agilent Tapestation 4200 (Agilent Technologies, Palo Alto, CA, USA), and quantified using Qubit 2.0 Fluorometer (ThermoFisher Scientific, Waltham, MA, USA) as well as by quantitative PCR (KAPA Biosystems, Wilmington, MA, USA). The sequencing libraries were multiplexed and clustered on one lane of a flowcell. After clustering, the flowcell was loaded on the Illumina HiSeq 4000 instrument according to manufacturer’s instructions. The samples were sequenced using a 2x150 Pair-End (PE) configuration.

Image analysis and base calling were conducted by the HiSeq Control Software (HCS). Raw sequence data (.bcl files) generated from Illumina HiSeq was converted into FASTQ files and de-multiplexed using Illumina's bcl2fastq 2.17 software. One mismatch was allowed for index sequence identification. After investigating the quality of the raw data, sequence reads were trimmed to remove possible adapter sequences and nucleotides with poor quality using Trimmomatic v.0.36. The trimmed reads were mapped to the GRCh38 reference genome available on ENSEMBL using the STAR aligner v.2.5.2b. BAM files were generated as a result of this step. Unique gene hit counts were calculated by using feature Counts from the Subread package v.1.5.2. Only unique reads that fell within exon regions were counted.

Immunohistochemistry
Tissue was cut at 6 µm on a Leica Microtome to for immunohistochemistry that was performed formalin-fixed paraffin-embedded (FFPE) brain sections as previously described \((46)\) (Iba-1 staining was conducted on a Ventana Benchmark using OptiView and UltraView detection kits provided by Roche (Roche Molecular Systems, Inc). Sections were deparaffinized in xylene and rehydrated through an ethanol series ending in distilled water. Heat-mediated antigen retrieval was carried out in a vacuum oven with Tris-EDTA buffer (10mM Trizma base, 1mM EDTA, 0.05% Tween 20, pH 9.0) or sodium citrate buffer (10mM sodium citrate, 0.05% Tween 20, pH 6.0). All washes were performed using tris buffered saline containing Tween 20 (TTBS; 0.1M Trizma base, 0.15M NaCl, 0.1% Tween 20, pH 7.4). Following antigen retrieval, tissues were blocked with 20% normal horse serum. Titrated primary antibodies included anti-HIF-1α (mouse mgc3, 1:400, Abcam, ab16066). Tissues were incubated with primary antibody overnight at room temperature and detected using the appropriate biotinylated secondary antibody (1:200, Vector Labs, BA-1100, BA-2000) and alkaline phosphatase-Vector Red according to manufacturer instructions (Vector Labs). Tissues were counterstained with Mayer’s hematoxylin and coverslipped.

**Imaging and quantitation**

Slides were scanned with the Axio Scan.Z1 digital slide scanner (Zeiss). Brightfield images were acquired using HALO (Indica Labs, v3.4.2986.151). Figures were created in Photoshop (Adobe, v23.5.1) by brightness and contrast adjustments applied to the entire image.
Threshold and multiplex analyses were performed with HALO algorithms for non-biased quantitation of proteins of interest without processing, as described previously (38). For microglia quantitation, hematoxylin-stained nuclei were used to quantify the total number of cells and those with Vector Red intensity above a rigorous threshold (Iba-1+). Microglia frequency is reported as the percentage of total nuclei in the tissue section assessed for each individual subject. To assess frequency of nodular lesions, all Iba-1-stained tissues were viewed in HALO and nodular lesions made up of 3 or more Iba-1+ microglia in contact with one another were counted as a single nodule. The total number of nodules for each tissue is reported per tissue area for each individual patient. Quantitation of HIF-1α was performed using an area quantification algorithm for Vector Red intensity. Annotations were drawn to outline the tissue and exclude empty spaces and glass. The annotated area was analyzed for overall quantity of Vector Red positivity per micron² of tissue and reported for each individual patient. Two-tailed Mann-Whitney U tests were performed with GraphPad Prism software, v9.3.1. Data are expressed as mean ± SEM. P values ≤0.05 are considered significant.

Bioinformatics Data Access and Analysis

Sample similarity was assessed with PCA analysis on VST-transformed expression values. Genes were filtered to remove lowly expressed genes, defined as fewer than 5 samples showing a minimum read count of 1 read, prior to performing differential analysis with DESeq2, in R (4.2.0) (47). The transcriptional profiles of SARS-CoV-2 and AD postmortem cases from the cortical BA9 and HF were assessed by pairwise comparison to the neurologically healthy control cases. Multiple testing correction was performed for all comparisons, however, due to the relatively small number of genes surpassing a threshold of 0.05, a nominal p-value was used in
the selection of differential genes for all downstream analyses. Genes were assigned as
differentially expressed if the nominal p-value was less than 0.05 and the absolute log2FC
exceeded 1. An additional round of differential expression testing was performed using a model
containing covariates (COVID status, Alzheimer’s status, gender, and brain region), to extract
the transcriptional effects of the individual covariates. Biological pathways and key regulators
impacted by disease were identified using QIAGEN Ingenuity Pathway Analysis (IPA)
(QIAGEN Inc., Version 73620684) (48) Genes with a threshold of p<0.05 were used as input for
IPA. The relative similarities of transcriptional changes in the DESeq2 comparisons were
assessed using Rank-rank hypergeometric overlap (RRHO2) analysis.
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Data and materials availability: The authors declare that the data supporting the findings of this study are available within the paper and its supplementary information files. The raw data discussed in this publication will be accessible through NCBI’s Gene Expression Ominubus (GEO) upon publication (49).
Figure 1. Gene expression in cortical BA9 and HF.

(A-C) Volcano plot distribution of gene transcripts of the cortical BA9 region in AD cases (A), SARS-CoV-2 cases (B), and SARS-CoV-2 infected AD cases (C) compared to neurological control cases. (E-G) Volcano plot distribution of gene transcripts of the cortical HF region in AD cases (E), SARS-CoV-2 cases (F), and SARS-CoV-2 infected AD cases (G) compared to neurological control cases. Volcano plots were generated from 39,901 genes (A, B, E and F), from 38,021 genes (C) and from 35,527 genes (G). Transcripts with nominal $p<0.01$ and an absolute log2 fold-change (log2 FC) greater than 1 are indicated in red. (D and H) Bar graphs, in which the number of upregulated and downregulated genes (with $p<0.01$ and absolute value of log2 FC>1), is indicated in red and blue, respectively.

Figure 2. Similarity of gene expression in cortical BA9 and HF in SARS-CoV-2 and SARS-CoV-2 infected AD relative to AD individuals based on log2FC.

(A-B) Genes known to be differentially expressed due to either SARS-CoV-2 infection or AD were examined with respect to their gene expression changes, with the given gene set being selected. The log2 FC of the genes for three individual differential comparisons (AD versus control, SARS-CoV-2 versus control, and SARS-CoV-2 infected AD versus control) are shown for cortical BA9 (A) and HF (B), respectively. (C, D) Using Ingenuity pathway analysis (IPA) as a filtering system for genes within the database, we compared the similarity in expression of genes present in all datasets (SARS-CoV-2, AD and SARS-CoV-2 infected AD cases) in both the cortical BA9 (C) and HF (D).
Figure 3. Changes in signaling pathways within the cortical BA9.

(A) The similarities of canonical pathways reveal the top regulated canonical pathways within the cortical BA9 in AD, SARS-CoV-2, and SARS-CoV-2 infected AD cases in reference to the control. Activation score (Z-score) is shown on the X-axis, while the pathways are indicated on the Y-axis. The color of the points indicates the IPA comparison, while the size of the point represents the -log10 \( p \)-value of the IPA comparison, with the larger points indicating the lowest \( p \)-values. (B) The predicted gene regulation of the Neuroinflammation, TREM1 and NGF pathways indicate that AD (red bar), SARS-CoV-2 (green bar) and SARS-CoV-2 infected AD groups (blue bar) have similar expression in key inflammatory and neuronal pathways, with Log2 FC shown on the X-axis.

Figure 4. Changes in signaling pathways within the HF.

(A) The similarities of canonical pathways reveal the top regulated canonical pathways were compared within the HF in AD, SARS-CoV-2 and SARS-CoV-2 infected AD cases in reference to the control. The X-axis represents the activation score (Z-score), while the Y-axis indicates the pathways. The color of the points reflects the IPA comparison, while the size of each point corresponds to the -log10 \( p \)-value of the IPA comparison, with larger points indicating lower \( p \)-values. (B) The predicted gene regulation of the Neuroinflammation, TREM1 and NGF pathways indicate that AD (red bar), SARS-CoV-2 (green bar) and SARS-CoV-2 infected AD groups (blue bar) have similar expression in key inflammatory and neuronal pathways. The X-axis represents Log2 FC.
Figure 5. Microgliosis and nodular lesions in neurological controls, SARS-CoV-2, AD and SARS-CoV-2 infected AD individuals.

The level of microglial activation was assessed by immunohistochemical staining using anti-Iba-1 antibody with Vector Red. The presence of SARS-CoV-2 was assessed using an antibody against the virus nucleocapsid and visualized with DAB. (B-D) Parenchymal microglia are more frequent and highly activated in the context of infection with and without AD, as shown by thickened processes, enlarged soma, and loss of individual cellular domain, as compared to age-matched controls (A and E). When present, SARS-CoV-2 localizes to the blood vessel endothelium (black arrows, G, H, K and L). (F, G and H) Microglia appear to gather around blood vessels in disease but do not form cuffs. (I-L) Nodular lesions are seen in most cases assessed, regardless of disease status, however, they appear larger and more frequent in the context of disease (J-L), as compared to controls (I). (M-O) A multiplex algorithm was used to count all cells, using DAPI+ nuclei, with HALO and calculate percent frequency of Iba-1+ microglia. Each point on the graphs shown represent the average finding per total brain area for each subject. (M) There was upregulation of Iba-1 in the hippocampus as compared to the cortical BA9 region for most cases, though there is a significantly higher ratio of microglia relative to other cells when comparing the AD and SARS-CoV-2 infected AD cases to the control group shown. (N) This trend seems to hold true for cortical BA9 region as well, where the only groups with a significant difference in microglia ratios are the control and SARS-CoV-2 infected AD cases shown. (O) When all cases have both regions averaged, the level of microgliosis is shown to be higher in the AD and SARS-CoV-2 infected AD cases when compared to control. (P-R) Graphs show the normalized counts of microglial nodule frequency. (P) A higher frequency of nodules is seen in the HF, however, the difference between groups did
not reach statistical significance. (Q) In contrast, fewer nodules are seen in cortical BA9, overall.

A statistically significant higher number of lesions are seen in SARS-CoV-2 infected AD cases, as compared to control cortical BA9, suggesting greater inflammation in the cortical BA9 of patients with both AD and SARS-CoV-2 infection. (R) Significance was maintained between the control and SARS-CoV-2 infected AD groups when the average was taken for both brain regions per case. Statistics were performed with a two-tailed Mann-Whitney U test. *p<0.05. Data are expressed as mean ± SEM.

**Figure 6. Hypoxia-inducible factor-1 alpha subunit (HIF-1α) in neurological controls, SARS-CoV-2, AD and SARS-CoV-2 infected individuals.**

(A-H) Representative images demonstrate HIF-1α immunopositivity around the vasculature that extends into the parenchyma in brain of patients with SARS-CoV-2 infection (C, D, G and H), regardless of AD status. Comparatively, this is seen less frequently and does not extend significantly into the parenchyma in brain of age-matched controls (A and E). In AD only, positivity was most often observed in epithelium with no or minimal extension into the brain parenchyma (B and F). (I-K) Graphs show the total average percentage of tissue positive for HIF-1α for each individual subject, as defined using a HALO area algorithm for detection of Vector Red intensity over the whole section. Although statistical significance between groups was not reached in the HF (I) or averaged group comparisons (K), increased HIF-1α expression is seen in the SARS-CoV-2 infected patients, with and without AD, as compared to age-matched controls. An increase in area positivity is seen in all groups in the cortical BA9 region (J), as compared to non-affected controls, however, statistical significance is only seen with the SARS-
CoV-2 group. Statistics were performed with a two-tailed Mann-Whitney U test. *$p<0.05$. Data are expressed as mean ± SEM.

Figure supplements 1. RRHO analysis of SARS-CoV-2/control and AD/control, each containing (39,902 DEGs) revealed that gene regulation between the AD/control and SARS-CoV-2/control groups showed a positive correlation in cortical BA9 region.

Figure supplements 2. RRHO analysis of SARS-CoV-2/control and AD/control, each containing (39,902 DEGs) revealed that gene regulation between the AD/control and SARS-CoV-2/control groups showed a positive correlation in HP region.

Supplementary file 1. Table 1. Demographics of Postmortem Cases.
Numbers presented are the average per group and standard deviation PMI is post mortem interval (n=4); OtD is onset to death. ABC is a multipoint measurement where A is a measure of amyloid beta deposition, B is a measure of neurofibrillary degeneration based on the Braak score, and C is scored based on neuritic plaques outlined by the Consortium to Establish a Registry for Alzheimer’s Disease diagnosis (CERAD).

Supplementary file 2. Table 2. Blood Chemistry and Symptoms.
Symptoms are presented as a percentage of the group (n=4); Chemistry numbers are presented as the average per group and standard deviation, based on maximum values collected for single values and the minimum and maximum range collected in cells with two values. Bolded values are implicated as indicators of severe SARS-CoV-2 outcomes; ALP is alkaline phosphatase,
ALT is alanine aminotransferase, AST is aspartate transferase, CRP is c-reactive protein, ESR is erythrocyte sedimentation, INR international normalized ratio, LDH is lactate dehydrogenase, WBC is white blood count, BUN is blood urea nitrogen.

Source data 1. Based on the deseq, and filtered using IPA, genes that were expressed in all SARS-CoV-2 and AD in cortical BA9 datasets with an absolute expression of 0.0001 or greater and \( p < 0.05 \) are presented.

Source data 2. Based on the deseq, and filtered using IPA, genes that were expressed in all SARS-CoV-2 and AD in HF datasets with an absolute expression of 0.0001 or greater and \( p < 0.05 \) are presented.
The images show histological sections from different groups: control, AD, SARS-CoV2, and AD+SARS-CoV2. The images are labeled as BA9 and HF, indicating different brain regions. The bar graphs below each group represent the percentage of HIF-1α positivity with error bars indicating variability. The graphs show a comparison between control, AD, SARS-CoV2, and AD+SARS-CoV2 conditions.