1	Meta-transcriptomic analysis reveals the gene expression and novel conserved
2	sub-genomic RNAs in SARS-CoV-2 and MERS-CoV
3	Lin Lyu ^{1†} , Ru Feng ^{1†} , Mingnan Zhang ^{1†} , Qiyu Gong ¹ , Yinjing Liao ² , Yanjiao Zhou ³ , Xiaokui
4	Guo ¹ , Bing Su ¹ , Yair Dorsett ³ *, Lei Chen ¹ *
5	
6	1 Shanghai Institute of Immunology, Shanghai Jiao Tong University School of Medicine,
7	Shanghai 200025, China
8	2 College of Chemistry, Sichuan University, Chengdu 610064, China
9	3 Department of Medicine, University of Connecticut Health, Farmington, CT, USA
10	
11	* Correspondence:
12	Lei Chen: <u>lei.chen@sjtu.edu.cn;</u>
13	Yair Dorsett: dorsett@uchc.edu
14	
15	[†] These authors contributed equally.
16	
17	
18	
19	
20	
21	
22	

2

23 ABSTRACT

24 **Background:** Fundamental to viral biology is identification and annotation of viral genes and 25 their function. Determining the level of coronavirus gene expression is inherently difficult 26 due to the positive stranded RNA genome and the identification of sub-genomic RNAs 27 (sgRNAs) that are required for expression of most viral genes. In the COVID-19 epidemic so 28 far, few genomic studies have looked at viral sgRNAs and none have systematically 29 examined the sgRNA profiles of large numbers of SARS-CoV2 datasets in conjuction with data for other coronaviruses. 30 31 **Results:** We developed a bioinformatic pipeline to analyze the sgRNA profiles of 32 coronaviruses and applied it to 588 individual samples from 20 independent studies, covering more than 10 coronavirus species. Our result showed that SARS-CoV, SARS-CoV-2 and 33 34 MERS-CoV each had a core sgRNA repertoire generated via a canonical mechanism. Novel 35 sgRNAs that encode peptides with evolutionarily conserved structures were identified in 36 several coronaviruses and were expressed *in vitro* and *in vivo*. Two novel peptides may have 37 direct functional relevance to disease, by alluding interferon responses and disrupting IL17E 38 (IL25) signaling. Relevant to coronavirus infectivity and transmission, we also observed that 39 the level of Spike sgRNAs were significantly higher *in-vivo* than *in-vitro*, while the opposite 40 held true for the Nucleocapside protein. 41 **Conclusions:** Our results greatly expanded the predicted number of coronaviruses proteins identified potential viral peptide suggested to be involved in viral virulence. These 42 and 43 methods and findings shed new light on coronavirus biology and provides a valuable resource 44 for future genomic studies of coronaviruses.

- 45
- 46 Key words: coronavirus, SARS-CoV-2, MERS-CoV, meta-transcriptome, sub-genomic
- 47 **RNA**
- 48

4

49 BACKGROUND

50	Corona virus disease 2019 (COVID-19) reached pandemic levels begining March 2020
51	and brought unprecedented devastation to human lives and the global economy [1]. The
52	causative agent is Severe Acute Respiratory Syndrome - Corona Virus - 2 (SARS-COV-2), a
53	beta coronavirus similar to MERS-CoV, the only other active virulent beta-coronavirus.
54	MERS-CoV is the causative agent of Middle Eastern Respiratory Syndrome (MERS) and is
55	more virulent but less infectious than SARS-CoV-2 and is phylogenetically different from
56	SARS-CoV-2 (less than 90% amino acid sequence homology). Both viruses have a positive
57	single-stranded RNA genome of approximately 30 kilobases that is polyadenylated that
58	encodes 4 structural proteins (spike (S), membrane (M), envelope (E) and nucleocapsid (N))
59	that play similar roles within each virus. The two viruses diverge with respect to the receptor
60	used for cell entry, their virlulent accessory proteins and the specific function(s) of the 16 non
61	structural proteins (nsp1 to nsp16). Nsp's are produced by viral proteinase cleavage of two
62	large polyproteins encoded by ORF1a and ORF1b. ORF1 is closest to the 5' end and is
63	directly translated from genomic RNA upon entrance into host cells and a ribosome skipping
64	mechanism divides it into ORF1a and ORF1b [2]. While MERS-CoV encodes at least 5
65	accessory proteins (ORF3, ORF4a, ORF4b, ORF5 and ORF8b), SARS-CoV-2 encodes at
66	least 6 (ORF3a, ORF6, ORF7a, ORF7b, ORF8, and ORF10 [3]. All proteins not encoded by
67	ORF1a or ORF1b, must be translated from sub-genomic RNAs (sgRNAs) [4, 5]. SgRNAs are
68	generated via a mechanism termed discontinuous extension that uses short sequences of
69	varying length (usually 6 to 12 nucleotides (nts)) termed Transcription Regulatory Sequences
70	(TRS's) spaced between genes to pair a 3' portion of the negative viral strand to a

5

71	complementary 5' leader sequence of around 70 nts. This is followed by extension of the
72	negative strand to the 5' end of the positive strand, generating a short negative strand sgRNA
73	intermediat. The RNA intermediary is then replicated to generate a positive strand sgRNA
74	that encodes viral protein(s) [6] (Fig. 1A).
75	Annotating viral transcriptomes is fundamental to understanding virus biology, which is
76	a key aspect in combating viral transmission, replication and pathogenesis. Prior coronavirus
77	outbreaks, such as the Severe Acute Respiratory Syndrome (SARS) outbreak in 2003 and the
78	MERS outbreak that began in 2012 and is still ongoing [7, 8], has increased research on these
79	viruses as well as coronaviruses of zoonotic orgin from which human coronaviruses are
80	thought to originate. Comparing transcriptional variation of different coronaviruses may
81	reveal mechanisms behind their distinct pathogenicity and infectivity, and potentially explain
82	the molecular etiology behind how species barriers are crossed. Systematically annotating
83	differences in the transcriptional profiles of virulent coronaviruses that is buried within
84	numerous metatranscriptomic data sets may shed new light on viral transmissibility and
85	virulence. However, even a simple systematic comparison of their in-vitro transcriptional
86	profiles is lacking.

For newly emerged SARS-CoV-2 virus, sequencing plays an essential role in diagnosis and monitoring of strain evolution [3, 9]. However, in general, sequencing data sets for SARS-CoV-2 and MERS-CoV are limited to the description of both viral and host transcripts generated during infection of *in-vitro* cell lines as well as model organisms. Analysis of viral transcriptomes orginating from different viral strains in humans is overlooked as suitable analysis tools are lacking.

6

93 Sequence homology plays an essential role in the functional annotation of viral genes. 94 However, sequence homology alone does not guarantee protein expression as rapidly 95 mutating RNA viruses can harbor sequence alterations that result in novel or mutated ORFs 96 that are not transcribed nor expressed. Therefore, direct profiling of viral RNAs is the step 97 toward understanding which viral products can actually be generated. For SARS-CoV-2, 98 direct profiling of viral RNAs produced in a cultured cell line was recently conducted using Oxford nanopore technology and identified the existence of a canonical and non-canonical 99 100 viral transcriptome. All three of these studies used isolated virus strains to infect the Vero cell line isolated from kidney epithelial cells of the African green monkey that does not initiate an 101 interferon (IFN) response upon infection. Although these studies establish a basic 102 103 characterization of virus transcription, individual studies only characterize viral gene 104 expression of a single viral strain and are unable to determine if viral transcriptional 105 responses are altered in response to even the most basic of immune responses (e.g. $IFN\gamma$) 106 [10-12].

107 We developed a bioinformatics pipeline CORONATATOR (CORONAvirus annoTATOR) 108 to quantify viral gene expression and identify bona-fide sgRNAs in numerous publicly 109 available meta-transcriptomic data sets. Beyond outlining the variation in sgRNA profiles and 110 their relative expression, our analysis identified novel sgRNAs for several different 111 coronaviruses. It also revealed the presence of a core sgRNA repertoire that is shared between 112 SARS and SARS-CoV-2 and one that is unique to MERS-CoV. A subset of novel sgRNAs for 113 SARS-CoV-2 and MERS-CoV appear to be evolutionarily conserved in related coronaviruses 114 found in bat and pangolin. Finally, we show that the transcription of specific sgRNAs differs

significantly in-vitro and in-vivo as well as between different coronaviruses.

116

117 **RESULTS**

118 CORONATATOR profiles viral sgRNAs via alignment breakpoint analysis

119 To systematically identify and compare coronavirus sgRNAs, we sought to identify 120 publicly available coronavirus transcriptomic data sets. As of 2021/09/10, more than 3410427 121 viral genome sequences were submitted to the Global Initiative on Sharing All Influenza Data 122 (GISAID) [13]. However, few data sets contain the raw sequencing reads. Using the search 123 term "coronavirus" along with manual curation, we located raw reads in a total of 19 124 Bioprojects within the NCBI Short Read Archive that contain 588 samples for SARS-CoV-2 125 as well as related coronaviruses, such as SARS and MERS (Table S1). We also used an 126 additional dataset with a single sample that was recently published [10].

To profile the sgRNAs present within these data sets, we developed an informatics pipeline (CORONATATOR). CORONATATOR is designed for the utilization of sequences produced by highly accurate second generation sequencing technology that permits identification of TRS sequences from individual reads. Direct RNA Sequencing on the Oxford Nanopore platform can also be used to profile viral sgRNAs but is currently not supported by CORONATATOR due to the limited data set availablility as well as it's restrictions in terms of sequencing accuracy and read length bias (see **Methods**).

Briefly, raw reads were first aligned to their respective viral references, i.e.
SARS-CoV-2 (GeneBank ID NC_045512.2), SARS-CoV (GeneBank ID NC_004718.3),

136	MERS-CoV (GeneBank ID NC_019843.3) or reference for other species of
137	coronaviruses(Table S1). Specific sgRNAs were inferred from alignment breakpoint analysis
138	that identified reads that spanned the junctions between the 5' leader sequence and more
139	distal genomic sequence (Fig. 1A and Supplementary Methods). The relative abundance of
140	a specific sgRNA to all sgRNA's in a particular sample is analogous to relative gene
141	expression. We constructed a heatmap to determine how viral genotype and viral orgin
142	(e.gin-vivo vs in-vitro) influences viral gene expression (Fig. 3 and discussion below).
143	CORONATATOR was designed to profile all possible breakpoints. However, to obtain
144	bona-fide sgRNAs, we removed both rare breakpoints and breakpoints that were inconsistent
145	across samples. A complete breakpoint consists of two separate genomic positions (Fig. 1A).
146	We also analyzed non-sgRNA breakpoints, for which the 5' position does not encompass the
147	leader TRS. Our data suggested that non-sgRNA breakpoints are very rare (usually below
148	0.05% of total sgRNA breakpoints) and inconsistent, as these breakpoints were never
149	identified in more than a single study. We therefore focused on sgRNAs formed with the
150	canonical 5' leader and a 3' body part.

151

152 Most predicted coronavirus ORFs can be validated by sgRNA analysis

Many ORFs are annotated for SARS-CoV-2 based on consensus sequence annotation and the existence of some are disputed by proteomics as well as sequencing studies [3, 11]. Only after examination of a large number of data sets from multiple studies were we able to confidently assign commonly annotated ORFs into one of three categories (core, low support and no support) (**Fig. 1B**). Identifying bonafide sgRNAs requires multistudy and multisample

158 analysis as unique artifactual sequences are often generated during sequence library 159 preparation or sequencing([14, 15]). Additionaly, many non-canocical sgRNAs found in low 160 abundance may be random aberrant transcripts without dedicated function (Kim et al, 2020). 161 Therefore, only sgRNAs that are present in multiple studies and data sets are true sgRNA 162 candidates. To classify each viral gene we considered factors such as sgRNA relative 163 abundance, TRS conservation and the potential for leaky ribosome scanning that can be 164 affected by start-codon hijacking[16]. 165 We first validated the commonly annotated ORFs for SARS-CoV-2, SARS-CoV and

166 MERS-CoV by looking at the extent of sequencing evidence that supports the existence of 167 specific sgRNAs. For SARS-CoV-2, thirty four samples were kept after removing those with 168 less than 20 sgRNA reads. To identify robust and consistent sgRNAs that represent the "core" 169 repertoire, which we assign to our first sgRNA catagory, we pooled all sgRNAs identified for 170 a specific virus using a weighted average approach (see Methods) and noted their relative 171 abundance. At a relative abundance of 0.5%, 8 canonical breakpoints emerged corresponding 172 to 8 sgRNA species that harbor 8 well-described ORFs for SARS-CoV-2: S, E, M, N, ORF3a, 173 ORF6, ORF7a and ORF8 (Fig.1B-C, Fig. S2A, Table S2, Table S3). The sgRNA breakpoints 174 for these ORFs are situated between 9 and 162 nt upstream of the start codon. N is the most 175 abundant core sgRNA, representing 54% of the core sgRNAs identified in all samples. The E 176 sgRNA is the least abundant at 1.5%, and the only core protein not identified in recent 177 proteomics studies [11, 17]. ORF7a, M, ORF3a, S, ORF8 and ORF6 are present at 10.6%, 178 8.4%, 6.9%, 6.1%, 5.9 and 2.7% respectively. Together, these 8 core sgRNAs account for 70% 179 to 100% of the total sgRNAs depending on sample type (e.g. *in-vivo vs in-vitro*), viral strain

180 and read coverage (**Fig. 2, Table. S3**).

181	Beside their high relative abundance, these 8 core sgRNA are also defined by a shared
182	canonical body TRS with a converseved core sequence of "ACGAAC", which is unique to
183	this group of sgRNAs. This core sequence could be necessary and sufficient for sgRNA
184	formation. Futhermore, the same 8 core sgRNAs, as well the core TRS sequence, were shared
185	by SARS (Fig. S2). The 7 core sgRNAs for MERS following (S, E, M, N, ORF3, ORF4a and
186	ORF5) (Fig. 1C) also utilize this core sequence, with the exception of N that has a TRS
187	which contains "ACGAA".
188	A second category of sgRNA was generally present at low relative abundance and does
189	not use this core sequence a conserved core TRS sequence. This category include ORF7b in
190	SARS-CoV-2 and SARS-CoV, ORF3b in SARS and ORF4b and ORF8b in MERS-CoV. For
191	SARS-CoV-2, E has an average relative abundance of 1.5% which is the lowest amoungst the
192	core ones, while ORF7b's is only 0.02% . This low abundance or low efficiency in sgRNAs
193	formation may result from the use of noncanonical TRSs. This group of sgRNAs do not use
194	the conserved core TRS sequence as core sgRNAs do, meaning the sequence homology they
195	rely on for recombination is always shifted a few bases from the core and quite often they
196	contain mismatches between leader and body TRS.
197	Other predicted ORFs fell into the third category with no sgRNA support, at least in the

data set we examined. When factor in evidence beyond sgRNA support. This category can be futher divided into two sub-categories. The first would be no sgRNA support but can potentially be translated. It has been observed that some coronavirus ORFs can be expressed via a leaky ribosome scanning mechanism [16]. ORF9b of SARS-CoV-2 falls into this

202	sub-category. Indeed, multiple recent proteomics studies showed support for the ORF9b
203	protein product in SARS-CoV-2[17, 18]. Its homolog in SARS-CoV, also named ORF9b,
204	falls in the same category. Interestingly, the ORF7b of SARS-CoV-2 and SARS-CoV were
205	mentioned in previous studies to be in this category (Schaecher et al., 2007), and indeed the
206	long stretch (362 nt in SARS-CoV-2 and 365 nt in SARS) between start codons of ORF7b
207	and preceding ORF7a are void of additional start codons. Yet, these gene products still form
208	their own sgRNAs at low abundance.

209 The second sub-category would contain the most suspicious ORFs, where sgRNA support cannot be found and intervening start codon between they and the closest sgRNA 210 breakpoint would make their expression very unlikely. This category includes a few 211 212 commonly annotated ORFs: ORF3b, ORF9c and ORF10 of SARS-CoV-2, ORF8b in 213 SARS-CoV. The several out of frame start codons between these ORFs and preceding ones, 214 along with the absence of corresponding sgRNAs and its absence from proteomic studies [10, 215 17, 18], strongly argues that these proteins are not generated. Indeed, the existence ORF10 216 was recently debated in recent manuscripts [11, 12]. The evidence described above indicates 217 the potential pitfalls of conducting experiments on viral products from putative ORFs with no 218 sgRNA or proteomic support. For example, a recent study that generated a synthetic version of the predicted truncated version of ORF3b in SARS-CoV-2 speculated that the putative 219 220 truncated version in SARS-CoV-2 had a stronger anti-IFN activity than the SARS version 221 [19].

222

223 Identification of novel sgRNAs with non-canonical TRSs in SARS-CoV-2, MERS-CoV

12

and SARS-CoV.

225	As mentioned before, during formation of the core sgRNA repertoire, a body TRS that
226	contains a minimal core sequence will pair with the leader TRS. For each particular core
227	sgRNA, the two TRS's used must be of the same length and sequence, although the length
228	can vary between sgRNAs (Fig. 1C). We found the average length of these canonical TRS's
229	for SARS-CoV-2 was ~9.6 nts. Interestingly, the same core sequence is used in SARS, while
230	MERS also uses a six nucleotide TRS with a different core sequence (Fig. S2A,B).
231	When we looked for sgRNAs that composed more than 0.2% of sgRNA transcripts, we
232	identified three additional sgRNAs that were present in at least two separate samples and
233	studies (Fig. 2A). All three novel sgRNAs contained breakpoints that did not utilize canonical
234	TRS sequences that are present in core sgRNAs. The three breakpoints support the
235	discontinuous extension model of sgRNA formation, as the sequence from the body strand
236	was found in the TRS sequences of the final transcript (Fig. 2B, Fig. S3A-C). On a separate
237	note, sequence analysis of stranded RNA library preps identified the presence of negative
238	strand sgRNAs, which were not described in the previous Nanopore sequencing manuscripts
239	[10-12]. As previously noted for artificial TRS's, analysis of these non-canonical breakpoint
240	sequences revealed that TRS's without perfect complementarity may pair, and/or that large
241	regions of complementarity around a core TRS between the body to itself, maybe used for the
242	formation of sgRNAs (Fig. 2B,). Our analysis confirmed that TRS sequences can vary
243	significantly between distantly related viruses and find that canonical TRS sequences can be
244	more than 30 nt in length in some coronaviruses (Fig. S2D).

245 The three novel TRS's generated three novel sgRNAs that we have termed putative

13

246	ORF2b (pORF2b), alternative M (aM) and truncated ORF7b (tORF7b). The longest novel
247	sgRNA, pORF2b, is within the S gene and has two alternative TRS's positioned around
248	22501. Interestingly, it encodes a novel peptide that has a domain structure that is conserved
249	in closely related coronaviruses, with at least one virus harboring and extended ORF (Fig.
250	2B,C). The second novel breakpoint is located at 26494, 31 nt downsteam of the canonical
251	breakpoint for M. The sgRNA would support M expression, but with an alternative 5' UTR
252	(Fig. S3A). The shortest of the three novel sgRNA's has its breakpoint positioned at 27761
253	and codes for a truncated version of ORF7b (tORF7b). The truncation removes the
254	extracellular domain and 14 of the 24 amino acids that comprise the transmembrane domain
255	(Fig. 2A, Fig. S3B). This sgRNA is expressed at relatively high levels both in-vivo and
255 256	(Fig. 2A, Fig. S3B). This sgRNA is expressed at relatively high levels both <i>in-vivo</i> and <i>in-vitro</i> and likely harbors novel functions (see discussion below).
256	<i>in-vitro</i> and likely harbors novel functions (see discussion below).
256 257	<i>in-vitro</i> and likely harbors novel functions (see discussion below). Translation of pORF2b results in a 36 amino acid peptide. It was predicated by PSIPRED
256 257 258	<i>in-vitro</i> and likely harbors novel functions (see discussion below).Translation of pORF2b results in a 36 amino acid peptide. It was predicated by PSIPRED[20] to have a intracellular protein binding coil and two short alpha-helixes that overlap a
256 257 258 259	 <i>in-vitro</i> and likely harbors novel functions (see discussion below). Translation of pORF2b results in a 36 amino acid peptide. It was predicated by PSIPRED [20] to have a intracellular protein binding coil and two short alpha-helixes that overlap a transmembrane domain, with the second alpha helix partially extracellular (Fig. 2A, C).
256 257 258 259 260	 <i>in-vitro</i> and likely harbors novel functions (see discussion below). Translation of pORF2b results in a 36 amino acid peptide. It was predicated by PSIPRED [20] to have a intracellular protein binding coil and two short alpha-helixes that overlap a transmembrane domain, with the second alpha helix partially extracellular (Fig. 2A, C). pORF2b was present in 4 samples in two separate studies. The highest expression of pORF2b

identified (Table. S3). The virus strains infecting these two patients differed by one
nucleotide. Five other patient samples from the same study with different viral strains (Table.
S3) did not yield sgRNAs for pORF2b. The low breakpoint read numbers for these samples
as well as viral strain may contribute to the variable detection of pORF2b *in-vivo*. This

indicates that the level of pORF2b transcripts maybe loosely correlated with viral strain and
further demonstrates that samples within this bioproject are not cross contaminated with an
artifactual pORF2b sgRNA. SgRNA pORF2b was also identified in a separate study
(PRJNA615032), in two in-vitro samples that used a different viral strain than any of those
identified in the *in-vivo* study (**Table. S3**).

273 We searched for sequence conservation of pORF2b in other related Sarbecoviruses, 274 including SARS-CoV, HKU3 (bat coronavirus), RaTG13 (a bat coronavirus proposed to be 275 directly related to SARS-CoV-2) and a coronavirus infecting pangolin (SRX7732088)[21]. A 276 corresponding ORF was identified in all four viruses, with the highest level of homology 277 found in RaTG13, with 91.89% nucleotide identity (Table S4 & Fig. 2C). Interestingly, 278 pORF2b and more so the pangolin version which has a C terminal extension, share high 279 similarity with the ligand binding domain of human IL17RB (Fig. 2D and see section 280 "Disscussion").

281 The third novel breakpoint was located at position 27761, within ORF7b, and encodes a 282 truncated version of ORF7b (tORF7b). We identified this transcript and its relative abundance 283 *in-vivo* and *in-vitro* in two separate bioprojects that included more than one viral strain. This transcript was also recently identified in a VERO cell line infected by a single viral strain 284 285 [10]. Interestingly, this novel sgRNA was expressed at relatively high levels both *in-vivo* and 286 in-vitro (Fig. 3), and a SARS-CoV homolog of this sgRNA was also present in several 287 samples across two studies. This truncated version of ORF7b is missing the intracellular 288 domain and more than half of its transmembrane domain, while retaining its hydrophilic 289 extracellular domain (Fig. S3D). ORF7b is present in the SARS-CoV virion particle and is

290	homologous ORF7b encoded by SARS-CoV-2 [16]. The portion of ORF7b encoded by
291	tORF7b is highly conserved in SARS (Table S4 & Fig. S3D). Intrigingly, a previous study
292	observed that a 45 nt deletion in SARS ORF7b that removes much of the transmembrane
293	domain lost in tORF7b, attentuated the induction of interferon-beta, provides a replicative
294	advantage in-vitro and in-vivo as well as to cells pretreated with interferon-beta [22]. Future
295	research will reveal if this novel sgRNA encodes a novel virulent peptide that has function(s)
296	antagonistic to IFN while subverting the initation of an interferon response.
297	We also obtained a significant amount of <i>in-vivo</i> and <i>in-vitro</i> sequence data sets for
298	MERS-CoV, allowing us to identify abundant non-canonical sgRNAs (Fig. S2B). This novel
299	sgRNA (putative ORF8c or pORF8c), is predicted to encode a ORF that translate into a novel
300	51 amino acid peptide. This novel sgRNA was identified in 5 separate studies, both in-vivo
301	and <i>in-vitro</i> , ranging in abundance from 0.03% to 1.0% of total sgRNAs. PSIPRED suggest
302	this novel peptide has a transmembrane domain connected to a cytoplasmic helix domain. We
303	also looked for its conservation in other Merbecoviruses, including HKU4, HKU5 and an
304	Erinaceus coronavirus. pORF8c could be found in all 3 with varying conservation (Fig. S3E,
305	Table. S4). The cytoplasmic N terminal was the most conserved across Merbecovriuses and

306 C terminal elongated versions were observed in HKU5 and Erinaceus (Fig. S3E).

To exhaust our search for novel sgRNAs, we lowered our threshold value to a relative abundance of 0.01%, while maintaining our other criteria. This analysis identified additional novel sgRNAs that appeared in more than one study for SARS-CoV-2, SARS-CoV and MERS-CoV (**Table S5**). Additional sequencing and future experiments will determine the significance of pORF2b, tORF7B and aM as well as the numerous other novel sgRNAs

16

312 present at extremely low abundance.

313

314 CORONATATOR detects experimentally induced alteration of novel pORF8c relative

315 abundance.

316 We next wished to validate the experimental utility of our pipline and validate that a 317 novel sgRNA responds to experimental stimuli in a manner similar to other established viral 318 genes. To accomplish this, we utilized an experimental data set that tested the effects of 319 Gleevec and IFN- β on host gene expression during treatment of MERS-CoV infection 320 in-vitro (PRJNA233943 & PRJNA233944) (Fig. S5). Specifically, we analyzed the effects on 321 viral load, viral gene expression and the expression of novel pORF8c. Initial analysis 322 demonstrated that decreased viral load broadened the expression of individual viral genes 323 (Fig. S5). Importantly, even at low viral loads, the ratio of N to S remained high, 324 demonstrating that this ratio is not influenced by viral abundance, but by *in-vitro* and *in-vivo* 325 context. The effect of both Gleevec and IFN- β on viral gene expression was not uniform, 326 having different effects on different viral genes. Interestingly, the expression profile of novel 327 pORF8c followed the same trend of N and E with respect to viral load in response to IFN- β 328 and Gleevec. This demonstrates that pORF8c has the same biological response in terms of 329 gene expression as some "core" sgRNAs in this context.

330

331 The relative abundance of Spike sgRNAs is elevated for SARS-CoV-2 *in-vivo*.

When processing the data sets, we noticed two distinct patterns of read coverage along the SARS-CoV-2 reference genome that suggested that viral reads originate from two sources.

334	Upon further examination, it was revealed the two sources were in-vivo and in-vitro derived
335	samples (Fig. S1). The former is composed of extracellular virion particles and infected host
336	cells present in BALF (human) and nasal washes (Ferret) or lung homogenate (MERS), while
337	the latter is composed of infected cells that are not subject to systemic or sometimes innate
338	(e.g. VERO cells do not produce IFN) anti-viral reponses. In-vivo derived viral sequences
339	obtained primarily from BALF for SARS-CoV-2 (primarily BALF) generally covered the
340	entire viral reference length, with little bias towards the sgRNA containing 3' end. In contrast,
341	highly elevated coverage at the 3' end of the viral genome was observed in the in-vitro
342	samples due to the formation of nested sgRNAs during viral transcription.
343	SARS-CoV-2 and MERS-CoV are the only active virulent coronaviruses and are present
344	in both <i>in-vivo</i> and <i>in-vitro</i> derived metatranscriptomic data sets. We analyzed the relative
345	abundance of sgRNAs generated <i>in-vivo</i> and <i>in-vitro</i> for both SARS-CoV-2 and MERS-CoV.
346	When comparing the relative abundance of viral sgRNAs generated in-vivo to those
347	generated <i>in-vitro</i> , it was evident that the ratio of S sgRNAs to N sgRNAs was significantly
348	higher in-vivo, especially for SARS-CoV-2 (0.04 in-vitro vs 0.69 in-vivo for SARS-CoV-2, p
349	value 0.0012 with Wilcoxon ranksum test) (Fig. 4A-B, Fig. S4). The differences in
350	environemental pressures that influence the requirement for these sgRNAs for viral
351	replication, provdes a general explanation for this striking variation in sgRNA levels. The
352	selective pressures may alter viral transcriptional responses that promote viral propogation.
353	For example, the primary function of the S protein centers around host cell recognition and
354	invasion while the primary function of the N protein centers around the regulation of viral
355	RNAs to promote viral replication. This is mediated by direct binding of the 3' end of the

356	viral genome, the viral packaging signal as well as TRS's [23-25]. Another explanation,
357	which is not mutually exclusive, is that the increased S/N ratio in-vivo is due to an altered
358	viral Replication Transcription Complex (RTC) that favors TRS read-through, preferentially
359	generating longer sgRNAs. Such a "global" alteration of viral transcription likely involves
360	host factors, as observed for Infectious bronchitis virus (IBV), a gamma coronavirus in which
361	the N protein in phosphorylated by cellular GSK3 to recruit the helicase DDX1 to promote
362	TRS read through during the formation of long sgRNAs [26]. In this regard, the N protein is
363	generated from the shortest sgRNA while the S protein is generated from the longest. Future
364	electron microscopy studies on in vivo and in-vitro viron particles will determine if Spike
365	sgRNA abundance in SARS-CoV-2 correlates with spike protein levels on viron surfaces.
366	Other examples of sgRNAs that are significantly differentially expressed in-vitro and in-vivo
367	include the overall increase in the levels of accessory sgRNAs that act via multiple pathways
368	to quell the immune response in both SARS-CoV-2 and MERS-CoV (Fig. S4, [27]).
369	To obtain a clearer perspective on how the relative abundance of SARS-CoV-2 sgRNAs
370	compares to other coronaviruses in-vivo and in-vitro as well as determine if additional novel
371	sgRNAs have been overlooked, CARONTATOR was utilized to analyze additional
372	coronaviruses. This analysis included OC43, NL63, HKU1 as well as bat and pangolin
373	viruses with high sequence homology to SARS-COV-2 [9, 21] (Fig 4C, Table S1 and Fig.
374	S2). Some datasets did not yield enough breakpoint reads to be informative. For example,
375	analysis of the the bat virus RaTG13, with the highest homology to SARS-COV-2, yielded
376	only 1 break point read and was therefore omitted from Fig. 4C.
277	Of the different corresponding as prefiled SADS COV2 stands out as having the highest

377 Of the different coronaviruses profiled, SARS-COV-2 stands out as having the highest

19

378	levels of S sgRNAs, especially <i>in-vivo</i> (see discussion below and Fig. 4C). Our analysis
379	indicates that this is independent of viral strain as it is present at high levels in different
380	strains identified in-vivo (Fig. 2). The high levels of Spike protein may play a role in the
381	viruses ability to cross the species barrier (see discussion below) and it's high rate of
382	infectivity. In agreement, we noted that the relative levels of the Spike sgRNA is positively
383	correlated with coronavirus infectivity. Viral infectivity and levels of S sgRNAs <i>in-vivo</i> are as
384	follows: SARS-COV-2> HKU1> MERS [28]. However, S protein levels alone are not
385	sufficient to cause high levels of SARS-CoV-2 transmissibility, as factors such as Spike
386	protein stability, receptor aviditiy [29] and viron stability[30], also contribute to viral
387	transmissibility.

388

389 Mutations in the RTC reverse the expression of N and S sgRNAs in vitro and in-vivo.

390 We also observed mutations in viral RTC components that altered the expression profile 391 of S to N. Specifically, the viral strain Kim 2020 had one unique non-synonymous mutation 392 in the RTC component nsp3, a papin protease that binds the N and M protein (Fig. 3). The 393 transcriptome generated *in-vitro* for this viral strain showed a dramatic increase in the S to N 394 ratio, mimicking the expression profile of viruses found *in-vivo* (Fig. 3, Fig. S6). Interestingly, a viral strain identified *in-vivo* (SRX7852918), had two non-synonymous mutations in nsp3, 395 396 as well as nsp6 and nsp12 and had an *in-vitro* like transcription profile, with a decreased S to 397 N ratio (Fig. S6).

The observation that mutations in nsp3 occur in the two viruses with altered gene expression is thought provoking. Nsp3 is reported to to bind TRS's, the 3' end of the viral

ylation of the N protein has been reported to alter it's conformation to
al RNA and as mentioned above for IBV, promote TRS readthrough
of long sgRNAs [26, 33]. This observation tentatively implies that
affect the relative abundance of sgRNAs by acting in a global
ces overall viral structure and may act in concert with the mechanism
V. Additionally, the altered relative abundance of N sgRNAs <i>in-vivo</i>
chanisms discussed above, may feedback on it's interaction with Nsp3
on of mutations in Nsp3 <i>in-vivo</i> and <i>in-vitro</i> (Fig. 4A-B).

409

410 **DISCUSSION**

The vast amount of sequence data generated for SARS-CoV-2 thus far has primarily been 411 412 used for the typing and following of emerging viral strains. Although this is important, we 413 felt such a focus could be an under-utilization of a valuable information. By developing the 414 Coronatator informatics pipeline, we took a step beyond the characterization of viral strains 415 and described coronavirus viral sgRNA expression and uncovered novel and conserved 416 sgRNAs with unknown function that are generated via a non-canonoical TRS pairing 417 mechanism (Fig. 2). Functional prediction for some of these novel putative proteins is still 418 ongoing. We tentatively show that a homolog of SARS-CoV-2 pORF2b in pangolin virus 419 shares extensive similarity with human IL17RB's ligand binding domain (Fig. 2D). It is 420 curious that a coronaviruse may generate a peptide that could theoretically disrupt IL17B and

421 IL17E (IL25) signaling as they are generally associated with promoting or inhibiting
422 inflammatory responses in specific contexts. Future proteomic studies and/or ribosome
423 sequencing studies will be required to verify the production of the protein products encoded
424 by the novel sgRNAs idenrified here.

The analysis presented here also implicates that different strains of SARS-CoV-2 express sgRNAs at different levels (**Table S3, Fig. 3**), especially for the newly discovered sgRNAs. Our findings underscore that a true understanding of viral pathogenesis in terms of sgRNA expression can only come from thorough sequencing of patient samples in which the virus is under selective pressure. This begs for in-depth case examination, in which thorough sequencing and analysis is conducted for different stages of COVID-19 on a strain by strain basis. This would result in truly individualized patient care.

432 Although other zoonotic viruses may share extensive sequence similarity to 433 SARS-CoV-2 at the gene or genomic level, similarity alone is not sufficient for the generation 434 of pathogenic human viruses. Generally not considered during discussion of zoonotic viral 435 orgins, the specific expression level of viral genes, such as the Spike protein, are likely 436 important for crossing the species barrier. For example, considering the vast number of un-sampled zoonotic viruses, it is likely Spike proteins capable of crossing the species barrier 437 438 already exist, yet are not expressed at sufficiently high levels to enable sustainable 439 inter-human transmission. However, low level Spike protein expression would allow sporadic 440 transmision from bat to human, yet would not be sustainable as human to human transmission 441 would be low due to low S protein expression as well sanitary environments that do not exist 442 for bats. In agreement, it has been observed that people living in proximity to bat caves

harbor virus specific antibody without ever experienceing severe disease [34].

444	Our analysis of the meta-transcriptomic data sets identified numerous sources of RNA,
445	such as host RNA as well as microbial RNA (although not optimally captured). In a time
446	when it is unclear why some people succumb to SARS-CoV-2 infection while others do not,
447	these valuable sequences should not be wasted and could be made more useful if more
448	clinical information is shared for these data sets. Most GISAID entries for SARS-CoV-2 have
449	a meta-transcriptomic dataset that supports it. However, current GISAID entries that simply
450	outline the viral genome sequence and strain far out-number the raw read entries we
451	identified in SRA. Sharing the raw read information will greatly help researchers study this
452	virus and ultimately curb it.

453 CONCLUSIONS

454 We developed a bioinformatics pipeline CORONATATOR that can take meta-transcriptomic 455 sequencing reads generated from coronavirus samples and analyze the sub-genomic RNA 456 profiles of the underlying virus, akeen to a transcriptome for the virus. For emergent viruses, 457 as in the case of SARS-CoV2, homology search was usually the first and only choice of 458 predicting viral ORFs after sequencing was done. Now our tool can provide additional 459 evidence. By applying it to large number of SARS-CoV2 and related viral datasets, 460 interesting biology about these coronaviruses were revealed. In addition to define core and 461 predict novel ORFs, our results suggested, for beta-coronaviruses, the spike to nucleocapsid 462 ratio to be a potential tunable in adjusting viral life style and the elevation of this ratio in 463 SARS-CoV2 may contribute to its strong transmissibility. The methods and findings

464 presented here provides a valuable resource for future genomic studies of coronaviruses.

465 METHODS

466 **Data collection**

467 All sequencing data used were collected from NCBI Short Reads Archive (SRA). Some 468 nanopore datasets were downloaded from online repository described in their respective 469 manuscripts [10]. The bioprojects were located by searching with key words "coronavirus" 470 and with manual curation, only meta-transcriptomic data were kept Raw reads files were 471 downloaded from SRA using wget with a customized script, SRAtoolkit were used to 472 generate compressed fastq files from downloaded sra files. After initial sequence alignment 473 using bwa with reference genome sequences of SARS-CoV, SARS-CoV-2 or MERS-CoV, 474 samples with too few viral reads were filtered out. CORONATATOR only uses reads 475 generated from second generation technologies (Illumina), nanopore data were used for 476 comparison.

477 Coronatator

478 CORONATATOR were a series of perl and bash scripts developed for profiling and analysis
479 of RNA-Seq data from coronavirus. It consists of 3 major steps, including preprocessing,
480 breakpoint identification, sgRNA calling and profiling, details below.

481 **Preprocessing**

BAM files were generated from sequence alignment with reference genomes of SARS-CoV ,
SARS-CoV-2 or MERS-CoV, for viruses from bat and pangolin, responsive genome
assemblies were obtained from NCBI as references. SNPs were called and filtered with

487	Breakpoint identification
486	sequences were also generated with filtered SNPs for further analysis.
485	beftools [35] and annotated with vef-annotator [36]. In addition, consensus genome

Breakpoints were identified from alignments with soft or hard clips, these alignments were all partial alignments largely caused by reads with recombination joints, which was generated by the mechanism through which coronavirus produce their sgRNA. In this step, a matrix of reads' information, breakpoint sites, CIGAR strings together with possible TRS sequences was generated.

493 sgRNA calling and profiling

494 Typical sgRNAs were identified and defined by two breakpoint coordinates on a reference 495 genome sequence, these sites were obtained by extracting breakpoints from partial alignments, 496 i.e. one from primary alignment and the other from supplementary alignment. To recognize 497 possible TRS pattern, sequences between breakpoint pairs were extracted from previous 498 generated consensus genome sequences. After that, corresponding genes of called sgRNAs 499 were identified by manually comparing the distances between start codons of known viral 500 genes and their breakpoints. Biosamples with more than 20 sgRNAs were used for further 501 analysis, in these samples, sgRNAs were counted by genes and normalized by total sgRNA 502 count to obtain a transcription profile matrix.

503 Novel ORF identification

Potential ORFs were predicted using Prodigal [37] with -s arguments to write all potential genes. An in-house python script was also used to identify very short ORFs. Then for sgRNAs with multiple bioproject support, we calculated and sorted the distances between

- 507 their breakpoints and all identified start codon sites. ORFs that start closest to upstream
- 508 breakpoints were bookmarked and manually checked for verification.

509 Sequence alignment and phylogenetic analysis

- 510 Consensus genome sequences of SARS-CoV-2, SARS-CoV, MERS-CoV and biosamples
- from bat or pangolin or other human coronavirus with more than 20 sgRNAs were used for
- 512 phylogenetic analysis. Multi-sequence alignment were performed with MAFFT [38],
- 513 Maximum likelihood consensus trees were constructed using IQ-TREE [39] with 1000
- 514 bootstrap times.

515 Converting Nanopore sgRNA proportion to short reads'

516 Kim et al included both nanopore data and short read data. The ratios between the two were 517 used to convert the other nanopore data sets to proportions comparable with others in this 518 study.

519 Plots and statistical analysis

Heatmaps showing gene expression profile were produced using 'heatmap.plus' package.
SgRNA expression dot plots and boxplots were made with 'ggplot2' package to compare
difference between gene expression among different sample origin, T-test and wilcoxon test
were used for statistical analysis.

524 **Function annotation**

Novel peptide sequences were aligned with EMBL online tool FASTA
(https://www.ebi.ac.uk/Tools/sss/fasta/) against UniProtKB/Swiss-Prot database with default
arguments. NCBI CD Blast online service was used to identify protein domains.

528 Sequence conservation

529	To check for sequence conservation of putative peptides in related viral species, we generated
530	a reference database containing all predicted ORFs from related viral genomes. DC
531	MegaBlast (DisContinuous MegaBlast) was used to search for inter-species homologs.
532	Arguments were set as follows: window_size 0, gapopen 0, gapextend 2, penalty -1, reward 1,
533	num_alignments 1. A group of homologous ORFs were then subjected to multiple sequence
534	alignment (MSA) using MAFFT. After that CLUSTAO (Clustal Omega) was used to
535	calculate an identity matrix for the MSA result. The same procedure was performed for both
536	nucleotide and amino acid sequences.
537	ABBREVIATIONS
538	sgRNA: sub-genomic RNA
539	ORF: open reading frame
540	TRS: Transcription Regulatory Sequences
541	COVID: Corona virus disease
542	MERS: Middle Eastern Respiratory Syndrome
543	SARS: Severe Acute Respiratory Syndrome
544	CORONATATOR: CORONAvirus annoTATOR
545	DECLARATIONS

- 546 **Ethics approval and consent to participate:** Not applicable.
- 547 **Consent for publication:** Not applicable.
- 548 Availability of Data and Materials: All used sequenceing data are accessible with accession

549	number provided in supplementary table 1, code of CORONATATOR is accessible at:
550	https://github.com/15274972986/CORONATATOR.
551	Competing interests: The authors declare that they have no competing interests.
552	Funding: This study was supported by Shanghai Institute of Immunology COVID-19 Special
553	Fund.
554	Authors' contributions: Conceptualization: Lei Chen; Methodology: Lei Chen, Lyu Lin;
555	Investigation: Lyu Lin, Ru Feng, Mingnan Zhang, Yinjing Liao; Visualization: Lyu Lin, Qiyu
556	Gong; Supervision: Lei Chen, Xiaokui Guo, Bing Su, Yanjiao Zhou; Writing-original draft:
557	Lei Chen, Yair Dorsett, Lyu Lin; Writing-review & editing: Lei Chen, Yair Dorsett, Lyu Lin,
558	Ru Feng.
559	Acknowledgements: We thank Dr. Qiming Liang of Shanghai Institute of Immunology for
560	his insightful suggestion.
561	
562	
563	REFERENCES
564	1. WHO: World Health Organization:Coronavirus disease 2019 (COVID-19) Situation
565	Report. In., vol. 2020: World Health Organization; 2020.
566	2. Knipe DM, Howley PM: Coronaviridae. In: Fields virology. vol. 28, sixth edn.
567	Philadelphia: Wolters Kluwer Health/Lippincott Williams & Wilkins; 2013: 825-859.
568	3. Wu F, Zhao S, Yu B, Chen YM, Wang W, Song ZG, Hu Y, Tao ZW, Tian JH, Pei YY et al:

569	A new coronavirus associated with human respiratory disease in China. Nature 2020,
570	579 (7798):265-269.

- 571 4. Brian DA, Baric RS: Coronavirus genome structure and replication. Coronavirus
- 572 *Replication and Reverse Genetics* 2005, **287**:1-30.
- 573 5. Yount B, Curtis KM, Fritz EA, Hensley LE, Jahrling PB, Prentice E, Denison MR,
- 574 Geisbert TW, Baric RS: Reverse genetics with a full-length infectious cDNA of severe
- 575 acute respiratory syndrome coronavirus. Proceedings of the National Academy of
- 576 *Sciences of the United States of America* 2003, **100**(22):12995-13000.
- 577 6. Sola I, Almazan F, Zuniga S, Enjuanes L: Continuous and Discontinuous RNA
 578 Synthesis in Coronaviruses. *Annual Review of Virology, Vol 2* 2015, 2:265-288.
- 579 7. Peiris JSM, Yuen KY, Osterhaus ADME, Stohr K: Current concepts: The severe acute

respiratory syndrome. *New England Journal of Medicine* 2003, **349**(25):2431-2441.

- 581 8. Assiri A, McGeer A, Perl TM, Price CS, Al Rabeeah AA, Cummings DAT, Alabdullatif
- ZN, Assad M, Almulhim A, Makhdoom H et al: Hospital Outbreak of Middle East
- 583 Respiratory Syndrome Coronavirus. New England Journal of Medicine 2013,
 584 369(5):407-416.
- 585 9. Zhou P, Yang XL, Wang XG, Hu B, Zhang L, Zhang W, Si HR, Zhu Y, Li B, Huang CL et
- 586 al: A pneumonia outbreak associated with a new coronavirus of probable bat origin.
 587 Nature 2020, 579(7798):270-273.
- 588 10. Kim D, Lee J-Y, Yang J-S, Kim JW, Kim VN, Chang H: The Architecture of
 589 SARS-CoV-2 Transcriptome. *Cell* 2020, 181(4):914-921.e910.
- 590 11. Davidson AD, Williamson MK, Lewis S, Shoemark D, Carroll MW, Heesom K, Zambon

									29
591	М,	Ellis	J,	Lewis	PA,	Hiscox	JA	et	al:
592	2020: <u>ht</u>	tps://www.	.biorxiv.o	rg/content/10).1101/2020).1103.1122.00)2204v002	201.	
593	12. Taiaro	a G, Rawl	linson D,	Featherstone	e L, Pitt M	, Caly L, Dru	ce J, Purc	ell D, Ha	urty L,
594	Tran		Τ,	Robe	orts	J	et		al:
595	2020: <u>ht</u>	tps://www.	.biorxiv.o	rg/content/10).1101/2020).1103.1105.97	<u>/6167v976</u>	<u>5162</u> .	
596	13. Shu Y	L, McCau	ley J: GI	SAID: Globa	al initiativ	e on sharing a	ll influen	za data -	from
597	vision t	o reality. I	Eurosurve	eillance 2017	, 22 (13):2-	4.			
598	14. Lebrig	gand K, M	agnone V	/, Barbry P,	Waldmann	R: High thro	ughput e	rror cori	rected
599	Nanopo	ore single (cell trans	scriptome se	quencing.	Nat Commun 2	2020, 11 (1)):4025.	
500	15. Peng (Q, Vijaya S	Satya R, I	.ewis M, Ran	ndad P, War	ig Y: Reducin	g amplific	cation ar	tifacts
601	in high	multiplex	amplico	on sequencin	ng by using	g molecular ba	arcodes. <i>I</i>	BMC Gen	nomics
502	2015, 1	6 :589.							
503	16. Schae	cher SR,	Mackenz	zie JM, Pek	osz A: T	ne ORF7b p	rotein of	severe	acute
504	respira	tory synd	rome co	ronavirus (S	SARS-CoV) is expressed	l in virus	-infected	l cells
505	and inc	corporated	l into SA	RS-CoV par	ticles. Jour	rnal of Virolog	y 2007, 81	(2):718-7	31.
506	17. Bojko	va D, Kla	nn K, K	och B, Wide	era M, Kra	use D, Ciesek	x S, Cinat	l J, Mün	ich C:
507	Proteor	mics of SA	ARS-CoV	-2-infected	host cells	reveals theraj	py targets	. Nature	2020,
508	583 :469	9-472.							
509	18. Gordo	on DE, Jan	ng GM, l	Bouhaddou N	M, Xu J, C	Dbernier K, W	hite KM,	O'Mear	a MJ,
510	Rezelj	VV, Guo Jž	Z, Swane	ey DL et al: A	A SARS-C	oV-2 protein	interactio	n map r	eveals
11	targets	for drug r	repurpos	ing . <i>Nature</i> 2	2020, 583 :4	59-468.			

612 19. Konno Y, Kimura I, Uriu K, Fukushi M, Irie T, Koyanagi Y, Nakagawa S, Sato K:

613	2020: <u>htt</u>	ps://www.biorxiv.org	g/content/10	.1101/2020.	1105.1111	.088179v088171.

- 614 20. Buchan DWA, Jones DT: The PSIPRED Protein Analysis Workbench: 20 years on.
- 615 *Nucleic Acids Research* 2019, **47**(W1):W402-W407.
- 616 21. Lam TT-Y, Shum MH-H, Zhu H-C, Tong Y-G, Ni X-B, Liao Y-S, Wei W, Cheung WY-M,
- 617 Li W-J, Li L-F et al: Identifying SARS-CoV-2 related coronaviruses in Malayan
- 618 **pangolins**. *Nature* 2020, **583**:282-285.
- 619 22. Pfefferle S, Krähling V, Ditt V, Grywna K, Mühlberger E, Drosten C: Reverse genetic

620 characterization of the natural genomic deletion in SARS-Coronavirus strain

- Frankfurt-1 open reading frame 7b reveals an attenuating function of the 7b protein
- 622 **in-vitro and in-vivo**. *Virology journal* 2009, **6**:131-131.
- 23. Liang Y, Wang M-L, Chien C-S, Yarmishyn AA, Yang Y-P, Lai W-Y, Luo Y-H, Lin Y-T,
- 624 Chen Y-J, Chang P-C et al: Highlight of Immune Pathogenic Response and
- 625 Hematopathologic Effect in SARS-CoV, MERS-CoV, and SARS-Cov-2 Infection.
- 626 *Frontiers in Immunology* 2020, **11**:1022.
- 627 24. Molenkamp R, Spaan WJ: Identification of a specific interaction between the
- 628 coronavirus mouse hepatitis virus A59 nucleocapsid protein and packaging signal.
- 629 *Virology* 1997, **239**(1):78-86.
- 630 25. Fan H, Ooi A, Tan YW, Wang S, Fang S, Liu DX, Lescar J: The nucleocapsid protein of
- 631 coronavirus infectious bronchitis virus: crystal structure of its N-terminal domain
- and multimerization properties. Structure (London, England : 1993) 2005,
 13(12):1859-1868.
- 634 26. Wu C-H, Chen P-J, Yeh S-H: Nucleocapsid Phosphorylation and RNA Helicase DDX1

635	Recruitment Enables Coronavirus Transition from Discontinuous to Continuous
636	Transcription. Cell Host & Microbe 2014, 16(4):462-472.
637	27. Canton J, Fehr AR, Fernandez-Delgado R, Gutierrez-Alvarez FJ, Sanchez-Aparicio MT,
638	Garcia-Sastre A, Perlman S, Enjuanes L, Sola I: MERS-CoV 4b protein interferes with
639	the NF-kappaB-dependent innate immune response during infection. PLoS Pathog
640	2018, 14 (1):e1006838.
641	28. Kissler SM, Tedijanto C, Goldstein E, Grad YH, Lipsitch M: Projecting the
642	transmission dynamics of SARS-CoV-2 through the postpandemic period. Science
643	2020, 368 :860-868.
644	29. Wrobel AG, Benton DJ, Xu P, Roustan C, Martin SR, Rosenthal PB, Skehel JJ, Gamblin
645	SJ: SARS-CoV-2 and bat RaTG13 spike glycoprotein structures inform on virus
646	evolution and furin-cleavage effects. Nat Struct Mol Biol 2020, 27(8):763-767.
647	30. Aboubakr HA, Sharafeldin TA, Goyal SM: Stability of SARS-CoV-2 and other
648	coronaviruses in the environment and on common touch surfaces and the influence of
649	climatic conditions: A review. Transbound Emerg Dis 2020, 68:296-312.
650	31. Hurst KR, Koetzner CA, Masters PS: Characterization of a critical interaction
651	between the coronavirus nucleocapsid protein and nonstructural protein 3 of the
652	viral replicase-transcriptase complex. Journal of virology 2013, 87(16):9159-9172.
653	32. Lei J, Kusov Y, Hilgenfeld R: Nsp3 of coronaviruses: Structures and functions of a
654	large multi-domain protein. Antiviral Research 2018, 149:58-74.

- 655 33. Chang C-k, Hou M-H, Chang C-F, Hsiao C-D, Huang T-h: The SARS coronavirus
- nucleocapsid protein Forms and functions. *Antiviral Research* 2014, **103**:39-50.

3	2
-	~

- 657 34. Wang N, Li S-Y, Yang X-L, Huang H-M, Zhang Y-J, Guo H, Luo C-M, Miller M, Zhu G,
- 658 Chmura AA *et al*: Serological Evidence of Bat SARS-Related Coronavirus Infection in
- 659 **Humans, China**. *Virologica Sinica* 2018, **33**(1):104-107.
- 660 35. Li H: A statistical framework for SNP calling, mutation discovery, association
- 661 mapping and population genetical parameter estimation from sequencing data.
- 662 *Bioinformatics* 2011, **27**(21):2987-2993.
- 663 36. vcf-annotator [https://github.com/rpetit3/vcf-annotator]
- 664 37. Hyatt D, Chen GL, Locascio PF, Land ML, Larimer FW, Hauser LJ: Prodigal:
- 665 prokaryotic gene recognition and translation initiation site identification. *BMC*
- 666 *Bioinformatics* 2010, **11**:119.
- 38. Katoh K, Misawa K, Kuma K, Miyata T: MAFFT: a novel method for rapid multiple
- sequence alignment based on fast Fourier transform. Nucleic Acids Res 2002,
- **669 30**(14):3059-3066.
- 670 39. Minh BQ, Schmidt HA, Chernomor O, Schrempf D, Woodhams MD, von Haeseler A,
- 671 Lanfear R: IQ-TREE 2: New Models and Efficient Methods for Phylogenetic
- 672 **Inference in the Genomic Era**. *Mol Biol Evol* 2020, **37**(5):1530-1534.

33

674 FIGURE LEGEND

675 Fig. 1: Study overview and the sgRNA profile of SARS-CoV-2. (A) Study overview; Top 676 panel, datasets used in this study came from different hosts infected by different 677 coronaviruses, for in-vitro studies, a cell line culturing step was added. These samples were 678 subjected to meta-transcriptomic sequencing and reads were collected from Short Read 679 Archive. Bottom panel, reads mapped to viruses' genome. Alignment of reads spanning the 680 genome shows breakpoints sites. As viral sgRNAs were formed via the recombination of 681 transcript body and a fixed 5' leader via TRS homology, breakpoints with 5' position close to 682 leader TRS were sent to sub-genomic RNA profiling. (B) The canonical breakpoints plot of 683 SARS-CoV-2. The ratio of putative sub-genomic RNA, black bar indicates relative 684 abundance of profiled sgRNA. (C) ORF annotation and comparison among 3 coronaviruses. 685 Different color of blocks represent sgRNA supportive stat of that gene. Specially, red blocks 686 demonstrate novel sgRNA the algorithm identified.

687 Fig. 2: Novel sgRNAs and responsive translational product for SARS-CoV-2. (A) 688 Breakpoints plot for SARS-CoV-2 showing the three novel breakpoints at relative abundance 689 cut-off of 0.1%, putative TRS sequences were shown below. Count of classical breakpoints 690 were shown in grey as background. Peptides of novel ORFs, i.e. putative ORF2b (pORF2b) 691 and truncated ORF7b (tORF7b), were shown in inlets, secondary structures of these peptides 692 were predicted and shown in different color. Specially, a complete ORF7b peptide was shown 693 in grey as a reference for the truncated one; (B) Sequence homology between leader TRS 694 (top), sgRNA (middle) and body TRS (bottom) for novel sgRNAs, TRS core were shown in 695 (C) Structural conservation of novel peptide translated from newly discovered sgRNA blue.

696	putative ORF2b, left panel demonstrates consensus phylogenetic tree of responsive
697	coronavirus determined by genomic sequences, right panel compared second structure of
698	novel peptide predicted by PSIPRED Workbench. (D) Putative ORF2b in pangolin CoV
699	shows homology with IL17RB's fibronectin III like domain, which is also a ligand binding
700	domain.
701	Fig. 3: Heatmap of sgRNA expression profile of SARS-CoV-2 with SNP annotation. Left
702	panel shows sgRNA expression profile of SARS-CoV-2 in the transcriptomic or
703	meta-transcriptomic dataset profiled. Right panel shows all the SNP sites with annotation of
704	the responsive biosamples. Interestingly, virus strains from SRX7852918 and Kim et al had
705	distinctive SNP pattern as well as characteristic expression profiles.
706	Fig.4: Comparison of <i>in-vivo</i> and <i>in-vitro</i> sgRNA expression. (A) and (B) Expression
707	profile of SARS-CoV-2 and MERS-CoV, both in-vivo and in-vitro datasets were included. It
708	should be noticed that two third generation sequencing technology data were added as
709	complementary datasets to SARS-CoV-2 in-vitro plot, a math model was applied to adjust
710	long read expression ratio into an adapted version which was comparable with short read
711	archive datasets. Interestingly, higher levels of S and M expression ratio and lower level N
712	expression ratio were observed in <i>in-vivo</i> sample versus <i>in-vitro</i> sample in these two
713	coronaviruses. (C) Phylogenetic tree of involved coronaviruses (left), scale bar indicates
714	phylogenetic distance which were calculated as the ratio of nonidentical base positions to all
715	base positions, taxonomic classification at genus level were indicated at left part. Expression
716	ratio of Spike (S) genes in vivo and in vitro in different coronaviruses (right), each dot
717	represent a biosample, black bars indicate average expression level of responsive virus.

35

718

719 SUPPLEMENTARY INFORMATION

Additional file 1: Fig. S1: Coverage plot for in-vivo and in-vitro datasets. (A) Coverage plot
for SRX8089279, which is a representative of in-vitro sample. In-vivo RNA-Seq reads
relatively evenly mapped to viral genome, indicating a genomic RNA dominated sample. (B)
Coverage plot for SRX7736886, which is a representative of in-vivo sample. In-vitro reads
resulted a dense mapping at 3' and 5' end of the genome, which revealed active viral
transcription and replication in cultured cells.
Additional file 2: Fig. S2 : Breakpoint profile of 4 coronaviruses. (A) Breakpoints profile of

727 SARS-CoV; (B) Breakpoints profile of MERS-CoV; (C) Breakpoints profile of
728 Pangolin-CoV; (D) Breakpoints profile of HKU1, interestingly, this virus uses long TRS for
729 discontinuous sgRNA production.

730 Additional file 3: Fig. S3: Novel sgRNA breakpoint and TRS sequence. (A) Alternative TRS 731 of M in SARS-CoV-2. The canonical TRS region (upper panel) has 12 bases while the novel 732 one has only 6 (lower panel), start codon of M was shown in red. (B) TRS for tORF7b in 733 SARS-CoV-2, which has 7 bases overlapped with leader TRS. (C) TRS for pORF8c in 734 MERS-CoV. (D) and (E) Structural conservation of novel peptide translated from newly 735 discovered sgRNA truncated ORF7b and putative ORF8c, left panel demonstrates consensus 736 phylogenetic tree of responsive coronavirus determined by genomic sequences, right panel 737 compared second structure of novel peptide predicted by PSIPRED Workbench.

738 Additional file 4: Fig. S4: Detailed gene expression profile of SARS-CoV-2 and MERS-CoV.

36

739	(A) and (B) Detailed accessory gene expression profile of SARS-CoV-2 and MERS-CoV,
740	between in-vivo and iv-vitro datasets. Remarkably, MERS-CoV had higher ORF4a in-vivo
741	expression level while lower in-vivo ORF5 expression level. (C) and (D) Gene expression
742	level of SARS-CoV-2 and MERS-CoV among different hosts.
743	Additional file 5: Fig. S5: Detailed gene expression profile of MERS-CoV in PRJNA233943
744	& PRJNA233944.In this study, cells infected with MERS-CoV were treated with different
745	drugs, i.e. Gleevec and IFN- β , after 24 or 48h post-infection, distinct pattern can be observed
746	from viral gene expression profile as condition alters, especially for IFN- β , treated 24 hpi and
747	48 hpi resulted in distinct expression levels among several structural and accessory genes. It
748	also indicates that our analytical pipeline CORONATATOR is a powerful and sensitive tool
749	for analyzing how experimental manipulation effects the relative expression of specific
750	sgRNAs.
751	Additional file 6: Fig. S6 : Outliers in expression profiles habour interesting SNPs. (A) and
752	(B) The in-vivo sample from Kim et al 2020 had S expression level similar to that of in-vitro
753	samples. While also have a few mutations in ORF1a that's not found in other viral strains.
754	Mirroring this, the in-vitro sample SRX7852918 have S expression level similar to that of
755	in-vivo samples, and hold several private mutations in ORF1a as well.
756	Additional file 7: Table S1 : Meta information of samples collected.
757	Additional file 8: Table S2 : Annotation of SARS-CoV-2, SARS-CoV and MERS-CoV.
758	Additional file 9: Table S3: SARS-CoV-2 sgRNA abundance across samples.

759 Additional file 10: Table S4: Conservation of selected novel proteins.

760 Additional file 11: Table S5: List of novel sgRNAs.

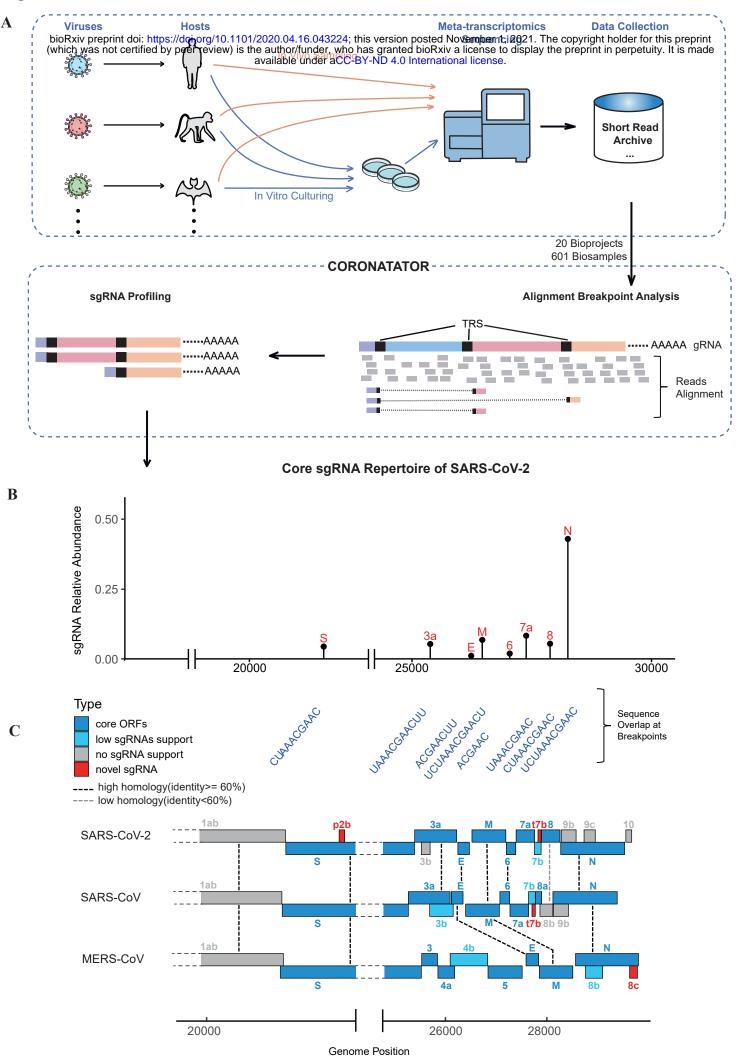
bioRxiv preprint doi: https://doi.org/10.1101/2020.04.16.043224; this version posted November 1, 2021. The copyright holder for this preprint (which was not certified by peer review) is the author/funder, who has granted bioRxiv a license to display the preprint in perpetuity. It is made available under aCC-BY-ND 4.0 International license.

37

761

762

Fig. 1

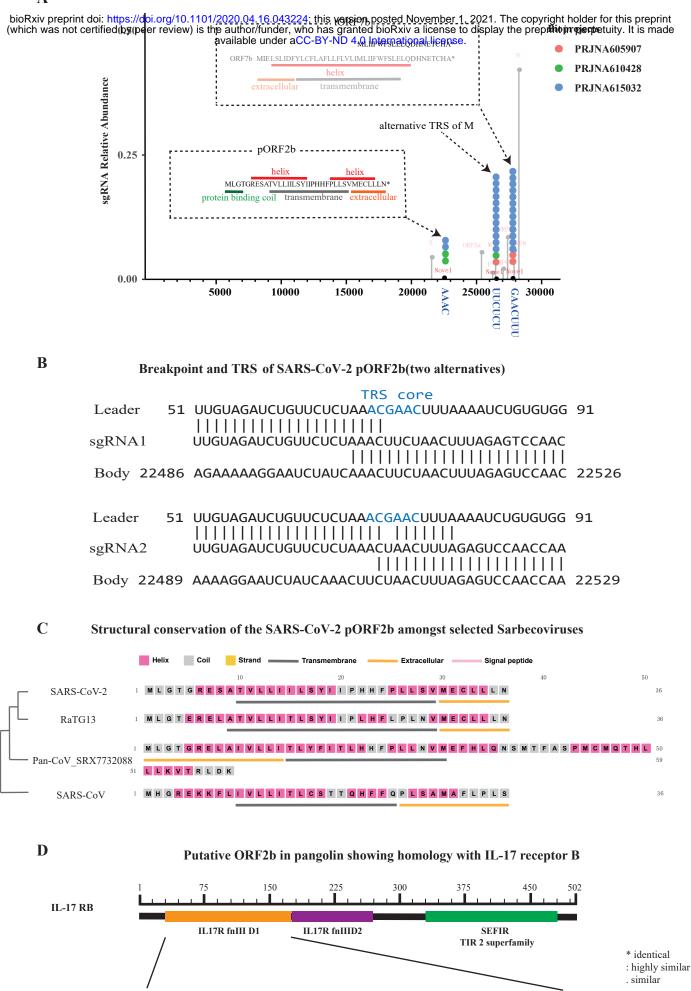


IL-17RB:33-175

Pan-CoV_SRX7732088

1 --ML--

: :



33 EWMLQHDLIPGDLRDLRVEPVT.....YIGFPVELNTVYFIGAHNIPNANMNEDGPSMSVNFTSPGCLD-HIMKYKKKCVK 176

* * * ****

----GTGRELAI-----VLLITLYFITLHHFPLLNVMEFHLQNSMTFASPMCMQTHLLLKVTRLDK 144

::

*: *

. *:.*:** *:: *::

. :

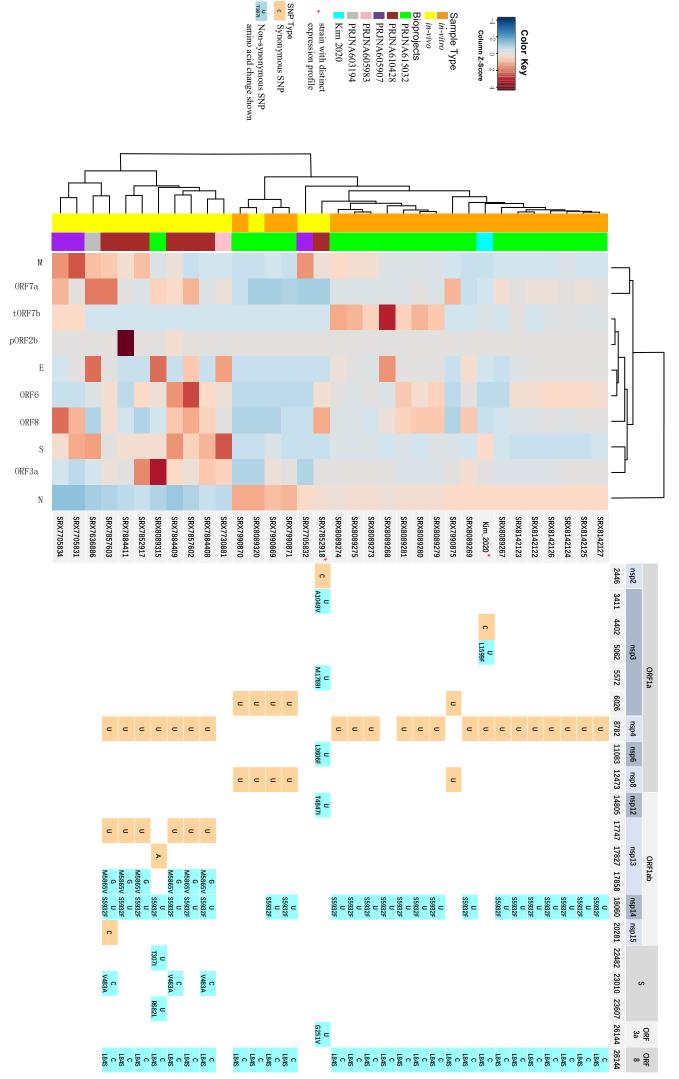
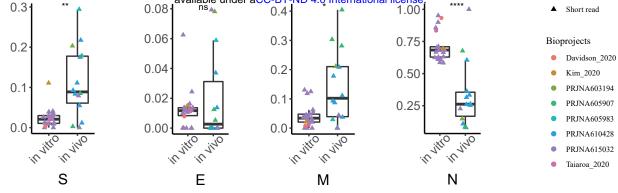


Fig. 3

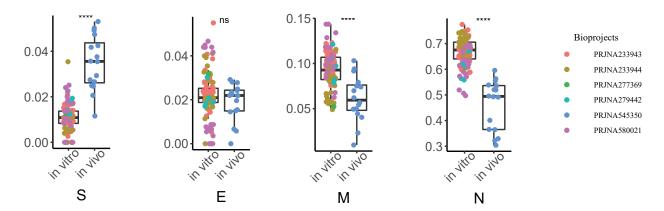
In-vivo versus in-vitro, SARS-CoV-2

A bioRxiv preprint doi: https://doi.org/10.1101/2020.04.16.043224; this version posted November 1, 2021. The copyright holder for this preprint (which was not certified by peer review) is the author/funder, who has granted bioRxiv a license to display the preprint in perpetuity. It is made available under aCC-BY-ND 4.0 International license.



B

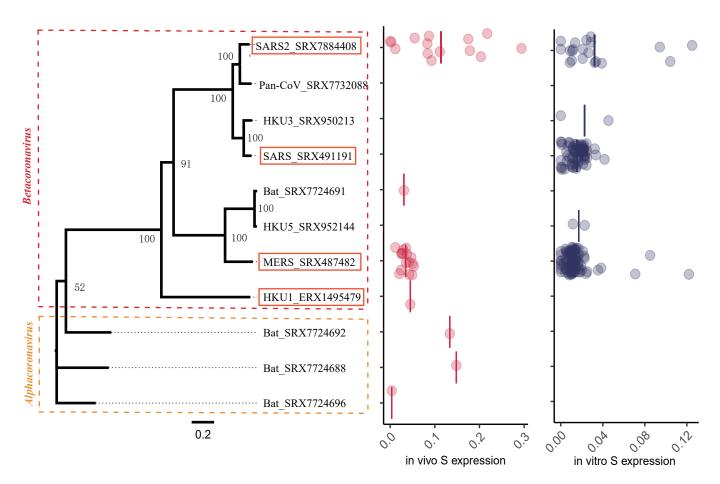
In-vivo versus in-vitro, MERS-CoV



С

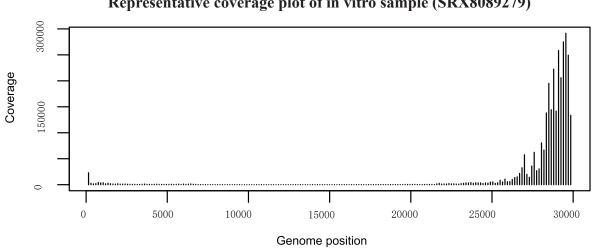
Phylogenetic Tree of Involved Coronaviruses

Expression of S in Vivo Across Coronaviruses



bioRxiv preprint doi: https://doi.org/10.1101/2020.04.16.043224; this version posted November 1, 2021. The copyright holder for this preprint (which was not certified by peer review) is the author/funder, who has granted bioRxiv a license to display the preprint in perpetuity. It is made available under aCC-BY-ND 4.0 International license.

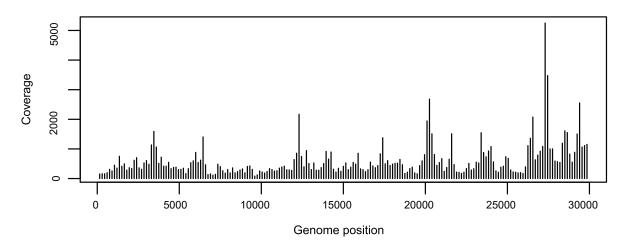
A

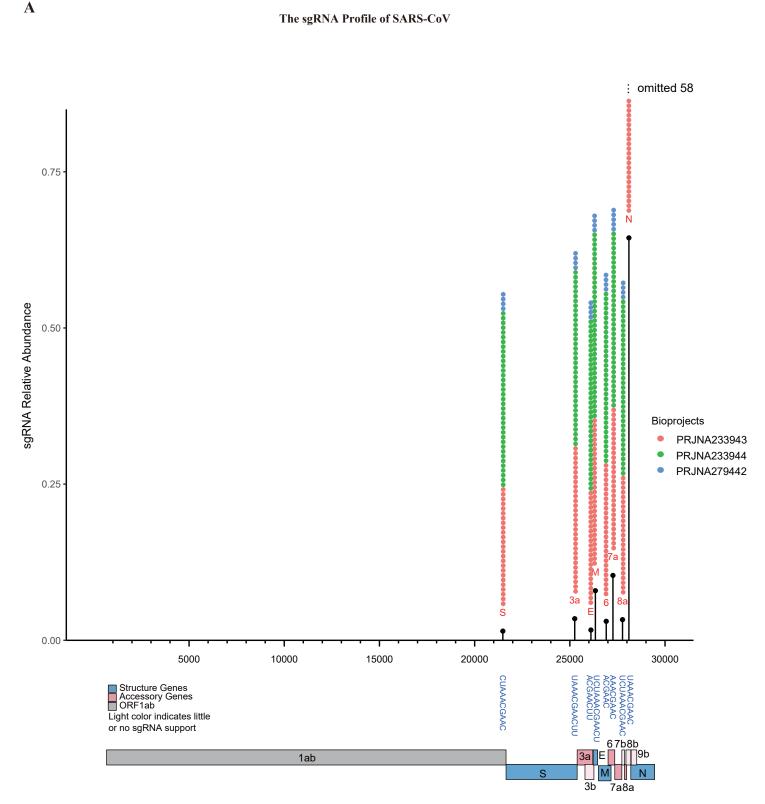


Representative coverage plot of in vitro sample (SRX8089279)

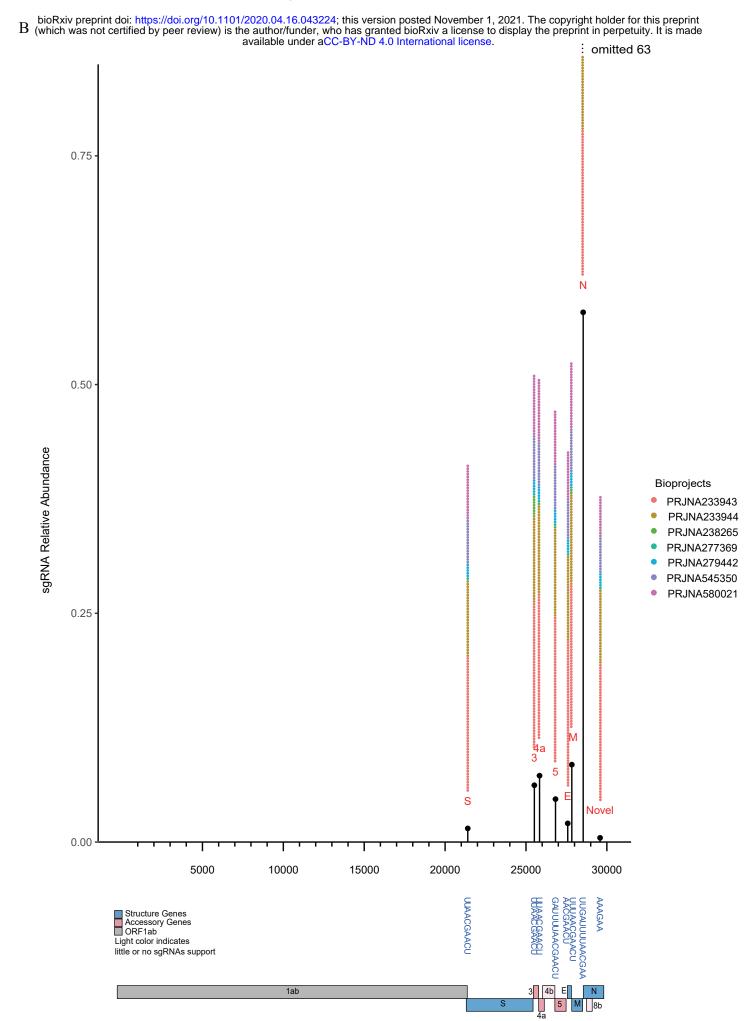
B

Representative coverage plot of in vivo sample (SRX7736886)

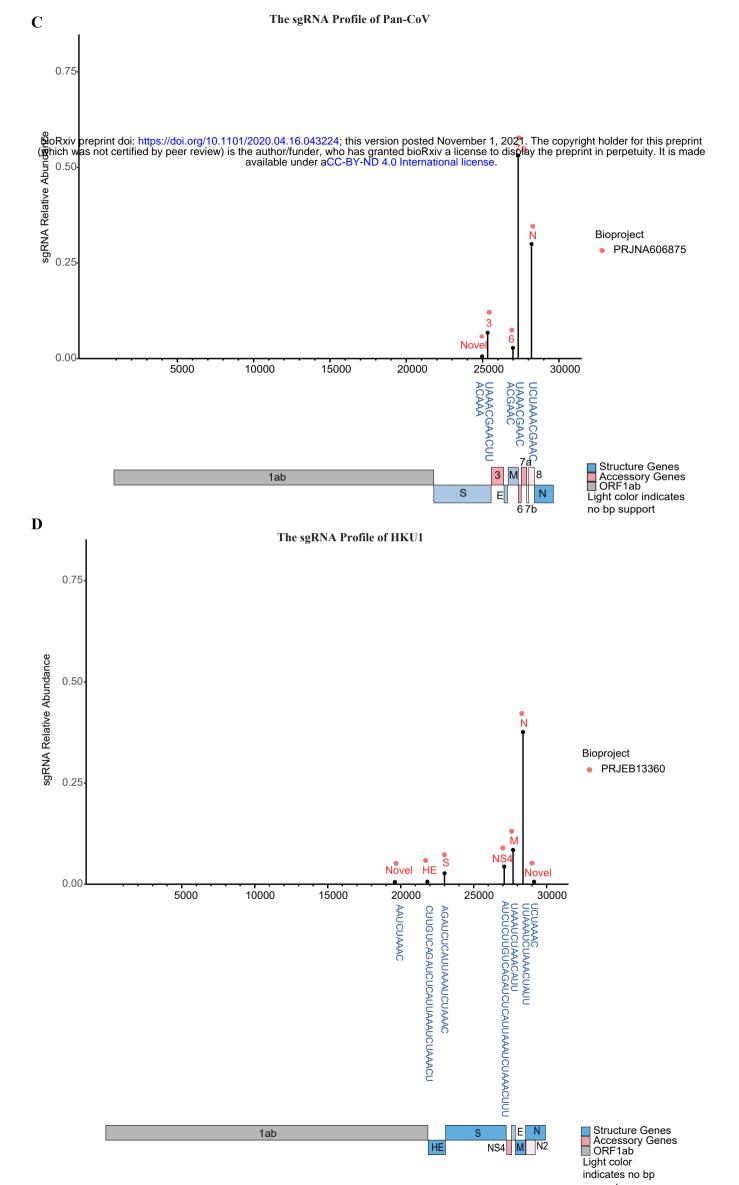




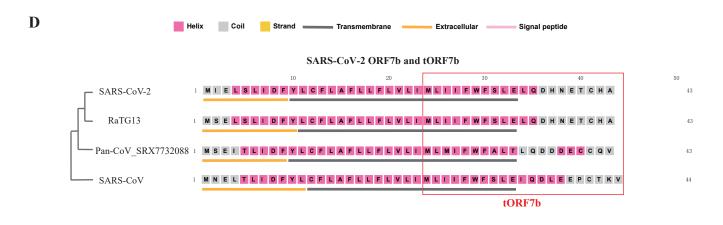
bioRxiv preprint doi: https://doi.org/10.1101/2020.04.16.043224; this version posted November 1, 2021. The copyright holder for this preprint (which was not certified by peer review) is the author/funder, who has granted bioRxiv a license to display the preprint in perpetuity. It is made available under aCC-BY-ND 4.0 International license.



The sgRNA Profile of MERS-CoV



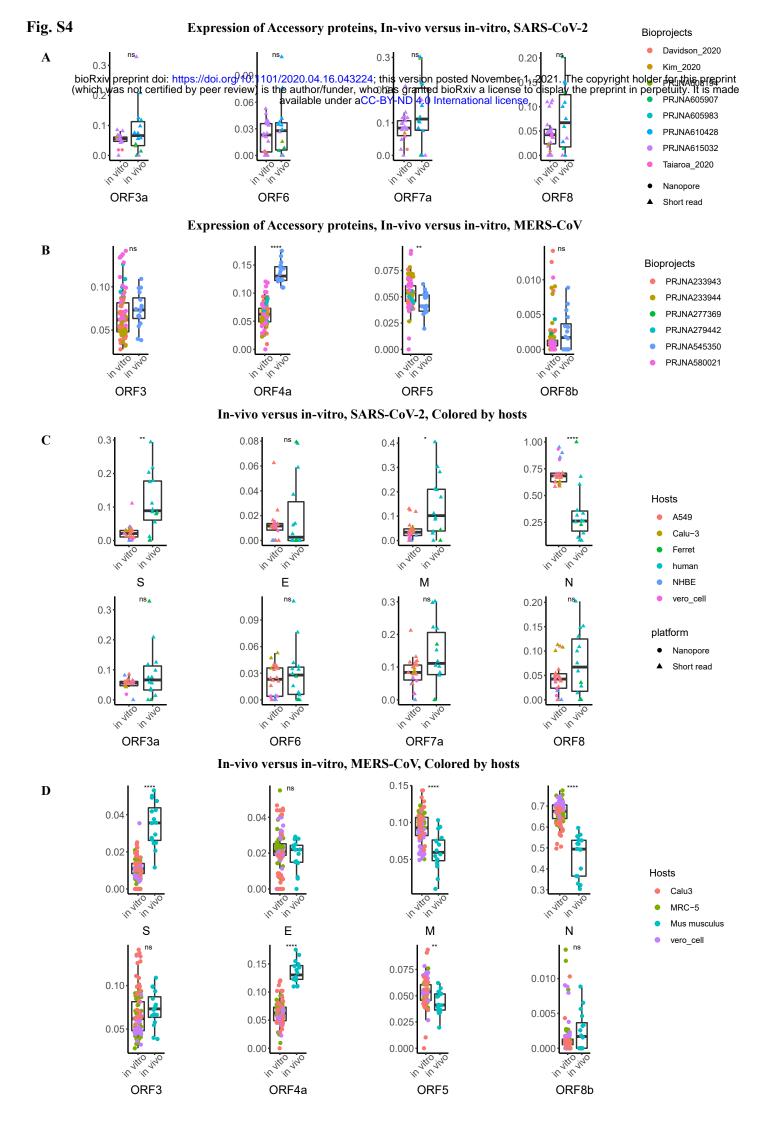
bioRxiv preprint doi: https://doi.org/10.1101/2020.04.16.043224; this version posted November 1, 2021. The copyright holder for this preprint A(which was not certified by peer review) is the autho Static Cono Basi gramed bida au d alternatio with preprint in perpetuity. It is made available under aCC-BY-ND 4.0 International license.	
Leader	51 UUGUAGAUCUGUUCUCUAAACGAACUUUAAAAUCUGUGUGG 91
sgRNA.ca	no UUGUAGAUCUGUUCUCUAAACGAACUAAAUAUUAUAUUA
Body 264	54 UCCUGAUCUUCUGGUCUAAACGAACUAAAUAUUAUAUUA
Leader	1 UUGUAGAUCUGUUCUCUAAACGAACUUUAAAAUCUGUGUGG 91
sgRNA.al	
Body 264	36 UAUAUUAGUUUUUCUGU-UUGGAACUUUAAUUUUAGCCAUG 26525 M start
_	
В	SARS-CoV-2 tORF7b
Leader	51 UUGUAGAUCUGUUCUCUAAACGAACUUUAAAAUCUGUGUGG 91
sgRNA	UUGUAGAUCUGUUCUCUAAACGAACUUUCAUUAAUUGACUU
Body 277	41 CAAAAGAAAGACAGAAUGAUUGAACUUUCAUUAAUUGACUU 27781
С	MERS-CoV pORF8c
e	51 ACUUUGAUUUUAACGAACUUAAAUAAAAGCCCUGUUGUUUA 91
sgRNA	ACUUUGAUUUUAACGAACUUAAAGAAUCCCAACUACAAUAA
Body 295	52 GGAGCCAUUAAACUUGACCCAAAGAAUCCCAACUACAAUAA 29602

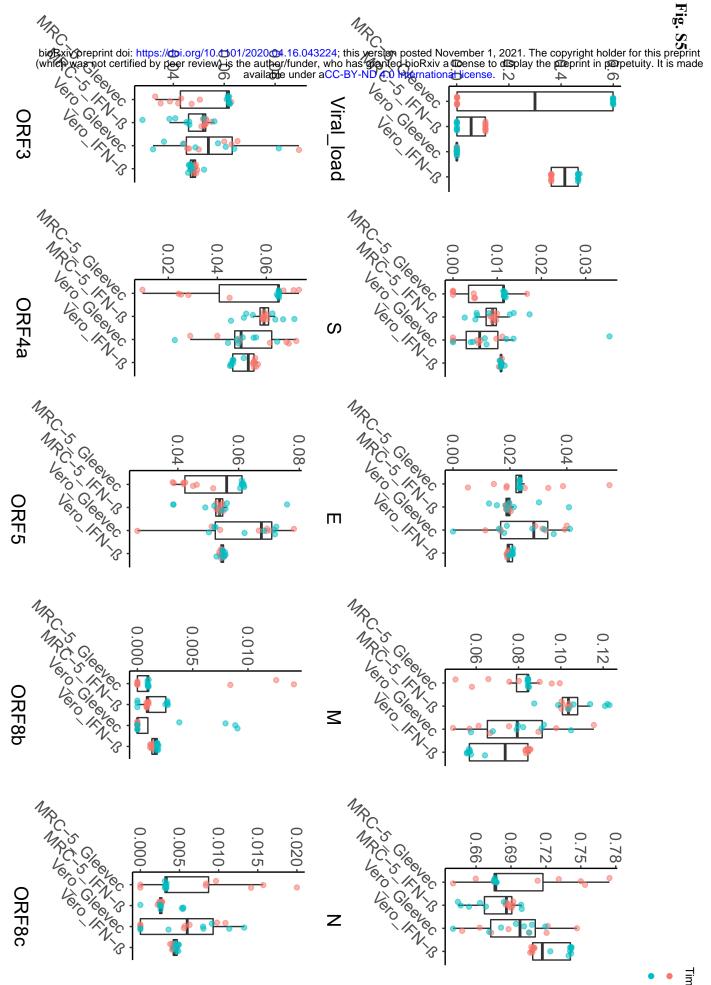


E

MERS-CoV pORF8c amongst selected Merbecoviruses







Time post-infection 24hpi 48hpi

Fig. S6

