Early temporal dynamics of cellular responses to SARS-CoV-2

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Abstract

Two highly pathogenic human coronaviruses that cause severe acute respiratory syndrome
(SARS) and Middle East respiratory syndrome (MERS) have evolved proteins that can inhibit
host antiviral responses, likely contributing to disease progression and high case-fatality rates.
SARS-CoV-2 emerged in December 2019 resulting in a global pandemic. Recent studies have
shown that SARS-CoV-2 is unable to induce a robust type I interferon (IFN) response in human
cells, leading to speculation about the ability of SARS-CoV-2 to inhibit innate antiviral
responses. However, innate antiviral responses are dynamic in nature and gene expression levels
rapidly change within minutes to hours. In this study, we have performed a time series RNA-seq
and selective immunoblot analysis of SARS-CoV-2 infected lung (Calu-3) cells to characterize
early virus-host processes. SARS-CoV-2 infection upregulated transcripts for type I IFNs and
interferon stimulated genes (ISGs) after 12 hours. Furthermore, we analyzed the ability of
SARS-CoV-2 to inhibit type I IFN production and downstream antiviral signaling in human
cells. Using exogenous stimuli, we discovered that SARS-CoV-2 is unable to modulate *IFNβ*
production and downstream expression of ISGs, such as *IRF7* and *IFIT1*. Thus, data from our
study indicate that SARS-CoV-2 may have evolved additional mechanisms, such as masking of
viral nucleic acid sensing by host cells to mount a dampened innate antiviral response. Further
studies are required to fully identify the range of immune-modulatory strategies of SARS-CoV-
2.

Significance

Highly pathogenic coronaviruses that cause SARS and MERS have evolved proteins to
shutdown antiviral responses. The emergence and rapid spread of SARS-CoV-2, along with its
relatively low case-fatality rate have led to speculation about its ability to modulate antiviral

responses. We show that SARS-CoV-2 is unable to block antiviral responses that are mounted by

exogenous stimuli. Data from our study provide promising support for the use of recombinant

type I IFN as combination therapy to treat COVID-19 patients. Furthermore, our data also

suggest that the inability of SARS-CoV-2 to efficiently modulate antiviral responses may be

associated with its low case-fatality rate compared to other pathogenic CoVs that cause SARS

and MERS.

Main Text

Introduction

Severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) emerged in December

2019 to cause a global pandemic of coronavirus disease 2019 (COVID-19) (1). SARS-CoV-2

causes a respiratory infection with acute respiratory distress syndrome (ARDS) in severe cases.

Innate antiviral responses, which include type I interferons (IFNs) are the first line of defense

after a virus enters a cell (2). Cellular pattern recognition receptors (PRRs) recognize viral

nucleic acids and activate key cellular kinases, such as inhibitor of nuclear factor kappa-B kinase

subunit epsilon (IKKε) and TANK-binding kinase 1 (TBK1). These kinases activate

transcription factors, such as interferon regulatory factor 3 (IRF3) to stimulate downstream

production of type I IFNs (3).

To counteract host antiviral responses, viruses encode proteins that can modulate type I

IFN production and signaling (4, 5). Emerging pathogenic human coronaviruses, such as SARS-

CoV and Middle East respiratory syndrome (MERS)-CoV have evolved multiple proteins that

can inhibit type I IFN responses in human cells (6-10). Thus, to better understand SARS-CoV-2

pathogenesis, it is critical to identify the early dynamic interactions of SARS-CoV-2 and the type

I IFN response.

Data from *in vitro* and *in vivo* work have demonstrated the lack of induction of type I IFN

responses following SARS-CoV-2 infection (11). Interestingly, on the contrary, emerging data

from patients with mild and moderate cases of COVID-19 have demonstrated the presence of

type I IFN (12, 13). Thus, the inability to mount an effective IFN response to SARS-CoV-2 may

also be associated with underlying host factors, along with the duration and extent of viral

infection. Furthermore, it is unclear if SARS-CoV-2 is unable to stimulate a type I IFN response

or actively suppresses the response after initiating it in infected cells.

In this study, we have identified global early transcriptional responses that are initiated

during SARS-CoV-2 infection of human lung epithelial (Calu-3) cells at 0, 1, 2, 3, 6, and 12

hours post infection. SARS-CoV-2 infected cells mounted a type I IFN response between 6 and

12 hours post infection (hpi) and the degree of this response correlated with virus replication and

transcription. However, a high dose infection of SARS-CoV-2 is unable to modulate poly (I:C)-

induced IFNβ production and signaling. Furthermore, SARS-CoV-2 is unable to modulate

interferon stimulated gene (ISG) expression in response to exogenous IFNβ. Our study provides

insights into early host responses that are generated on infection with SARS-CoV-2 and the

inability of the virus to efficiently modulate these responses, which may explain the low case-

fatality rate of COVID-19. Furthermore, it is likely that comorbidities and deficiencies in type I

IFN responses are associated with severe outcomes in COVID-19 patients. In summary, our data

indicate that SARS-CoV-2 is inefficient in modulating type I IFN production and signaling when

cells are exogenously stimulated. Further investigations into the ability of SARS-CoV-2 to mask

its nucleic acid pathogen associated molecular pattern (PAMP) from cellular PRRs to generate a

dampened innate antiviral response is warranted.

Results

SARS-CoV-2 replication proceeds in a directional manner. The replication cycle of CoVs is

complex and involves the generation of sub-genomic RNA molecules, which in turn code for

mRNA that are translated into proteins (14, 15). To determine SARS-CoV-2 replication kinetics

in human cells using RNA-seq, we infected human lung epithelial cells (Calu-3) at a multiplicity

of infection (MOI) of 2. One hour post incubation, virus inoculum was replaced with cell growth

media and the clock was set to zero hours. We extracted and sequenced poly-A enriched RNA at

0, 1, 2, 3, 6 and 12 hours post infection (hpi). SARS-CoV-2 genome, sub-genomic RNA and

transcripts were detected in infected samples; viral transcript expression clustered based on post-

infection time using PCA (*SI Appendix,* Fig. S1). From our RNA-seq analysis, we were able to

detect high levels of expression of SARS-CoV-2 structural and accessory genes at the 3' end of

the genome as early as 0 hpi (Fig. 1A). Significant expression of *ORF1ab*, relative to 0 hpi was

detected at 6 hpi (Fig. 1A). SARS-CoV-2 nucleocapsid (*N*) gene was highly expressed relative to

other genes as early as 0 hpi (Fig 1B), with relative expression significantly increasing over time

($p = 1.4e-16$). The absolute expression of other genes increased over time with levels of $N > M >$

ORF10 > S > ORF1ab > ORF7a > ORF8 > ORF3a > ORF6 > E > ORF7b > ORF1a at 12 hpi

(Fig. 1B; *SI Appendix*, Table S1).

SARS-CoV-2 induces a mild type I IFN response. We analyzed the early host response

mounted by Calu-3 cells that were infected with SARS-CoV-2. Gene expression levels in these

cells clustered based on time-points via PCA (*SI Appendix*, Fig. S2). One hundred and twenty-

four genes were significantly (FDR-adjusted $p < 0.05$) differentially expressed in infected cells,

relative to mock infected cells in at least one time point post infection (absolute $\log_2$ fold-change

> 1) (Fig. 1D; *SI Appendix*, Table S2 and Fig. S3). The extent of antiviral gene expression at 12

hpi correlated with an increase in viral transcripts (*SI Appendix*, Fig. S3). Interestingly, at early

time points of 2 and 3 hpi, pathway enrichment analysis revealed numerous cellular processes

that were significantly downregulated in SARS-CoV-2 infected cells, relative to mock infected

cells (FDR-adjusted $p < 0.05$). These processes included RNA splicing, apoptosis, ATP

synthesis, and viral and host translation, while genes associated with viral processes, cell

adhesion and double-stranded RNA binding were upregulated in infected cells relative to mock

infected cells (Fig. 1C and *SI Appendix*, Figs. S4 and S5, Table S3). Cellular pathways associated

with type I IFN production and signaling, along with OAS/TRAF-mediated antiviral responses

were upregulated at 12 hpi (Figs. 1C and 1D). Consistent with other reports, transcript levels for

IFNβ1 and *IFNλ1* were significantly upregulated at 12 hpi with SARS-CoV-2 at a high MOI of 2

(Fig. 1E) (11). Transcript levels of *IFNλ2* and *IFNλ3* also increased at 6 and 12 hpi, but the

levels did not reach significance relative to mock infected cells at these time points (Fig. 1E). At

least 19 well-studied antiviral ISGs were upregulated in infected cells, relative to mock infected

cells at 12 hpi, including interferon induced protein with tetratricopeptide repeats 1 (*IFIT1*),

interferon regulatory factor 7 (*IRF7*), 2'-5-oligoadenylate synthetase 2 (*OAS2*) and MX dynamin

GTPase 1 (*MX1*) (Fig. 1F and *SI Appendix*, Fig. S6 and Table S2). Genes associated with

structural molecule activity, cell adhesion and exocytosis were downregulated in SARS-CoV-2

infected cells, relative to uninfected cells at 12 hpi (*SI Appendix*, Fig. S5).

SARS-CoV-2 is unable to modulate type I IFN gene expression induced by an exogenous

stimulus. Coronaviruses, such as those that cause SARS and MERS have evolved multiple

proteins that can modulate type I IFN expression (7-10, 16, 17). To confirm that SARS-CoV-2

infection is sufficient to induce type I IFN and ISG responses in Calu-3 cells, we infected the

cells with SARS-CoV-2 and assessed transcript levels of *IFNβ*, *IRF7* and *IFIT1* by quantitative

polymerase chain reaction (qPCR). *IFNβ* induction was observed 12 hpi in SARS-CoV-2

infected cells, relative to mock-infected cells (Fig. 2A). Consistent with the upregulation of *IFNβ*

transcripts in SARS-CoV-2 infected cells, transcript levels for ISGs, such as *IRF7* and *IFIT1*

were also upregulated 12 hpi (Figs. 2B and 2C).

Next, to identify if SARS-CoV-2 is able to modulate type I IFN responses mounted

against an exogenous stimulus, we infected Calu-3 cells with SARS-CoV-2 for 12 hours at a

MOI of 2 and stimulated these cells with exogenous double-stranded RNA [poly(I:C)] for 6

hours. We measured the levels of *IFNβ* transcripts in these cells by qPCR. Poly(I:C) transfection

alone induced significantly higher levels of *IFNβ* transcripts relative to mock transfected cells

(Fig. 2D). Similar to that shown in Fig. 2A, SARS-CoV-2 infection alone also induced high

levels of *IFNβ* transcripts relative to mock infected cells (Fig. 2D). However, SARS-CoV-2

infection-induced levels of *IFNβ* transcripts were significantly lower compared to both poly(I:C)

transfected cells and SARS-CoV-2 infected + poly(I:C) transfected cells. Interestingly, there was

no significant difference in *IFNβ* transcript levels between poly(I:C) transfected and SARS-CoV-

2 infected + poly(I:C) transfected cells (Fig. 2D). In fact, there was an increasing trend in *IFNβ*

transcript levels in SARS-CoV-2 infected + poly(I:C) transfected cells relative to cells that were

transfected with poly(I:C) alone; however, the data were not significant at this time point.

To support our observations with *IFNβ* transcripts in SARS-CoV-2 infected and/or

poly(I:C) transfected cells, we also quantified the levels of ISG transcripts, such as *IRF7* and

IFIT1 in these cells. Poly(I:C) transfection alone induced significantly higher levels of *IRF7* and

IFIT1 transcripts relative to mock transfected cells (Figs. 2E and 2F). Similar to that shown in

Figs. 2B and 2C, SARS-CoV-2 alone also induced high levels of *IRF7* and *IFIT1* transcripts

relative to mock infected cells (Figs. 2E and 2F). However, SARS-CoV-2 infection-induced

levels of *IRF7* and *IFIT1* transcripts were significantly lower compared to both poly(I:C)

transfected cells and SARS-CoV-2 infected + poly(I:C) transfected cells. Notably, *IRF7* and

IFIT1 transcript levels in SARS-CoV-2 infected + poly(I:C) transfected cells were significantly

higher than levels in cells that were transfected with poly(I:C) alone (Figs. 2E and 2F).

To corroborate our gene expression studies, we repeated our experiments and performed

immunoblots for SARS-CoV-2 N, IFIT1 and glyceraldehyde 3-phosphate dehydrogenase

(GAPDH). Poly(I:C) transfection induced low levels of IFIT1 in Calu-3 cells, while SARS-CoV-

2 infection alone was unable to induce detectable levels of IFIT1 in our immunoblots (Fig. 2G).

SARS-CoV-2 infection + poly(I:C) transfection also induced low, but detectable levels of IFIT1

(Fig. 2G). We confirmed SARS-CoV-2 infection in these cells by detecting N protein in the

samples (Fig. 2G).

SARS-CoV-2 is unable to modulate type I IFN signaling. SARS-CoV and MERS-CoV

proteins can also modulate downstream IFN signaling to restrict the production of ISGs (6). To

determine if SARS-CoV-2 can modulate type I IFN signaling in response to exogenous IFNβ

treatment, we infected Calu-3 cells for 12 hours at a MOI of 2 and stimulated these cells with

recombinant human IFNβ for 6 hours. We monitored gene expression levels of *IRF7* and *IFIT1*

in these cells by qPCR. For this assay, we developed and utilized recombinant human IFNβ1. To

demonstrate the antiviral efficacy of our recombinant IFN, we pre-treated human fibroblast

(THF) cells with IFNβ1, followed by RNA and DNA virus infections. Pre-treatment of THF

cells with IFNβ1 inhibited the replication of herpes simplex virus (HSV), vesicular stomatitis

virus (VSV) and H1N1 in a dose-dependent manner (*SI Appendix*, Fig. S7).

Exogenous IFNβ treatment alone significantly upregulated transcript levels of *IRF7* and

IFIT1 relative to mock treated Calu-3 cells (Figs. 2H and 2I). Consistent with our RNA-seq data,

SARS-CoV-2 infection alone induced significant levels of *IRF7* and *IFIT1* transcripts (Figs. 2H

and 2I). However, SARS-CoV-2 induced *IRF7* and *IFIT1* transcript levels were significantly

lower compared to levels in both IFNβ treated cells and SARS-CoV-2 infected + IFNβ treated

cells (Figs. 2H and 2I). Transcript levels of *IRF7* and *IFIT1* in IFNβ treated cells and SARS-

CoV-2 infected + IFNβ treated cells were not significantly different (Figs. 2H and 2I).

Finally, we repeated the experiments with exogenous IFNβ treatment and performed

immunoblots to determine if SARS-CoV-2 can modulate type I IFN-mediated upregulation of

IFIT1. Exogenous IFNβ treatment alone induced a robust expression of IFIT1 (Fig. 2J). SARS-

CoV-2 infection alone was not sufficient for a visible increase in IFIT1 expression in our

immunoblots (Fig. 2J). Interestingly, IFNβ treatment after 12 hours of high dose infection (MOI

= 2) of SARS-CoV-2 also induced a robust expression of IFIT1 (Fig. 2J). We confirmed SARS-

CoV-2 infection in these cells by detecting N protein (Fig. 2J).

Discussion

Discussion

SARS-CoV-2 emerged in December 2019 and has since caused a global pandemic of COVID-19

(1, 18). Clinical observations and emerging data from *in vitro* and *in vivo* studies have

demonstrated the limited ability of SARS-CoV-2 to induce type I IFNs (11). However, the ability

of SARS-CoV-2 to modulate IFN production and signaling remains unknown. Furthermore, gene

expression kinetics of SARS-CoV-2, along with time-associated host responses have not been

described. In this study, we have identified early virus-host interactions using a time-series

RNA-seq experiment. Consistent with other studies (11), we demonstrate that a high dose of

SARS-CoV-2 induces a type I IFN response; however, our data show that SARS-CoV-2 is

unable to modulate cellular type I IFN production and signaling that are mounted in response to

exogenous stimuli.

RNA-seq analysis of poly(A)-enriched RNA allowed us to map the progression of SARS-

CoV-2 replication and transcription in Calu-3 cells. As observed with other coronaviruses (19-

21), SARS-CoV-2 replicated and transcribed sub-genomic RNA and mRNA in a directional

manner (Figs. 1A and 1B). SARS-CoV-2 *N* gene was highly expressed as early as 0 hpi. High

MOI of SARS-CoV-2 produced cytopathic effects (CPE) in Calu-3 cells at later time points,

which made it difficult to reliably assess host gene expression relative to unstable levels of

house-keeping genes.

Coronaviruses, including highly pathogenic SARS-CoV, MERS-CoV and porcine

epidemic diarrhea virus (PEDV) have evolved proteins that can efficiently modulate type I IFN

responses (7-10, 16, 17, 22, 23). The recently demonstrated inability of SARS-CoV-2 to

stimulate the expression of robust amounts of type I IFNs (11) may be associated with its ability

to mask the detection of viral RNA by cellular PRRs and/or its ability to inactivate cellular

mechanisms involved in type I IFN upregulation. Data from our studies show that SARS-CoV-2

is indeed unable to stimulate high levels of *IFNβ* transcripts relative to poly(I:C) (Fig. 2D).

However, SARS-CoV-2 is unable to efficiently shutdown poly(I:C)-mediated upregulation of

IFNβ transcripts and downstream ISGs (Figs. 2D-F). In fact, poly(I:C) + SARS-CoV-2 induced

higher levels of ISG transcripts relative to poly(I:C) alone. Thus, our data hint at additional

mechanisms that SARS-CoV-2 may have evolved to mitigate the recognition of viral PAMPs by

cellular PRRs. MERS-CoV protein 4a interferes with RIGI and MDA5-mediated sensing of viral

RNA (7). Murine hepatitis virus (MHV) encodes an endoribonuclease that cleaves poly-uridine

residues in the viral genome, thus limiting the activation of cellular PRRs (24). Endoribonuclease

deficient mouse CoVs induce a robust type I IFN response and can only replicate in cells that are

IFN deficient (25, 26). It is possible that SARS-CoV-2 uses a similar strategy to limit the

detection of its nucleic acid by cellular PRRs, thus leading to a dampened antiviral IFN response

in these cells. Future studies will identify the full breadth of strategies deployed by SARS-CoV-2

to modulate innate antiviral responses.

A recent study has identified the ability of SARS-CoV-2 to replicate to higher titers in the

upper respiratory tract, including nasal cells (27). Hou *et al.* have shown that high levels of virus

replication in nasal cells is associated with high levels of angiotensin-converting enzyme 2

(ACE2) receptor expression in these cells, relative to cells in the lower respiratory tract (27).

Studies have also shown that rhinovirus (common cold virus) replicates to higher titers in nasal

cells due to diminished temperature-dependent innate antiviral responses in these cells (28).

Thus, the inability of SARS-CoV-2 to induce a robust type I IFN response, coupled with the

dampened ability of nasal cells to potentiate an innate immune response may lead to high levels

of virus replication in the upper respiratory tract, as observed in COVID-19 patients (29).

In conclusion, our study demonstrates that SARS-CoV-2 is a weak stimulator of type I

IFN responses in infected human cells, relative to the more potent form of PAMP, poly(I:C).

However, our data suggest that the lack of a robust type I IFN response in SARS-CoV-2 infected

cells is likely due to the inability of the cells to recognize viral PAMPs, such as double-stranded

RNA. The inability of SARS-CoV-2 to modulate downstream IFN responses is promising for the

development of IFNβ as a treatment or post-exposure prophylactic. Clinical trials for

combination IFNβ therapy against MERS-CoV are currently ongoing (30). IFNβ, in combination

with lopinavir-ritonavir and ribavirin has been used with promising results in COVID-19 patients

(31). While it is possible that over-expressing viral proteins may identify interactions that can

modulate type I IFN production in human cells, we did not observe these effects when cells were

infected with a high MOI of SARS-CoV-2 and stimulated exogenously. Future studies will shed

more light on the full breadth of immune modulatory capabilities of SARS-CoV-2.

Materials and Methods

Cells and viruses. Vero E6 cells (African green monkey cells; ATCC, Manassas, VA, USA)

were maintained in Dulbecco's modified Eagle's media (DMEM) supplemented with 10% fetal

bovine serum (FBS; Sigma-Aldrich), 1x L-Glutamine and Penicillin/Streptomycin (Pen/Strep;

Corning, VWR, Mississauga, ON, Canada). Calu-3 cells (human lung adenocarcinoma derived;

ATCC) were cultured as previously mentioned (32). THF cells (human telomerase life-extended

cells; from Dr. Victor DeFilippis' lab) were cultured as previously mentioned (33). *Drosophila*

S2 cells (ThermoFisher Scientific, Waltham, MA, USA) were cultured in Schneider's

Drosophila medium supplemented with 10% FBS (Sigma-Aldrich) as recommended by the

manufacturer and cells were incubated at 28°C. Stocks of genetically engineered vesicular

stomatitis virus (VSV-GFP) carrying a green fluorescent protein (GFP) cassette (34) were stored

at -80°C. H1N1 (A/Puerto Rico/8/1934 mNeon – 2A-HA) stocks were obtained from Dr.

Matthew Miller's laboratory. HSV-GFP stocks were generated and maintained as mentioned

previously (35). SARS-CoV-2/SB3 virus stocks were propagated on Vero E6 cells and validated

by next generation sequencing (36). Virus stocks were thawed once and used for an experiment.

A fresh vial was used for each experiment to avoid repeated freeze-thaws. VSV-GFP, HSV-GFP

and H1N1 infections were performed at an MOI of 1. SARS-CoV-2 infections were performed at

an MOI of 2. Experiments with SARS-CoV-2 were performed in a BSL3 laboratory and all

procedures were approved by institutional biosafety committees at McMaster University and the

University of Toronto.

RNA-Seq

RNA was isolated from cells using RNeasy Mini kit (Qiagen, Hilden, Germany). Sequencing

was conducted at the McMaster Genomics Facility, Farncombe Institute at McMaster University.

Sample quality was first assessed using a Bioanalyzer (Agilent 2100 Bioanalyzer G2938C,

Aligent RNA 6000 Nano Kit, Agilent; Santa Clara, CA, USA), then enriched (NEBNext

Poly(A) mRNA Magnetic Isolation Module; NEB, Ipswich, MA, USA). Library preparations

were conducted (NEBNext Ultra II Directional RNA Library Prep Kit; NEB, Ipswich, MA,

USA) and library fragment size distribution was verified (Agilent TapeSection D1000; Agilent,

Santa Clara, CA, USA). Libraries were quantified by qPCR, pooled in equimolar amounts, and

qPCR and fragment size distribution verification was conducted again. Libraries were then

sequenced on an Illumina HiSeq 1500 across 3 HiSeq Rapid v2 flow cells in 6 lanes (Illumina;

San Diego, CA, USA) using a paired-end, 2x50 bp configuration, with onboard cluster

generation averaging 30.8M clusters per replicate (minimum 21.9M, maximum 46.0M).

Transcript quantification and differential expression analysis

Sequence read quality was checked with FastQC

(https://www.bioinformatics.babraham.ac.uk/projects/fastqc/), with reads subsequently aligned to

the human reference transcriptome (GRCh37.67) obtained from the ENSEMBL database (37) ,

indexed using the 'index' function of Salmon (version 0.14.0) (38) with a k-mer size of 31.

Alignment was performed using the Salmon 'quant' function with the following parameters: "-l

A --numBootstraps 100 --gcBias --validateMappings". All other parameters were left to defaults.

Salmon quantification files were imported into R (version 3.6.1) (39) using the tximport library

(version 1.14.0) (40) with the 'type' option set to 'salmon'. Transcript counts were summarized

at the gene-level using the corresponding transcriptome GTF file mappings obtained from

ENSEMBL. Count data was subsequently loaded into DESeq2 (version 1.26.0) (41) using the

'DESeqDataSetFromTximport' function. In order to determine time/treatment dependent

expression of genes, count data was normalized using the 'estimateSizeFactors' function using

the default 'median ratio method' and output using the 'counts' function with the 'normalized'

option.

For subsequent differential-expression analysis, a low-count filter was applied prior to

normalization, wherein a gene must have had a count greater than 5 in at least three samples in

order to be retained. Using all samples, this resulted in the removal of 12,980 genes for a final set

of 15,760 used. Principal Component Analysis (PCA) of samples across genes was performed

using the 'vst' function in DESeq2 (default settings) and was subsequently plotted with the

ggplot2 package in R (42). Differential expression analyses were carried out with three designs:

(a) the difference between infection/control status across all timepoints, (b) considering the

effects of post-infection time (i.e. the interaction term between time and infection status) and (c)

the difference between infection/control status at individual timepoints. (a) and (b) were

performed using the 'DESeq' function of DESeq2 using all samples, with results subsequently

summarized using the 'results' function with the 'alpha' parameter set to 0.05; p-values were

adjusted using the Benjamini-Hochberg FDR method (43), with differentially expressed genes

filtered for those falling below an adjusted p-value of 0.05. For (c), infected/mock samples were

subset to individual timepoints, with differential expression calculated using DESeq as described

above. Additionally, given the smaller number of samples at individual time-points, differential-

expression analysis was also performed with relaxation of the low-count filter described above,

with results and p-value adjustments performed as above.

Viral transcript quantification

Paired-end sequencing reads were mapped to CDS regions of the SARS-CoV-2 genomic

sequence (Assembly ASM985889v3 - GCF_009858895.2) obtained from NCBI, indexed using

the 'index' function of Salmon (version 0.14.0) (38) with a k-mer size of 31. Subsequently, reads

were aligned using the Salmon 'quant' function with the following parameters: "-l A --

numBootstraps 100 --gcBias --validateMappings". All other parameters were left to defaults.

Salmon quantification files were imported into R (version 3.6.1) (39) using the tximport library

(version 1.14.0) (40) with the 'type' option set to 'salmon'. All other parameters were set to

default. Transcripts were mapped to their corresponding gene products via GTF files obtained

from NCBI. Count data was subsequently loaded into DESeq2 (version 1.26.0) (41) using the

'DESeqDataSetFromTximport' function. Principal Component Analysis (PCA) of samples

across viral genes was performed using the 'vst' function in DESeq2 (default settings) and was

subsequently plotted with the ggplot2 package in R (42). As viral transcript levels increased over

time post-infection, we first converted non-normalized transcript counts to a $\log_2$ scale, and

subsequently compared these across time-points (Fig. 1B; *SI Appendix*, Table S1). To look at the

changes in the expression of viral transcripts relative to total viral expression as a function of

post-infection time, normalized transcript counts were used to perform differential-expression

analysis with DESeq2. Results and p-value adjustments were performed as described above.

In order to compare host/viral expression patterns, normalized transcript counts from

infected samples were compared with either normalized or non-normalized viral transcript

counts (from the same sample) across time-points. For each viral transcript (n = 12), all host

genes (n = 15,760, after filtering described above) were tested for correlated expression changes

across matched infected samples (n = 18, across 5 time-points) using Pearson's correlation

coefficient (via the cor.test function in R). Correlation test *p*-values were adjusted across all-by-

all comparisons using the Benjamini-Hochberg FDR method, and gene-transcript pairs at

adjusted p-value < 0.05 were retained. To account for possible effects of cellular response to

plate incubation, viral transcript abundance was averaged at each time-point and compared to

host transcript abundance similarly averaged at each time-point for non-infected samples;

correlation testing was done all-by-all for n = 5 data-points. Host genes that correlated with viral

transcription in mock samples across time were removed from subsequent analyses; to increase

stringency, mock correlation was defined using un-adjusted p-value < 0.05. Host genes were

sorted by correlation coefficient (with any given viral transcript), with the top 100 unique genes

retained for visualization. Normalized host transcript counts were z-score transformed per-gene

using the 'scale' function in R, with normalized/un-normalized viral transcript counts similarly

transformed per-transcript. Resulting z-score expression heatmaps were generated using the

ComplexHeatmap library in R (version 2.2.0) (44). Heatmaps were generated for normalized/un-

normalized viral transcript counts, given the different information revealed by absolute and

relative viral expression patterns.

Viral genome mapping

Paired-end RNA-seq reads were filtered for quality control with Trim Galore! (version

0.6.4_dev) (45) and mapped to the SARS-CoV-2 reference sequence (NC_045512.2) with the

Burrow-Wheeler Aligner (46), using the BWA-MEM algorithm (47). Output SAM files were

sorted and compressed into BAM files using Samtools (version 1.10) (48). Read coverage

visualization was performed from within the R statistical environment (version 4.0.0) (39) using

the "scanBam" function from the Rsamtools R package (version 1.32.0) to extract read coverage

data and the ggplot2 R package (version 3.3.0) (42) to plot read coverage histograms (using 300

bins across the SARS-CoV-2 sequence).

Cellular pathway enrichment analysis

To determine cellular pathways that were associated with differentially expressed genes (DEGs),

the ActivePathways R (version 1.0.1) (49) package was utilized to perform gene-set based

pathway enrichment analysis. DEGs at each time point were treated as an independent set for

enrichment analysis. Fisher's combined probability test was used to enrich pathways after *p*-

value adjustment using Holm-Bonferroni correction. Pathways of gene-set size less than 5 and

greater than 1000 were excluded. Only pathways enriched at individual time-points were

considered for downstream analysis; pathways enriched across combined timepoints as

determined by ActivePathways Brown's *p*-value merging method were filtered out. Visualization

of enriched pathways across timepoints was done using Cytoscape (version 3.8.0) (50) and the

EnrichmentMap plugin (version 3.2.1) (51), as outlined by Reimand *et al*. (52). Up-to-date Gene-

Matrix-Transposed (GMT) files containing information on pathways for the Gene Ontology

(GO) Molecular Function (MF), GO Biological Process (BP) (53) and REACTOME (54)

pathway databases were utilized with ActivePathways. Only pathways that were enriched at

specific time points were considered. Bar plots displaying top ActivePathway GO terms and

REACTOME enrichments for infection versus mock were plotted using the ggplot2 R package

(version 3.2.1) for 1, 2, 3, and 12 hour time points. Zero and 6 hour time points were omitted due

to a lack of sufficient numbers of differentially expressed genes required for functional

enrichment analysis.

Poly(I:C) transfection and IFNβ treatment. Calu-3 cells were mock transfected with 4 μl of

lipofectamine 3000 (ThermoFisher Scientific) only or transfected with 100 ng of poly(I:C)

(InvivoGen, San Diego, CA, USA). Recombinant human IFNβ1 was generated using *Drosophila*

Schneider 2 (S2) cells following manufacturer's recommendation and by using ThermoFisher

Scientific's *Drosophila* Expression system. As a control, recombinant GFP was also generated

using the same protocol and used for mock treated cells. For VSV-GFP, HSV-GFP and H1N1-

mNeon infections, cells were treated with increasing concentrations of IFNβ1 or GFP (control).

SARS-CoV-2 infected cells were treated with 200 μg/ml of IFNβ1 or GFP.

Quantitative PCR. Calu-3 cells were seeded at a density of 3 x 10^5 cells/well in 12-well plates.

Cells were infected with SARS-CoV-2 and RNA extraction was performed using RNeasy® Mini

Kit (Qiagen, Hilden, Germany) according to manufacturer's protocol. 200 ng of purified RNA

was reverse transcribed using iScript™ gDNA Clear cDNA Synthesis Kit (Bio-Rad, Hercules,

CA, USA). Quantitative PCR reactions were performed with TaqMan™ Universal PCR Master

Mix (Applied Biosystems, Foster City, CA, USA) using pre-designed Taqman gene expression

assays (ThermoFisher Scientific) for *IFNβ1* (catalog #4331182), *IRF7* (catalog #4331182), *IFIT1*

(catalog #4331182) and *GAPDH* (catalog #4331182) according to manufacturer's protocol.

Relative mRNA expression was normalized to *GAPDH* and presented as 1/ΔCt.

Immunoblots. Calu-3 cells were seeded at a density of 3 x 10^5 cells/well in 12-well plates. Cells

were infected with SARS-CoV-2 at an MOI of 2. Control cells were sham infected. Twelve

hours post infection, cells were transfected or treated with poly(I:C) or IFNβ, respectively. Cell

lysates were harvested for immunoblots and analyzed on reducing gels as mentioned previously

(33). Briefly, samples were denatured in a reducing sample buffer and analyzed on a reducing

gel. Proteins were blotted from the gel onto polyvinylidene difluoride (PVDF) membranes

(Immobilon, EMD Millipore, Burlington, MA, USA) and detected using primary and secondary

antibodies. Primary antibodies used were: 1:1000 mouse anti-GAPDH (EMD Millipore;

Catalogue number: AB2302; RRID: AB_10615768), 1:1000 mouse anti-SARS-CoV-2 N

(ThermoFisher Scientific; Catalogue number: MA5-29981; RRID: AB_2785780 and 1:1000

rabbit anti-IFIT1 (ThermoFisher Scientific; Catalogue number: PA3-848; RRID: AB_1958733).

Secondary antibodies used were: 1:5000 donkey anti-rabbit 800 (LI-COR Biosciences, Lincoln,

NE, USA; Catalogue number: 926-32213; RRID: 621848) and 1:5000 goat anti-mouse 680 (LI-

COR Biosciences; Catalogue number: 925-68070; RRID: AB_2651128). Blots were observed

and imaged using Image Studio (LI-COR Biosciences) on the Odyssey CLx imaging system (LI-

COR Biosciences).

Antiviral bioassay. THF cells were pre-treated or mock treated with recombinant human IFNβ,

followed by VSV-GFP, HSV-GFP or H1N1-mNeon infection at an MOI of 1. Infected cells were

incubated at 37°C for 1 hour with gentle rocking every 15 minutes. After 1 hr, virus inoculum

was aspirated and Minimum Essential Medium (MEM) with Earle's salts (Sigma) containing 2%

FBS and 1% carboxymethyl cellulose (CMC; Sigma) was added on the cells. Cells were

incubated for 19 hours at 37°C and green fluorescent protein (GFP) or mNeon levels were

measured using a typhoon scanner (Amersham, GE Healthcare, Chicago, IL, USA).

Statistics. Statistical analyses for RNA-seq data were performed in R and are mentioned under

the respective RNA-seq analyses sections. All other statistical calculations were performed in

GraphPad Prism (version 8.4.2; www.graphpad.com) using two-tailed paired t-test. Significance

values are indicated in the figures and figure legends. p*<0.05, **<0.01, ***<0.001 and

****<0.0001.

Data Availability

456   The DESeq2 normalized transcript counts for all genes with RNA-Seq data, significant or

457 otherwise, plus the raw sequencing FASTQ reads have been deposited into the Gene Expression

458 Omnibus (GEO) database with NCBI GEO accession number GSE151513.

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Figures and Figure Legends

Figure 1. SARS-CoV-2 and cellular gene expression. Calu-3 cells were infected with SARS-CoV-2 at an MOI of 2. Virus and host gene expression were determined using time-series RNA-seq analyses. **(A)** SARS-CoV-2 gene expression over 12 hours (n = 3/time point). The genome organization of SARS-CoV-2 is indicated above in pink. **(B)** Major SARS-CoV-2 gene expression levels at different times post infection (n = 3/time point). **(C)** Cellular processes that are down or upregulated at different times post infection. The size of the circles represents the number of genes that are down or upregulated at different times after infection (n = 3/time point). **(D)** Cellular genes (n = 124) that are significantly up or downregulated (FDR-adjusted $p<0.05$; $|\log_2 FC| > 1$) in SARS-CoV-2 infected cells, relative to mock infected cells at different times post infection. Transcript levels are shown as z-score normalized expression (scaled by gene). **(E)** Transcript abundance of type I interferon (IFN) genes (*IFNβ* and *IFNλ1-3*) in mock infected and SARS-CoV-2 infected Calu-3 cells at different times (Mean ± SD; n = 3). **(F)** Transcript abundance of representative interferon stimulated genes (ISGs) in mock infected and SARS-CoV-2 infected Calu-3 cells at different times (Mean ± SD; n = 3). For Figs. A-D, statistical analysis was performed in R (see methods). For Figs. E and F, statistical significance was calculated using two-tailed paired t-test. hpi, hours post infection. $p^*<0.05$, $^{**}<0.01$.

Figure 2. SARS-CoV-2 is unable to modulate type I IFN gene expression and downstream ISG production. To determine if SARS-CoV-2 is able to induce type I IFN production, Calu-3 cells were infected at an MOI of 2 for 12 hours. Transcript levels for *IFNβ* were quantified using qPCR. To assess if SARS-CoV-2 can modulate *IFNβ* gene expression and downstream stimulation of ISGs, Calu-3 cells were infected with SARS-CoV-2 at an MOI of 2 for 12 hours, following which cells were treated or transfected with recombinant IFNβ or poly(I:C), respectively for 6 hours. Mock infected and mock treated cells served as controls. **(A)** *IFNβ* transcript levels in SARS-CoV-2 infected or mock infected Calu-3 cells, normalized to *GAPDH* (Mean ± SD; n = 6). **(B)** *IRF7* transcript levels in SARS-CoV-2 infected or mock infected Calu-3 cells, normalized to *GAPDH* (Mean ± SD; n = 6). **(C)** *IFIT1* transcript levels in SARS-CoV-2 infected or mock infected Calu-3 cells, normalized to *GAPDH* (Mean ± SD; n = 6). **(D)** *IFNβ* transcript levels in SARS-CoV-2 infected or mock infected, and poly(I:C) transfected or mock transfected Calu-3 cells, normalized to *GAPDH* (Mean ± SD; n = 6). **(E)** *IFIT1* transcript levels in SARS-CoV-2 infected or mock infected, and poly(I:C) transfected or mock transfected Calu-3 cells, normalized to *GAPDH* (Mean ± SD; n = 6). **(F)** *IRF7* transcript levels in SARS-CoV-2 infected or mock infected, and poly(I:C) transfected or mock transfected Calu-3 cells, normalized to *GAPDH* (Mean ± SD; n = 6). **(G)** SARS-CoV-2 N, IFIT1 and GAPDH protein expression in SARS-CoV-2 infected or mock infected, and poly(I:C) transfected or mock transfected Calu-3 cells (Mean ± SD; n = 3). **(H)** *IRF7* transcript levels in SARS-CoV-2 infected or mock infected, and recombinant IFNβ treated or mock treated Calu-3 cells, normalized to *GAPDH* (Mean ± SD; n = 6). **(I)** *IFIT1* transcript levels in SARS-CoV-2 infected or mock infected, and recombinant IFNβ treated or mock treated Calu-3 cells, normalized to *GAPDH* (Mean ± SD; n = 6). (J) SARS-CoV-2 N, IFIT1 and GAPDH protein expression in SARS-CoV-2

636   infected or mock infected, and recombinant IFNβ treated or mock treated Calu-3 cells (Mean ±

637   SD; n = 3). Statistical significance was calculated using two-tailed paired t-test. Ct, cycle

638   threshold. p*<0.05, **<0.01, ***<0.001 and ****<0.0001.

SI Appendix

Tables

Table S1. Mean raw read counts for SARS-CoV-2 genes. INF, SARS-CoV-2 infected; H, hours; SD, standard deviation.

Mean INF 0H	SD INF 0H	Mean INF 1H	SD INF 1H	Mean INF 2H	SD INF 2H	Mean INF 3H	SD INF 3H	Mean INF 6H	SD INF 6H	Mean INF 12H	SD INF 12H	SARS-CoV-2 gene	Transcript
257.67	38.59	285.33	56.13	243.67	39.25	278.00	23.00	12173.33	3006.93	25827.33	2054.93	ORF1ab	lcl|NC_045512.2_cds_YP_009724389.1_1
1.00	1.73	0.00	0.00	0.33	0.58	0.00	0.00	33.67	6.81	1061.00	468.03	ORF1a	lcl|NC_045512.2_cds_YP_009725295.1_2
500.67	94.52	491.33	86.19	378.00	61.39	521.67	49.69	19232.33	3952.46	26903.33	3860.82	spike	lcl|NC_045512.2_cds_YP_009724390.1_3
173.67	24.99	172.33	43.68	127.33	17.16	203.33	26.50	9995.00	1736.00	13976.33	2233.55	ORF3a	lcl|NC_045512.2_cds_YP_009724391.1_4
43.67	5.51	44.67	13.65	39.00	2.65	63.00	11.53	2903.33	485.15	4086.33	627.70	envelope	lcl|NC_045512.2_cds_YP_009724392.1_5
199.67	27.02	196.00	37.32	162.33	28.87	298.67	19.60	22344.33	3354.18	31200.33	4915.23	membrane	lcl|NC_045512.2_cds_YP_009724393.1_6
34.67	2.08	32.33	10.50	25.00	7.81	45.33	1.53	3508.00	509.12	4704.67	886.56	ORF6	lcl|NC_045512.2_cds_YP_009724394.1_7
107.33	19.50	102.33	23.35	94.00	22.61	173.67	34.00	14834.00	2357.53	21920.67	3441.71	ORF7a	lcl|NC_045512.2_cds_YP_009724395.1_8
10.33	2.52	11.67	2.31	15.33	1.53	20.67	1.15	1516.33	241.00	2191.33	526.17	ORF7b	lcl|NC_045512.2_cds_YP_009725318.1_9
109.33	22.19	107.00	27.22	98.00	21.70	189.00	14.00	14651.33	2136.80	21518.67	3992.04	ORF8	lcl|NC_045512.2_cds_YP_009724396.1_10
1251.00	230.97	1157.33	247.52	1067.67	144.58	2945.67	402.61	258553.00	34843.96	393221.67	62159.07	nucleocapsid	lcl|NC_045512.2_cds_YP_009724397.2_11
112.33	27.57	97.00	22.52	94.67	10.69	250.00	19.00	18385.33	2239.71	27679.00	5406.01	ORF10	lcl|NC_045512.2_cds_YP_009725255.1_12

		0H INF (N=3)	0H MOCK (N=3)	1H INF (N=3)	1H MOCK (N=3)	2H INF (N=3)	2H MOCK (N=3)	3H INF (N=3)	3H MOCK (N=3)	6H INF (N=3)	6H MOCK (N=3)	12H INF (N=3)	12H MOCK (N=3)
IFNs	IFNB1	1.35	0.00	1.21	0.41	1.48	0.97	0.57	1.93	6.40	0.30	21.23	0.89
	IFNL1	3.49	3.45	2.20	4.80	4.93	5.46	3.17	1.90	7.00	2.66	15.07	0.73
	IFNL2	0.00	0.00	0.36	0.96	4.11	0.35	0.28	0.00	4.66	0.00	8.61	0.00
	IFNL3	0.35	0.00	0.58	0.44	2.38	0.31	0.88	0.00	3.02	0.00	8.46	0.00
ISGs	IFIT1	388.42	358.77	370.80	487.33	447.59	590.32	425.31	498.05	463.17	408.65	2790.57	367.50
	IRF7	278.50	283.73	320.43	284.00	339.99	383.89	399.07	363.67	399.93	432.29	966.54	305.31
	OAS2	172.67	236.24	178.18	222.85	287.61	208.20	252.85	296.10	292.36	378.90	2979.22	303.60
	MX1	588.48	620.75	624.79	647.52	758.95	800.13	839.47	867.29	728.29	811.68	3922.41	546.94
	RSAD2	204.76	216.53	228.73	272.67	313.84	348.31	365.12	393.68	274.53	269.56	948.75	210.54
	SLC44A4	1247.82	1171.72	1218.77	1046.17	1138.09	1128.19	1129.60	1106.06	1010.30	1142.19	1032.09	1298.09
	IFIH1	1052.81	1100.39	1134.78	1163.76	1235.36	1164.31	1223.66	1371.55	1189.70	1191.00	2492.69	1087.88
	GBP1	506.79	512.73	503.57	608.29	496.74	485.28	458.14	509.15	530.04	509.53	1151.35	488.92
	IFI44	689.16	741.40	789.19	803.61	963.68	1113.99	997.06	1052.67	785.42	782.39	1889.54	671.51
	IFI27	311.49	318.74	302.63	399.59	343.37	472.30	328.28	361.48	333.63	351.85	921.55	342.54
	IFI6	592.82	612.04	599.90	697.80	673.06	1010.20	692.26	752.25	729.19	775.17	2066.30	709.85
	ISG15	430.95	447.57	443.60	533.02	465.88	704.43	490.49	554.07	473.88	502.97	1260.48	435.91
	IFIT2	657.23	698.02	676.46	795.49	645.57	732.08	455.75	504.29	493.48	422.04	1465.16	413.27
	USP18	212.27	217.53	218.01	257.03	253.55	301.50	266.17	297.44	243.57	232.66	873.18	218.27
	IFIT3	648.15	656.89	747.61	858.17	810.13	1069.67	567.26	668.13	458.25	428.90	1900.07	420.64
	CMPK2	163.89	179.41	169.11	182.05	219.35	244.03	235.97	265.54	172.78	201.60	906.22	153.23
	XAF1	58.53	82.76	73.40	53.61	69.79	60.14	79.67	55.09	86.30	91.97	513.01	90.51
	IFITM1	27.68	34.25	21.94	27.89	28.53	53.49	26.88	34.91	34.59	35.75	182.01	34.33
	MX2	82.11	87.24	69.22	81.96	100.75	83.43	87.84	87.48	108.05	78.88	547.98	64.92

683 **Table S3.** Pathway enrichment analysis. Significance was determined after FDR correction. H,
684 hour; 0, non-significant; 1, significant.
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term.id	term.name	adjusted.p.val	1H	2H	3H	12H
GO:0000976	transcription regulatory region sequence-specific DNA binding	0.004824255	0	1	0	0
GO:0001067	regulatory region nucleic acid binding	0.004203707	0	1	0	0
GO:0001816	cytokine production	0.005529472	0	0	0	1
GO:0001817	regulation of cytokine production	0.001829233	0	0	0	1
GO:0002230	positive regulation of defense response to virus by host	0.002197834	0	0	0	1
GO:0002831	regulation of response to biotic stimulus	8.60E-08	0	0	0	1
GO:0002833	positive regulation of response to biotic stimulus	0.008687053	0	0	0	1
GO:0003690	double-stranded DNA binding	0.000112873	0	1	0	0
GO:0003712	transcription coregulator activity	1.30E-06	0	1	0	0
GO:0003713	transcription coactivator activity	2.39E-05	0	1	0	0
GO:0005178	integrin binding	0.013874905	0	0	1	0
GO:0008270	zinc ion binding	0.000103938	0	1	0	0
GO:0009615	response to virus	1.39E-35	0	0	0	1
GO:0010810	regulation of cell-substrate adhesion	0.008350323	0	1	0	0
GO:0016482	cytosolic transport	0.011086056	0	1	0	0
GO:0019058	viral life cycle	3.92E-11	0	0	0	1
GO:0019079	viral genome replication	3.87E-15	0	0	0	1
GO:0019221	cytokine-mediated signaling pathway	8.45E-16	0	0	0	1
GO:0019900	kinase binding	0.003539788	0	1	0	0
GO:0019901	protein kinase binding	0.012867428	0	1	0	0
GO:0030099	myeloid cell differentiation	0.011382292	0	1	0	0
GO:0031347	regulation of defense response	2.16E-05	0	0	0	1
GO:0031589	cell-substrate adhesion	0.002867293	0	1	0	0
GO:0032020	ISG15-protein conjugation	0.008627708	0	0	0	1
GO:0032069	regulation of nuclease activity	1.26E-06	0	0	0	1
GO:0032479	regulation of type I interferon production	4.92E-06	0	0	0	1
GO:0032480	negative regulation of type I interferon production	0.005210998	0	0	0	1
GO:0032481	positive regulation of type I interferon production	0.00531473	0	0	0	1
GO:0032606	type I interferon production	6.14E-06	0	0	0	1
GO:0032607	interferon-alpha production	0.005237546	0	0	0	1
GO:0032647	regulation of interferon-alpha production	0.00400414	0	0	0	1
GO:0032727	positive regulation of interferon-alpha production	0.001567461	0	0	0	1
GO:0034340	response to type I interferon	9.21E-31	0	0	0	1
GO:0034341	response to interferon-gamma	1.44E-10	0	0	0	1
GO:0034504	protein localization to nucleus	0.00295333	0	1	0	0
GO:0035455	response to interferon-alpha	3.29E-10	0	0	0	1
GO:0035456	response to interferon-beta	2.08E-07	0	0	0	1

GO:0042393	histone binding	0.002987285	0	1	0	0
GO:0043900	regulation of multi-organism process	2.03E-17	0	0	0	1
GO:0043901	negative regulation of multi-organism process	3.86E-17	0	0	0	1
GO:0043902	positive regulation of multi-organism process	0.008274484	0	0	0	1
GO:0043903	regulation of symbiosis encompassing mutualism through parasitism	6.66E-20	0	0	0	1
GO:0044212	transcription regulatory region DNA binding	0.004047416	0	1	0	0
GO:0045069	regulation of viral genome replication	1.01E-16	0	0	0	1
GO:0045071	negative regulation of viral genome replication	3.61E-17	0	0	0	1
GO:0045088	regulation of innate immune response	5.98E-06	0	0	0	1
GO:0045089	positive regulation of innate immune response	0.005979802	0	0	0	1
GO:0046596	regulation of viral entry into host cell	0.048025337	0	0	0	1
GO:0048525	negative regulation of viral process	3.26E-20	0	0	0	1
GO:0050657	nucleic acid transport	0.048485615	0	0	1	0
GO:0050658	RNA transport	0.048485615	0	0	1	0
GO:0050688	regulation of defense response to virus	0.002163216	0	0	0	1
GO:0050691	regulation of defense response to virus by host	0.009541892	0	0	0	1
GO:0050792	regulation of viral process	1.31E-20	0	0	0	1
GO:0051056	regulation of small GTPase mediated signal transduction	0.026048495	0	1	0	0
GO:0051607	defense response to virus	1.25E-37	0	0	0	1
GO:0060333	interferon-gamma-mediated signaling pathway	1.36E-13	0	0	0	1
GO:0060337	type I interferon signaling pathway	3.69E-31	0	0	0	1
GO:0060700	regulation of ribonuclease activity	6.89E-07	0	0	0	1
GO:0060759	regulation of response to cytokine stimulus	0.000740173	0	0	0	1
GO:0060760	positive regulation of response to cytokine stimulus	0.007105564	0	0	0	1
GO:0061629	RNA polymerase II-specific DNA-binding transcription factor binding	0.011126656	0	1	0	0
GO:0070566	adenylyltransferase activity	0.006545402	0	0	0	1
GO:0071346	cellular response to interferon-gamma	1.05E-09	0	0	0	1
GO:0071357	cellular response to type I interferon	3.69E-31	0	0	0	1
GO:0098586	cellular response to virus	0.0037813	0	0	0	1
GO:1903900	regulation of viral life cycle	1.50E-18	0	0	0	1
GO:1903901	negative regulation of viral life cycle	1.15E-18	0	0	0	1
GO:1990837	sequence-specific double-stranded DNA binding	0.002945526	0	1	0	0
GO:2001251	negative regulation of chromosome organization	0.039979672	0	1	0	0
REAC:R-HSA-1169408	ISG15 antiviral mechanism	5.61E-12	0	0	0	1
REAC:R-HSA-1169410	Antiviral mechanism by IFN-stimulated genes	5.77E-19	0	0	0	1
REAC:R-HSA-1280215	Cytokine Signaling in Immune system	1.52E-19	0	0	0	1
REAC:R-HSA-168928	DDX58/IFIH1-mediated induction of interferon-alpha/beta	0.001851135	0	0	0	1
REAC:R-HSA-2990846	SUMOylation	0.000289223	0	1	0	0
REAC:R-HSA-3108214	SUMOylation of DNA damage response and repair proteins	0.023406467	0	1	0	0
REAC:R-HSA-	SUMO E3 ligases SUMOylate target proteins	0.000850049	0	1	0	0

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REAC:R-HSA-3247509	Chromatin modifying enzymes	0.016088428	0	1	0	0
REAC:R-HSA-4839726	Chromatin organization	0.016088428	0	1	0	0
REAC:R-HSA-6806834	Signaling by MET	2.89E-05	0	1	0	0
REAC:R-HSA-877300	Interferon gamma signaling	2.97E-09	0	0	0	1
REAC:R-HSA-8874081	MET activates PTK2 signaling	0.000994797	1	0	0	0
REAC:R-HSA-8934593	Regulation of RUNX1 Expression and Activity	0.000745328	0	1	0	0
REAC:R-HSA-8983711	OAS antiviral response	3.29E-08	0	0	0	1
REAC:R-HSA-9006934	Signaling by Receptor Tyrosine Kinases	0.017643755	0	1	0	0
REAC:R-HSA-909733	Interferon alpha/beta signaling	2.97E-31	0	0	0	1
REAC:R-HSA-913531	Interferon Signaling	4.75E-36	0	0	0	1
REAC:R-HSA-918233	TRAF3-dependent IRF activation pathway	0.000139967	0	0	0	1
REAC:R-HSA-933541	TRAF6 mediated IRF7 activation	0.018776243	0	0	0	1
REAC:R-HSA-936440	Negative regulators of DDX58/IFIH1 signaling	0.000931238	0	0	0	1

Figures

Figure S1. SARS-CoV-2 transcripts clustering. To determine SARS-CoV-2 replication
kinetics in human cells using RNA-seq, we infected human lung epithelial cells (Calu-3) at a
multiplicity of infection (MOI) of 2. One hour post incubation, virus inoculum was replaced with
cell growth media and the clock was set to zero hours. We extracted and sequenced poly-A
enriched RNA at 0, 1, 2, 3, 6 and 12 hours post infection (hpi). SARS-CoV-2 genome, sub-
genomic RNA and transcripts were detected in infected samples. PCA clustering was performed
on quantified SARS-CoV-2 transcript levels in infected samples across time-points. Axes labels
indicate the proportion of between-samples variance explained by the first two principal
components. H, hours post infection.

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Figure S2. Host-cell transcripts clustering. To determine gene expression in human cells using
RNA-seq, we infected human lung epithelial cells (Calu-3) at a multiplicity of infection (MOI)
of 2. One hour post incubation, virus inoculum was replaced with cell growth media and the
clock was set to zero hours. We extracted and sequenced poly-A enriched RNA at 0, 1, 2, 3, 6
and 12 hours post infection (hpi). PCA clustering was performed on quantified and filtered host
gene transcripts in both SARS-CoV-2 infected (blue) and mock infected (red) samples across
time-points (indicated in text for each data-point). Axes labels indicate the proportion of
between-samples variance explained by the first two principal components. H, hours post
infection; Mock, mock infected; INF, SARS-CoV-2 infected.
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741 **Figure S3. Virus-host correlated transcriptional profiles.** Host gene expression that correlated
742 with one or more viral transcripts over the course of infection are shown as z-score normalized
743 expression (bottom), along with viral transcripts (top). Top 100 strongly-correlated genes are
744 represented here. H, hour.

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Figure S4. Cytokine signaling in immune system (REAC:R-HSA-1280215). **(A)** Pathway
schematic of REACTOME cytokine signalling pathway involving interferon alpha/beta/gamma
signalling, and OAS antiviral response mediated by interferon stimulated genes. **(B)** Heatmap of
genes within REACTOME cytokine signalling pathway and their $\log_2$ transformed fold-change
(FC) between SARS-CoV-2 infected and mock infected samples across all timepoints (0, 1, 2, 3,
6, 12 hrs). H, hours.

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Figure S5. Top functional enrichments over time. Top significantly (adjusted p<0.05)
enriched ActivePathway GO terms and REACTOME enrichments for infection vs. mock at 1, 2,
3 and 12 hrs post infection with SARS-CoV-2. Orange bars represent enriched terms associated
with genes upregulated in infection vs. mock. Blue bars represent enriched terms associated with
genes downregulated in infection vs. mock. 0 and 6 hr time points were omitted due to lack of
sufficient numbers of differentially expressed genes.

Figure S6. Infection-responsive gene expression profiles for ISGs. ISGs with significantly different levels of transcript expression between mock (blue) and SARS-CoV-2 infected (red) samples at 12 hpi are shown. Normalized read counts per gene, across six time-points are represented here. Time indicated is in hours. Mock, mock infected; INF, SARS-CoV-2 infected.

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Figure S7. Recombinant human IFNβ1 inhibits VSV, HSV and H1N1 replication. Human
fibroblast (THF) cells were treated with increasing concentrations of recombinant human IFNβ1
or mock treated with GFP containing media (control) for 6 hrs. Cells were then infected with
vesicular stomatitis virus (VSV-GFP), herpes simplex virus (HSV-KOS-GFP) or H1N1
influenza virus (H1N1-mNeon). VSV and HSV were engineered to express green fluorescent
protein (GFP). H1N1 expressed mNeon that is detectable in the same wavelength as GFP.
Nineteen hours post infection, GFP/mNeon levels were measured in mock infected and virus
infected cells as a surrogate for virus replication. **(A)** VSV-GFP replication in THF cells treated
or mock treated with IFNβ1, normalized to mock infection (Mean±SD; n=3). **(B)** HSV-KOS-
GFP replication in THF cells treated or mock treated with IFNβ1, normalized to mock infection
(Mean±SD; n=3). **(C)** H1N1-mNeon replication in THF cells treated or mock treated with
IFNβ1, normalized to mock infection (Mean±SD; n=3).
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