Analysis

Title: Re-insights into origin and adaptation of SARS-CoV-2

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Abstract

Following the recent report of a furin-like cleavage site unique to SARS-CoV-2, we characterized

additional amino acid residues, positioned within the crown of the spike (S) glycoprotein, under

diversifying selection along the ancestral lineage of currently circulating pandemic strains, and

affecting enhanced ability to infect host cells. Molecular dynamic simulations revealed long-

range covariant movements of SARS-CoV-2 S glycoprotein monomer in pre-fusion conformation

with furin cleavage, and residue T333, under purifying selection, hinge of such movement.

Evolutionary analysis revealed that the current human lineage likely emerged after a

recombination event between ancestors in bat and pangolin. Recombination impacted residues

located in the binding receptor region of S glycoprotein, identical to pangolin ancestor.

Interestingly, several recombination hotspots were detected throughout the genome, with the

exception of the S glycoprotein, which exhibits a coldspot in the N terminal region, possibly the

result of negative selection against changes that may reduce its optimal fitness.

Main – With a toll of confirmed cases rapidly increasing globally, the COVID-19 pandemic is likely

to meet the millions in matter of weeks. One of the features that may explain why this novel

betacoronavirus (SARS-CoV-2) has been successfully spreading around the globe, is the newly

acquired furin-like cleavage site[1]. Furin is ubiquitously expressed at different levels in all

tissues[2,3]. However, kinetics of furin-like enzyme activity are different among species – e. g.

human lung cells display quicker kinetic than fruit bat (Pteropus *alecto*) lung cells [4] – and may not

fully explain SARS-CoV-2 enhanced infectivity. On the other hand, we detected several unique

amino acid residues that distinguish the spike (S) glycoprotein from ancestral bat and pangolin

strains, tested for selective pressure acting on these residues, and further performed homology

modeling and molecular dynamic simulations to understand the impact of the cleavage site on the

S glycoprotein structure, as it might be modulating infectivity and pathogenicity of the virus[7]. Our

results may provide important insights on SARS-CoV-2 infectivity and pathogenesis, as well as

for the development of targeted therapies and a vaccine.

SARS-CoV-2 spike (S) glycoprotein features unique to COVID-19

We defined all residues in the new SARS-CoV-2 S glycoprotein unique to the strains circulating

among humans by comparing SARS-CoV-2 strains obtained from human, bat, and two pangolin

lineages (lineage 'a' strains isolated in 2017 in China; 'b' strains isolated in 2019 in China) (Table

S1). Before diving into the new features, we define major domains of SARS-CoV-2 S

glycoprotein, based on the available SARS-CoV-1 structure [5]: S_1 subunit, that binds to the

receptor on the host cell surface[6], and S_2 subunit which plays a role in membrane fusion [7]. In

particular, S_1 is composed of an N-terminal domain (NTD, residues 14-294) and three C-terminal

domains (CTD1, CTD2, and CTD3) (Figure 1). The CTD1 (residues 320-516 for SARS-CoV and

residue 333-529 for SARS-CoV-2, see Methods) functions as the receptor-binding domain (RBD),

which allows SARS-CoV-2 to directly bind to the peptidase domain (PD) of angiotensin-

converting enzyme 2 (ACE2) [8] (Figure 1). Within this region, several binding receptor residues

and different snapshots of SARS-CoV-2 S glycoprotein have been reported[1,9] that reflect the

dynamic. The recognition step is a dynamic process in which the protein surface of both partners

adapt one to the other, in line with cryo-EM evidences that shows how the receptor binding can

open up CTD1[10]. For simplicity, we show canonical ACE2 binding residues F486, Q493, S494

and N501, as reported by Andersen *et al*[1] (Figure 1). The S_2 subunit is located downstream of S_1

after the S_1/S_2 cleavage site (residue 667 for SARS-CoV and residue 660 for SARS-CoV-2). The

majority of the residues that differentiate human S glycoproteins from bat or pangolin were found

within the S_1 subunit (Figure 1a). The S glycoprotein was under strong negative/purifying

selection with 775 out of 1273 sites were under negative/purifying selection (Table S3), which

suggests the protein has reached a peak on the fitness landscape in terms of structure, stability

and folding[11]. We detected six amino acid positions within the S_1 subunit under

positive/diversifying selection, likely involved in the adaptation to new host environments (Figure

1 a, b): positions 32 (S32F), 50 (L50S), 483 (Q483V), and 519 (N519H). Residues 112 (A112S

and 352 (K253D) were also under positive selection, but emerged before the split between

SARS-CoV-2 human and bat (Figure 1a). Residue mutation Q483V folds in near proximity to the

ACE2 binding domain residues (F496-Q493-S493-N501). It is likely that V483 affects the

enhanced ability of currently circulating strains to target human cells (Figure 1a). *In silico*

prediction of the three-dimensional structure of SARS-CoV-2 S glycoprotein by homology

modeling shows that residue mutation Q483V folds in close proximity with residues in the CTD1

domain binding to ACE2 (Figure 1). All other residues are likely to be positioned in the outside of

the S_2 subunit crown, and therefore potentially interacting with receptor of the host cell membrane

(Figure 1 c, d). Interestingly, canonical ACE2 binding residues, conserved between human and

pangolin lineage b strains, do not appear to be under negative selection. These mutations may be

involved in host adaption, thus leaving with the question of when and how these residues have

been acquired.

Polybasic furin cleavage and prefusion conformation

While the S glycoproteins of MERS-CoV, HcoV-OC43, and HKU1 display the canonical (R/K)-

(2X)n-(R/K)* (or RXXR*SA) motif for the S1/S2 cleavage[12,13], the S glycoprotein of the recently

emerged SARS-CoV-2 presents a solvent exposed furin-like cleavage sequence (PRRARS*V) in

positions 681-684[1]. We carried out a 90 ns-long simulation to assess the stability of the trimeric

structure of SARS-CoV-2 and whether the new gained favorable furin-like cleavage site alters the

overall structure of the S glycoprotein.

As recently reported [14], the CTD1 domain of SARS-CoV-2 S glycoprotein for the pre-fusion

conformation rotates outward from center of the triangular head to an "up" position, while this

conformation for SARS-CoV is "down" [5] (Figure 1f). One monomer of the SARS-CoV-2 S

glycoprotein, in fact, acquired the ability of being already in a pre-fusion conformation pointing the

ACE2 binding site in a different angle than its predecessor, thus giving SARS-CoV-2

advantageous ability to bind and enter into the cell. To test whether specific evolutionary

pressures allowed SARS-CoV-2 to display enhanced binding to the human binding receptor, we

performed molecular dynamic (MD) simulations. Projection of the MD trajectory along the first

eigenvector, carried out by Essential Dynamics analysis [15], shows a clear up/down

(opening/closing) mechanism of the SARS-CoV-2 monomer by means of the rotation of CTD1

(see Figure 1e and Suppl. Movies 1-2). MD simulations highlighted T333 (under

negative/purifying selection), its first residue, as the hinge of the movement. It is worth noting that

the region 325-344 (mostly under negative/purifying selection) is completely conserved between

SARS-CoV-2 and closely related bat and pangolin strains (Figure 1a), in line with a key role of

this region in guiding the opening/closing of CTD1. The same analysis shows a correlated

movement between CTD1 (residues between red lines in Figure 1f) and the region 828-852

(highlighted by cyan lines in Figure 1 f), encompassing a portion of the fusion peptide (Suppl.

Movie 3, in which residues 828-852 are highlighted in cyan). This co-variant movement between

CTD1 and the fusion peptide region is indicative of long range coordinated movements, functional

to the protein rearrangement during the fusion step (Figure 1f, Suppl. Movie 3). A correlated

movement (peak of fluctuation) is also observed in the residues 699-718 (orange lines in Figure

1f) that comprehend furin-like cleavage site typical of SARS-CoV-2. This region shows a

covariance in its movement with residues 460-496 in CTD1 and 250-263 in NTD (Suppl. movie 4,

in which residues 828-852 are highlighted in orange color).

On the origin of SARS-CoV-2 lineage: recombination

While the origin of this virus strain remains unknown, numerous scenarios have been

proposed[1,16,17]. Recombination is an important factor driving viral diversity [18,19], and has been

reported as a possible mechanism favoring cross-species transmission of CoV and increasing its

adaptability to a new host [20,21]. Recombination was a critical factor in the emergence of Middle

East respiratory syndrome coronavirus (MERS-CoV) in 2012 from bat reservoirs [22,23]. There is

currently conflicting evidence on whether or not the new coronavirus originated from

recombination events [24-26]. Since the human lineages is 96% identical to bat-SARS-CoV-2 [1], but

ACE2 binding residues F486, Q493, S494 and N501 (Figure 1a) are conserved between

pangolin-SARS-CoV-2 lineage b strains, we investigated the evolutionary relationships among

available full genome sequences (bat and pangolin-related SARS-CoV-2 strains, bat-SL-CoVs,

SARS-CoV-1 and SARS-like strains) using a neighbor network (NNet) based algorithm [27]. NNets

have the advantage over phylogeny inference algorithms of allowing internal nodes to be

cyclically connected (nets) to represent the presence of conflicting phylogenetic signal in the data

(i.e. data support for different tree topologies), as in fact we found in the analysis of the

coronavirus genomes (Fig. 2a). Of course, conflicting phylogenetic signal in the data is not

sufficient to infer recombination [28]. To test whether the NNet topology is due to recombination we

used the pairwise homoplasy index (PHI) test[29], which detected highly significant signal for

recombination ($p < 10^{-99}$) among all included strains (Figure 1a). SARS-CoV-1 and human SARS-

CoV-2 monophyletic clades do not display net-like topology, and the PHI test on a sub-dataset

including only these lineages was not significant ($p = 0.08$). SARS-like and bat-SL-CoVs, on the

other hand, show a high degree of conflicting phylogenetic signals reflected by intricate networks

(Fig. 1a). While SARS-like strains are ancestors of SARS-CoV-1 strains, bat-SL-CoVs stand as

143 intermediate link between SARS-CoV-1 and SARS-CoV-2 strains. Bat-SARS-CoV-2

144 (EPI_ISL_402131, isolated from bat in 2013 in China) and pangolin-SARS-CoV-2 strains

145 (EPI_ISL_410538-43 strains isolated in 2017 in China, and EPI_ISL_412860 and

146 EPI_ISL_410721 strains isolated in 2019 in China, indicated with "a" and "b", respectively, in

147 Figure 2a) are ancestors of SARS-CoV-2 circulating in humans. When testing sub-sets of the

148 alignment including only SARS-CoV-2 and its ancestors, the PHI test still indicated strong

149 evidence of recombination (p = 1.9×10^{-8}), although a weaker signal for recombination was

150 detected for SARS-CoV-2 with pangolin-SARS-CoV-2 ancestral strains (p = 0.013). Significant

151 recombination signal was also detected (p = 6.6×10^{-4}) testing SARS-CoV-2 bat

152 (EPI_ISL_402131) and pangolin strains. As it is not possible to conclude whether pangolin, bat or

153 human SARS-CoV-2 strains have been recombining based on PHI test, we scanned whole

154 genomes for putative recombination breakpoints using six algorithms[30] that allow the identification

155 of specific sets of recombinant as well as parental strains (when present in the sampled

156 sequences) . A total of 68 potential recombination events distributed along the genome were

157 found to affect bat-SL-CoVs, SARS-CoVs and SARS-like CoVs strains (Table S4). Human SARS-

158 CoV-2 strains were identified as potential parental sequences recombining with pangolin-SARS-

159 CoV-2 isolated in 2019 and bat-SARS-CoV-2 strains; with a recombination event involving the

160 region of the S glycoprotein that binds to the cell receptor (Table 1). This finding matches the

161 previous observation that ACE2 binding residues are shared between pangolin and human, and

162 the recent analysis by Li *et al* [26]. While our analysis could not confidently discern the recombinant

163 from the parent sequences, signature pattern analysis (Figure 1a) and a Maximum likelihood tree

164 calculated using this recombinant region only (data not shown), clearly showed pangolin and

165 COV-2 clustering together. Our overall findings point to the conclusion that a recombination event

166 between ancestors of human SARS-CoV-2, or a closely related to bat-SARS-CoV-2, and

167 ancestors of pangolin-SARS-CoV-2 lineage b strains, is the likely origin of the currently circulating

168 pandemic strains.

Using coordinates of identified recombination breakpoints, we plotted their distribution and

analyzed it through permutations to predict hot- and coldspots of recombination across the

coronavirus genomes (Figure 2b). A first iteration, performed with a sliding window of size 100,

identified nine potential recombination hotspots. Seven of them were found in the polyprotein

segment (Figure 2b), which spans three quarters of the viral RNA; the remaining two involved

protein E, *orf8*, and protein N. It has been reported that the *orf8* gene plays a role in regulating

the viral replication in the host cell[31], and was probably acquired by the whole lineage of SARS-

CoV-1 from the SARS-related CoV from Greater Horseshoe bats (SARSr-Rf-BatCoVs) following

a recombination event with a Chinese horseshoe bat coronavirus (SARSr-Rs-BatCoV)[32]. A

second analysis, with a sliding window of size 200, detected two recombination coldspots on the

N terminal region of the glycoprotein (between residues S31 to V47, and Y145 to G184, $p \leq 0.01$)

(Figure 2b). Under more relax assumptions ($0.01 \leq p \leq 0.05$) coldspot region spans from Q14 to

S71, and from V127 to T240 (Figure 2b). Residues S32F, A112S, N519H (Figure 1) are outside

the coldspot region, however L50S is inside the first coldspot region of SARS-CoV-2 S

glycoprotein. Our finding seems to be in conflict with high recombination rates being found in non-

human coronaviruses[33], yet, it is not surprising that part of a gene playing a fundamental role in

cell entry and fusion would sit into a recombination coldspot. While the majority of the CoV

genome seems prone to recombination, the S glycoprotein is the only one that harbored

recombination coldspots. This might depend on functional costraints on specific domains of the

protein [34], or other factors impacting the probability of recombination events, such as RNA

secondary structure [35,36]. Another possibility is that the N terminal, under negative selection,

reached a peak on the adaptive landscape and new mutations/rearrangements are unlikely to be

fixed in the next generations due to negative selection. These findings correlate with mobility of

MD simulations, where flexible stretches are less conserved, while hinge regions remained

unchanged during evolution of the lineage.

4 – Conclusions

We provided molecular insights on SARS-CoV-2 infectivity and pathogenesis and identified

potential targets for therapies, in the context of coronaviruses evolutionary trajectory that has led

to the current pandemic.

SARS-CoV-2 S glycoprotein presents two notable changes: a favorable furin-like cleavage site [1],

and CTD1 up in pre-fusion mode [14]. On the contrary S glycoprotein of previously circulating

SARS-CoV strains do not have a favorable furin-like cleavage site [1] and remain un-cleaved after

biosynthesis[37]. Furin has the potential to cleave specifically viral envelope glycoproteins, to

enhance viral fusion with host cell membranes[38], thereby any variation/mutational changes in

amino acid composition at cleavage sites may impact tissue and cell tropism, host range, and

pathogenesis[39], as reported for other respiratory viruses[40-42]. Moreover, our MD simulations

correlated furin-like cleavage site correlated with pre-fusion conformation. Simulations indicated

that the motion of CTD1 is quite visible and is the most relevant motion in the protein: a

conserved region that acts as a hinge in the opening/closing mechanism. We described residue

T333, under purifying selection, as the hinge of the rotation movement that gives further

advantage to bind cell receptors facilitating fusion and entry. Further analysis will be done to

analyze the conformation change of the S glycoprotein and the residues interface in nearly real

conditions. Development of inhibitors of such motion may reduce substantially pre-fusion

conformation and therefore the infectivity of SARS-CoV-2.

SARS-CoV-2 S glycoprotein gained specific features that have facilitated the successful spread

of the virus as compared to its predecessors. The assumption that the insertion of a furin

cleavage site allowed a bat CoV to gain the ability to infect humans facilitating the host specie

jump once the conditions were favorable cannot be excluded. By structure similarity with S

glycoprotein of mouse hepatitis virus, it is possible that SARS-CoV-2 also uses distal loops of the

N-terminal domain to interact with cell receptor [43]. Although more evidence is needed, it is

conceivable that residues in this region that are under positive selection may have been selected

to immune-modulate the host. Several recombination events involving ancestral pangolin- and

bat- SARS-CoV-2 lineages. According to our results, one of these events led to the current

configuration of SARS-CoV-2 ACE2 binding residues, including recently described residues

forming hydrogen bonds with the host cell receptor [9]. In spite of the frequency of recombination

events involving the S glycoprotein, the 5' end of the gene might sit into a recombination coldspot.

While mutation/recombination events in the glycoprotein RBD might expand the virus tropism,

recombination coldspots and negative/diversifying selective pressure limit its variability to

maintain correct structure, and therefore function. RDB is definitively a good target for therapeutic

development, while regions in coldspots or undergoing negative/diversifying selective pressure

may be helpful for successful vaccine design. Our candidate *in silico* recombination hotspots and

coldspots findings need to be further investigated with *in vitro* experiments.

We conclude that SARS-CoV-2 likely emerged from recombination events that involves ancestral

strains circulating among bat and pangolin. However, we could not completely distinguish the

recombinant sequence from the parental ones in other cases. Recombination signal and selective

pressures can interfere with each other [44], with positive selected mutations near potential

recombination breakpoints [45]. More extensive sampling of wildlife CoVs would help in clarifying

the events that led to the jump to humans. Future emergence of novel recombinant CoVs, and

other viruses as well, should be a constant concern; strengthening surveillance of viral

populations and improving our understanding of their dynamics will help the prevention of new

zoonotic outbreaks and spread of emerging pathogens.

Methods

Genetic data

Publicly available genomic CoVs sequences were obtained from GenBank and GISAID (Table S1

and S5). To gather closely related CoVs BLASTn was performed using the ViPR database and

keeping the first 100 hits. Genome sequences were aligned using MAFFT and refined manually[49].

Alignments for the glycoprotein were extracted from the genome alignment.

Protein selection and signature pattern

Because of the fast rate at which new genomes have been deposited in GISAID, this analysis

was updated to 728 genomes (Table S5). While in the recombination analysis the number of

genomes is not relevant, as they are very similar to each other, for selection analysis and

identification of new features, a higher number of strains permits to explore a larger evolutionary

landscape. HyPhy [46] algorithm mixed effects model of evolution (MEME) [47] and Fast

Unconstrained Bayesian AppRoximation (FUBAR)[48] were used to estimate the ratio of non-

synonymous (dN) to synonymous (dS) substitution rates for each codon. MEME was employed to

identify individual sites subject to episodic diversifying selection, and FUBAR for inferring site-

specific pervasive selection. Sites were considered to have experienced statistically significant

positive or negative selection based on the following cut-offs: LRT ≤ 0.05 or posterior probability

(PP) > 0.90, for MEME and FUBAR respectively. VESPA algorithm[49] was used to detect the

residue signature pattern distinguishing SARS-CoV-2 strains from the closely related bat

(EPI_ISL_402131, 2013, Accession MN996532.1) and pangolin (EPI_ISL_410538, 2017;

EPI_ISL_410539, 2017; EPI_ISL_410540, 2017; EPI_ISL_410541, 2017; EPI_ISL_410542,

2017; EPI_ISL_410543, 2017; EPI_ISL_412860, 2019; EPI_ISL_410721, 2019) strains. Positions

are numbered with SARS-CoV-2 isolate Wuhan-Hu-1 (Accession MN908947.3) as reference.

Structure homology modeling

As a first step, the amino acid sequences of SARS-CoV-2 Wuhan-Hu-1 (Accession QHD43416)

and EPI_ISL_402131 (bat, 2013, Accession MN996532.1) were submitted to Swiss-model [50]. The

best models showing high values of Global Model Quality Estimation (GMQE) were selected and

used as templates. Amino acid alignments of targets and templates were performed using

EMBOSS Needle [51]. The 3D homology homo-trimer models were created with the Modeller v9.23

[52,53] *automodel* class using the alignment as a guide. Then Discreet Optimised Protein Energy

Score (DOPE) based model selection [54] and refinements were conducted using Modeller v9.23

scripts. The final structure model validation was conducted with ProSA [55,56] and QMEAN servers

[57,58]. Visualization of the atomic models, including figures and movies, is made with Chimera

v1.12[59].

Molecular Dynamics simulation

The structure of SARS-CoV-2, obtained for homology modeling, was used as starting structure

for Molecular Dynamics (MD) simulations. In detail, topology file was generated with the pdb2gmx

Gromacs tool[60], using the amber99sb forcefield [61].The protein was embedded in a triclinic box,

extending up to 15 Å from the solute, and immersed in TIP3P water molecules [62]. Counter ions

were added to neutralize the overall charge with the genion gromacs tool. After energy

minimizations, the systems were slowly relaxed for 5 ns by applying positional restraints of 1000

kJ mol^{-1} nm^{-2} to the protein atoms. Then unrestrained MD simulations were carried out for a

length of 90 ns with a time step of 2 fs with GROMACS 2018.3 simulation package (Pall et al.,

2015) (supercomputer Galileo, CINECA, Bologna, Italy). V-rescale temperature coupling was

employed to keep the temperature constant at 300 K [63]. The Particle-Mesh Ewald method was

used for the treatment of the long-range electrostatic interactions [64]. The first 5 ns portion of the

trajectory was excluded from the analysis.

We carried out the Essential Dynamics analysis to identify the main 3N directions along which the

majority of the protein motion is defined [15]. This analysis is based on the diagonalization of the

covariance matrix built from the atomic fluctuations after the removal of the translational and

rotational movement and we applied it on the 1120 c-alpha atoms of each monomer (residue 27-

1146 in our model), after concatenating their trajectories (255 ns total sampling).

Recombination breakpoint and hotspot analyses

The presence of conflicting phylogenetic signals (i.e. distinct tree topologies compatible with the

same set of aligned sequences) in the full genome data set was investigated, first, by inferring

networks using neighbor-net [27] implemented in SplitsTree4 [28]. Significant presence of

recombination signal was then assessed with the pairwise homoplasy index (Phi) test [29].

Identification of putative recombinant events and associated breakpoints was performed using the

RDP, GENECONV, MaxChi, CHIMAERA, SIScan, and 3Seq algorithms implemented in the

RDP4 [30] software (http://web.cbio.uct.ac.za/~darren/rdp.html). Statistical evidence of

recombination was indicated by *p*-values < 0.05, and a putative recombination events were

considered as such if supported by at least five of the six algorithms used. Full table with

recombination results id given in Table S4.

Recombination breakpoints hotspot analysis was also performed using RDP4 through a

permutation test, which takes into account number of variable positions between the three

sequences involved in the event when performing simulations. Two iteration were performed, with

sliding windows of size 100 and 200, with different sensitivity to hotspots and coldspots.

Segments that had more observed breakpoints than the local maximum number of breakpoints of

99% of 1,000 permutaions were considered hotspots (p=0.01); in the same fashion, segments

that had less breakpoints than the local minimum of 99% of permutations were considered

coldspots [65].

Figures and Tables

Figure 1. Signature residues in the S glycoprotein of SARS-CoV-2 under positive selection, ribbon representations of the S protomer and MD simulations. (a) Signature residues that distinguish human SARS-COV-2 from closely related bat and pangolin (lineage 'a' strains isolated in 2017 in China; 'b' strains isolated in 2019 in China) strains; residues are colored according their charge: hydrophilic (RKDENQ) residues in blue, neutral (SGHTAP) in green, and hydrophobic (YVMCLFIW) in black. Numbers correspond to the residue position in Wuhan-Hu-1 MN908947.3 strain. In the box below we show a schematic representation of the spike glycoprotein is given: S₁ and the S₂ subunits of the S glycoprotein are indicated in grey and fuchsia, while CTD1 is shown green, CTD2 in red, CTD3 in blue. Fuchsia box around the residues between 333-516 are indicating the residues involved into rotation of CTD1. Residues

under positive selection are shaded in cyan, residues that form the ACE receptor binding are

shaded in orange, and the acquired polybasic residues (PRRA) in the S_1/S_2 cleavage site in

yellow. (b) The monomer of SARS-CoV-2 S glycoprotein and its three S1 domains (CTD1, CTD2,

and CTD3) and S2 subunit (pink) are shown in different colors: CTD1 in green, CTD2 in red,

CTD3 in blue. The N-terminal domain was shown is colored tan. The residues under selection,

ACE2 biding site residues, and S1/S2 cleavage sites were shown with cyan, orange and yellow

spheres, respectively. (c) Homo-trimer structure of the modeled SARS-CoV-2. Residues, S_1

(NTD, CTD1, CTD2, and CTD3) and the S_2 subunits are colored corresponding to panel b. (d)

The superposition of the modeled S protomer of SARS-CoV-2 with the S protomer of SARS-CoV

(PDB 5WRG). The missing NTD domain for SARS-CoV was not shown whereas the NTD domain

of SARS-CoV-2 is shown by tan color. Residues and domains are colored corresponding to panel

b. (e) Projection of the SARS-CoV-2 S glycoprotein MD trajectory along Essential Dynamics

eigenvector 1 (see also Supplementary movies 1-2). The two extreme conformations are shown

in i and ii. Domain colors as in panel b. The opening/closing mechanism of the SARS-CoV-2 S

glycoprotein monomer by means of the rotation of CTD1 is evident. (f) Per-residue Root Mean

Square Fluctuations in nm of the filtered S glycoprotein trajectory along eigenvectors 1-3. The

region with higher fluctuations along eigenvector 1 is the CTD1, highlighted by red lines, since its

rotation in the opening/closing mechanism is the most correlated motion in the protein. The peak

in fluctuations observed in eigenvectors 1 and 2, involving residues 828-852, is highlighted by

cyan lines and is indicative of a correlated motion of this region with the opening/closing motion of

CTD1. Note that this region includes a portion of the fusion peptide (Suppl. Movie 3). The peak in

fluctuations observed in eigenvector 3 involving residues 699-718 is highlighted by orange lines.

Note that this region includes all the furin-like cleavage site typical of SARS-CoV-2. Fluctuations

along eigenvector 3 shows a correlated motion between the furin-like cleavage site, residues 460-

496 in CTD1, and 250-263 in NTD (Suppl. Movie 4).

Figure 2. Recombination hot- and cold- spots prediction across CoV genomes. (a) NNet

graphs obtained using split-decomposition and based on the full genome sequence to detect

recombination. P-value of PHI test of recombination is reported. Pangolin taxa: (a)

EPI_ISL_410538-43 strains isolated in 2017 in China; (b) EPI_ISL_412860 and EPI_ISL_410721

strains isolated in 2019 in China. (b) Hotspot (upper graph) or coldspot (lower graph) presence

across the genome. P-values ≤ 0.01 indicate presence of either hotspots (upper graph) or

coldspots (lower graph).

Breakpoint in codon coordinates in aligned SARS-COV-2		Breakpoints codon coordinates in recombinant sequence				
Start	End	Start	End	Recombinant Sequence	Minor Parental Sequence	Major Parental Sequence
in poly-protein	A626	A6892 (poly-protein)	A601	^MG772934_Bat_coV_SARS-like_bat-SL-CoVZXC21_Host:Ba	KY417146_Bat_coV_SARS-like_Rs4231_Host:Bat	Unknown (KP886809_Bat_coV_SARS-likeYNLF_34C_Host:Bat)
D111	K304	D109	K298	EPI_ISL_410721_BetaCoV_pangolin_Guandong_1_2019	Unknown (EPI_ISL_410543_BetaCoV_pangolin_Guangxi_P3B_2017)	EPI_ISL_402131_BetaCoV_bat_Yunnan_RaTG13_2013
G339	P579	H335	P556	MG772933_Bat_coV_SARS-like_bat-SL-CoVZC45_Host:Bat	GQ153542_Bat_coV_SARS_HKU3-7_Host:Unknown	EPI_ISL_410721_BetaCoV_pangolin_Guandong_1_2019
T430	V511	T429	V510	^EPI_ISL_402131_BetaCoV_bat_Yunnan_RaTG13_2013	Unknown (EPI_ISL_412860_BetaCoV_pangolin_China_MP789_2019)	EPI_ISL_410535_BetaCoV_Singapore_4_2020
T768	I993	S741	I966	^MG772933_Bat_coV_SARS-like_bat-SL-CoVZC45_Host:Bat	Unknown (EPI_ISL_410538_BetaCoV_pangolin_Guangxi_P4L_2017)	EPI_ISL_412860_BetaCoV_pangolin_China_MP789_2019
L849	Q992	L818	Q961	^KF294457_SARS-relatedHost:Bat	MG772934_Bat_coV_SARS-like_bat-SL-CoVZXC21_Host:Ba	KJ473814_BtRs-BetaCoV_HuB2013Host:Bat

Table 1. Potential recombination breakpoints of CoVs detected using the RDP, GENECONV, BootScan, MaxChi, CHIMAERA, SIScan, and 3Seq

algorithms affecting ancestors of SARS-CoV-2 S protein. Significant p-values are indicated (alpha = 0.05). ~ = It is possible that this apparent

recombination signal could have been caused by an evolutionary process other than recombination. * = The actual breakpoint position is

undetermined (it was most likely overprinted by a subsequent recombination event). ^ = The recombinant sequence may have been misidentified

(one of the identified parents might be the recombinant) Minor Parent = Parent contributing the smaller fraction of sequence. Major Parent =

Parent contributing the larger fraction of sequence. Unknown = Only one parent and a recombinant need be in the alignment for a recombination

event to be detectable. The sequence listed as unknown was used to infer the existence of a missing parental sequence. NS = No significant P-

value was recorded for this recombination event using this method.

Supplementary information

Table S1. List of strains sequences from GISAID and GenBank and used for recombination

analysis. See Excel file

Table S2. Signature pattern analyses for Figure 1 panel a. See Excel file

Table S3. List of protein S glycoprotein codons under negative selection. See Excel file

Table S4. Full list of potential recombination breakpoints of CoVs detected by the RDP,

GENECONV, BootScan, MaxChi, CHIMAERA, SIScan, and 3Seq algorithms. See Excel file

Table S5. List of strains sequences from GISAID and GenBank and used for selection and

signature pattern analyses. See Excel file

Supplementary Movie S1. Projection of the S glycoprotein monomer trajectory along

eigenvector 1. Tube visualization of c-alpha atoms. Domain colors as in Figure 1. The

opening/closing mechanism of the SARS-CoV-2 monomer by means of the rotation of CTD1 is

evident.

Supplementary Movie S2. Projection of the S glycoprotein monomer trajectory along

eigenvector 1 in a different orientation.

Supplementary Movie S3. Projection of the S glycoprotein monomer trajectory along

eigenvector 1. Residues 828-852 (cyan color), including a portion of the fusion peptide, show

correlated motion with the opening/closing rotation of CTD1.

Supplementary Movie S4. Projection of the S glycoprotein monomer trajectory along

eigenvector 3. Residues 699-718 (orange color), comprehending all the furin-like cleavage site

typical of SARS-CoV-2, show correlated motion with residues 460-496 in CTD1, and 250-263 in

NTD.

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