Characterization and structural prediction of the putative ORF10 protein in SARS-CoV-2

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Abstract

Upstream of the 3'-untranslated region in the SARS-CoV-2 genome is ORF10 which has been proposed to encode for the ORF10 protein. Although current research is unclear on whether this protein is synthesized, further investigations are still warranted. Herein, this study uses multiple bioinformatic tools to biochemically and functionally characterize the ORF10 protein along with predicting its structure. Results indicate this protein is highly ordered, hydrophobic, thermally stable, and contains at least one transmembrane region. This protein possesses high protein-binding propensity, primarily in the N-terminal half. An assessment of twenty-two mutations reveal slight changes in flexibility, mainly in the C-terminal half. However, mutations do not seem to inflict significant changes on protein stability and other biochemical features. The predicted model suggests the ORF10 protein contains a β-α-β motif with a β-molecular recognition feature occurring in the first β-strand. Functionally, the ORF10 protein could be a membrane protein that forms a viroporin. The ORF10 consists of two distinct lineages: the SARS-CoV lineage and the SARS-CoV-2 lineage. Evidence of positive selection was found in the SARS-CoV-2 lineage whereas purifying selection was found in the SARS-CoV lineage. The ORF10 in SARS-CoV-2 is highly similar to ORF10 sequences in closely related bat and pangolin coronaviruses. Collectively, these results help continue to assess the biological relevance of ORF10 and its putatively encoded protein, thereby aiding in diagnostic and possibly vaccine development.

1. Introduction

An outbreak of coronavirus disease 2019 (COVID-19) in Wuhan, China has resulted in a global pandemic causing, at the time this manuscript was prepared, well over 70,475,000 cases and 1,599,000 deaths.[1] The virus responsible for causing COVID-19, severe acute respiratory syndrome coronavirus type 2 (SARS-CoV-2), contains an individual copy of a single-stranded positive-sense RNA genome that is ~29.8 kilobases in length.[2] In particular, the 3'-terminus encompasses multiple open reading frames (ORFs) that encode four main structural proteins: the spike (S), membrane (M), envelope (E), and nucleocapsid (N) proteins.[2] In addition, shorter length ORFs have been detected and proposed to encode for approximately nine different accessory proteins (3a, 3b, 6, 7a, 7b, 8, 9b, 9c, 10) in the SARS-CoV-2 genome.[2,3] Past research indicates these accessory proteins are dispensable for viral growth *in vitro*; however, the genes encoding for these proteins are maintained in the coronavirus (CoV) genome, suggesting they might play important roles within the environment of the infected host.[4]

To some degree, all of these accessory proteins have been structurally and/or functionally characterized. For example, the 3a protein was shown to induce apoptosis whereas the 9b and 9c (i.e. ORF14) proteins were suggested to be involved within membrane interactions during virion assembly and host-virus interactions, respectively.[5,6] Prior work has detected expression of RNA transcripts corresponding to these proteins in SARS-CoV-2; however, RNA transcripts belonging to ORF10 have not been observed in large amounts.[3,7] Pancer *et al.* 2020 had also identified several SARS-CoV-2 variants with prematurely terminated ORF10 sequences, finding that disease was not attenuated, transmission was not hindered, and replication proceeded similarly to strains possessing intact ORF10 sequences.[8] In light of these findings, there is some

uncertainty regarding the biological relevance of ORF10 in SARS-CoV-2, with suggestions that its genome annotation should be amended. As an unintended consequence, this putatively encoded protein in SARS-CoV-2 has been somewhat understudied.

Found upstream of the 3'-untranslated region (3'-UTR), ORF10 (117 nt long) encodes for a protein that is 38 amino acids in length.[9] Although there exists sequences homologous to the ORF10 protein in other closely related bat and pangolin CoVs, there is no experimentally derived crystallographic structure for the ORF10 protein.[10,11] Studies attempting to predict the secondary structural elements of this protein indicate one α-helix and, depending on the study, two β-strands.[12,13] The ORF10 protein also lacks significant levels of disorder; however, a short molecular recognition feature (MoRF) likely spans residues 3-7.[12] This protein is hydrophobic with the α-helix having been identified as a possible transmembrane (TM) helix as well.[14]

The exact implications are unclear, but the ORF10 protein contains high numbers of promiscuous cytotoxic T lymphocyte (CTL) epitopes, primarily on the α-helix.[12,15] A recent study by Liu *et al.* 2020 found that in severe cases of COVID-19, the ORF10 was overexpressed when compared against ORF10 expression levels in moderate cases.[16] If the ORF10 protein is synthesized, this may help to partially explain the nature behind severe cases of COVID-19. The large number of CTL epitopes in conjunction with the overproduction of the ORF10 protein could result in elevated immune responses towards SARS-CoV-2, leading to some potentially fatal immunopathological outcomes.

The ORF10 protein was shown to associate with the $CUL2^{ZYG11B}$ complex by interacting with the substrate adapter ZYG11B, thereby overtaking the complex to modulate aspects of ubiquitination and enhance viral pathogenesis.[13] However, the mechanisms and residues involved

in this interaction have not been described in detail. Alternatively, ORF10 may act itself or serve as a precursor for other RNAs in regulating gene expression/replication, translation efficiency, or interfering with cellular antiviral pathways.[7]

Two lineages for the ORF10 have been observed. The SARS-CoV lineage consists of sequences interrupted by a stop codon (TAA) encoding for a truncated protein and was found to be undergoing neutral evolution (dN/dS = 1.18).[14] The SARS-CoV-2 lineage consists entirely of uninterrupted sequences encoding for a complete protein and was found to be undergoing strong positive selection (dN/dS = 3.82) without relaxed constraints.[14] This has led to the possibility that the ORF10 in SARS-CoV-2 and closely related CoVs might encode a conserved and functional protein. Furthermore, the ORF10 present in SARS-CoV-2 and closely related CoVs is thought to have evolved from a mutation occurring within the ancestral stop codon originally found in the SARS-CoV lineage at nucleotide position 76.[17]

Collectively, there still remains minimal efforts spent on investigating the ORF10 protein within SARS-CoV-2. Herein, this current study uses a bioinformatic approach to characterize this protein, along with providing an updated phylogenetic and evolutionary analysis. This study also further examines the structural and biochemical effects of mutations found in this protein.

2. Materials and Methods

2.1 Sequence, Phylogenetic, and Evolutionary Analysis

The nucleotide (NC_045512.2) and protein (YP_009725255.1) reference sequences for ORF10 in SARS-CoV-2 were acquired from the NCBI RefSeq Database. BLASTn was used to collect additional sequences from both distantly and closely related CoVs (Table S1). Sequence alignments were performed using MUSCLE on the MEGA-X v10.1.7 software.[18,19] Alignment reliability and total similarity were determined by overall mean distance and calculated using the *p-distance* substitution model. If the p-distance was found to be ≤ 0.7 then the alignment was considered reliable and suitable for subsequent analyses.[20]

A single phylogenetic gene tree was constructed using the neighbor-joining method and visualized on MEGA-X.[21] Uniform rates among sites and complete deletion were used. The phylogeny was tested using the bootstrap method.

The average nonsynonymous/synonymous substitution (dN/dS) rate was calculated using the single-likelihood ancestor counting (SLAC) method.[22] To determine either relaxed (k < 1) or intensified (k > 1) selection, the RELAX method was utilized.[23] Both methods were performed using the datamonkey webtool (http://datamonkey.org/). Wu-Kabat variability coefficients were calculated using the protein variability server (http://imed.med.ucm.es/PVS/).[24]

2.2 Protein Characterization and Secondary Structure Prediction

Per-residue disorder scores were determined using several online tools as previously described.[25] Scores were averaged to calculate overall protein disorder. To detect phosphorylation sites, the DEPP server (http://www.pondr.com/cgi-bin/depp.cgi) was used. A hydrophobicity plot was generated using ProtScale.[26] The grand average of hydropathicity (GRAVY), aliphatic index,

and instability index were determined using ProtParam.[26] The propensity of residues involved in protein-binding were evaluated by SCRIBER.[27] Secondary structural elements were predicted using PSIPRED v4.0.[28] TM regions were predicted using TMPred.[29]

2.3 Protein Modeling and Evaluation

The webserver IntFOLD was employed to make use of an *ab initio* modeling approach in constructing the ORF10 protein.[30] Models were evaluated based on quality and confidence scoring. The best model was then refined using the 3Drefine webserver.[31] Of the models generated post-refinement, the one with a higher qualitative model energy analysis (QMEAN) Z-score was viewed as most favorable.[32] This final model was evaluated by PROCHECK and ERRAT on the SAVES v6.0 webtool (https://saves.mbi.ucla.edu/).[33-35] Quality comparisons against two previously constructed ORF10 protein models was also performed.[36,37] Access to this protein model can be found on ModelArchive (https://modelarchive.org/).

The ProFunc server predicted the likely function of the ORF10 protein based on the newly produced model.[38] UCSF Chimera was used to visualize the model.[39] Electrostatic surfaces were generated based on the AMBER ff14SB charge model. Hydrophobicity surfaces were produced according to the default Kyte-Doolittle scale.

2.4 Mutation Analysis

Using twenty-two previously collected ORF10 variants, the effect these mutations had on protein stability and biochemical features were determined.[12] The same methods used to determine the GRAVY, instability index, aliphatic index, and disorder scores were used to evaluate the effects of these mutations. The effect each mutation had on protein stability was determined by using the I-MUTANT v2.0 webtool.[40]

3. Results

3.1 Sequence, Phylogenetic, and Evolutionary Analysis

The whole nucleotide alignment (p-distance = 0.08) had 92% similarity. The phylogenetic gene tree revealed two lineages (Figure 1a). The SARS-CoV lineage (p-distance = 0.04) had 96% similarity whereas the SARS-CoV-2 lineage (p-distance = 0.08) had 92% similarity. An alignment of protein sequences taken from the SARS-CoV-2 lineage revealed a total of thirty-one conserved residues (Figure 1b). A Wu-Kabat plot displayed higher variability coefficients in the C-terminal half (20-38) rather than the N-terminal half (1-19) of the ORF10 protein, with amino acid position twenty-five presenting with the highest variability coefficient (Figure 1c).

The SARS-CoV lineage had a dN/dS = 0.713 to indicate purifying selection. Oppositely, the SARS-CoV-2 lineage had a dN/dS = 4.01 (Figure 1a). To determine if this dN/dS ratio was either the result of positive or relaxed purifying selection, the RELAX method was used. In doing so, a k = 6.79 was obtained; therefore, the hypothesis of a relaxed constraint was rejected in favor of positive selection acting upon the SARS-CoV-2 lineage.

3.2 Protein Characterization and Secondary Structure Prediction

A per-residue disorder plot for the ORF10 protein in SARS-CoV-2 showed that residues within both termini held the highest disorder scores whereas moving inwards resulted in gradually decreasing scores (Figure 2). The average disorder score for the ORF10 protein was 0.095, which indicates a highly ordered protein. No specific phosphorylation sites had been detected within this protein (Table S2). The GRAVY score was 0.637 to indicate a very hydrophobic protein. The aliphatic and instability indices were 107.03 and 16.06, respectively. Both of these values suggest a thermally stable protein. The hydrophobicity plot uncovered two hydrophobic regions covering

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residues 3-19 and 28-36, along with a hydrophilic region covering residues 20-27 (Figure 3a). As for protein-binding propensity, residues in the N-terminal half had a tendency to possess greater scores than residues found in the C-terminal half (Figure 3b). In particular, residues 1-14 presented with a relatively consistent stretch of high propensity scores to implicate this region's possible involvement in protein-binding (Figure 3b).

According to secondary structure predictions, there exists a single α-helix (11-21) and two β-strands (4-8 & 26-34) in the ORF10 protein (Table S3). Furthermore, a TM region was predicted to span residues 3-19, corresponding to the exact same hydrophobic region shown in the hydrophobicity plot (Figure 3a).

3.3 Protein Modeling and Predictive Function

IntFOLD produced five different ORF10 protein models (Table S4). The model having the lowest p-value (3.33e-03) and highest quality score (0.3714) was subjected to refinement. Of the five protein models generated post-refinement, the model with a QMEAN Z-score of -0.66 was selected as most favorable (Table S5). This final model had an ERRAT score of 93.333 and a Ramachandran plot indicated that 94% of residues (32 total) existed in the favorable region while only 6% of residues (2 total) existed in the allowed region (Figure S1). Functional analysis results by ProFunc have been summarized (Table 1). The predicted ontological terms were based on "long shot" hits using a reverse template comparison against known Protein Data Bank structures; therefore, these functional characteristics should not be viewed as definitive.

Structurally, the model presented with a β-α-β motif spanning residues 3-31, along with a 3/10-helix spanning residues 34-37 (Figure 4a). An electrostatic surface map revealed two regions of positive charge and one region of negative charge (Figure 4b). Electropositive regions

were influenced by R20 and R24 whereas the electronegative region was influenced by D31. As expected, a majority of the protein's surface was hydrophobic; however, hydrophilic regions did appear (Figure 4c). For example, residues spanning 20-27 presented with surface coloring that reflected hydrophilic character, as shown in the hydrophobicity plot as well (Figure 3a).

Prior to this study, two ORF10 protein models were built using different methods. The first model was constructed using the QUARK web server whereas the second model was built using the I-TASSER web server.[36,37] Both models were different in terms of secondary structure and topology (Figure S2). The QMEAN Z-scores (QUARK: −3.88 / I-TASSER: −2.63) obtained for both models were lower, suggesting the model produced in this study had higher quality.

3.4 Mutation Analysis

One missense mutation per variant was observed. The longest conserved region spanned residues 15-19, followed by two shorter regions spanning residues 11-12 and 33-34 (Figure 5). For mutations within the N-terminal half, disorder scores did not fluctuate; however, mutations within the C-terminal half did present with greater fluctuations in scoring (Figure 2). Regardless, average disorder scores for each variant still indicated highly ordered proteins. The effects each mutation had on protein stability and other biochemical features have been summarized (Table 2).

4. Discussion

In this study, the evolutionary history for ORF10 is similar to the history previously described by Cagliani *et al.* 2020.[14] Currently, there exists two distinct and conserved lineages for the ORF10 in CoVs: the SARS-CoV and SARS-CoV-2 lineages. Both lineages are easily distinguishable based upon whether they encode for either a truncated (SARS-CoV lineage) or full-length (SARS-CoV-2 lineage) protein. The SARS-CoV lineage is undergoing purifying selection whereas the SARS-CoV-2 lineage is undergoing positive selection without any relaxed constraints. A bat-CoV (SARS-related-CoV Rc-o319) isolated from Japan in 2013 possessed an uninterrupted ORF10 sequence with approximately 95% nucleotide and 87% amino acid similarity to ORF10 in SARS-CoV-2.[10] Following the addition of this new sequence, this study provides an updated phylogenetic and evolutionary look at ORF10 and continues to reinforce the current hypothesis that positive selection is occurring within the SARS-CoV-2 lineage. These data also suggest that ORF10 might encode a functional and conserved protein in SARS-CoV-2 and other closely related bat and pangolin CoVs.

The prediction of secondary structural elements for the ORF10 protein describe the presence of one α-helix and two β-strands. When compared to the structural assignment of residues in the ORF10 protein model, only 76% of residues match. One significant difference was the model contained a 3/10-helix. Since this model is predicative and no experimentally derived structure exists for the ORF10 protein, this 3/10-helice's existence along with its structural and functional role remains unclear. According to Hassan *et al.* 2020, a MoRF was predicted to occur in residues 3-7.[12] A unique feature associated with MoRFs is their ability to exist in semi-disordered (i.e. flexible) states until binding, after which they can assume a designated

secondary structure.[41] Altogether, it is possible residues 3-7 (or 3-8) function as a β-MoRF that upon binding to its corresponding protein ligand, will assume the structure of a β-strand. Residues found in this region are associated with a β-strand, and also have suitable residue protein-binding propensity and per-residue disorder scores to support this notion.

Protein disorder predictions show the ORF10 protein in SARS-CoV-2 is highly ordered. It is only near each terminus that residues begin to indicate some degree of flexibility. These results contradict those by Hassan *et al.* 2020.[12] Their study indicates more flexible residues, including several disordered residues within the C-terminus, exist in the ORF10 protein.[12] This disagreement is likely attributed to the method of prediction, where Hassan *et al.* 2020 used a single prediction algorithm unlike this study which used multiple prediction algorithms to evaluate disorder.[12] By doing so, per-residue disorder scores were not over estimated by a single algorithm; therefore, more conservative but reliable predictions were achieved.[25] The data in this study also indicates that mutations occurring in the C-terminus instill changes in residue flexibility; however, such changes did not result in major shifts towards disorder. Conversely, mutations in the N-terminus had caused minimal changes in residue flexibility, but as noted in the C-terminus, changes did not result in shifts towards disorder. Overall, the mutations analyzed in this study do not pose any drastic changes on the average disorder score for the ORF10 protein in SARS-CoV-2.

Mutations in the ORF10 protein decrease protein stability; however they do not result in a completely unstable protein according to the instability index and aliphatic index determined by ProtParam. As such, these mutations are not likely to have major effects on protein structure or function; however, Hassan *et al.* 2020 does indicate these mutations, regardless of their change in

polarity, are deleterious and will decrease the protein's structural integrity.[12] Modeling of these mutations could help determine if they would negatively impact protein structure.

The ORF10 protein in SARS-CoV-2 might be a membrane protein. Whether this protein is integrated within or peripheral to the membrane remains unclear, but given the extent of nonpolar residues it is reasonable to consider this protein as integrated. Furthermore, the TM predictions suggest residues 3-19 span the membrane with the N-terminus exposed to the extracellular space or, if within an organelle, the cytoplasmic space. With additional evidence coming from ProFunc and the hydrophobicity plot, it appears residues 3-19 might compose a TM region in the ORF10 protein. Residues 28-36 might be another TM region; however, TMPred did not implicate these residues and scores for these residues in the hydrophobicity plot were closer to threshold.

In particular, CoVs encode multiple viroporins.[42] For example, in SARS-CoV the 8a protein oligomerizes to form ion channels inside mitochondrial membranes that induce apoptosis by depolarizing membrane potential.[14,42-44] The ORF10 protein in SARS-CoV-2 may oligomerize in such a way that polar and electrically charged residues coat the inside of a pore that facilitates transport of ions or small molecules for viral replication, virulence, and/or pathogenicity. Based on the ORF10 protein model, it consists almost entirely of one β-α-β motif. Furthermore, these motifs are found in proteins associated with porins as well. A prior study found the ORF10 protein colocalized with ORF8 and ORF7b proteins in the endoplasmic reticulum (ER); however, its role in the ER was not described.[45] In this case, the ORF10 protein could associate with ER membranes as a means to assist in viral replication and/or processing of other viral proteins. Experimental work is necessary to further evaluate this finding. Alternatively, the ORF10 protein could interact with another protein. As previously stated, the ORF10 protein was suggested to associate with the

substrate adapter ZYG11B.[13] This interaction is likely mediated by residues in the N-terminal half on account of the β-MoRF and high protein-binding propensity scores occurring throughout residues 1-14. Altogether, it is tempting to think the ORF10 protein may be multifunctional; however, further experimental studies are necessary to explore this idea.

A limitation to this study is the quality of sequences due to the varying methods used to acquire them. Although unlikely, there is always the possibility of errors and missing information that may affect the results obtained herein. In addition, with only six sequences used, the evidence of positive selection should not be regarded as conclusive but as additional support for the hypothesis that positive selection is acting on the SARS-CoV-2 lineage for ORF10. Additional sequences of good quality from other closely related CoVs can help to better resolve this issue.

As previously mentioned, overexpression of ORF10 has been shown to occur in severe cases of COVID-19 whereas in milder cases expression seems to be minimal.[16] Therefore, knowing the structure and biochemical characteristics associated with the ORF10 protein might aid in developing diagnostic tests that detect for the ORF10 protein, thereby helping to determine the likely progression the disease may take in patients.[16] Hopefully, this study can provide a foundation that drives further work assessing the biological relevance of ORF10 and its putatively encoded protein in SARS-CoV-2, thereby aiding in diagnostic and possibly vaccine development.

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Conflict of Interest

The author declares that no conflicts of interest exist.

Tables and Figures

Gene Ontology Terminology		
Cellular Component	**Biological Process**	**Biochemical Function**
cell (0.87), cell part (0.87), membrane (0.50), integral to membrane (0.50)	metabolic process (1.84), oxidation reduction (1.17), cellular process (0.85), cellular metabolic process (0.85)	catalytic activity (2.40), transferase activity (1.29), oxidoreductase activity (1.17), metal ion binding (0.98)

Table 1. Predicted gene ontology terms for the ORF10 protein in SARS-CoV-2. These terms are the most commonly appearing terms found in the hits obtained by using the ProFunc analysis method. Each of these terms' scores are based on the number of times they independently occur during the analysis and are shown within parentheses and in red.

SARS-CoV-2 ORF10 Variant	Nucleotide Accession	Protein Accession	Missense Mutation	Change in Polarity	Change in Stability	GRAVY Score	Average Disorder	Instability Index	Aliphatic Index
var1	MT259226.1	QIS29991.1	V6I	NP → NP	-1.14	0.645	0.094	27.62	110.26
var2	MT632596.1	QKU54102.1	Y26H	NP → P	-1.68	0.587	0.105	27.62	107.63
var3	MT632956.1	QKV08176.1	L37P	NP → NP	-2.25	0.495	0.108	15.30	97.37
var4	MT641656.1	QKV37245.1	T38I	P → NP	-1.25	0.774	0.086	16.06	117.89
var5	MT679202.1	QLA48060.1	R20K	P → P	-2.39	0.653	0.095	24.64	107.63
var6	MT745708.1	QLG76514.1	N22T	P → P	-1.34	0.711	0.093	15.30	107.63
var7	MT750342.1	QLG99793.1	I3M	NP → NP	-0.17	0.568	0.098	22.29	97.37
var8	MT764166.1	QLI33453.1	Y26C	NP → P	0.41	0.737	0.094	16.06	107.63
var9	MT252755.1	QLJ57416.1	A8V	NP → NP	-0.01	0.700	0.092	16.06	112.63
var10	MT806775.1	QLY88596.1	A28V	NP → NP	-0.54	0.700	0.087	13.83	112.63
var11	MT831635.1	QMT54534.1	R24L	P → NP	-1.40	0.855	0.089	7.74	117.89
var12	MT834229.1	QMT94417.1	D31Y	P → NP	-1.48	0.695	0.081	18.03	107.63
var13	MT834697.1	QMT97141.1	S23F	P → NP	0.29	0.732	0.085	7.04	107.63
var14	MT846564.1	QMU93213.1	R24C	P → P	-1.66	0.821	0.086	16.32	107.63
var15	MT876527.1	QNA70543.1	L37F	NP → NP	-0.09	0.611	0.090	12.11	97.37
var16	MT876555.1	QNB17780.1	F35S	NP → P	-3.45	0.542	0.121	20.02	107.63
var17	MT877418.1	QNC04532.1	R20I	P → NP	-0.31	0.874	0.089	16.06	117.89
var18	MT879619.1	QNC49349.1	V30L	NP → NP	-1.31	0.626	0.097	22.00	110.26
var19	MT892994.1	QNG41574.1	I4L	NP → NP	0.52	0.618	0.098	16.06	107.63
var20	MT893673.1	QNG42985.1	P10S	NP → P	-2.23	0.658	0.096	5.92	107.63
var21	MT911797.1	QNI23218.1	G2D	NP → P	-0.71	0.555	0.093	18.29	107.63
var22	MT913142.1	QNI25281.1	V30A	NP → NP	-2.55	0.574	0.106	19.77	102.63

Table 2. Twenty-two ORF10 protein variants from SARS-CoV-2, along with their corresponding mutations and predicted structural and biochemical effects. NP - Nonpolar; P - Polar.

Figure 1. (A) Phylogenetic analysis of ORF10 sequences. The SARS-CoV lineage has branches shown in red whereas the SARS-CoV-2 lineage has branches shown in blue. Their corresponding dN/dS ratios calculated by SLAC are also shown. The phylogeny was inferred using the Jukes-Cantor method. The sum of branch lengths was 0.192 and the tree was produced out of 1,000 replicates. **(B)** Screen capture of the MUSCLE alignment for ORF10 proteins in the SARS-CoV-2 lineage. Conserved positions are indicated with an asterisk. **(C)** Wu-Kabat plot highlighting amino acid variations found in the ORF10 protein taken from the SARS-CoV-2 lineage. The consensus sequence was used as the reference sequence and the variability threshold was set to 1.0 with all other default settings maintained.

Figure 2. Per-residue disorder plot for the ORF10 protein in SARS-CoV-2, along with twenty-two variants. Scores ≥ 0.5 indicate disorder whereas scores between 0.25 and 0.5 indicate highly flexible residues. Any score that exists between 0.1 and 0.25 suggests moderate flexibility. Scores ≤ 0.1 indicate rigidity.

Figure 3. (A) Hydrophobicity plot for the SARS-CoV-2 ORF10 protein. In generating this plot, scores were normalized on a 0-1 scale and a 5-residue window was used. The red line displays the threshold where values above this line indicate hydrophobic regions and values below this line indicate hydrophilic regions. Residues 1-2 and 37-38 are not shown on this plot. **(B)** Protein-binding propensity scores for amino acids within the SARS-CoV-2 ORF10 protein. The value of each score is shown above its corresponding box. As a general rule, a higher score reflects an increased likelihood that a specific residue is going to engage in a protein-binding interaction.

Figure 4. (A) Ribbon diagram of the ORF10 protein in SARS-CoV-2. **(B)** Electrostatic surface map of the ORF10 protein. Red indicates regions having negative charge whereas blue indicates regions having positive charge. White indicates regions having neutral charge. **(C)** Hydrophobicity surface map of the ORF10 protein. Blue indicates regions of hydrophilic character and orange-red indicates regions of hydrophobic character. White indicates regions of amphipathic character.

Figure 5. Screen capture of the MUSCLE alignment consisting of the ORF10 protein in SARS-CoV-2 and twenty-two ORF10 protein variants. The red boxes indicate conserved regions with more than one residue.

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