

# A transcriptome analysis identifies potential preventive and therapeutic approaches towards COVID-19.

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## Abstract

The recent outbreak of Coronavirus Disease 2019 (COVID-19) is a major threat to human health and the global economy. In addition to the development of vaccines there is an urgent need for preventive and therapeutic strategies towards severe COVID-19. Key to this would be a better understanding of the molecular mechanisms affected by SARS-COV-2. To address this, we performed a systems biology approach by integrating available RNA-seq datasets from post-mortem lung tissue of COVID-19 patients and cell culture models infected with SARS-COV-2, Respiratory Syncytial virus or influenza virus. We identified two gene-expression modules that are commonly regulated by the three viral diseases and one that is specific to COVID-19. All 3 gene-expression modules represent key inflammatory processes. We identified several proteins within these networks that can be targeted by FDA approved drugs. Key examples are *TNF*, *NFkB*, *INTERLEUKIN-1* and *ALOX5* signaling pathways. Our data also suggest that Vitamin D supplementation and a ketogenic diet should be further analyzed as preventive strategies. In conclusion, our data highlights the potential of transcriptomics to unravel the pathological processes related to COVID-19 and guide the initiation of clinical trials.

**Keywords:** COVID-19, gene-expression, TNF, IL-1, Alox5, NFkB, Vitamin D, therapy

## 41 Introduction

42 Novel coronavirus disease 2019 (COVID-2019) is caused by the severe acute  
43 respiratory syndrome coronavirus 2 (SARS-CoV-2). In humans SARS-Cov2 can lead  
44 to acute respiratory syndromes eventually causing end stage lung injury, failure of  
45 multiple organs and death<sup>1</sup>. The current SARS-Cov2 pandemic has spread across the  
46 globe and has already killed thousands of individuals since the first case was reported  
47 in late December, 2019 in Wuhan, China. On 30<sup>th</sup> January, 2020 WHO declared the  
48 outbreak of COVID-2019 a public health emergency of international concern<sup>2</sup>. There  
49 are enormous international research efforts to better understand SARS-Cov2 and  
50 important findings such as the structural analysis of the Spike (s) protein<sup>3</sup> by which the  
51 virus gains entry to cells or the viral RNA polymerase<sup>4</sup> have been reported. There is  
52 also evidence that SARS-Cov2 recognizes the receptor angiotensin converting  
53 enzyme II (ACE2) for cell entry<sup>5</sup>. Interestingly, while most research focuses on the  
54 effect of SARS-Cov19 on the general organs such as the lung, a neuro-invasive  
55 potential has been observed for SARS-Cov2 and patients often present with  
56 neurological symptoms<sup>6</sup>. These data highlight the urgent need to better understand  
57 Covid-19 and develop preventive and therapeutic strategies. Several research  
58 institutes, biotech and pharma companies are currently developing vaccines against  
59 COVID-19. A recent release from WHO (4<sup>th</sup> of April, 2020) listed 63 of such projects<sup>7</sup>  
60 that are however in preclinical or early clinical stages. In addition to vaccination, there  
61 is an urgent need for preventive and therapeutic strategies to reduce the rising death  
62 toll due to COVID-19. Thus, a better understanding of the molecular pathology  
63 associated with SARS-CoV-2 infections is of utmost importance since it also offers the  
64 chance for drug repurposing. Some treatment options using several broad-spectrum  
65 antivirals such as favipiravir, remdesivir<sup>8</sup>, and anti-malaria drug chloroquine<sup>9</sup> have  
66 been suggested and await further validation. In addition, severe COVID-19 is  
67 associated with a major immune inflammatory response and anti-inflammatory  
68 therapies are discussed as a promising therapeutic avenue<sup>10,11</sup>.

69  
70 The analysis of gene-expression networks is a suitable approach to elucidate patho-  
71 mechanisms and identify drug targets<sup>12</sup>. In this study, we use a systems biology  
72 approach to integrate the available transcriptomic data on COVID-19 patients and  
73 corresponding cellular models. Our data confirm that COVID-19 infection induces  
74 transcriptional changes associated with a major inflammatory response. Further

75 analysis revealed that COVID-19 shares a conserved inflammation related gene  
76 regulatory network with two other respiratory viruses namely Respiratory Syncytial  
77 virus (RSV) and Influenza virus. We also detected a network linked to inflammation  
78 that is specific to Covid-19. On the basis of these networks, we provide evidence that  
79 anti-inflammatory drugs such as inhibitors of tumor necrosis factor (TNF), NFkB, IL-1  
80 and Alox5 signaling would be *bona fide* drug targets to treat severe COVID-19. Our  
81 data moreover suggest that a ketogenic diet and vitamin D supplementation could be  
82 a preventive approach for a severe course of COVID-19.

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## 85 **Results**

### 86 **Conserved gene-expression networks in COVID-19 are linked to inflammation**

87 Starting point of our study were RNA-seq datasets obtained from post-mortem lung  
88 tissue of COVID-19 patients and corresponding lung-biopsy tissue from uninfected  
89 healthy subjects and data from corresponding cell lines. Since data from patients is  
90 still rare we decided to start our analysis with more robust data from cell culture models  
91 of COVID-19 and related diseases and later cross-correlate these data to findings from  
92 patients (**Fig. 1A**). Our first aim was to identify gene-expression modules that are  
93 observed in COVID-19 and related diseases. Thus, we analyzed RNA sequencing data  
94 from experiments that employed the same experimental setting using lung epithelial  
95 cell line A549 and Calu3 infected with SARS-CoV-2<sup>13</sup> and the A549 cell line infected  
96 with Respiratory Syncytial Virus (RSV)<sup>13</sup>. As control, we employed lung epithelial cell  
97 lines subjected to mock infection. Unbiased weighted gene co-expression (WGCNA)  
98 analysis revealed 9 different expression modules in the data (**Fig. 1B, Table S1**) of  
99 which 5 modules displayed high correlation (correlation coefficient > 0.75, \*P<0.05)  
100 with the status of viral infection in all experimental conditions (**Fig. 1B**). While the other  
101 four modules would represent cell type or viral infection specific modules, we first  
102 decided to further analyze the five modules commonly affected following SARS-CoV-  
103 2 and RSV infections. Three of these five modules showed decreased expression  
104 pattern in both A549 and Calu3 cells while the other two modules displayed increased  
105 expression pattern following SARS-CoV-2 infection (**Fig. 1C**). These changes were  
106 consistent in RSV infected cells and data suggest that COVID-19 shares pathological  
107 gene-expression changes with related viral diseases. GO analysis of the 3 down-

108 regulated modules revealed that the MEDarkred module plays a role in mitochondrial  
109 function, cellular respiration, and protein folding (**Fig. 1D**). In line with this, our data  
110 suggest that also MEyellow module is linked to cellular respiration and mitochondrial  
111 functions (**Fig. 1E**), while the MEgrey60 module regulates cell cycle check point and  
112 genomic repair mechanisms (**Fig. 1G**). Interestingly, genes of the two up-regulated  
113 modules were linked to inflammatory processes (**Fig. 1G, 1H**). Thus, the MELightcyan  
114 module represents biological processes related to increased innate immune response  
115 and increased signaling via tumor necrosis factor (TNF), interleukin 1 (IL-1) and NF-kb  
116 (**Fig. 1G**). Genes of the MEmidnightblue module are also involved in immune  
117 responses, NF-kb, TNF signaling and viral life cycle (**Fig. 1H**). In conclusion, the RNA-  
118 seq data analyses suggest that similar inflammatory responses are induced by COVID-  
119 19 and Respiratory Syncytial Virus pointing to the possibility that existing anti-  
120 inflammatory therapies may help in the case of severe COVID-19 cases.

121

122 To further support this hypothesis, we decided to compare our data to findings obtained  
123 from an independent study on the influenza virus (InV) that can also cause respiratory  
124 syndromes. To this end we employed RNA-seq data generated from lung epithelial  
125 A549 cells that were exposed to InV. In line with the original study<sup>14</sup>, we observed a  
126 substantial number of deregulated genes (**Table S2**). Gene enrichment analysis  
127 revealed that the up-regulated genes are linked to inflammatory processes including  
128 activation of the innate immune response, cellular responses to virus and interferons  
129 and activation of T-cells (**Fig. 2A**). Next, we asked whether the expression changes of  
130 two up-regulated modules linked to inflammation (i.e. MELightcyan and  
131 MEmidnightblue) that were identified in lung cells treated with COVID-19 are also  
132 affected in cells infected with influenza virus. To this end we constructed eigenvalues  
133 of the corresponding genes and compared their expression between control and  
134 influenza infected samples. Interestingly, the expression of genes within the  
135 MELightcyan and MEmidnightblue modules was substantially increased in influenza  
136 infected cells (**Fig. 2B, 2C**). These data further support the view that SARS-COV-2  
137 infection drives a conserved gene regulatory network linked to inflammatory processes  
138 that is similar to respiratory syndromes caused by RSV and Influenza viruses, pointing  
139 to common patho-mechanisms. Moreover, our data provide evidence for the  
140 hypothesis that common anti-inflammatory therapies might be a suitable approach  
141 towards severe COVID-19.

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143 To substantiate this view we retrieved available RNA-seq dataset on post-mortem  
144 lung samples obtained from COVID-19 patients<sup>13</sup>. This dataset also included lung  
145 biopsies from uninfected healthy subjects that were processed similarly and used as  
146 controls. Although care has to be taken since biopsy and post-mortem data is  
147 compared and moreover the data set consist of only 2 patients and 2 controls, a  
148 differential expression analysis revealed deregulation of 774 genes (**Fig. 3A, Table**  
149 **S3**). Gene ontology analysis suggest that the upregulated genes are involved in  
150 processes related to inflammation and viral interaction, which is in line with the data  
151 obtained from cell culture models (**Fig. 3B**, also see **Fig. 1**). Indeed, when we  
152 compared the MElightcyan and MEmidnightblue gene-expression modules linked to  
153 inflammatory processes, we found both modules to be up-regulated in COVID-19  
154 patients (**Fig. 3C, 3D**).

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### 157 **Transcriptomics point to preventive and therapeutics strategies in COVID-19.**

158 The finding that COVID-19 seems to be linked to the up-regulation of 2 conserved gene  
159 expression modules related to inflammation allowed us to screen if existing drugs  
160 might counteract COVID-19 related pathology. To this end, we constructed the  
161 regulatory networks of the 2 modules and asked whether any currently available and  
162 FDA approved drugs would target the proteins in these regulatory networks. For  
163 MElightcyan module we found a number of proteins (e.g. IFNAR2, XDHH, PTK2B,  
164 INSR, JANK2, TNF, ADORA2A, CREB, NFKB1, NFKB2, PTG32) that can be targeted  
165 by available drugs. The most important targets are tumor necrosis factor (TNF) and  
166 NF-KB1 (**Fig. 4A**) as they represent the hub genes of the network. This is interesting  
167 since various drugs that target these two proteins are available and inhibition of TNF  
168 has been recently suggested in a commentary article as a promising strategy to treat  
169 severe COVID-19<sup>15</sup>. In line with these data, chloroquine can inhibit TNF signaling and  
170 is already tested in COVID-19 patients<sup>9</sup>. Of particular interest might also be Pranlukast  
171 and Thalidomide since they are able to suppress both TNF and NFKB1 signaling and  
172 especially Pranlukast has been used to treat asthma in children and adults with no  
173 severe side effects<sup>16</sup>.

174

175 In the MEmidnightblue module we also find several target proteins that can be  
176 modulated by already available drugs (**Fig. 4B**). For example, immune modulators like  
177 IL6, JUN, CD55, TNFS13B, SH2B3, CASP1 and IFNGR2 are potential candidates for  
178 which drugs are available and a recent study demonstrated promising results related  
179 to COVID-19 management via IL-6 inhibitors<sup>17</sup>. It is also noteworthy that Vitamin D  
180 Receptor (VDR) is observed as a potential target within this inflammation regulatory  
181 network, supporting recent ideas that Vitamin D supplementation might be a preventive  
182 strategy towards severe COVID-19<sup>18</sup>. Another interesting observation is the presence  
183 of CASP1 as it plays a key role in the inflammasome signaling and control for example  
184 IL-1 $\beta$  signaling which might be suppressed by Minocycline<sup>19</sup>. The data described  
185 above are based on the 2 identified conserved gene-expression modules. We also  
186 analyzed the dataset from human patient material separately. Although care has to be  
187 taken when interpreting this limited dataset alone, we observed IL-1 $\beta$  and TNF as key  
188 targets for therapeutic intervention (**Fig. 4C**).

189

190 So far, we focused our analysis on gene-expression modules commonly de-regulated  
191 in COVID-19 and other viral diseases such as influenza (See Fig. 2). The reverse  
192 approach would be to screen for gene-expression changes specific to COVID-19 that  
193 are not observed in RSV or influenza. Indeed, we found that out of the 9 originally  
194 identified gene-expression modules, the MEBrown module (see Fig 1) was increased  
195 in A549 cells treated with COVID-19, while it was decreased in A549 cells treated with  
196 RSV or influenza (**Fig. 5A**). In line with this observation genes of the MEBrown module  
197 were increased in postmortem lung tissue of COVID-19 patient (**Fig. 5B**). GO-analysis  
198 of this module indicates that the corresponding genes are linked to inflammation  
199 associated processes (e.g. autophagy and Wnt-signaling) and gene-expression  
200 control via histone-modifications and non-coding RNA processes (**Fig. 5C**). When we  
201 generated the network of the MEBrown module and asked for potential targets for FDA  
202 approved drugs, we identified ALOX5 a promising target (**Fig. 5D**). ALOX5 codes for  
203 the Arachidonate 5-lipoxygenase, a well-known target for inflammatory processes  
204 including respiratory diseases<sup>20</sup>. The complete list of potential drugs identified via the  
205 approaches described above is listed in supplemental file.

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208 **High-fat low-carbohydrate ketogenic diet may reduce COVID-19 related**  
209 **inflammation**

210  
211 So far, our data point to a number of interesting FDA approved drugs that could be  
212 considered for further testing but also for epidemiological analysis with the aim to  
213 identify factors that may lower the risk for a severe course of COVID-19. For example,  
214 a number of the identified drugs are used in chronic inflammatory diseases such as  
215 rheumatoid arthritis. There is evidence that patient treated with such drugs can have a  
216 lower risk for other inflammatory diseases. For example, Alzheimer's disease (AD) is  
217 linked to neuroinflammatory processes and there is substantial evidence from  
218 epidemiological studies that patients chronically treated with anti-inflammatory drugs  
219 for other reasons have a lower risk to develop AD at old age<sup>21</sup>. On these bases we  
220 wondered about general environmental factors that could be associated with  
221 inflammatory processes and may therefore provide preventive strategies for severe  
222 COVID-19. One interesting example is the finding that a ketogenic diet was shown to  
223 reduce inflammation. To address this issue we used a dataset from Goldberg et al.<sup>22</sup>.  
224 Here, the authors employed a mouse model of influenza infection and performed RNA-  
225 seq from lung tissue extracted from mice fed ketogenic diet (KD) or normal chow. We  
226 re-analyzed this data (**Fig. 6A**) and confirmed the previously reported reduced  
227 inflammatory responses in KD mice (**Fig. 6B**). Interestingly, while ketogenic diet had  
228 no effect on MElightcyan and MEbrown gene-expression modules (**Fig. 6C, 6D**), this  
229 diet reduced the expression of MEmidnightblue inflammatory module that we found to  
230 be increased COVID-19, RSV and InV (**Fig. 6E**). Moreover, expression of several viral  
231 responsive related genes in COVID-19 were downregulated in mice fed a ketogenic  
232 diet (**Fig. 6F**). In sum, these data indicate the potential of ketogenic diet as a preventive  
233 approach to affect the risk for a severe course COVID-19.

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## 240 Discussion

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242 This study uses available transcriptome data from cell culture models and postmortem  
243 material from COVID-19 patients. Although especially the human data is preliminary  
244 due to the small sample size, we aimed to overcome this issue by using an integrative  
245 approach by combining the various datasets from cell culture models. A robust  
246 observation is the finding that COVID-19 leads to the up-regulation of gene-expression  
247 modules that all represent key inflammatory processes. Two conserved gene-  
248 expression modules are also regulated in response to RSV or influenza while one  
249 module was specific to COVID-19. Overall, these findings are in line with the  
250 hypothesis that an aberrant inflammatory response plays a role in severe COVID-19  
251 cases<sup>10,11</sup>. Our data allowed us to identify therapeutic targets within these modules and  
252 screen for potential FDA approved drugs. As most promising targets we identify  
253 processes related to TNF, NFκB, IL-1β and ALOX5 signaling. These data are in line  
254 with recent proposals to test TNF-related therapies in COVID-19<sup>15</sup> and also hint to  
255 additional approaches that can be tested immediately. For example, our data suggest  
256 that Pranlukast which affects TNF and NFκB-signaling<sup>23</sup> could be a suitable approach  
257 to attenuate COVID-19 related inflammation. Equally interesting is Minocycline  
258 towards ALOX5, since this pathway was in our dataset specific to COVID-19 and not  
259 observed in datasets from RVS or influenza. Additionally, Minocycline can suppress  
260 IL-1β signaling that was also detected as a key process affected by COVID-19. A  
261 potential role of IL-1β is also noteworthy, since ongoing clinical trials are rather  
262 focusing on IL-6 as therapeutic target (see <https://dzif.clinicalsite.org/de/cat/2084#c2084>).  
263 A recent study reported that high glucose levels in blood promote the release of inflammatory  
264 cytokines following influenza infection. This was mediated in part via TNF receptor–associated  
265 factor 6 (TRAF6) mediated activation of interferon regulatory factor–5 (IRF5). In line with these  
266 data, we also observed TRAF6 as one of the hub genes in our gene regulatory networks (see  
267 Fig. 3A). In addition to TRAF6, the Insulin receptor (INSR) was present in the same conserved  
268 network indicating that a similar glucose dependent increased inflammatory mechanism may  
269 persist in COVID-19. In line with this hypothesis, patients with high blood glucose levels (e.g.  
270 individuals suffering from diabetes) might be more susceptible to develop severe COVID-19<sup>24</sup>.

271 Another interesting observation is the presence of Vitamin D receptor as a key target  
272 in the COVID-19 related gene-expression network. This data is in line with reports  
273 showing that vitamin D supplements exhibit anti-inflammatory effects a by reducing  
274 interleukins and NFκB activity<sup>25,26</sup>. Dietary vitamin D supplementation may be a simple,



275 safe and inexpensive preventive measure to reduce the risk for severe COVID-19, a  
276 hypothesis that should be tested in epidemiological studies. In fact, recent reports have  
277 already pointed to a potential beneficial effect of Vitamin D in COVID-19<sup>18</sup>. Moreover,  
278 Vitamin E can suppress ALOX5 to limit inflammation<sup>27</sup>, providing further evidence that  
279 strategies towards ALOX5 inhibition should be further tested in COVID-19. It would be  
280 important to screen for additional strategies that may impact on the course of COVID-  
281 19. As a first approach we analyzed the effect of ketogenic diet and observed that it  
282 might have a beneficial effect since in counteracted the expression changes of at least  
283 1 of the 3 gene-expression modules we identified as a consequence of COVID-19.

284  
285 Although the available data on COVID-19 is still limited, our study highlights the power  
286 of gene-expression analysis to identify potential preventive strategies and drug targets  
287 for repurposing to treat severe COVID-19. Taking into account that at least 657 clinical  
288 studies towards COVID-19 are currently ongoing or in preparation (clinicaltrials.gov),  
289 we suggest that the continued analysis of available multi-omics data should become  
290 an important approach to guide the development and testing of drugs against COVID-  
291 19.

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## 296 **Methods**

### 297 **Data acquisition and analysis**

298 RNAseq data were retrieved from three studies. The NCBI GEO accession of these  
299 studies are as follow: GSE147507, GSE136536 and GSE121155. Compressed  
300 FASTQ files were downloaded from corresponding SRA accessions using fastq-dump  
301 of sratoolkit (version 2.8.0). The quality of the FASTQ files were inspected using  
302 FASTQC (version 1.0). Transcriptomic reads were mapped to corresponding genome  
303 using STAR (version 2.5.2b). For human and mouse, hg38 and mm10 genome were  
304 used for mapping. The mapped files (BAM) were used to enumerate the reads for each  
305 gene using featureCounts of Subread (version 1.5.1). Downstream analyses were  
306 performed in R (version 3.6.1). These raw reads were normalized to the library size  
307 and differential expression analysis was performed in DESeq2<sup>28</sup>. Except otherwise  
308 mentioned, genes with adjusted p value < 0.05 were considered as differentially

309 expressed. Gene ontology analysis was performed using gene ontology  
310 (<http://geneontology.org/>).

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### 316 **Weighted gene co-expression analysis**

317 Weighted gene co-expression network analysis was performed using (WGCNA)  
318 package<sup>29</sup> in R (version 1.61). Briefly, after normalizing expression data, gene  
319 expression counts were log (base 2) transformed to calculate pair-wise correlations  
320 between genes. A signed gene expression network was constructed using soft  
321 threshold power of 17. Modules of co-expressed microRNAs with a minimum module  
322 size of 200 with deepsplit 1 was later identified. Similar modules were merged using  
323 dissimilarity correlation threshold of 0.25. Different modules were summarized as  
324 network of modular eigengenes, (MEs) which were then correlated with infection  
325 status. Pearson correlation of MEs and each of these features was plotted as heat  
326 map. Eigenvalues of the modules in *in vivo* samples were determined after limiting the  
327 module genes to be expressed and deregulated therein. Plots are generated using  
328 ggplot2 in R. Gene ontology analyses were performed as described above. Bar plots  
329 of gene ontology are plotted using GraphPad Prism (version 7.0).

330

### 331 **Gene-Drug interactions analysis**

332 Gene regulatory network was constructed using STRING (v 11.0) based on the  
333 evidence of protein-protein interactions. Network was visualized in Cytoscape (version  
334 3.7.1). Hub genes of the network was identified using CytoHubba. Top 10 Hubba  
335 nodes were ranked using degree of interactions and their shortest paths were  
336 displayed. The network was extended and integrated to available FDA approved drugs  
337 using CyTargetLinker. The DrugBank database (version 4.0) was used for this  
338 purpose.

339

### 340 **Statistics**

341 Statistical analysis was performed using either R or GraphPad Prism 7. Statistical test  
342 details are given in the figure legends.

343

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351

352 **Figure legends**

353 **Fig 1. Weighted co-expression analysis.** **A.** Outline of the experimental approach. **B.**  
354 Signed correlation of identified modules to viral infection. A total of 9 modules were identified.  
355 Heatmap showing Pearson's correlation and statistical significance. Numbers in parenthesis  
356 indicate p value while numbers on top represent the correlation co-efficient. **C.** Five highly  
357 significant clusters ( $cor > 0.75$ ,  $*P < 0.05$ ). Gene expression changes of MEDarkred, MEyellow  
358 and MEgrey modules are negatively correlated to infection, while the MElightcyan and  
359 MEMidnightblue modules are positively correlated. **(D-H)** Expression of MEDarkred, MEyellow,  
360 MEgrey60, MElightcyan and MEMidnightblue across datasets. In vitro data from two cell lines  
361 namely A549 and Calu3 were used for COVID-19 infection. For RSV infection, in vitro  
362 transcriptomic data from A549 cell line has been used. Effect on gene expression after viral  
363 treatment were compared to that treated with mock of the same experiment. Bar plot shows  
364 biological processes. X-axis represent  $-\log_{10}$  of adjusted p value. Two-tailed, unpaired t-test;  
365 p value is indicated on the figure panel. N = 2-3 per group.

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369 **Fig 2. MElightcyan and MEMidnightblue modules are up-regulated after infections with**  
370 **influenza virus.** **A.** Gene ontology analysis of the influenza viral induced upregulated genes.  
371 Boxplots show that the MElightcyan **(B)** and MEMidnightblue **(C)** modules are increased in  
372 A549 cells treated with influenza viral titers compared to mock treated cells. Unpaired t-test,  
373 two-tailed, N =3; each group.

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377 **Fig 3. Case vs. control lung RNAseq data links SARS-CoV-2 infection towards**  
378 **inflammation. A.** Volcano plot showing differentially expressed genes in COVID-19 patients  
379 compared to controls. Magenta color indicates the up-regulated genes while the darkgreen  
380 color represents down-regulated genes. X-axis represents fold change (log<sub>2</sub> scale) and Y-axis  
381 represent the significance level. A dashed line is drawn to indicate the significance cutoff. **B.**  
382 Gene-ontology analysis of up- and down-regulated genes in COVID-19 patients compared to  
383 healthy subjects. Statistically significant (p value after multiple adjustments < 0.05) biological  
384 processes are represented. Gene members in the given processes are indicated as counts. N  
385 = 2 in both cases. Expression of MElightcyan (**C**) and MEMidnightblue (**D**) between COVID-  
386 19 patient and healthy control. Expression of both MElightcyan and MEMidnightblue modules  
387 are substantially increased in patient samples and statistically significant at P = 0.06 and 0.07  
388 respectively. N = 2 each group. Unpaired t-test, two-tailed.

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392 **Fig 4. Integrative analysis of Inflammation related gene regulatory network and FDA**  
393 **approved drugs. A.** Gene regulatory network for MElightcyan module based on protein-  
394 protein interactions. Several therapeutic targets including insulin receptor (INSR), tumor  
395 necrosis factor (TNF), Prostaglandin-endoperoxide synthase 2 (PTGS2), Nuclear Factor  
396 Kappa B Subunit 1 (NFKB1) are present in the network. TNF and NFKB1 are two top hubs in  
397 the protein-protein interaction network of MElightcyan module. Several FDA approved drugs  
398 are available for these two therapeutic targets. Among them Pranlukast can inhibit both TNF  
399 and NFKB1. **B.** Vitamin D receptor (VDR), Caspase-1 (CASP1), JUN, CD55, TNFS13B,  
400 SH2B3, and Interferon Gamma Receptor 2 (IFNGR2), and IL6 are among the therapeutic  
401 targets in MEMidnightblue module. CASP1 is a key component of inflammasome and plays  
402 central role in pro-inflammatory cytokine IL-1 $\beta$  production and its associated response.  
403 Antibiotic Minocycline can suppress IL-1 $\beta$  production by blocking CASP1. **C.** In human COVID-  
404 19 patients, TNF and IL-1 $\beta$  are two hubs in the inflammation regulatory network. Among the  
405 inhibitors of TNF, Pranlukast acts also upon NFKB1. The antibiotic Minocycline can affect  
406 inflammasome and IL-1 $\beta$  signaling and may reduce the inflammation in COVID-19.  
407 Combination of TNF inhibitor and the antibiotic Minocycline may suppress the inflammation  
408 more efficiently. List of the therapeutic targets and the available drugs are summarized in the  
409 supplementary file.

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414 **Fig 5. MEbrown module represents a unique cluster associated with SARS-CoV-2**  
415 **infection.**

416 A549 cells were infected with either SARS-CoV-2, Respiratory Syncytial Virus (RSV) or  
417 Influenza Virus (InV). In all conditions, A549 cells were infected at multiplicity of infection (MOI)  
418 of 2 for 24 hours. **A.** MEbrown module shows increased expression in SARS-CoV-2 infected  
419 cells compared to mock ( $p = 0.0045$ ). Expression of the same module is decreased in RSV ( $p$   
420  $= 0.069$ ) and InV ( $p = 0.00012$ ) infected cells. Unpaired t-tests, two-tailed. Bar plot shows  
421 significant representative gene ontology terms of the MEbrown module. **B.** Expression of  
422 MEbrown genes in COVID-19 patients. **C.** Integrative network analysis displays key  
423 therapeutic targets and corresponding drugs in the MEbrown module. Potential therapeutic  
424 targets in COVID-19 include Peptidyl Arginine Deiminase 2 (PADI2) and 5-Lipoxygenase  
425 (ALOX5). One of the drugs that target ALOX5 is antibiotic Minocycline.

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429 **Fig 6. Ketogenic diet may attenuate COVID-19 induced inflammatory responses.**

430 **A.** Heatmap displays differentially deregulated genes between influenza virus infected mice  
431 either fed ketogenic diet (KD) or chow. Gene expression changes with adjusted pvalue  $< 0.1$   
432 were considered as differentially deregulated. **B.** Gene ontology of the downregulated genes  
433 in KD mice compared to chow. Comparison of expression of human orthologs of **(C)**  
434 MElightcyan, **(D)** MEbrown and **(E)** ME midnightblue modules between chow and KD groups.  
435 Changes in expression of MElightcyan and MEbrown are not significant between groups. **F.**  
436 Reduced expression of ME midnightblue is noted in KD mice that is statistically significant at P  
437 value 0.07. **G.** Human orthologs of viral response related genes are down-regulated in KD  
438 group compared to chow. N = 3 each group. Unpaired t-test, two-tailed, \* $p < 0.05$ , \*\* $p < 0.01$ .

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440

## 441 Literature

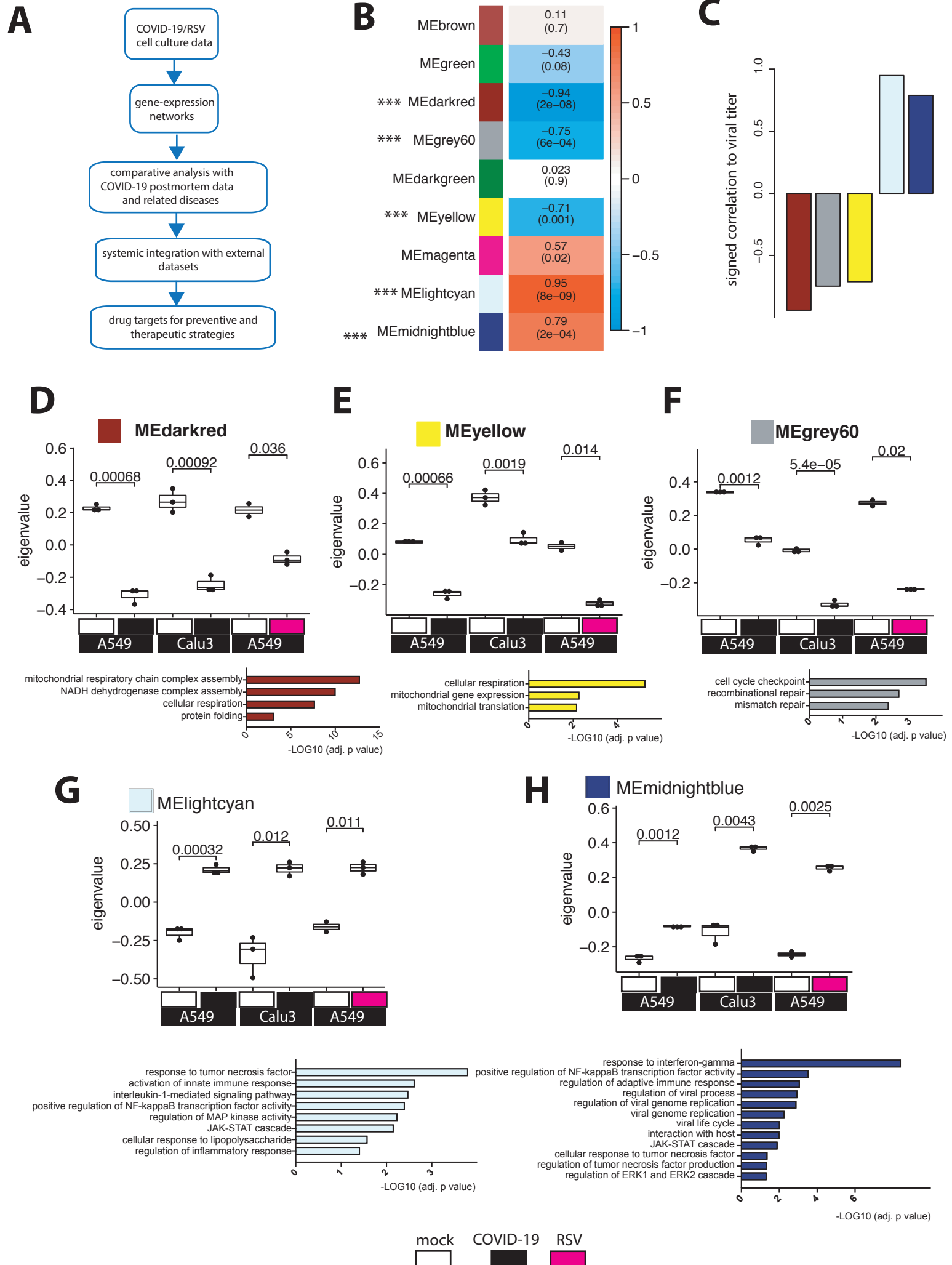
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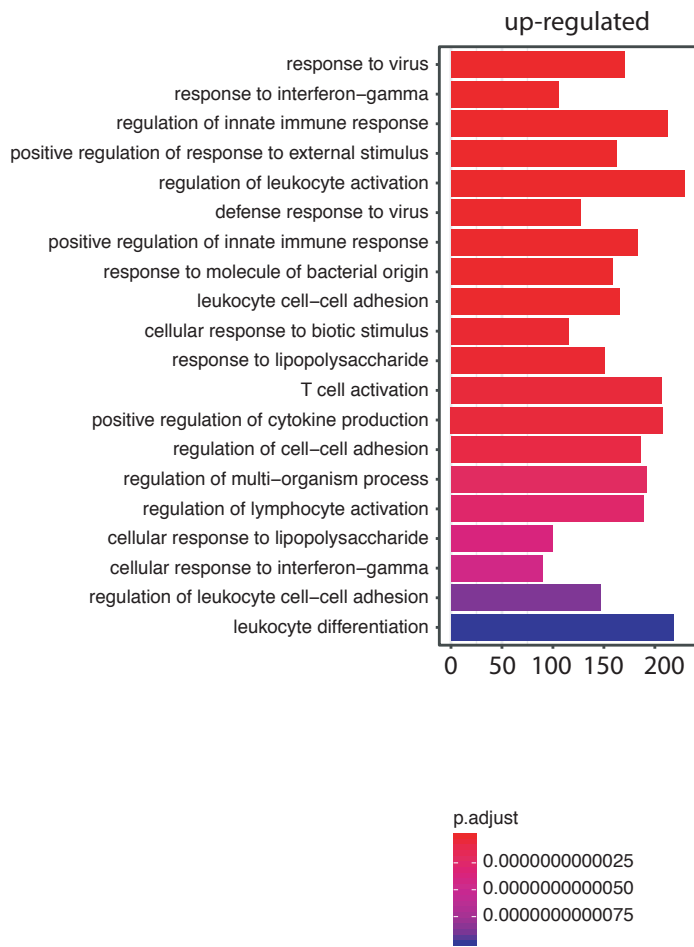
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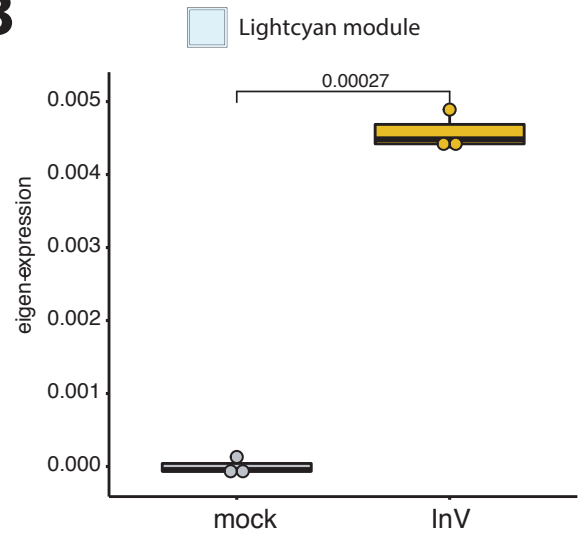




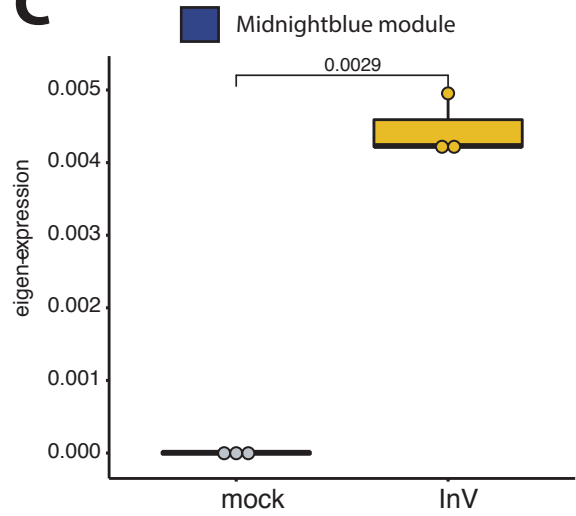
**A**



**B**



**C**



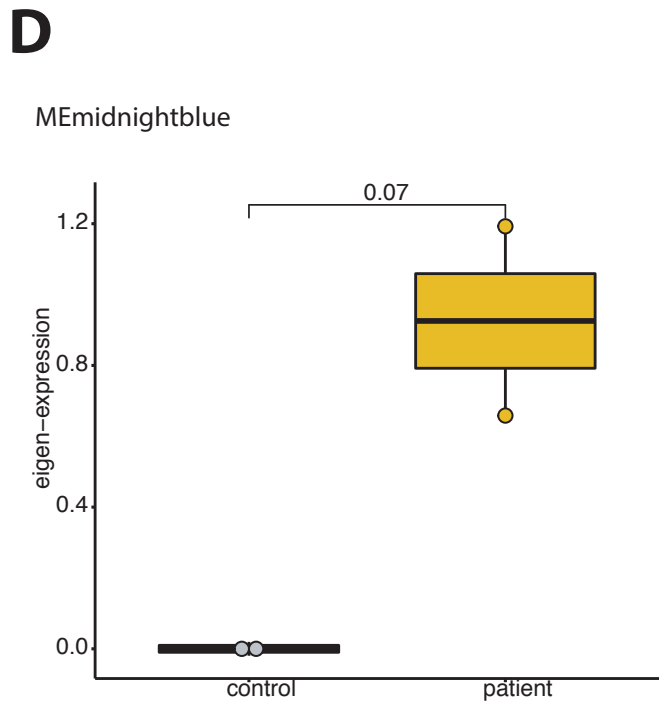
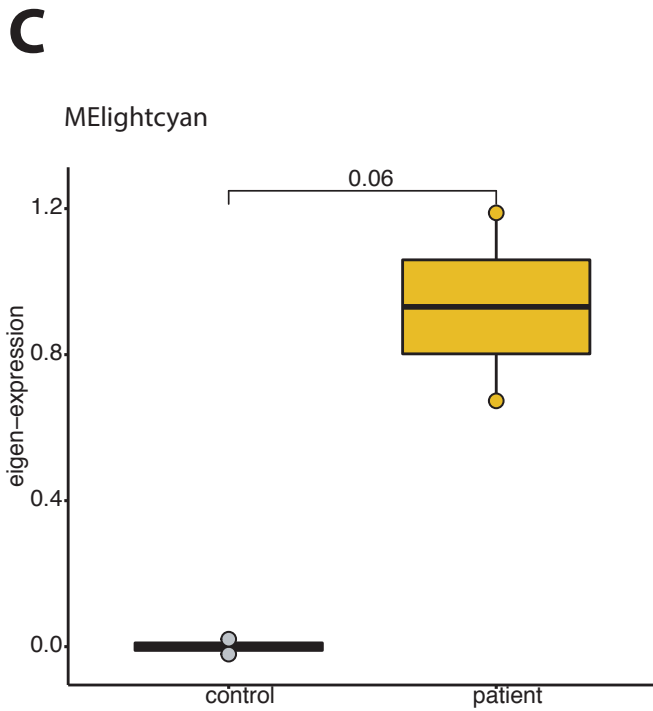
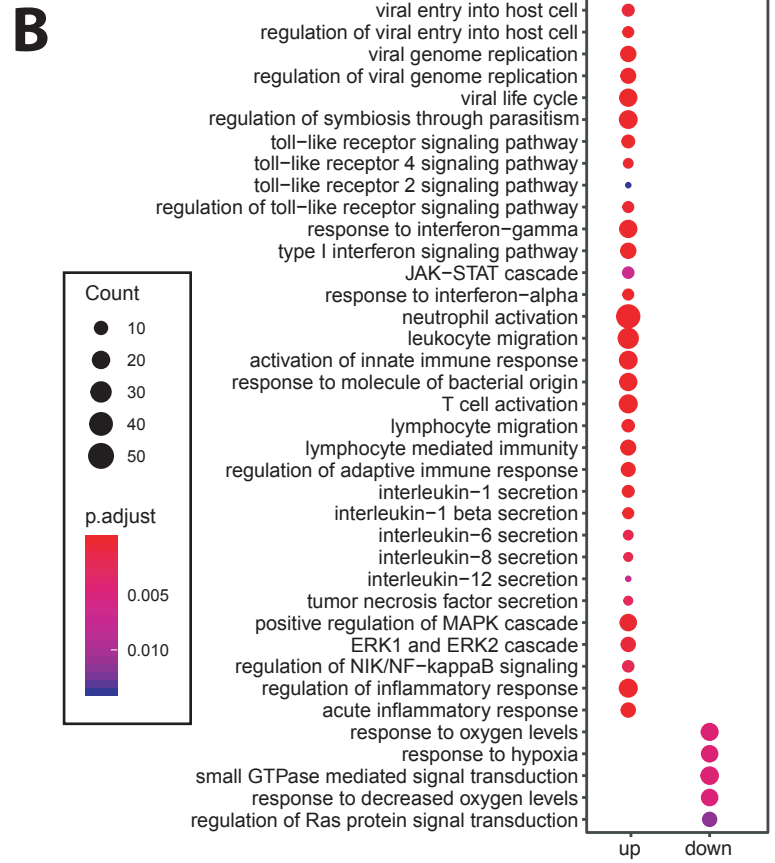
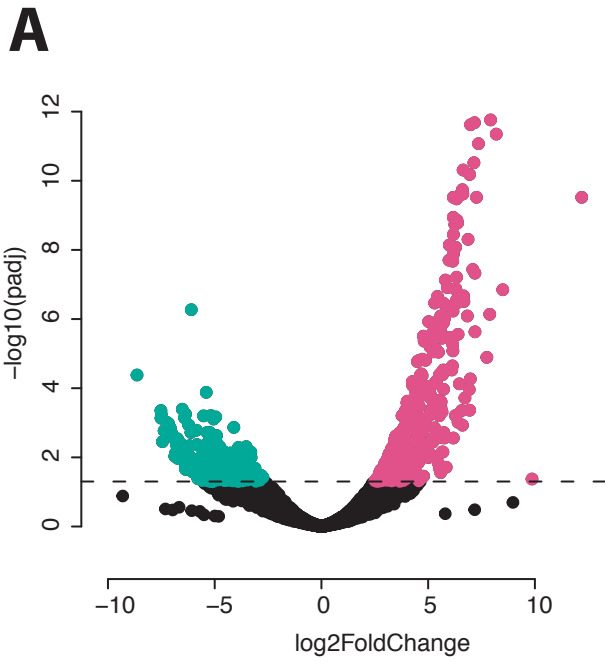
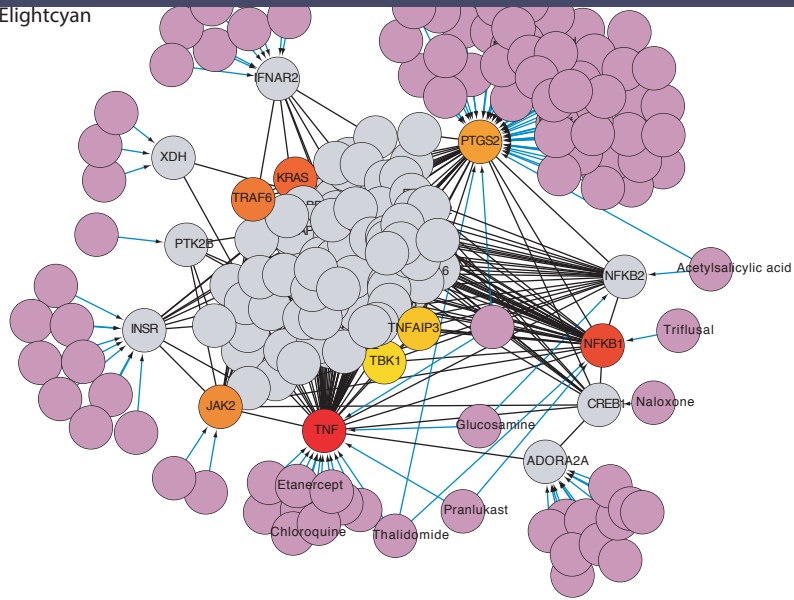


Fig. 4

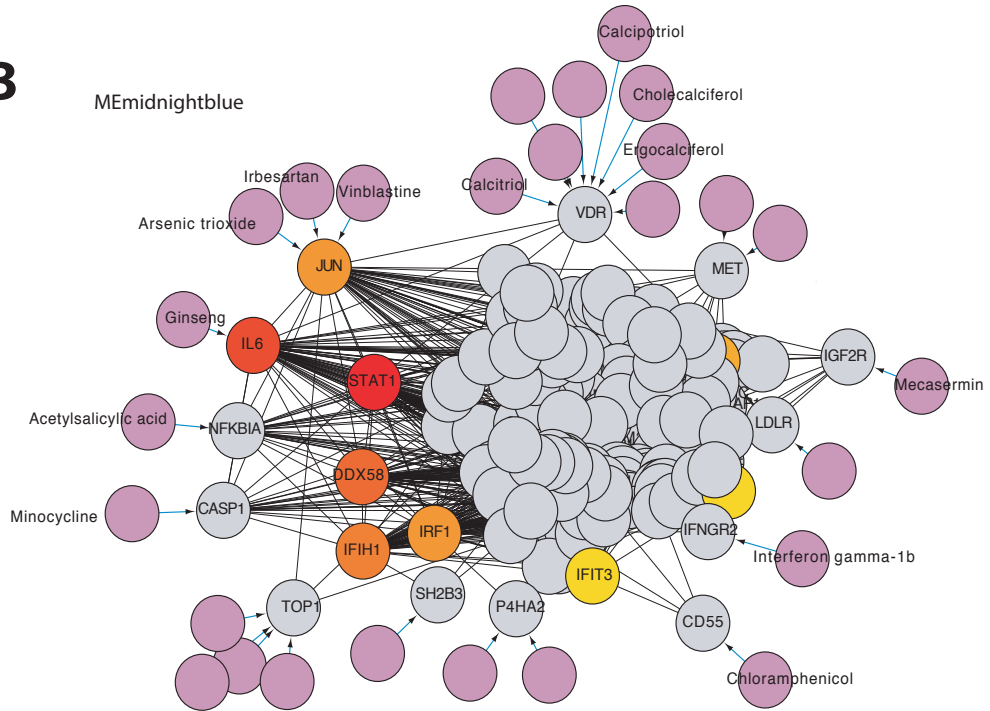
**A**

MElightcyan



**B**

MEmidnightblue



**C**

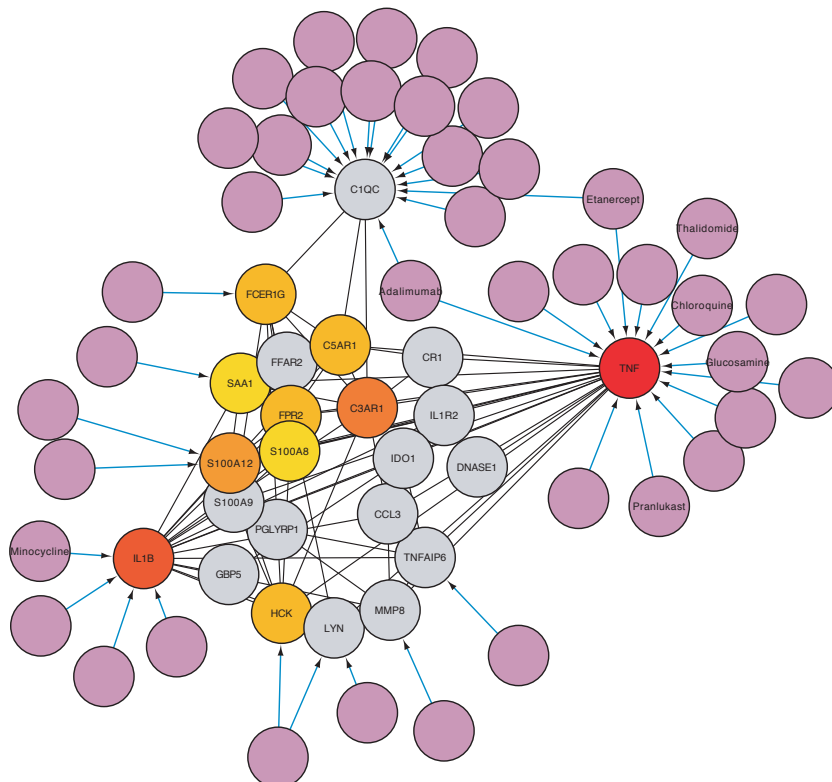


Fig. 5

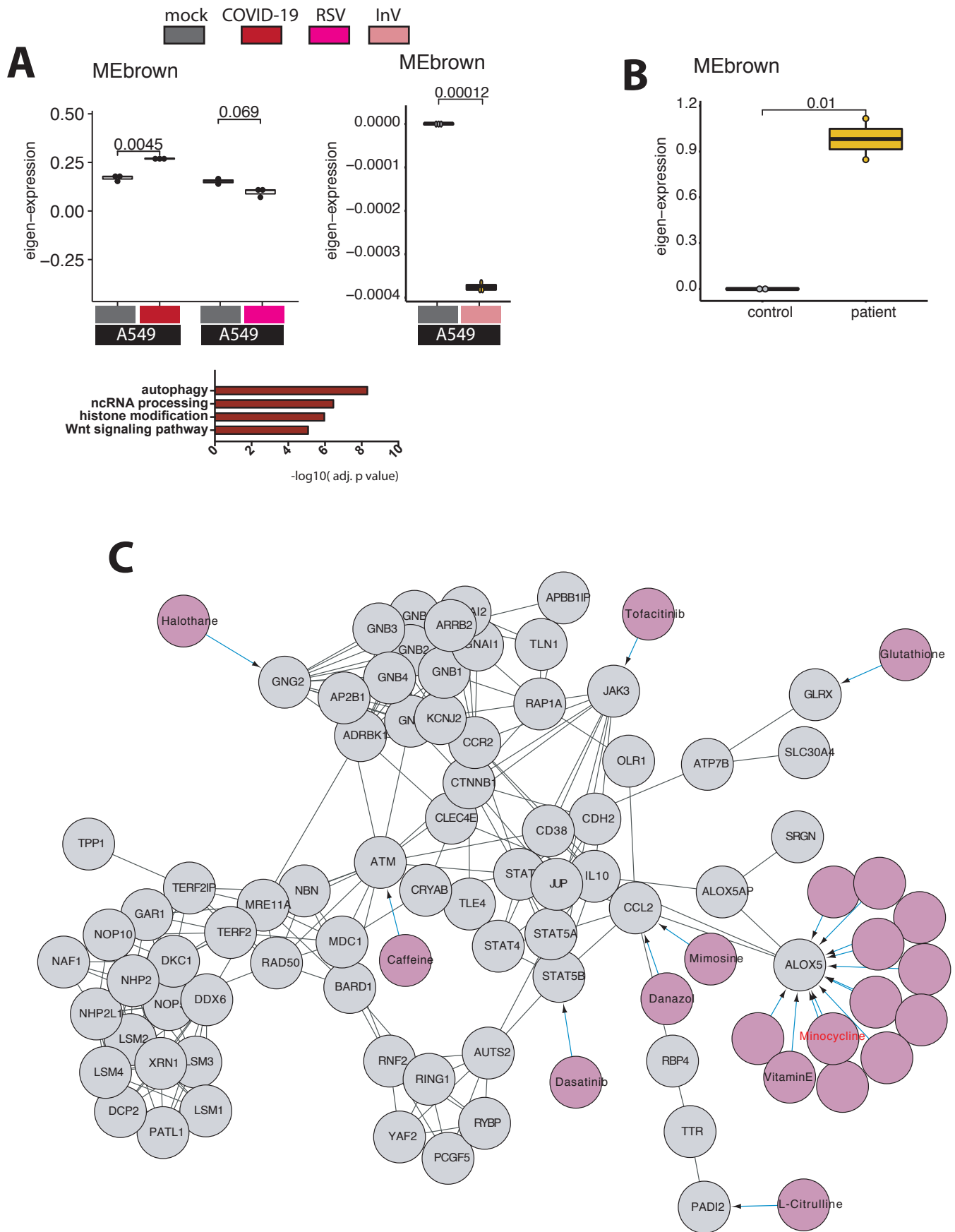


Fig. 6

