Molecular Architecture of Early Dissemination and Massive Second Wave of the SARS-CoV-2 Virus in a Major Metropolitan Area

Running Title: Two waves of COVID-19 disease in Houston, Texas

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ABSTRACT We sequenced the genomes of 5,085 SARS-CoV-2 strains causing two COVID-19 disease waves in metropolitan Houston, Texas, an ethnically diverse region with seven million residents. The genomes were from viruses recovered in the earliest recognized phase of the pandemic in Houston, and an ongoing massive second wave of infections. The virus was originally introduced into Houston many times independently. Virtually all strains in the second wave have a Gly614 amino acid replacement in the spike protein, a polymorphism that has been linked to increased transmission and infectivity. Patients infected with the Gly614 variant strains had significantly higher virus loads in the nasopharynx on initial diagnosis. We found little evidence of a significant relationship between virus genotypes and altered virulence, stressing the linkage between disease severity, underlying medical conditions, and host genetics. Some regions of the spike protein - the primary target of global vaccine efforts - are replete with amino acid replacements, perhaps indicating the action of selection. We exploited the genomic data to generate defined single amino acid replacements in the receptor binding domain of spike protein that, importantly, produced decreased recognition by the neutralizing monoclonal antibody CR30022. Our study is the first analysis of the molecular architecture of SARS-CoV-2 in two infection waves in a major metropolitan region. The findings will help us to understand the origin, composition, and trajectory of future infection waves, and the potential

effect of the host immune response and therapeutic maneuvers on SARS-CoV-2 evolution.

IMPORTANCE **There is concern about second and subsequent waves of COVID-19 caused by the SARS-CoV-2 coronavirus occurring in communities globally that had an initial disease wave. Metropolitan Houston, Texas, with a population of 7 million, is experiencing a massive second disease wave that began in late May 2020. To understand SARS-CoV-2 molecular population genomic architecture, evolution, and relationship between virus genotypes and patient features, we sequenced the genomes of 5,085 SARS-CoV-2 strains from these two waves. Our study provides the first molecular characterization of SARS-CoV-2 strains causing two distinct COVID-19 disease waves.**

KEYWORDS: SARS-CoV-2, COVID-19 disease, genome sequencing, molecular population genomics, evolution

[Introduction]

Pandemic disease caused by the severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) virus is now responsible for massive human morbidity and mortality worldwide (1-5). The virus was first documented to cause severe respiratory infections in Wuhan, China, beginning in late December 2019 (6-9). Global dissemination occurred extremely rapidly and has affected major population centers on most continents (10, 11). In the United States, the Seattle and the New York City (NYC) regions have been especially important centers of COVID-19 disease caused by SARS-CoV-2. For example, as of August 19, 2020, there were 227,419 confirmed SARS-CoV-2 cases in NYC, causing 56,831 hospitalizations and 19,005 confirmed fatalities and 4,638 probable fatalities (12). Similarly, in Seattle and King County, 17,989 positive patients and 696 deaths have been reported as of August 18, 2020 (13).

The Houston metropolitan area is the fourth largest and most ethnically diverse city in the United States, with a population of approximately 7 million (14, 15). The 2,400-bed Houston Methodist health system has seven hospitals and serves a large, multiethnic, and socioeconomically diverse patient population throughout greater Houston (13, 14). The first COVID-19 case in metropolitan Houston was reported on March 5, 2020 with community spread occurring one week later (16). Many of the first cases in our region were associated with national or international travel in areas known to have SARS-CoV-2 virus outbreaks (16). A central molecular diagnostic laboratory serving all Houston Methodist hospitals and our very early adoption of a molecular test for the SARS-

CoV-2 virus permitted us to rapidly identify positive patients and interrogate genomic variation among strains causing early infections in the greater Houston area. Our analysis of SARS-CoV-2 genomes causing disease in Houston has continued unabated since early March and is ongoing. Genome sequencing and related efforts were expanded extensively in late May as we recognized that a prominent second wave was underway (**Figure 1**).

Here, we report that SARS-CoV-2 was introduced to the Houston area many times, independently, from diverse geographic regions, with virus genotypes representing genetic clades causing disease in Europe, Asia, South America and elsewhere in the United States. There was widespread community dissemination soon after COVID-19 cases were reported in Houston. Strains with a Gly614 amino acid replacement in the spike protein, a polymorphism that has been linked to increased transmission and *in vitro* cell infectivity, increased significantly over time and caused virtually all COVID-19 cases in the massive second disease wave. Patients infected with strains with the Gly614 variant had significantly higher virus loads in the nasopharynx on initial diagnosis. Some naturally occurring single amino acid replacements in the receptor binding domain of spike protein resulted in decreased reactivity with a neutralizing monoclonal antibody, consistent with the idea that some virus variants arise due to host immune pressure.

RESULTS

Description of metropolitan Houston. Houston, Texas, is located in the southwestern United States, 50 miles inland from the Gulf of Mexico. It is the most ethnically diverse city in the United States (14). Metropolitan Houston is comprised predominantly of Harris County plus parts of eight contiguous surrounding counties. In the aggregate, the metropolitan area includes 9,444 square miles**.** The estimated population size of metropolitan Houston is 7 million (https://www.houston.org/houston-data).

Epidemic curve characteristics over two disease waves. The first confirmed case of COVID-19 in the Houston metropolitan region was reported on March 5, 2020 (16), and the first confirmed case diagnosed in Houston Methodist hospitals was reported on March 6, 2020. The epidemic curve indicated a first wave of COVID-19 cases that peaked around April 11-15, followed by a decline in cases until May 11. Soon thereafter, the slope of the case curve increased with a very sharp uptick in confirmed cases beginning on June 12 (**Figure 1B**). We consider May 11 as the transition between waves, as this date is the inflection point of the cumulative new cases curve and had the absolute lowest number of new cases in the mid-May time period. Thus, for the data presented herein, wave 1 is defined as March 5 through May 11, 2020, and wave 2 is defined as May 12 through July 7, 2020. Epidemiologic trends within the Houston Methodist Hospital population were mirrored by data from Harris County and the greater metropolitan Houston region (**Figure 1A**). Through the 7th of July, 25,366 COVID-19 cases were reported in Houston, 37,776 cases in Harris County, and 53,330 in metropolitan Houston, including 9,823 cases in Houston Methodist

facilities (inpatients and outpatients) (https://www.tmc.edu/coronavirus-updates/infection-rate-in-the-greater-houston-area/ and https://harriscounty.maps.arcgis.com/apps/opsdashboard/index.html#/c0de71f8ea484b85bb5efcb7c07c6914).

During the first wave (early March through May 11), 11,476 COVID-19 cases were reported in Houston, including 1,729 cases in the Houston Methodist Hospital system. Early in the first wave (from March 5 through March 30, 2020), we tested 3,080 patient specimens. Of these, 406 (13.2%) samples were positive for SARS-CoV-2, representing 40% (358/898) of all confirmed cases in metropolitan Houston during that time period. As our laboratory was the first hospital-based facility to have molecular testing capacity for SARS-CoV-2 available on site, our strain samples are likely representative of COVID-19 infections during the first wave.

For the entire study period (March 5 through July 7, 2020), we tested 68,418 specimens from 55,800 patients. Of these, 9,121 patients (16.4%) had a positive test result, representing 17.1% (9,121/53,300) of all confirmed cases in metropolitan Houston. Thus, our strain samples are also representative of those responsible for COVID-19 infections in the massive second wave.

To test the hypothesis that, on average, the two waves affected different groups of patients, we analyzed individual patient characteristics (hospitalized and non-hospitalized) in each wave. Consistent with this hypothesis, we found significant differences in the COVID-19 patients in each wave (**Table S1**). For example, patients in the second wave were significantly younger, had fewer

196 comorbidities, were more likely to be Hispanic/Latino (by self-report), and lived in
197 zip codes with lower median incomes (**Table S1**). A detailed analysis of the
198 characteristics of patients hospitalized in Houston Methodist facilities in the two
199 waves has recently been published (17).

200 **SARS-CoV-2 genome sequencing and phylogenetic analysis**. To
201 investigate the genomic architecture of the virus across the two waves, we
202 sequenced the genomes of 5,085 SARS-CoV-2 strains dating to the earliest time
203 of confirmed COVID-19 cases in Houston. Analysis of SARS-CoV-2 strains
204 causing disease in the first wave (March 5 through May 11) identified the
205 presence of many diverse virus genomes that, in the aggregate, represent the
206 major clades identified globally to date (**Figure 1B**). Clades G, GH, GR, and S
207 were the four most abundantly represented phylogenetic groups (**Figure 1B**).
208 Strains with the Gly614 amino acid variant in spike protein represented 82% of
209 the SARS-CoV-2 strains in wave 1, and 99.9% in wave 2 (p<0.0001; Fisher's
210 exact test) (**Figure 1B**). This spike protein variant is characteristic of clades G,
211 GH, and GR. Importantly, strains with the Gly614 variant represented only 71%
212 of the specimens sequenced in March, the early part of wave 1 (**Figure 1B**). We
213 attribute the decrease in strains with this variant observed in the first two weeks
214 of March (**Figure 1B**) to fluctuation caused by the relatively fewer COVID-19
215 cases occurring during this period.

216 **Relating spatiotemporal genome analysis with virus genotypes over**
217 **two disease waves**. We examined the spatial and temporal mapping of genomic
218 data to investigate community spread during wave 1 (**Figure 2**). Rapid and

widespread community dissemination occurred soon after the initial COVID-19 cases were reported in Houston. The heterogenous virus genotypes present very early in wave 1 indicate that multiple strains independently entered metropolitan Houston, rather than introduction and spread of a single strain. An important observation was that strains of most of the individual subclades were distributed over broad geographic areas (**Figure S1**). These findings are consistent with the known ability of SARS-CoV-2 to spread very rapidly from person to person.

Relationship between virus clades, clinical characteristics of infected patients, and additional metadata. It is possible that SARS-CoV-2 genome subtypes have different clinical characteristics, analogous to what is believed to have occurred with Ebola virus (18-20) and known to occur for other pathogenic microbes (21). As an initial examination of this issue in SARS-CoV-2, we tested the hypothesis that patients with disease severe enough to warrant hospitalization were infected with a non-random subset of virus genotypes. We also examined the association between virus clades and disease severity based on overall mortality, highest level of required care (intensive care unit, intermediate care unit, inpatient or outpatient), need for mechanical ventilation, and length of stay. There was no simple relationship between virus clades and disease severity using these four indicators. Similarly, there was no simple relationship between virus clades and other metadata, such as sex, age, or ethnicity (**Figure S2**).

Machine learning analysis. Machine learning models can be used to identify complex relationships not revealed by statistical analyses. We built

machine learning models to test the hypothesis that virus genome sequence can predict patient outcomes including mortality, length of stay, level of care, ICU admission, supplemental oxygen use, and mechanical ventilation. Models to predict outcomes based on virus genome sequence alone resulted in low F1 scores less than 50% (0.41 – 0.49) and regression models showed similarly low R^2 values (-0.01 – -0.20) (**Table S2**). F1 scores near 50% are indicative of classifiers that are performing similarly to random chance. The use of patient metadata alone to predict patient outcome improved the model's F1 scores by 5-10% (0.51 – 0.56) overall. The inclusion of patient metadata with virus genome sequence data improved most predictions of outcomes, compared to genome sequence alone, to 50% to 55% F1 overall (0.42 – 0.55) in the models (**Table S2**). The findings are indicative of two possibilities that are not mutually exclusive. First, patient metadata, such as age and sex, may provide more signal for the model to use and thus result in better accuracies. Second, the model's use of single nucleotide polymorphisms (SNPs) may have resulted in overfitting. Most importantly, no SNP predicted a significant difference in outcome. A table of classifier accuracy scores and performance information is provided in **Table S2**.

Patient outcome and metadata correlations. Overall, very few metadata categories correlated with patient outcomes (**Table S3**). Mortality was independently correlated with both Rh factor-positive blood type and increasing age, with Pearson correlation coefficients (PCC) equal to 0.22 and 0.27, respectively. This means that 22-27% of the variation in mortality can be predicted from Rh factor-positive blood type and patient age, respectively. Length

of stay correlated independently with increasing age (PCC=0.20). Positive Rh factor also correlated with ICU length of stay (PCC=0.20). All other patient metadata correlations to outcomes had PCC less than 0.20 (**Table S3**).

We further analyzed outcomes correlated to isolates from wave 1 and 2, and the presence of the Gly614 variant in spike protein. Being in wave 1 was independently correlated with mechanical ventilation days, overall length of stay, and ICU length of stay, with PCC equal to 0.20, 0.18, and 0.14, respectively. Importantly, the presence of the Gly614 variant did not correlate with patient outcomes (**Table S3**).

Analysis of the *nsp12* polymerase gene. The SARS-CoV-2 genome encodes an RNA-dependent RNA polymerase (RdRp, also referred to as Nsp12) used in virus replication (22-25). Two amino acid substitutions (Phe479Leu and Val556Leu) in RdRp each confer significant resistance *in vitro* to remdesivir, an adenosine analog (26). Remdesivir is inserted into RNA chains by RdRp during replication, resulting in premature termination of RNA synthesis and inhibition of virus replication. This compound has shown prophylactic and therapeutic benefit against MERS-CoV and SARS-CoV-2 experimental infection in rhesus macaques (27, 28). Recent reports indicate that remdesivir has therapeutic benefit in some COVID-19 hospitalized patients (29-33), leading it to be now widely used in patients worldwide. Thus, it may be important to understand variation in RdRp in large strain samples.

To acquire data about allelic variation in the *nsp12* gene, we analyzed our 5,085 virus genomes. The analysis identified 265 SNPs, including 140

nonsynonymous (amino acid-altering) SNPs, resulting in amino acid replacements throughout the protein (**Table 1, Figure 3, Figure 4, Figure S3,** and **Figure S4**). The most common amino acid change was Pro322Leu, identified in 4,893 of the 5,085 (96%) patient isolates. This amino acid replacement is common in genomes from clades G, GH, and GR, which are distinguished from other SARS-CoV-2 clades by the presence of the Gly614 amino acid change in the spike protein. Most of the other amino acid changes in RdRp were present in relatively small numbers of strains, and some have been identified in other isolates in a publicly available database (34). Five prominent exceptions included amino acid replacements: Ala15Val in 138 strains, Met462Ile in 59 strains, Met600Ile in 75 strains, Thr907Ile in 45 strains, and Pro917Ser in 80 strains. All 75 Met600Ile strains were phylogenetically closely related members of clade G, and also had the Pro322Leu amino acid replacement characteristic of this clade (**Figure S3**). These data indicate that the Met600Ile change is likely the evolved state, derived from a precursor strain with the Pro322Leu replacement. Similarly, we investigated phylogenetic relationships among strains with the other four amino acid changes noted above. In all cases, the vast majority of strains with each amino acid replacement were found among individual subclades of strains (**Figure S3**).

Importantly, none of the observed amino acid polymorphisms in RdRp were located precisely at two sites known to cause *in vitro* resistance to remdesivir (26). Most of the amino acid changes are located distantly from the RNA-binding and catalytic sites (**Figure S4** and **Table 1**). However,

311 replacements at six amino acid residues (Ala442Val, Ala448Val, Ala553Pro/Val,
312 Gly682Arg, Ser758Pro, and Cys812Phe) may potentially interfere with either
313 remdesivir binding or RNA synthesis. Four (Ala442Val, Ala448Val,
314 Ala553Pro/Val, and Gly682Arg) of the six substitution sites are located
315 immediately above the nucleotide-binding site, that is comprised of Lys544,
316 Arg552, and Arg554 residues as shown by structural studies (**Figure 4**). The
317 positions of these four variant amino acid sites are comparable to Val556 (**Figure**
318 **4**), for which a Val556Leu mutation in SARS-CoV was identified to confer
319 resistance to remdesivir *in vitro* (26). The other two substitutions (Ser758Pro and
320 Cys812Phe) are inferred to be located either at, or in the immediate proximity of,
321 the catalytic active site, that is comprised of three contiguous residues (Ser758,
322 Asp759, and Asp760). A proline substitution we identified at Ser758 (Ser758Pro)
323 is likely to negatively impact RNA synthesis. Although Cys812 is not directly
324 involved in the catalysis of RNA synthesis, it is only 3.5 Å away from Asp760.
325 The introduction of the bulkier phenylalanine substitution at Cys812 (Cys812Phe)
326 may impair RNA synthesis. Consequently, these two substitutions are expected
327 to detrimentally affect virus replication or fitness.
328 **Analysis of the gene encoding the spike protein**. The densely glycosylated
329 spike protein of SARS-CoV-2 and its close coronavirus relatives binds directly to
330 host-cell angiotensin-converting enzyme 2 (ACE2) receptors to enter host cells
331 (35-37). Thus, the spike protein is a major translational research target, including
332 intensive vaccine and therapeutic antibody (35-64). Analysis of the gene
333 encoding the spike protein identified 470 SNPs, including 285 that produce

amino acid changes (**Table 2, Figure 5**). Forty-nine of these replacements (V11A, T51A, W64C, I119T, E156Q, S205A, D228G, L229W, P230T, N234D, I235T, T274A, A288V, E324Q, E324V, S325P, S349F, S371P, S373P, T385I, A419V, C480F, Y495S, L517F, K528R, Q628E, T632I, S708P, T719I, P728L, S746P, E748K, G757V, V772A, K814R, D843N, S884A, M902I, I909V, E918Q, S982L, M1029I, Q1142K, K1157M, Q1180R, D1199A, C1241F, C1247G, and V1268A) are not represented in a publicly available database (34) as of August 19, 2020. Interestingly, 25 amino acid sites have three distinct variants (that is, the reference amino acid plus two additional variant amino acids), and five amino acid sites (amino acid positions 21, 27, 228, 936, and 1050) have four distinct variants represented in our sample of 5,085 genomes (**Table 2, Figure 5**).

We mapped the location of amino acid replacements onto a model of the full-length spike protein (35, 65) and observed that the substitutions are found in each subunit and domain of the spike (**Figure 6**). However, the distribution of amino acid changes is not uniform throughout the protein regions. For example, compared to some other regions of the spike protein, the receptor binding domain (RBD) has relatively few amino acid changes, and the frequency of strains with these substitutions is low, each occurring in fewer than 10 isolates. This finding is consistent with the functional constraints on RBD to mediate interaction with ACE2. In contrast, the periphery of the S1 subunit NTD contains a dense cluster of substituted residues, with some single amino acid replacements found in 10–20 isolates (**Table 2, Figure 5, Figure 6**). Clustering of amino acid changes in a distinct region of the spike protein may be a signal of

positive selection. Inasmuch as infected patients make antibodies against the NTD, we favor the idea that host immune selection is one force contributing to some of the amino acid variation in this region. One NTD substitution, H49Y, was found in 142 isolates. This position is not well exposed on the surface of the NTD and is likely not a result of immune pressure. The same is true for another highly represented substitution, F1052L. This substitution was observed in 167 isolates, and F1052 is buried within the core of the S2 subunit. The substitution observed most frequently in the spike protein in our sample is D614G, a change observed in 4,895 of the isolates. As noted above, strains with the Gly614 variant significantly increased in wave 2 compared to wave 1.

As observed with RdRp, the majority of strains with each single amino acid change in the spike protein were found on a distinct phylogenetic lineage (**Figure S5**), indicating identity by descent. A prominent exception is the Leu5Phe replacement that is present in all major clades, suggesting that this amino acid change arose multiple times independently or very early in the course of SARS-CoV-2 evolution. Finally, we note that examination of the phylogenetic distribution of strains with multiple distinct amino acid replacements at the same site (e.g., Arg21Ile/Lys/Thr, Ala27Ser/Thr/Val, etc.) revealed that they were commonly found in different genetic branches, consistent with independent origin (**Figure S5**).

Cycle threshold (Ct) comparison of SARS-CoV-2 strains with either the Asp614 or Gly614 amino acid replacements in spike protein. It has been reported that patients infected with strains having spike protein Gly614 variant

have, on average, higher virus loads on initial diagnosis (66-70). To determine if this is the case in Houston strains, we examined the cycle threshold (Ct) for every sequenced strain that was detected from a patient specimen using the SARS-CoV-2 Assay done by the Hologic Panther instrument. We identified a significant difference (p<0.0001) between the mean Ct value for strains with an Asp614 (n=102) or Gly614 (n=812) variant of the spike protein (**Figure 7**). Strains with Gly614 had a Ct value significantly lower than strains with the Asp614 variant, indicating that patients infected with the Gly614 strains had, on average, higher virus loads on initial diagnosis than patients infected by strains with the Asp614 variant (**Figure 7**). This observation is consistent with the conjecture that, on average, strains with the Gly614 variant are better able to disseminate.

Characterization of recombinant proteins with single amino acid replacements in the receptor binding domain region of spike protein. The RBD of spike protein binds the ACE2 surface receptor and is also targeted by neutralizing (36, 37, 41, 43-46, 48-62, 71). Thus, single amino acid replacements in this domain may have functional consequences that enhance virus fitness. To begin to test this idea, we expressed spike variants with the Asp614Gly replacement and 13 clinical RBD variants identified in our genome sequencing studies (**Figure 8, Table S4, S5**). All RBD variants were cloned into an engineered spike protein construct that stabilizes the perfusion state and increases overall expression yield (spike-6P, here referred to as spike) (64).

We first assessed the biophysical properties of spike-Asp614Gly, an amino acid polymorphism that is common globally and increased significantly in our wave 2 strain isolates. Pseudotyped viruses expressing spike-Gly614 have higher infectivity for host cells *in vitro* than spike-Asp614 (66, 67, 69, 72, 73). The higher infectivity of spike-Gly614 is correlated with increased stability and incorporation of the spike protein into the pseudovirion (73). We observed a higher expression level (**Figure 8A, B**) and increased thermostability for the spike protein construct containing this variant (**Figure 8C, D**). The size exclusion chromatography (SEC) elution profile of spike-Asp614 was indistinguishable from spike-Gly614, consistent with a trimeric conformation (**Figure 8A**). These results are broadly consistent with higher-resolution structural analyses of both spike variants.

Next, we purified and biophysically characterized 13 RBD mutants that each contain Gly614 and one additional single amino acid replacement we identified by genome sequencing our clinical samples (**Table S6**). All variants eluted as trimers, indicating the global structure, remained intact (**Figure 8** and **Figure S6**). However, several variants had reduced expression levels and virtually all had decreased thermostability relative to the variant that had only a D614G single amino acid replacement (**Figure 8D**). The A419V and A522V mutations were especially deleterious, reducing yield and precluding further downstream analysis (**Figure 8B**). We next assayed the affinity of the 11 highest-expressing spike variants for ACE2 and the neutralizing monoclonal antibody CR3022 via enzyme-linked immunosorbent assays (ELISAs) (**Figure 8E-G** and

Table S6). Most variants retained high affinity for the ACE2 surface receptor. However, importantly, three RBD variants (F338L, S373P, and R408T) had substantially reduced affinity for CR3022, a monoclonal antibody that disrupts the spike protein homotrimerization interface (63, 74). Notably, the S373P mutation is one amino acid away from the epitope recognized by CR3022. These results are consistent with the interpretation that some RBD mutants arising in COVID-19 patients may have increased ability to escape humoral immune pressure, but otherwise retain strong ACE2 binding affinity.

DISCUSSION

In this work we analyzed the molecular population genomics, sociodemographic, and medical features of two waves of COVID-19 disease occurring in metropolitan Houston, Texas, between early March and early July 2020. We also studied the biophysical and immunologic properties of some naturally occurring single amino acid changes in the spike protein RBD identified by sequencing the 5,085 genomes. We discovered that the first COVID-19 wave was caused by a heterogenous array of virus genotypes assigned to several different clades. The majority of cases in the first wave are related to strains that caused widespread disease in European and Asian countries, as well as other localities. We conclude that the SARS-CoV-2 virus was introduced into Houston many times independently, likely by individuals who had traveled to or from different parts of the world, including other communities in the United States. In support of this conclusion, the first cases in metropolitan Houston were associated with a travel

history to a known COVID-19 region (16). The data are consistent with the fact that Houston is a large international city characterized by a multi-ethnic population and is a prominent transport hub with direct flights to major cities globally.

The second wave of COVID-19 cases also is characterized by SARS-CoV-2 strains with diverse genotypes. Virtually all cases in the second and ongoing disease wave were caused by strains with the Gly614 variant of spike protein (**Figure 1B**). Our data unambiguously demonstrate that strains with the Gly614 variant increased significantly in frequency in wave 2 relative to wave 1 in the Houston metropolitan region. This shift occurred very rapidly in a matter of just a few months. Amino acid residue Asp614 is located in subdomain 2 (SD-2) of the spike protein and forms a hydrogen bond and electrostatic interaction with two residues in the S2 subunit of a neighboring protomer. Replacement of aspartate with glycine would eliminate both interactions, thereby substantively weakening the contact between the S1 and S2 subunits. We previously speculated (75) that this weakening produces a more fusogenic spike protein, as S1 must first dissociate from S2 before S2 can refold and mediate fusion of virus and cell membranes. Stated another way, virus strains with the Gly614 variant may be better able to enter host cells, potentially resulting in enhanced spread. Consistent with this idea, Korber et al. (66) showed that the Gly614 variant grows to higher titer as pseudotyped virions. On initial diagnosis infected individuals had lower RT-PCR cycle thresholds suggesting higher upper respiratory tract viral loads. Our data (**Figure 7**) are fully consistent with that finding Zhang et al. (73)

reported that pseudovirus with the 614Gly variant infected ACE2-expressing cells more efficiently than the 614Asp. Similar results have been described by Hu et al. (67) and Lorenzo-Redondo et al. (68). Plante et al. (76) recently studied isogenic mutant SARS-CoV-2 strains with either the 614Asp or 614Gly variant and found that the 614Gly variant virus had significantly increased replication in human lung epithelial cells *in vitro* and increased infectious titers in nasal and trachea washes obtained from experimentally infected hamsters. These results are consistent with the idea that the 614Gly variant bestows increased virus fitness in the upper respiratory tract (76).

Additional work is needed to investigate the potential biomedical relevance and public health importance of the Asp614Gly polymorphism, including but not limited to virus dissemination, overall fitness, impact on clinical course and virulence, and development of vaccines and therapeutics. Although it is possible that stochastic processes alone may account for the rapid increase in COVID-19 disease frequency caused by viruses containing the Gly614 variant, we do not favor that interpretation in part because of the cumulative weight of the epidemiologic, human RT-PCR diagnostics data, *in vitro* experimental findings, and animal infection studies using isogenic mutant virus strains. In addition, if stochastic processes solely are responsible, we believe it is difficult to explain essentially simultaneous increase in frequency of the Gly614 variant in genetically diverse viruses in three distinct clades (G, GH, and GR) in a geographically large metropolitan area with 7 million ethnically diverse people. Regardless, more research on this important topic is warranted.

Several groups have analyzed the relationship between human genetics, blood types, and patient outcomes (33, 34, 77-80). In this regard, we identified a correlation between Rh factor-positive blood type and increased mortality and length of ICU stay. (Supplemental Table 3). Thus, our data are consistent with studies suggesting that there are host genetic factors that contribute to disease severity and outcome. More research is required to address the important possibility.

The diversity present in our 1,026 virus genomes from the first disease wave contrasts somewhat with data reported by Gonzalez-Reiche et al., who studied 84 SARS-CoV-2 isolates causing disease in patients in the New York City region (11). Those investigators concluded that the vast majority of disease was caused by progeny of strains imported from Europe. Similarly, Bedford et al. (10) reported that much of the COVID-19 disease in the Seattle, Washington area was caused by strains that are progeny of a virus strain recently introduced from China. Some aspects of our findings are similar to those reported recently by Lemieux et al. based on analysis of strains causing disease in the Boston area (81). Our findings, like theirs, highlight the importance of multiple importation events of genetically diverse strains in the epidemiology of COVID-19 disease in this pandemic. Similarly, Icelandic and Brazilian investigators documented that SARS-CoV-2 was imported by individuals traveling to or from many European and other countries (82, 83).

The virus genome diversity and large sample size in our study permitted us to test the hypothesis that distinct virus clades were nonrandomly associated

with hospitalized COVID-19 patients or disease severity. We did not find evidence to support this hypothesis, but our continuing study of COVID-19 cases accruing in the second wave will further improve statistical stratification.

We used machine learning classifiers to identify if any SNPs contribute to increased infection severity or otherwise affect virus-host outcome. The models could not be trained to accurately predict these outcomes from the available virus genome sequence data. This may be due to sample size or class imbalance. However, we do not favor this interpretation. Rather, we think that the inability to identify particular virus SNPs predictive of disease severity or infection outcome likely reflects the substantial heterogeneity in underlying medical conditions and treatment regimens among COVID-19 patients studied herein. An alternative but not mutually exclusive hypothesis is that patient genotypes play an important role in determining virus-human interactions and resulting pathology. Although some evidence has been presented in support of this idea (84, 85), available data suggest that in the aggregate, host genetics does not play an overwhelming role in determining outcome in the great majority of adult patients, once virus infection is established.

Remdesivir is a nucleoside analog reported to have activity against MERS-CoV, a coronavirus related to SARS-CoV-2. Recently, several studies have reported that remdesivir shows promise in treating COVID-19 patients (29-33), leading the FDA to issue an emergency use authorization. Because *in vitro* resistance of SARS-CoV to remdesivir has been reported to be caused by either of two amino acid replacements in RdRp (Phe479Leu or Val556Leu), we

interrogated our data for polymorphisms in the *nsp12* gene. Although we identified 140 different inferred amino acid replacements in RdRp in the 5,085 genomes analyzed, none of these were located precisely at the two positions associated with *in vitro* resistance to remdesivir. Inasmuch as remdesivir is now being deployed widely to treat COVID-19 patients in Houston and elsewhere, our findings suggest that the majority of SARS-CoV-2 strains currently circulating in our region should be susceptible to this drug.

The amino acid replacements Ala442Val, Ala448Val, Ala553Pro/Val, and Gly682Arg that we identified occur at sites that, intriguingly, are located directly above the nucleotide substrate entry channel and nucleotide binding residues Lys544, Arg552, and Arg554 (22, 23) (**Figure 4**). One possibility is that substitution of the smaller alanine or glycine residues with the bulkier side chains of Val/Pro/Arg may impose structural constraints for the modified nucleotide analog to bind, and thereby disfavor remdesivir binding. This, in turn, may lead to reduced incorporation of remdesivir into the nascent RNA, increased fidelity of RNA synthesis, and ultimately drug resistance. A similar mechanism has been proposed for a Val556Leu change (23).

We also identified one strain with a Lys477Asn replacement in RdRp. This substitution is located close to a Phe479Leu replacement reported to produce partial resistance to remdesivir *in vitro* in SARS-CoV patients from 2004, although the amino acid positions are numbered differently in SARS-CoV and SARS-CoV-2. Structural studies have suggested that this amino acid is surface-exposed, and distant from known key functional elements. Our observed

Lys477Asn change is also located in a conserved motif described as a finger domain of RdRp (**Figure 3** and **4**). One speculative possibility is that Lys477 is involved in binding a yet unidentified cofactor such as Nsp7 or Nsp8, an interaction that could modify nucleotide binding and/or fidelity at a distance. These data warrant additional study in larger patient cohorts, especially in individuals treated with remdesivir.

Analysis of the gene encoding the spike protein identified 285 polymorphic amino acid sites relative to the reference genome, including 49 inferred amino acid replacements not present in available databases as of August 19, 2020. Importantly, 30 amino acid sites in the spike protein had two or three distinct replacements relative to the reference strain. The occurrence of multiple variants at the same amino acid site is one characteristic that may suggest functional consequences. These data, coupled with structural information available for spike protein, raise the possibility that some of the amino acid variants have functional consequences, for example including altered serologic reactivity and shown here. These data permit generation of many biomedically relevant hypotheses now under study.

A recent study reported that RBD amino acid changes could be selected *in vitro* using a pseudovirus neutralization assay and sera obtained from convalescent plasma or monoclonal antibodies (86). The amino acid sites included positions V445 and E484 in the RBD. Important to note, variants G446V and E484Q were present in our patient samples. However, these mutations retain high affinity to CR3022 **(Figure 8F, G**). The high-resolution structure of the

586 RBD/CR3022 complex shows that CR3022 makes contacts to residues 369-386,
587 380-392, and 427-430 of RBD (74). Although there is no overlap between
588 CR3022 and ACE2 epitopes, CR3022 is able to neutralize the virus through an
589 allosteric effect. We found that the Ser373Pro change, which is located within the
590 CR3022 epitope, has reduced affinity to CR3022 (**Figure 8F, G**). The F338L and
591 R408T mutations, although not found directly within the interacting epitope, also
592 display reduced binding to CR3022. Other investigators (86) using *in vitro*
593 antibody selection identified a change at amino acid site S151 in the N-terminal
594 domain, and we found mutations S151N and S151I in our patient samples. We
595 also note that two variant amino acids (Gly446Val and Phe456Leu) we identified
596 are located in a linear epitope found to be critical for a neutralizing monoclonal
597 antibody described recently by Li et al. (87).

598 In the aggregate, these findings suggest that mutations emerging within
599 the spike protein at positions within and proximal to known neutralization
600 epitopes may result in escape from antibodies and other therapeutics currently
601 under development. Importantly, our study did not reveal that these mutant
602 strains had disproportionately increased over time. The findings may also bear
603 on the occurrence of multiple amino acid substitutions at the same amino acid
604 site that we identified in this study, commonly a signal of selection. In the
605 aggregate, the data support a multifaceted approach to serological monitoring
606 and biologics development, including the use of monoclonal antibody cocktails
607 (46, 47, 88).
608

CONCLUDING STATEMENT

Our work represents analysis of the largest sample to date of SARS-CoV-2 genome sequences from patients in one metropolitan region in the United States. The investigation was facilitated by the fact that we had rapidly assessed a SARS-CoV-2 molecular diagnostic test in January 2020, more than a month before the first COVID-19 patient was diagnosed in Houston. In addition, our large healthcare system has seven hospitals and many facilities (e.g., outpatient care centers, emergency departments) located in geographically diverse areas of the city. We also provide reference laboratory services for other healthcare entities in the Houston area. Together, our facilities serve patients of diverse ethnicities and socioeconomic status. Thus, the data presented here likely reflect a broad overview of virus diversity causing COVID-19 infections throughout metropolitan Houston. We previously exploited these features to study influenza and *Klebsiella pneumoniae* dissemination in metropolitan Houston (89, 90). We acknowledge that every “twig” of the SARS-CoV-2 evolutionary tree in Houston is not represented in these data. The samples studied are not comprehensive for the entire metropolitan region. For example, it is possible that our strain samples are not fully representative of individuals who are indigent, homeless, or of very low socioeconomic groups. In addition, although the strain sample size is relatively large compared to other studies, the sample represents only about 10% of all COVID-19 cases in metropolitan Houston documented in the study period. In addition, some patient samples contain relatively small amounts of virus nucleic acid and do not yield adequate sequence data for high-quality genome

analysis. Thus, our data likely underestimate the extent of genome diversity present among SARS-CoV-2 causing COVID-19 and will not identify all amino acid replacements in the virus in this geographic region. It will be important to sequence and analyze the genomes of additional SARS-CoV-2 strains causing COVID-19 cases in the ongoing second massive disease wave in metropolitan Houston, and these studies are underway. Data of this type will be especially important to have if a third and subsequent waves were to occur in metropolitan Houston, as it could provide insight into molecular and epidemiologic events contributing to them.

The genomes reported here are an important data resource that will underpin our ongoing study of SARS-CoV-2 molecular evolution, dissemination, and medical features of COVID-19 in Houston. As of August 19, 2020, there were 135,866 reported cases of COVID-19 in metropolitan Houston, and the number of cases is increasing daily. Although the full array of factors contributing to the massive second wave in Houston is not known, it is possible that the potential for increased transmissibility of SARS-CoV-2 with the Gly614 may have played a role, as well as changes in behavior associated with the Memorial Day and July 4th holidays, and relaxation of some of the social constraints imposed during the first wave. The availability of extensive virus genome data dating from the earliest reported cases of COVID-19 in metropolitan Houston, coupled with the database we have now constructed, may provide critical insights into the origin of new infection spikes and waves occurring as public health constraints are further relaxed, schools and colleges re-open, holidays occur, commercial air

travel increases, and individuals change their behavior because of COVID-19 "fatigue." The genome data will also be useful in assessing ongoing molecular evolution in spike and other proteins as baseline herd immunity is generated, either by natural exposure to SARS-CoV-2 or by vaccination. The signal of potential selection contributing to some spike protein diversity and identification of naturally occurring mutant RBD variants with altered serologic recognition warrant close attention and expanded study.

MATERIALS AND METHODS

Patient specimens. All specimens were obtained from individuals who were registered patients at Houston Methodist hospitals, associated facilities (e.g., urgent care centers), or institutions in the greater Houston metropolitan region that use our laboratory services. Virtually all individuals met the criteria specified by the Centers for Disease Control and Prevention to be classified as a person under investigation.

SARS-CoV-2 molecular diagnostic testing. Specimens obtained from symptomatic patients with a high degree of suspicion for COVID-19 disease were tested in the Molecular Diagnostics Laboratory at Houston Methodist Hospital using an assay granted Emergency Use Authorization (EUA) from the FDA (https://www.fda.gov/medical-devices/emergency-situations-medical-devices/faqs-diagnostic-testing-sars-cov-2#offeringtests). Multiple testing platforms were used, including an assay that follows the protocol published by

the WHO (https://www.who.int/docs/default-source/coronaviruse/protocol-v2-1.pdf) using the EZ1 virus extraction kit and EZ1 Advanced XL instrument or QIASymphony DSP Virus kit and QIASymphony instrument for nucleic acid extraction and ABI 7500 Fast Dx instrument with 7500 SDS software for reverse transcription RT-PCR, the COVID-19 test using BioFire Film Array 2.0 instruments, the Xpert Xpress SARS-CoV-2 test using Cepheid GeneXpert Infinity or Cepheid GeneXpert Xpress IV instruments, the SARS-CoV-2 Assay using the Hologic Panther instrument, and the Aptima SARS-CoV-2 Assay using the Hologic Panther Fusion system. All assays were performed according to the manufacturer's instructions. Testing was performed on material obtained from nasopharyngeal or oropharyngeal swabs immersed in universal transport media (UTM), bronchoalveolar lavage fluid, or sputum treated with dithiothreitol (DTT). To standardize specimen collection, an instructional video was created for Houston Methodist healthcare workers (https://vimeo.com/396996468/2228335d56).

Epidemiologic curve. The number of confirmed COVID-19 positive cases was obtained from USAFacts.org (https://usafacts.org/visualizations/coronavirus-covid-19-spread-map/) for Austin, Brazoria, Chambers, Fort Bend, Galveston, Harris, Liberty, Montgomery, and Waller counties. Positive cases for Houston Methodist Hospital patients were obtained from our Laboratory Information System and plotted using the documented collection time.

701 **SARS-CoV-2 genome sequencing**. Libraries for whole virus genome
702 sequencing were prepared according to version 1 or version 3 of the ARTIC
703 nCoV-2019 sequencing protocol (https://artic.network/ncov-2019). Long reads
704 were generated with the LSK-109 sequencing kit, 24 native barcodes (NBD104
705 and NBD114 kits), and a GridION instrument (Oxford Nanopore). Short reads
706 were generated with the NexteraXT kit and NextSeq 550 instrument (Illumina).
707
708 **SARS-CoV-2 genome sequence analysis**. Consensus virus genome
709 sequences from the Houston area isolates were generated using the ARTIC
710 nCoV-2019 bioinformatics pipeline. Publicly available genomes and metadata
711 were acquired through GISAID on August 19, 2020. GISAID sequences
712 containing greater than 1% N characters, and Houston sequences with greater
713 than 5% N characters were removed from consideration. Identical GISAID
714 sequences originating from the same geographic location with the same
715 collection date were also removed from consideration to reduce redundancy.
716 Nucleotide sequence alignments for the combined Houston and GISAID strains
717 were generated using MAFFT version 7.130b with default parameters (91).
718 Sequences were manually curated in JalView (92) to trim the ends and to
719 remove sequences containing spurious inserts. Phylogenetic trees were
720 generated using FastTree with the generalized time-reversible model for
721 nucleotide sequences (93). CLC Genomics Workbench (QIAGEN) was used to
722 generate the phylogenetic tree figures.
723

Geospatial mapping. The home address zip code for all SARS-CoV-2 positive patients was used to generate the geospatial maps. To examine geographic relatedness among genetically similar isolates, geospatial maps were filtered to isolates containing specific amino acid changes.

Time series. Geospatial data were filtered into wave 1 (3/5/2020-5/11/2020) and wave 2 (5/12/2020-7/7/2020) time intervals to illustrate the spread of confirmed SARS-CoV-2 positive patients identified over time.

Machine learning. Virus genome alignments and patient metadata were used to build models to predict patient metadata and outcomes using both classification models and regression. Metadata considered for prediction in the classification models included age, ABO and Rh blood type, ethnic group, ethnicity, sex, ICU admission, IMU admission, supplemental oxygen use, and ventilator use. Metadata considered for prediction in regression analysis included ICU length of stay, IMU length of stay, total length of stay, supplemental oxygen use, and ventilator use. Because sex, blood type, Rh factor, age, age decade, ethnicity, and ethnic group are features in the patient features and combined feature sets, models were not trained for these labels using patient and combined feature sets. Additionally, age, length of stay, IMU length of stay, ICU length of stay, mechanical ventilation days, and supplemental oxygen days were treated as regression problems and XGBoost regressors were built while the rest were treated as classification problems and XGBoost classifiers were built.

Three types of features were considered for training the XGBoost classifiers: alignment features, patient features, and the combination of alignment and patient features. Alignment features were generated from the consensus genome alignment such that columns containing ambiguous nucleotide bases were removed to ensure the models did not learn patterns from areas of low coverage. These alignments were then one-hot encoded to form the alignment features. Patient metadata values were one-hot encoded with the exception of age, which remained as a raw integer value, to create the patient features. These metadata values consisted of age, ABO, Rh blood type, ethnic group, ethnicity, and sex. All three types of feature sets were used to train models that predict ICU length of stay, IMU length of stay, overall length of stay, days of supplemental oxygen therapy, and days of ventilator usage while only alignment features were used to train models that predict age, ABO, Rh blood type, ethnic group, ethnicity, and sex.

A ten-fold cross validation was used to train XGBoost models (94) as described previously (95, 96). Depths of 4, 8, 16, 32, and 64 were used to tune the models, but accuracies plateaued after a depth of 16. SciKit-Learn's (97) classification report and r2 score were then used to access overall accuracy of the classification and regression models, respectively.

Patient metadata correlations. We encoded values into multiple columns for each metadata field. For example, the ABO column was divided into four columns for A, B, AB, and O blood type. Those columns were encoded with a 1

for the patients' ABO type, with all other columns encoded with 0. This was repeated for all non-outcome metadata fields. Age, however, was not re-encoded, as the raw integer values were used. Each column was then correlated to the various outcome values for each patient (deceased, ICU length, IMU length, length of stay, supplemental oxygen length, and ventilator length) to obtain a Pearson coefficient correlation value for each metadata label and outcome.

Analysis of the *nsp12* polymerase and S protein genes. The *nsp12* virus polymerase and S protein genes were analyzed by plotting SNP density in the consensus alignment using Python (Python v3.4.3, Biopython Package v1.72). The frequency of SNPs in the Houston isolates was assessed, along with amino acid changes for nonsynonymous SNPs.

Cycle threshold (Ct) comparison of SARS-CoV-2 strains with either Asp614 or Gly614 amino acid replacements in the spike protein. The cycle threshold (Ct) for every sequenced strain that was detected from a patient specimen using the SARS-CoV-2 Assay on the Hologic Panther instrument was retrieved from the Houston Methodist Hospital Laboratory Information System. Statistical significance between the mean Ct value for strains with an aspartate (n=102) or glycine (n=812) amino acid at position 614 of the spike protein was determined with the Mann-Whitney test (GraphPad PRISM 8).

Creation and characterization of spike protein RBD variants. Spike RBD variants were cloned into the spike-6P (HexaPro; F817P, A892P, A899P, A942P, K986P, V987P) base construct that also includes the D614G substitution (pIF638). Briefly, a segment of the gene encoding the RBD was excised with EcoRI and NheI, mutagenized by PCR, and assembled with a HiFi DNA Assembly Cloning Kit (NEB).

FreeStyle 293-F cells (Thermo Fisher Scientific) were cultured and maintained in a humidified atmosphere of 37°C and 8% CO_2 while shaking at 110-125rpm. Cells were transfected with plasmids encoding spike protein variants using polyethylenimine. Three hours post-transfection, 5μM kifunensine was added to each culture. Cells were harvested four days after transfection and the protein containing supernatant was separated from the cells by two centrifugation steps: 10 min at 500rcf and 20 min at 10,000rcf. Supernatants were kept at 4°C throughout. Clarified supernatant was loaded on a Poly-Prep chromatography column (Bio-Rad) containing Strep-Tactin Superflow resin (IBA), washed with five column volumes (CV) of wash buffer (100mM Tris-HCl pH 8.0, 150mM NaCl; 1mM EDTA), and eluted with four CV of elution buffer (100mM Tris-HCl pH 8.0, 150mM NaCl, 1mM EDTA, 2.5mM d-Desthiobiotin). The eluate was spin-concentrated (Amicon Ultra-15) to 600μL and further purified via size-exclusion chromatography (SEC) using a Superose 6 Increase 10/300 column (G.E.) in SEC buffer (2mM Tris pH 8.0, 200mM NaCl and 0.02% NaN_3). Proteins were concentrated to 300μL and stored in SEC buffer.

The RBD spike mutants chosen for analysis were all RBD amino acid mutants identified by our genome sequencing study as of June 15, 2020. We note that the exact boundaries of the RBD domain varies depending on the paper used as reference. We used the boundaries demarcated in Figure 1A of Cai et al. Science paper 21 July) (98) that have K528R located at the RBD-CTD1 interface.

Differential scanning fluorimetry. Recombinant spike proteins were diluted to a final concentration of 0.05mg/mL with 5X SYPRO orange (Sigma) in a 96-well qPCR plate. Continuous fluorescence measurements (λex=465nm, λem=580nm) were collected with a Roche LightCycler 480 II. The temperature was increased from 22°C to 95°C at a rate of 4.4°C/min. We report the first melting transition.

Enzyme-linked immunosorbent assays. ELISAs were performed to characterize binding of S6P, S6P D614G, and S6P D614G-RBD variants to human ACE2 and the RBD-binding monoclonal antibody CR3022. The ACE2-hFc chimera was obtained from GenScript (Z03484), and the CR3022 antibody was purchased from Abcam (Ab273073). Corning 96-well high-binding plates (CLS9018BC) were coated with spike variants at 2μg/mL overnight at 4°C. After washing four times with phosphate buffered saline + 0.1% Tween20 (PBST; 300μL/well), plates were blocked with PBS+2% milk (PBSM) for 2 h at room temperature and again washed four times with PBST. These were serially diluted in PBSM 1:3 seven times in triplicate. After 1 h incubation at room temperature,

plates were washed four times in PBST, labeled with 50µL mouse anti-human IgG1 Fc-HRP (SouthernBlots, 9054-05) for 45 min in PBSM, and washed again in PBST before addition of 50µL 1-step Ultra TMB-ELISA substrate (Thermo Scientific, 34028). Reactions were developed for 15 min and stopped by addition of 50µL 4M H_2SO_4. Absorbance intensity (450nm) was normalized within a plate and EC_{50} values were calculated through 4-parameter logistic curve (4PL) analysis using GraphPad PRISM 8.4.3.

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Data and material availability: The spike-6P (“HexaPro”) plasmid is available from Addgene (ID: 154754) or from I.J.F. under a material transfer agreement with The University of Texas at Austin. Additional plasmids are available upon request from I.J.F.

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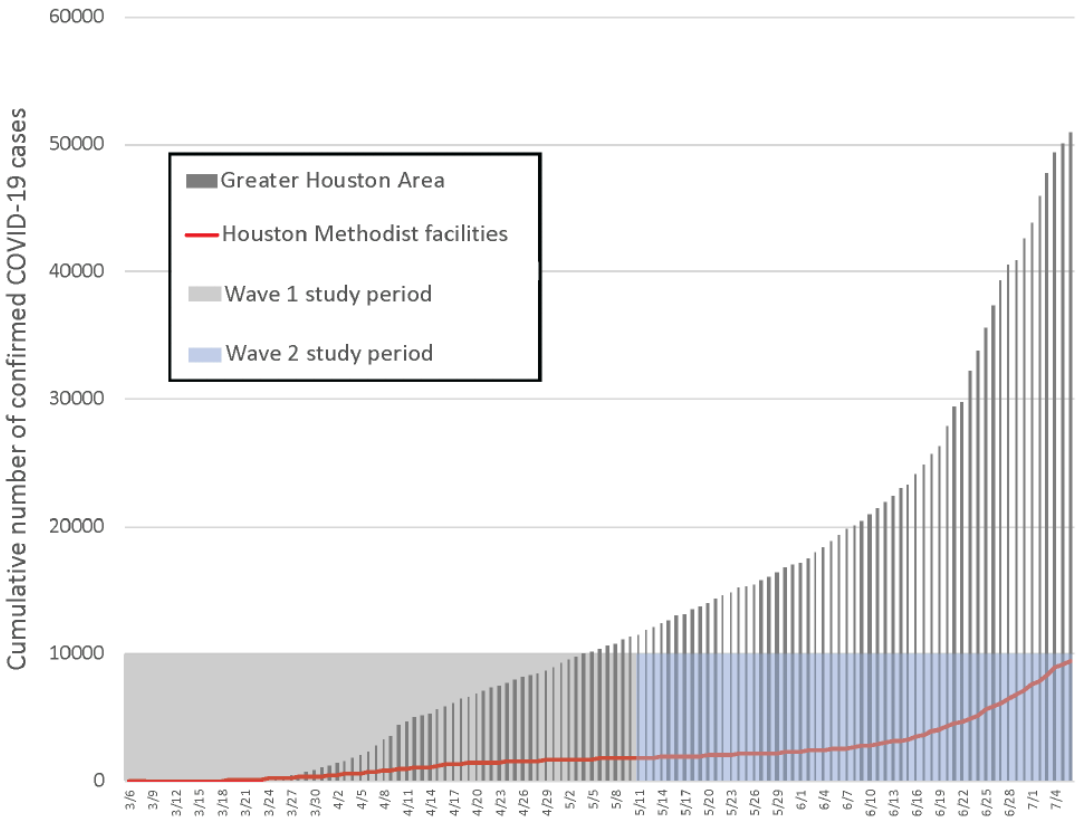
A



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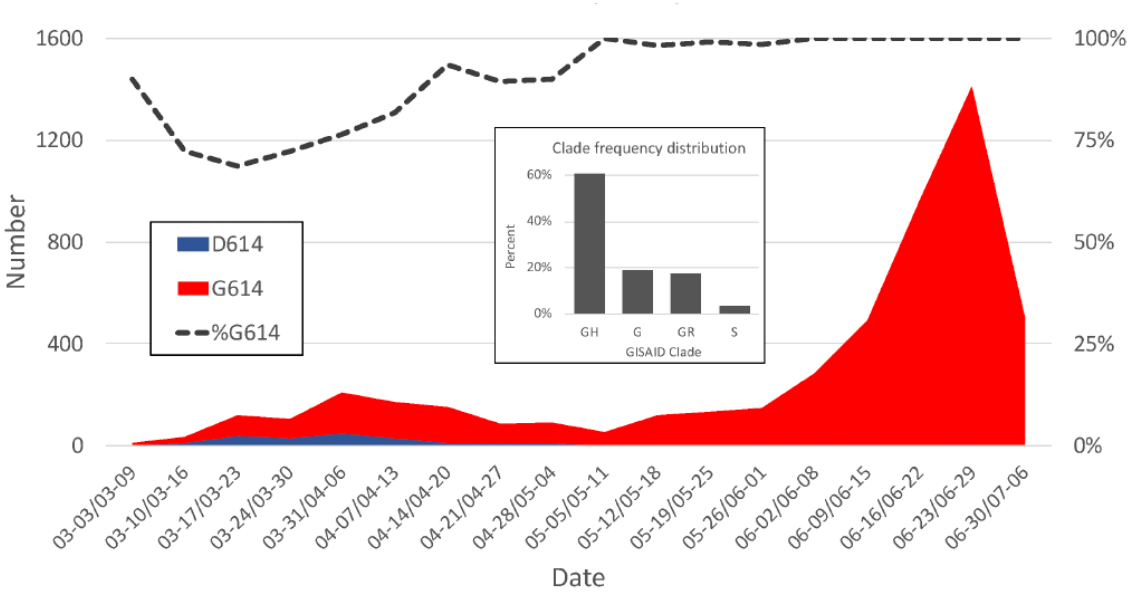
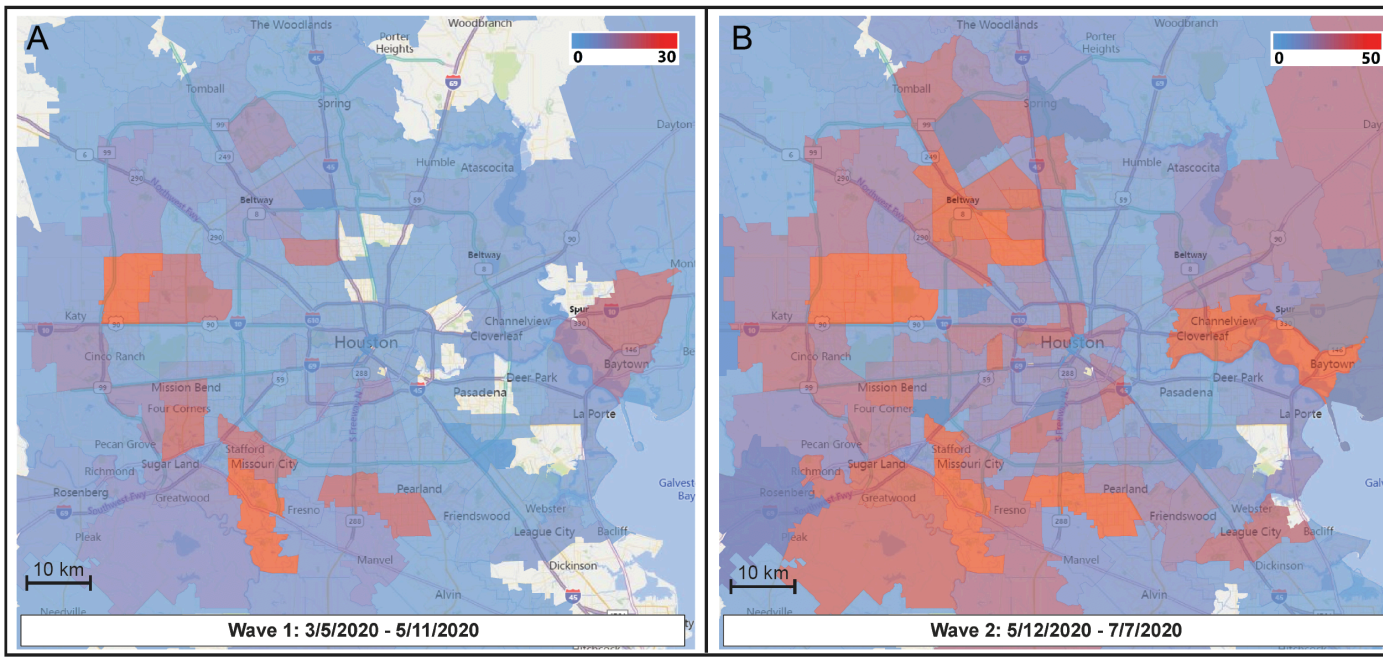


1330

FIG 1 (A) Confirmed COVID-19 cases in the Greater Houston Metropolitan region. Cumulative number of COVID-19 patients over time through July 7, 2020. Counties include Austin, Brazoria, Chambers, Fort Bend, Galveston, Harris, Liberty, Montgomery, and Waller. The shaded area represents the time period during which virus genomes characterized in this study were recovered from COVID-19 patients. The red line represents the number of COVID-19 patients diagnosed in the Houston Methodist Hospital Molecular Diagnostic Laboratory. (B) Distribution of strains with either the Asp614 or Gly614 amino acid variant in spike protein among the two waves of COVID-19 patients diagnosed in the Houston Methodist Hospital Molecular Diagnostic Laboratory. The large inset shows major clade frequency for the time frame studied.



1342

1343 **FIG 2** Sequential time-series heatmaps for all COVID-19 Houston Methodist patients
1344 during the study period. Geospatial distribution of COVID-19 patients is based on zip
1345 code. Panel A (left) shows geospatial distribution of sequenced SARS-CoV-2 strains in
1346 wave 1 and panel B (right) shows wave 2 distribution. The collection dates are shown at
1347 the bottom of each panel. The insets refer to numbers of strains in the color spectrum
1348 used. Note difference in numbers of strains used in panel A and panel B insets.

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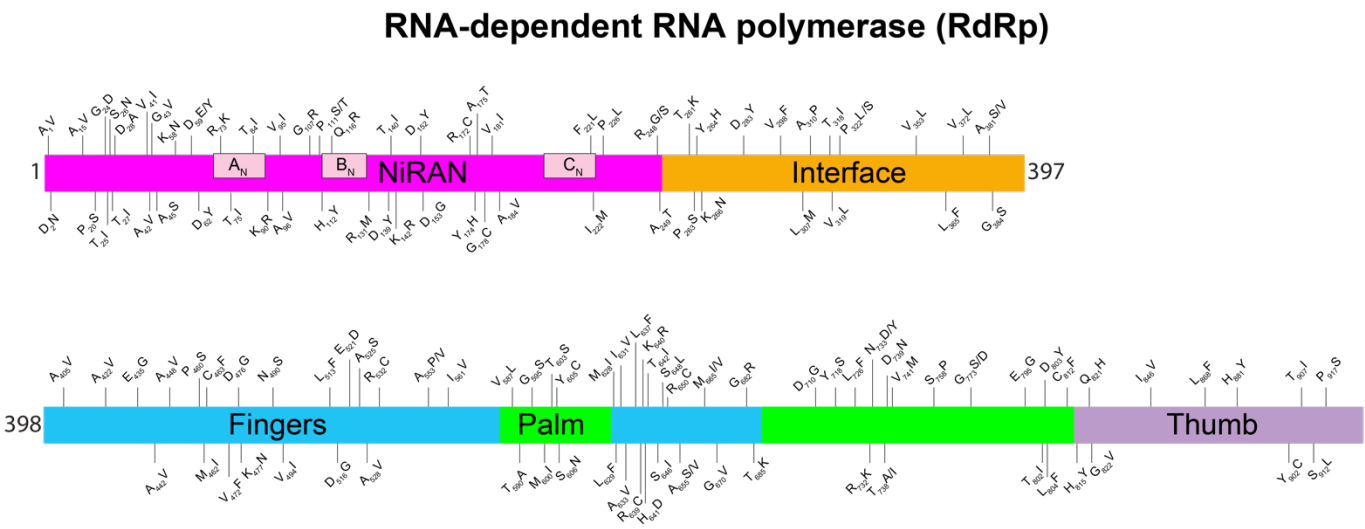

1350

1351 **FIG 3** Location of amino acid replacements in RNA-dependent RNA polymerase

1352 (RdRp/Nsp12) among the 5,085 genomes of SARS-CoV-2 sequenced. The various

1353 RdRp domains are color-coded. The numbers refer to amino acid site. Note that several

1354 amino acid sites have multiple variants identified.

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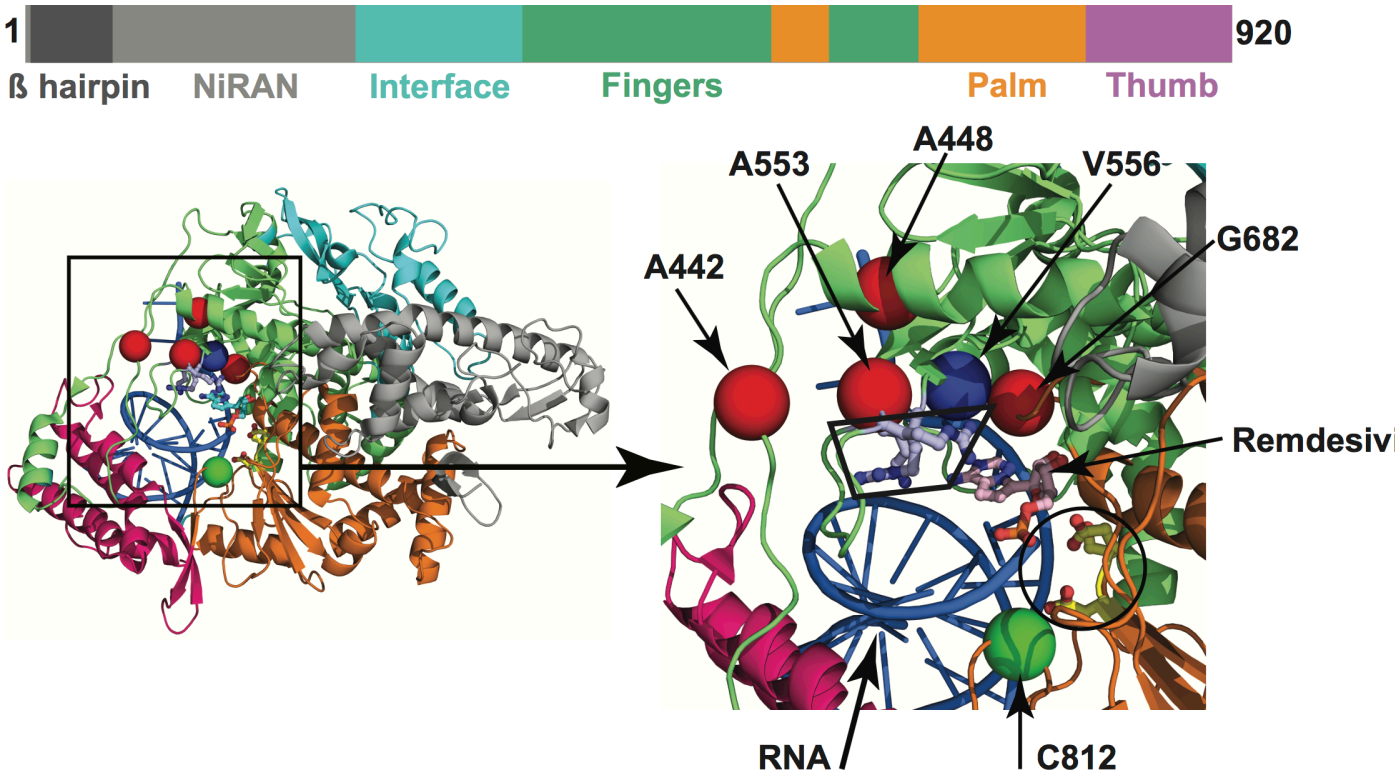


FIG 4 Amino acid changes identified in Nsp12 (RdRp) in this study that may influence interaction with remdesivir. The schematic at the top shows the domain architecture of Nsp12. (Left) Ribbon representation of the crystal structure of Nsp12-remdesivir monophosphate-RNA complex (PDB code: 7BV2). The structure in the right panel shows a magnified view of the boxed area in the left panel. The Nsp12 domains are colored as in the schematic at the top. The catalytic site in Nsp12 is marked by a black circle in the right panel. The side chains of amino acids comprising the catalytic site of RdRp (Ser758, Asp759, and Asp760) are shown as balls and stick and colored yellow. The nucleotide binding site is boxed in the right panel. The side chains of amino acids participating in nucleotide binding (Lys544, Arg552, and Arg554) are shown as balls and sticks and colored light blue. Remdesivir molecule incorporated into the nascent RNA is shown as balls and sticks and colored light pink. The RNA is shown as a blue cartoon and bases are shown as sticks. The positions of Cα atoms of amino acids identified in

this study are shown as red and green spheres and labeled. The amino acids that are shown as red spheres are located above the nucleotide binding site, whereas Cys812 located at the catalytic site is shown as a green sphere. The side chain of active site residue Ser758 is shown as ball and sticks and colored yellow. The location of Cα atoms of remdesivir resistance conferring amino acid Val556 is shown as blue sphere and labeled.

FIG 5 Location of amino acid replacements in spike protein among the 5,085 genomes of SARS-CoV-2 sequenced. The various spike protein domains are color-coded. The numbers refer to amino acid site. Note that many amino acid sites have multiple variants identified.

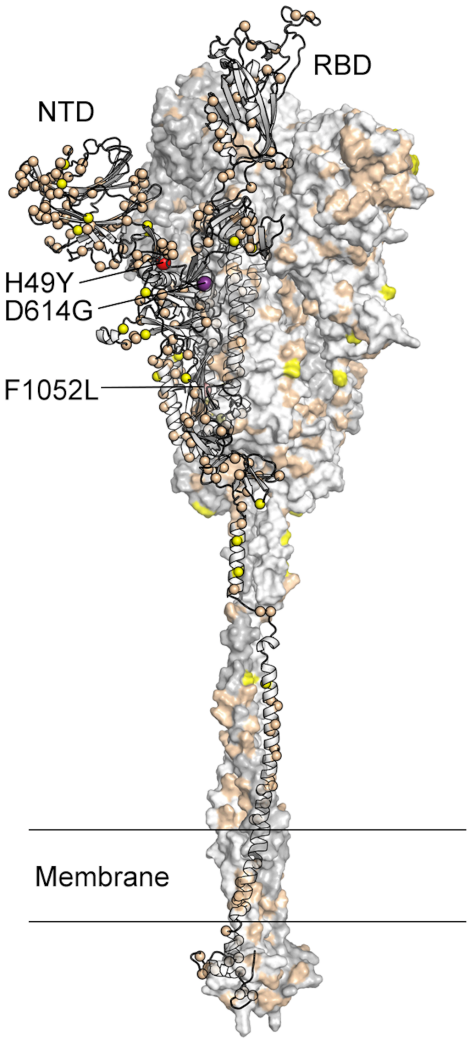
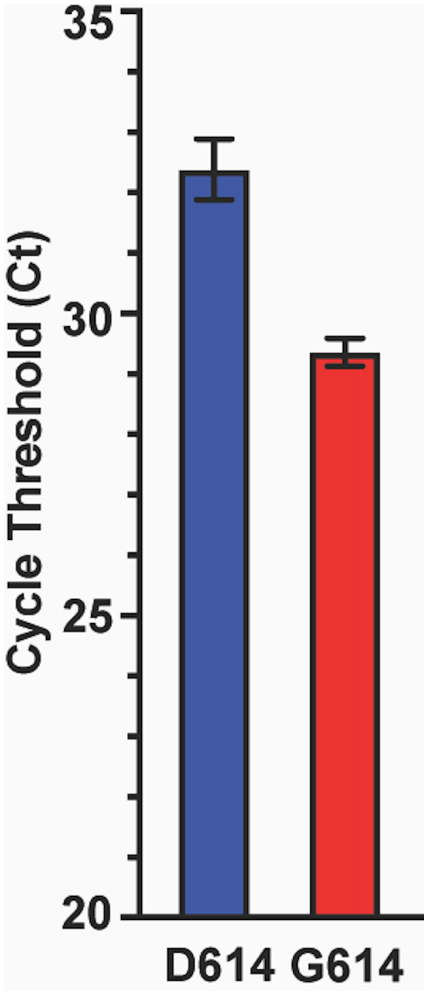


FIG 6 Location of amino acid substitutions mapped on the SARS-CoV-2 spike protein. Model of the SARS-CoV-2 spike protein with one protomer shown as ribbons and the other two protomers shown as a molecular surface. The Cα atom of residues found to be substituted in one or more virus isolates identified in this study is shown as a sphere on the ribbon representation. Residues found to be substituted in 1–9 isolates are colored tan, 10–99 isolates yellow, 100–999 isolates colored red (H49Y and F1052L), and >1000 isolates purple (D614G). The surface of the aminoterminal domain (NTD) that is distal to

1391 the trimeric axis has a high density of substituted residues. RBD, receptor binding

1392 domain.



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1394 **FIG 7** Cycle threshold (Ct) for every SARS-CoV-2 patient sample tested using the

1395 Hologic Panther assay. Data are presented as mean +/- standard error of the mean for

1396 strains with an aspartate (D614, *n*=102 strains, blue) or glycine (G614, *n*=812 strains,

1397 red) at amino acid 614 of the spike protein. Mann-Whitney test, **P*<0.0001.

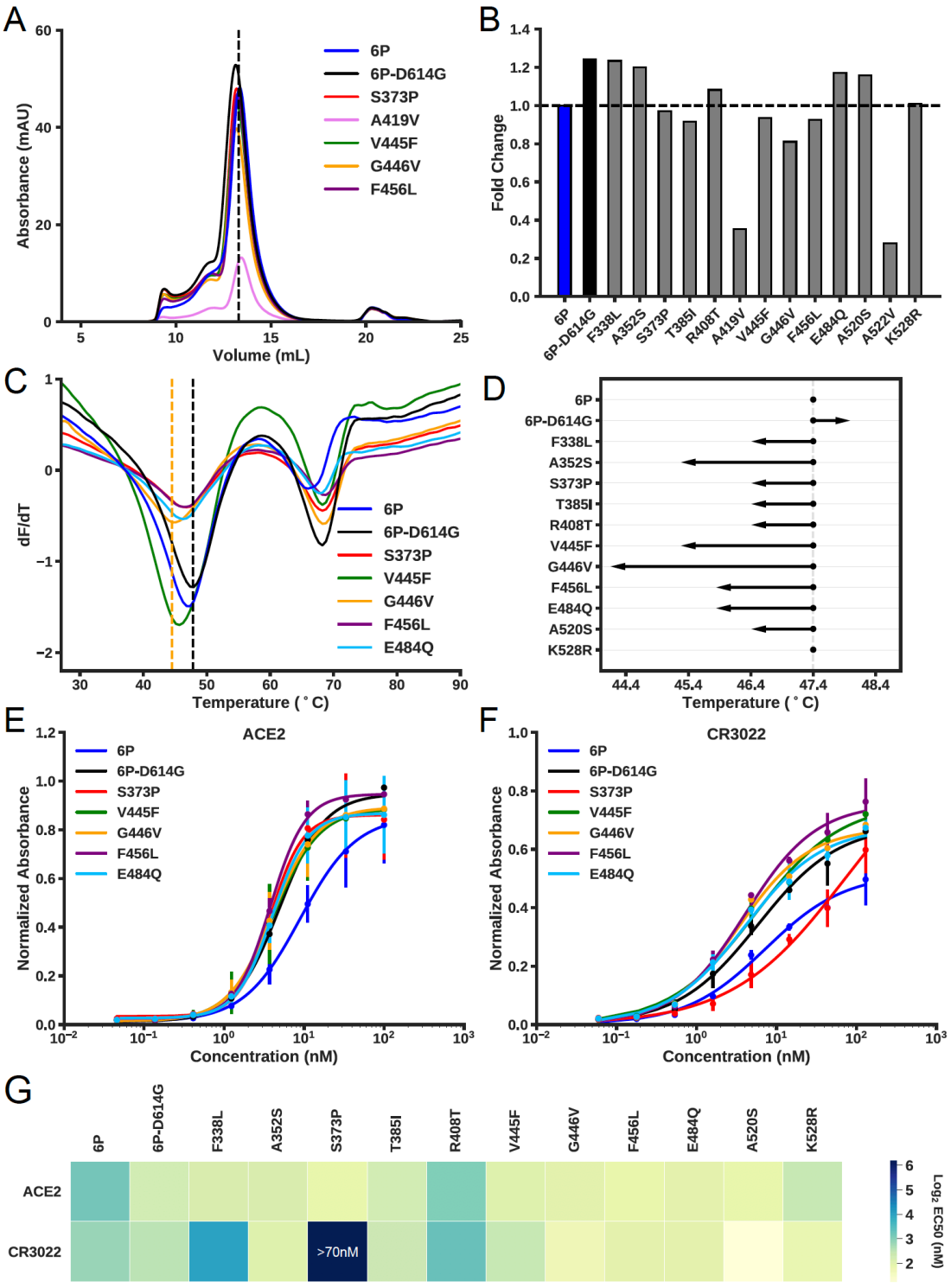

FIG 8 Biochemical characterization of spike RBD variants. (A) Size-exclusion chromatography (SEC) traces of the indicated spike-RBD variants. Dashed line indicates the elution peak of spike-6P. (B) The relative expression of all RBD variants as

1402 determined by the area under the SEC traces. All expression levels are normalized
1403 relative to spike-6P. (C) Thermostability analysis of RBD variants by differential scanning
1404 fluorimetry. Each sample had three replicates and only mean values were plotted. Black
1405 vertical dashed line indicates the first melting temperature of 6P-D614G and orange
1406 vertical dashed line indicates the first melting temperature of the least stable variant
1407 (spike-G446V). (D) First apparent melting temperature of all RBD variants. (E) ELISA-
1408 based binding affinities for ACE2 and (F) the neutralizing antibody CR3022 to the
1409 indicated RBD variants. (G) Summary of EC50s for all measured RBD variants.
1410
1411

1412 **Table 1.** Nonsynonymous SNPs of SARS-CoV-2 *nsp12.*

Genomic Locus	Gene Locus	Amino Acid Change	Domain	Wave 1 (n=1026)	Wave 2 (n=4059)	Total (n=5085)
13446	C3T	A1V	N-terminus		2	2
13448	G5A	D2N	N-terminus	1		1
13487	C44T	A15V	N-terminus		138	138
13501	C58T	P20S	N-terminus		1	1
13514	G71A	G24D	N-terminus		3	3
13517	C74T	T25I	N-terminus		4	4
13520	G77A	S26N	N-terminus		1	1
13523	C80T	T27I	N-terminus		1	1
13526	A83C	D28A	N-terminus		1	1
13564	G121A	V41I	B hairpin		1	1
13568	C125T	A42V	B hairpin	1		1
13571	G128T	G43V	B hairpin	1		1
13576	G133T	A45S	B hairpin		12	12
13617	G174T	K58N	NiRAN		1	1
13618	G175T	D59Y	NiRAN		24	24
13620	C177G	D59E	NiRAN		1	1
13627	G184T	D62Y	NiRAN	1		1
13661	G218A	R73K	NiRAN		1	1
13667	C224T	T75I	NiRAN		2	2
13694	C251T	T84I	NiRAN		1	1
13712	A269G	K90R	NiRAN		1	1
13726	G283A	V95I	NiRAN		1	1
13730	C287T	A96V	NiRAN	2	2	4
13762	G319C	G107R	NiRAN		1	1
13774	C331A	P111T	NiRAN		1	1
13774	C331T	P111S	NiRAN		15	15
13777	C334T	H112Y	NiRAN		1	1
13790	A347G	Q116R	NiRAN		2	2
13835	G392T	R131M	NiRAN		1	1
13858	G415T	D139Y	NiRAN		3	3
13862	C419T	T140I	NiRAN	1	5	6
13868	A425G	K142R	NiRAN	1		1
13897	G454T	D152Y	NiRAN		4	4
13901	A458G	D153G	NiRAN		2	2
13957	C514T	R172C	NiRAN		2	2
13963	T520C	Y174H	NiRAN		1	1
13966	G523A	A175T	NiRAN		1	1
13975	G532T	G178C	NiRAN		4	4

Genomic Locus	Gene Locus	Amino Acid Change	Domain	Wave 1 (n=1026)	Wave 2 (n=4059)	Total (n=5085)
13984	G541A	V181I	NiRAN		1	1
13994	C551T	A184V	NiRAN		8	8
14104	T661C	F221L	NiRAN	2		2
14109	A666G	I222M	NiRAN	1		1
14120	C677T	P226L	NiRAN		2	2
14185	A742G	R248G	NiRAN		1	1
14187	G744T	R248S	NiRAN		1	1
14188	G745A	A249T	NiRAN		1	1
14225	C782A	T261K	Interface		4	4
14230	C787T	P263S	Interface		1	1
14233	T790C	Y264H	Interface		1	1
14241	G798T	K266N	Interface		1	1
14290	G847T	D283Y	Interface		1	1
14335	G892T	V298F	Interface		8	8
14362	C919A	L307M	Interface		2	2
14371	G928C	A310P	Interface	1		1
14396	C953T	T318I	Interface		1	1
14398	G955T	V319L	Interface		1	1
14407	C964T	P322S	Interface		2	2
14408	C965T	P322L	Interface	843	4050	4893
14500	G1057T	V353L	Interface		5	5
14536	C1093T	L365F	Interface		1	1
14557	G1114T	V372L	Fingers		4	4
14584	G1141T	A381S			1	1
14585	C1142T	A381V	Fingers		10	10
14593	G1150A	G384S	Fingers	1		1
14657	C1214T	A405V	Fingers		1	1
14708	C1265T	A422V			1	1
14747	A1304G	E435G	Fingers		2	2
14768	C1325T	A442V	Fingers		21	21
14786	C1343T	A448V	Fingers	3	6	9
14821	C1378T	P460S			1	1
14829	G1386T	M462I	Fingers		59	59
14831	G1388T	C463F	Fingers		3	3
14857	G1414T	V472F			1	1
14870	A1427G	D476G	Fingers		5	5
14874	G1431T	K477N	Fingers	1		1
14912	A1469G	N490S	Fingers	1	1	2
14923	G1480A	V494I	Fingers		2	2

Genomic Locus	Gene Locus	Amino Acid Change	Domain	Wave 1 (n=1026)	Wave 2 (n=4059)	Total (n=5085)
14980	C1537T	L513F	Fingers	1	1	2
14990	A1547G	D516G			1	1
15006	G1563C	E521D	Fingers	2	3	5
15016	G1573T	A525S	Fingers		3	3
15026	C1583T	A528V	Fingers	5	1	6
15037	C1594T	R532C	Fingers		1	1
15100	G1657C	A553P	Fingers		1	1
15101	C1658T	A553V	Fingers		1	1
15124	A1681G	I561V	Fingers		2	2
15202	G1759C	V587L	Palm		7	7
15211	A1768G	T590A			1	1
15226	G1783A	G595S	Palm		1	1
15243	G1800T	M600I	Palm	71	4	75
15251	C1808G	T603S	Palm		1	1
15257	A1814G	Y605C			1	1
15260	G1817A	S606N	Palm		1	1
15327	G1884T	M628I	Fingers	3	1	4
15328	C1885T	L629F	Fingers	1		1
15334	A1891G	I631V	Fingers		1	1
15341	C1898T	A633V	Fingers		1	1
15352	C1909T	L637F	Fingers		1	1
15358	C1915T	R639C	Fingers		1	1
15362	A1919G	K640R	Fingers		1	1
15364	C1921G	H641D	Fingers	1		1
15368	C1925T	T642I	Fingers	1		1
15380	G1937T	S646I	Fingers	1		1
15386	C1943T	S648L	Fingers		2	2
15391	C1948T	R650C	Fingers		1	1
15406	G1963T	A655S	Fingers		3	3
15407	C1964T	A655V	Fingers	1		1
15436	A1993G	M665V	Fingers		2	2
15438	G1995T	M665I	Fingers		24	24
15452	G2009T	G670V	Fingers		28	28
15487	G2044C	G682R	Palm		1	1
15497	C2054A	T685K	Palm		1	1
15572	A2129G	D710G	Palm		1	1
15596	A2153G	Y718S	Palm	2		2
15619	C2176T	L726F	Palm	1		1
15638	G2195A	R732K	Palm	1		1

Genomic Locus	Gene Locus	Amino Acid Change	Domain	Wave 1 (n=1026)	Wave 2 (n=4059)	Total (n=5085)
15640	A2197G	N733D	Palm	1		1
15640	A2197T	N733Y	Palm	1		1
15655	A2212G	T738A	Palm		2	2
15656	C2213T	T738I	Palm		2	2
15658	G2215A	D739N	Palm		2	2
15664	G2221A	V741M	Palm		1	1
15715	T2272C	S758P	Palm		1	1
15760	G2317A	G773S	Palm	1		1
15761	G2318A	G773D	Palm		1	1
15827	A2384G	E795G	Palm	1		1
15848	C2405T	T802I	Palm		1	1
15850	G2407T	D803Y	Palm		1	1
15853	C2410T	L804F	Palm		2	2
15878	G2435T	C812F	Palm		1	1
15886	C2443T	H815Y	Palm		1	1
15906	G2463T	Q821H	Thumb	1	1	2
15908	G2465T	G822V	Thumb		1	1
15979	A2536G	I846V	Thumb	4		4
16045	C2602T	L868F	Thumb		1	1
16084	C2641T	H881Y	Thumb		1	1
16148	A2705G	Y902C	Thumb		1	1
16163	C2720T	T907I	Thumb		45	45
16178	C2735T	S912L	Thumb		2	2
16192	C2749T	P917S	Thumb		80	80

1413
1414

14**Table 2.** Nonsynonymous SNPs in SARS-CoV-2 spike protein.

Genomic Locus	Gene Locus	Amino Acid Change	Domain	Wave 1 (n=1026)	Wave 2 (n=4059)	Total (n=5085)
21575	C13T	L5F	S1	11	25	36
21578	G16T	V6F	S1		1	1
21587	C25T	P9S	S1	2		2
21588	C26T	P9L	S1	1	1	2
21594	T32C	V11A	S1		1	1
21597	C35T	S12F	S1		6	6
21604	G42T	Q14H	S1		1	1
21614	C52T	L18F	S1 - NTD	1	11	12
21618	C56T	T19I	S1 - NTD	1	1	2
21621	C59T	T20I	S1 - NTD		1	1
21624	G62T	R21I	S1 - NTD		6	6
21624	G62A	R21K	S1 - NTD		1	1
21624	G62C	R21T	S1 - NTD		3	3
21627	C65T	T22I	S1 - NTD	2	4	6
21638	C76T	P26S	S1 - NTD		17	17
21641	G79T	A27S	S1 - NTD	1	1	2
21641	G79A	A27T	S1 - NTD	1		1
21642	C80T	A27V	S1 - NTD		1	1
21648	C86T	T29I	S1 - NTD	1	4	5
21707	C145T	H49Y	S1 - NTD		142	142
21713	A151G	T51A	S1 - NTD		1	1
21724	G162T	L54F	S1 - NTD		11	11
21754	G192T	W64C	S1 - NTD		1	1
21767	C205T	H69Y	S1 - NTD	1	7	8
21770	G208A	V70I	S1 - NTD		1	1
21770	G208T	V70F	S1 - NTD	1		1
21774	C212T	S71F	S1 - NTD		1	1
21784	T222A	N74K	S1 - NTD	1		1
21785	G223C	G75R	S1 - NTD		1	1
21793	G231T	K77N	S1 - NTD		1	1
21824	G262A	D88N	S1 - NTD		1	1
21834	A272T	Y91F	S1 - NTD		1	1
21846	C284T	T95I	S1 - NTD	1	10	11
21852	A290G	K97R	S1 - NTD		1	1
21855	C293T	S98F	S1 - NTD	1	2	3
21861	T299C	I100T	S1 - NTD		2	2

Genomic Locus	Gene Locus	Amino Acid Change	Domain	Wave 1 (n=1026)	Wave 2 (n=4059)	Total (n=5085)
21918	T356C	I119T	S1 - NTD	1		1
21930	C368T	A123V	S1 - NTD		1	1
21941	G379T	V127F	S1 - NTD		1	1
21942	T380C	V127A	S1 - NTD		4	4
21974	G412T	D138Y	S1 - NTD	2		2
21985	G423T	L141F	S1 - NTD		1	1
21986	G424A	G142S	S1 - NTD		2	2
21993	A431G	Y144C	S1 - NTD	1		1
21995	T433C	Y145H	S1 - NTD	2		2
21998	C436T	H146Y	S1 - NTD	1	2	3
22014	G452A	S151N	S1 - NTD		1	1
22014	G452T	S151I	S1 - NTD		2	2
22017	G455T	W152L	S1 - NTD	1	1	2
22021	G459T	M153I	S1 - NTD		1	1
22021	G459A	M153I	S1 - NTD		1	1
22022	G460A	E154K	S1 - NTD		1	1
22028	G466C	E156Q	S1 - NTD	2		2
22037	G475A	V159I	S1 - NTD	1		1
22097	C535T	L179F	S1 - NTD		1	1
22104	G542T	G181V	S1 - NTD		1	1
22107	A545G	K182R	S1 - NTD		1	1
22135	A573T	E191D	S1 - NTD		1	1
22139	G577T	V193L	S1 - NTD		1	1
22150	T588G	N196K	S1 - NTD	1		1
22175	T613G	S205A	S1 - NTD		1	1
22205	G643T	D215Y	S1 - NTD		1	1
22206	A644G	D215G	S1 - NTD		2	2
22214	C652G	Q218E	S1 - NTD		1	1
22227	C665T	A222V	S1 - NTD		1	1
22241	G679A	V227I	S1 - NTD		2	2
22242	T680C	V227A	S1 - NTD	1		1
22244	G682C	D228H	S1 - NTD		2	2
22245	A683G	D228G	S1 - NTD	1		1
22246	T684G	D228E	S1 - NTD	2		2
22248	T686G	L229W	S1 - NTD	1		1
22250	C688A	P230T	S1 - NTD	1		1
22253	A691G	I231V	S1 - NTD	1		1

Genomic Locus	Gene Locus	Amino Acid Change	Domain	Wave 1 (n=1026)	Wave 2 (n=4059)	Total (n=5085)
22254	T692C	I231T	S1 - NTD	1		1
22259	A697G	I233V	S1 - NTD	1		1
22260	T698C	I233T	S1 - NTD	1		1
22262	A700G	N234D	S1 - NTD	1		1
22266	T704C	I235T	S1 - NTD	1		1
22281	C719T	T240I	S1 - NTD		5	5
22286	C724T	L242F	S1 - NTD		1	1
22295	C733T	H245Y	S1 - NTD		2	2
22304	T742C	Y248H	S1 - NTD		3	3
22311	C749T	T250I	S1 - NTD	1	4	5
22313	C751T	P251S	S1 - NTD		2	2
22320	A758G	D253G	S1 - NTD		2	2
22320	A758C	D253A	S1 - NTD	1		1
22323	C761T	S254F	S1 - NTD		3	3
22329	C767T	S256L	S1 - NTD	1		1
22335	G773T	W258L	S1 - NTD	1		1
22344	G782T	G261V	S1 - NTD	3		3
22346	G784T	A262S	S1 - NTD		4	4
22350	C788T	A263V	S1 - NTD	1		1
22382	A820G	T274A	S1 - NTD		1	1
22398	A836T	Y279F	S1 - NTD	1		1
22408	T846G	N282K	S1 - NTD		1	1
22425	C863T	A288V	S1 - NTD		1	1
22430	G868T	D290Y	S1 - NTD	1		1
22484	G922T	V308L	S1		3	3
22487	G925C	E309Q	S1		1	1
22532	G970C	E324Q	S1		1	1
22533	A971T	E324V	S1		1	1
22535	T973C	S325P	S1		1	1
22536	C974T	S325F	S1		1	1
22550	C988T	P330S	S1 - RBD		2	2
22574	T1012C	F338L	S1 - RBD	1		1
22608	C1046T	S349F	S1 - RBD		1	1
22616	G1054T	A352S	S1 - RBD		7	7
22661	G1099T	V367F	S1 - RBD		1	1
22673	T1111C	S371P	S1 - RBD		3	3
22679	T1117C	S373P	S1 - RBD		1	1

Genomic Locus	Gene Locus	Amino Acid Change	Domain	Wave 1 (n=1026)	Wave 2 (n=4059)	Total (n=5085)
22712	C1150T	P384S	S1 - RBD		1	1
22716	C1154T	T385I	S1 - RBD	3		3
22785	G1223C	R408T	S1 - RBD		1	1
22793	G1231T	A411S	S1 - RBD		1	1
22818	C1256T	A419V	S1 - RBD	1		1
22895	G1333T	V445F	S1 - RBD		1	1
22899	G1337T	G446V	S1 - RBD	2		2
22928	T1366C	F456L	S1 - RBD	1		1
23001	G1439T	C480F	S1 - RBD		1	1
23012	G1450C	E484Q	S1 - RBD	1		1
23046	A1484C	Y495S	S1 - RBD		1	1
23111	C1549T	L517F	S1 - RBD		1	1
23120	G1558T	A520S	S1 - RBD	1	6	7
23121	C1559T	A520V	S1 - RBD		1	1
23127	C1565T	A522V	S1 - RBD	1	1	2
23145	A1583G	K528R	S1 - RBD		2	2
23149	G1587T	K529N	S1		1	1
23170	C1608A	N536K	S1		1	1
23202	C1640A	T547K	S1		2	2
23202	C1640T	T547I	S1		1	1
23223	A1661T	E554V	S1		2	2
23224	G1662T	E554D	S1	4	31	35
23270	G1708T	A570S	S1		3	3
23277	C1715T	T572I	S1	5	5	10
23282	G1720T	D574Y	S1		1	1
23292	G1730T	R577L	S1	1		1
23311	G1749T	E583D	S1		6	6
23312	A1750G	I584V	S1		1	1
23315	C1753T	L585F	S1	1	7	8
23349	G1787A	S596N	S1		1	1
23373	C1811T	T604I	S1		2	2
23380	C1818A	N606K	S1		2	2
23403	A1841G	D614G	S1	841	4054	4895
23426	G1864T	V622F	S1		2	2
23426	G1864C	V622L	S1		2	2
23435	C1873T	H625Y	S1		1	1
23439	C1877T	A626V	S1		1	1

Genomic Locus	Gene Locus	Amino Acid Change	Domain	Wave 1 (n=1026)	Wave 2 (n=4059)	Total (n=5085)
23444	C1882G	Q628E	S1		7	7
23453	C1891T	P631S	S1	1		1
23457	C1895T	T632I	S1		1	1
23481	C1919T	S640F	S1	1	42	43
23486	G1924T	V642F	S1		1	1
23502	C1940T	A647V	S1		1	1
23536	C1974A	N658K	S1		4	4
23564	G2002T	A668S	S1		1	1
23586	A2024G	Q675R	S1		14	14
23587	G2025C	Q675H	S1		1	1
23587	G2025T	Q675H	S1		4	4
23589	C2027T	T676I	S1	1	2	3
23593	G2031T	Q677H	S1	1	1	2
23595	C2033T	T678I	S1	1		1
23624	G2062T	A688S	S2		4	4
23625	C2063T	A688V	S2		16	16
23655	C2093T	S698L	S2		1	1
23664	C2102T	A701V	S2		21	21
23670	A2108G	N703S	S2		1	1
23679	C2117T	A706V	S2		1	1
23684	T2122C	S708P	S2		1	1
23709	C2147T	T716I	S2		1	1
23718	C2156T	T719I	S2		1	1
23745	C2183T	P728L	S2	1		1
23755	G2193T	M731I	S2	3	1	4
23798	T2236C	S746P	S2		1	1
23802	C2240T	T747I	S2		1	1
23804	G2242A	E748K	S2		1	1
23832	G2270T	G757V	S2		1	1
23856	G2294T	R765L	S2		1	1
23868	G2306T	G769V	S2		3	3
23873	G2311T	A771S	S2		8	8
23877	T2315C	V772A	S2		1	1
23895	C2333T	T778I	S2		1	1
23900	G2338C	E780Q	S2		1	1
23936	C2374T	P792S	S2		1	1

Genomic Locus	Gene Locus	Amino Acid Change	Domain	Wave 1 (n=1026)	Wave 2 (n=4059)	Total (n=5085)
23948	G2386T	D796Y	S2		2	2
23955	G2393T	G798V	S2	1		1
23987	C2425T	P809S	S2		2	2
23988	C2426T	P809L	S2		1	1
23997	C2435T	P812L	S2		1	1
24003	A2441G	K814R	S2		1	1
24014	A2452G	I818V	S2 - FP		5	5
24026	C2464T	L822F	S2 - FP		97	97
24041	A2479T	T827S	S2 - FP		4	4
24077	G2515T	D839Y	S2	2		2
24089	G2527A	D843N	S2	1	1	2
24095	G2533T	A845S	S2		5	5
24099	C2537T	A846V	S2		1	1
24129	A2567G	N856S	S2		7	7
24138	C2576T	T859I	S2		5	5
24141	T2579C	V860A	S2		1	1
24170	A2608G	I870V	S2		3	3
24188	G2626T	A876S	S2		1	1
24197	G2635T	A879S	S2		31	31
24198	C2636T	A879V	S2		1	1
24212	T2650G	S884A	S2		11	11
24237	C2675T	A892V	S2		1	1
24240	C2678T	A893V	S2	1		1
24268	G2706T	M902I	S2		1	1
24287	A2725G	I909V	S2 - HR1		2	2
24314	G2752C	E918Q	S2 - HR1	1		1
24328	G2766C	L922F	S2 - HR1		2	2
24348	G2786T	S929I	S2 - HR1		1	1
24356	G2794T	G932C	S2 - HR1		1	1
24357	G2795T	G932V	S2 - HR1		1	1
24368	G2806A	D936N	S2 - HR1		3	3
24368	G2806C	D936H	S2 - HR1		1	1
24368	G2806T	D936Y	S2 - HR1	3	4	7
24374	C2812T	L938F	S2 - HR1		3	3
24378	C2816T	S939F	S2 - HR1		4	4
24380	T2818G	S940A	S2 - HR1		5	5
24389	A2827G	S943G	S2 - HR1		6	6

Genomic Locus	Gene Locus	Amino Acid Change	Domain	Wave 1 (n=1026)	Wave 2 (n=4059)	Total (n=5085)
24463	C2901A	S967R	S2 - HR1	2		2
24507	C2945T	S982L	S2 - HR1		1	1
24579	C3017T	T1006I	S2 - CH		1	1
24588	C3026G	T1009S	S2 - CH		1	1
24621	C3059T	A1020V	S2 - CH	1		1
24638	G3076T	A1026S	S2 - CH		2	2
24642	C3080T	T1027I	S2 - CH	5		5
24649	G3087T	M1029I	S2 - CH		1	1
24710	A3148T	M1050L	S2		1	1
24710	A3148G	M1050V	S2	1	1	2
24712	G3150T	M1050I	S2		2	2
24718	C3156A	F1052L	S2	1	166	167
24770	G3208T	A1070S	S2		2	2
24794	G3232T	A1078S	S2 - CD	3	2	5
24812	G3250T	D1084Y	S2 - CD	1	29	30
24834	G3272T	R1091L	S2 - CD	1		1
24867	G3305T	W1102L	S2 - CD		1	1
24872	G3310T	V1104L	S2 - CD		1	1
24893	G3331C	E1111Q	S2 - CD		2	2
24897	C3335T	P1112L	S2 - CD	2	2	4
24912	C3350T	T1117I	S2 - CD		1	1
24923	T3361C	F1121L	S2 - CD		2	2
24933	G3371T	G1124V	S2 - CD	1	2	3
24959	G3397T	V1133F	S2 - CD		1	1
24977	G3415T	D1139Y	S2 - CD		1	1
24986	C3424A	Q1142K	S2	1		1
24998	G3436T	D1146Y	S2		4	4
24998	G3436C	D1146H	S2		13	13
25019	G3457T	D1153Y	S2		11	11
25032	A3470T	K1157M	S2	1		1
25046	C3484T	P1162S	S2		5	5
25047	C3485T	P1162L	S2		3	3
25050	A3488T	D1163V	S2		2	2
25088	G3526T	V1176F	S2		18	18
25101	A3539G	Q1180R	S2		1	1
25104	A3542G	K1181R	S2		4	4
25116	G3554A	R1185H	S2		2	2

Genomic Locus	Gene Locus	Amino Acid Change	Domain	Wave 1 (n=1026)	Wave 2 (n=4059)	Total (n=5085)
25121	A3559T	N1187Y	S2		1	1
25135	G3573T	K1191N	S2		1	1
25137	A3575C	N1192T	S2	1		1
25158	A3596C	D1199A	S2		1	1
25160	C3598T	L1200F	S2		1	1
25163	C3601A	Q1201K	S2		1	1
25169	C3607T	L1203F	S2	1		1
25183	G3621T	E1207D	S2		1	1
25186	G3624T	Q1208H	S2	1		1
25217	G3655T	G1219C	S2	1	3	4
25234	G3672T	L1224F	S2		1	1
25241	A3679G	I1227V	S2	1		1
25244	G3682T	V1228L	S2		2	2
25249	G3687T	M1229I	S2		1	1
25249	G3687C	M1229I	S2		2	2
25250	G3688A	V1230M	S2		1	1
25266	G3704T	C1235F	S2		4	4
25273	G3711T	M1237I	S2		2	2
25284	G3722T	C1241F	S2		1	1
25287	G3725T	S1242I	S2		4	4
25297	G3735T	K1245N	S2		1	1
25301	T3739G	C1247G	S2		1	1
25302	G3740T	C1247F	S2		4	4
25305	G3743T	C1248F	S2		2	2
25317	C3755T	S1252F	S2		1	1
25340	G3778T	D1260Y	S2		2	2
25350	C3788T	P1263L	S2	1	2	3
25352	G3790T	V1264L	S2		1	1
25365	T3803C	V1268A	S2		1	1

The domain region of RBD is based on structural information found in Cai et al. 2020 (98).

Forty-nine of these amino acid replacements (V11A, T51A, W64C, I119T, E156Q, S205A, D228G, L229W, P230T, N234D, I235T, T274A, A288V, E324Q, E324V, S325P, S349F, S371P, S373P, T385I, A419V, C480F, Y495S, L517F, K528R, Q628E, T632I, S708P, T719I, P728L, S746P, E748K, G757V, V772A,

K814R, D843N, S884A, M902I, I909V, E918Q, S982L, M1029I, Q1142K, K1157M, Q1180R, D1199A, C1241F, C1247G, and V1268A) were not represented in a publicly available database (34) as of August 19, 2020

1428