

Do Phenotypically and Genetically Similar Animals have Similar TNF Genes?

Yvonna Leung

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Background

Tumor Necrosis Factor–alpha (TNF α), is known to regulate a multitude of cellular events, such as cell survival, proliferation, differentiation, and cell death [7]. It is a pro-inflammatory cytokine that was originally identified as a cytokine to suppress tumor cell proliferation [7]. Since its early days of discovery, other roles for TNF have emerged. The primary role of TNF is in its immunological response. TNF is able to induce fever, apoptosis, and inflammation [4]. These characteristics can help reduce tumor growth and viral replication. However, dysregulation of TNF can lead to an excess production of TNF. This phenomenon can be detrimental and be a cause of many diseases, including cancer, diabetes, and autoimmune diseases. Patients with these health conditions often have to be treated with TNF inhibitor drugs to suppress the inflammatory response [5].

Since TNF plays a major role in the host response to many diseases, it would be valuable to further understand the TNF gene by analyzing its evolutionary aspects. Evolutionary studies of genes help uncover what role some evolutionary forces, such as natural selection, have on those genes [1]. While most genes are inherited by the organism from its ancestors, other evolutionary forces can effectively change the composition of the gene over time in order for the organism to adapt to a particular environment [6]. By studying the evolutionary relationships of these genes, we can better understand how genetic changes occur within that organism and also in closely related taxa [1].

This study aims to understand how similar or dissimilar the TNF gene is across humans and primates, and whether other mammals have a similar genetic disposition to humans and primates. A total of 50 TNF sequences will be obtained, 25 of which will be primates, and the other 25 of which will be mammals from other taxa. A multiple sequence alignment (MSA) will

be built to infer the presence of ancestral relationships between the sequences. A phylogenetic tree will also be drawn to visually represent the evolutionary relationships among the species analyzed. Furthermore, these sequences will be trained through a Hidden Markov Model (HMM) that then later computes a log-Viterbi score. These analyses will help determine the similarity between the species among the primate clade and mammals from other taxa. Because humans are most similar to primates, it is to be expected that the MSA will align similarly and result in low log-Viterbi scores. Since primates may not be genetically similar with other mammals, it is to be expected that there can be a significant difference between the MSA alignments and could result in much higher log-Viterbi scores. We hypothesize that phenotypically and genetically similar animals will have genotypically similar TNF genes.

II. Methods

Data

The TNF sequences of both the primate set, labeled as the “basic set”, and the mammals set, labeled as the “related set”, were obtained from NCBI Genbank database. A Python script was written to grab these files from GenBank [2] using the accession numbers and start and stop locations from the TNF sequences. The names of each species and their accession number are represented in Table 3 in the Appendix as the basic set and Table 4 in the Appendix as the related set. Each individual sequence was then saved in individual FASTA files and then later combined into one separate file for the “basic set” and another file for the “related set”. The links for these files are located in Table 5 of the Appendix.

Procedure

After saving the sequences into a FASTA file, the next step was to align the sequences using CLUSTALW as an MSA tool. BioPython’s library, AlignIO, is a tool used to align the

sequences from both the basic set and the related set. These alignment results were then compared with the results from Clustal Omega [3], a publicly available bioinformatics tool. The alignment from BioPython's library was very similar to Clustal Omega's alignment. Clustal Omega's alignment was used to train the HMM since it had easier readability.

BioPython's library, Phylo, was used to draw phylogenetic trees for both the basic set and the related set to visually inspect the evolutionary relationships between the species. The branch lengths were used to determine which species were more similar to each other. BioPython's trees were then compared with the trees from Clustal Omega. Clustal Omega's cladograms and phylogenetic trees were used in this report since they have better visualization.

We then wrote a custom HMM builder class. This allowed us to easily specify a complex HMM using simple terms. The builder used multi-level dictionaries to store probabilities, which allowed us to easily specify transmissions and emissions. After building the HMM specification, the builder computes appropriate A, B, and pi matrices from the built-up dictionaries. Finally, the builder validates these matrices by checking that all rows sum to 1.

We then wrote a function to build our alignment HMM. Here, we incorporated match states, delete states, and insert states at every column position in the alignment training data. current state. Pseudocounts were used where appropriate to handle situations that didn't occur in the training data. Since the probabilities of a sequence containing amino acids that do not occur in the training set is 0, we have to add pseudocounts to every match and delete state for the transition probabilities, as well as emission probabilities for base pairs that weren't seen in the training data. The transition probability from a match state to an insert state is 0.01, the probability from an insert state to itself is 0.01, and the probability from an insert state back to a match state is 0.99. We added other transition probabilities, such as a match state to another

match state, by incorporating them into a builder function that then takes in the previous state, the current state, and the probability of that transition.

We also had to consider instances where there are conserved regions, which means that there are no deletions present in these regions. In this instance, the delete probabilities are 0.01. The probability of going from the previous match state to the current match state will be 0.5, while the probability of going from the previous match state to the current delete state will also be 0.5 because there is equal probability of going to either state.

Emission probabilities were calculated for each match state by dividing the total count of each nucleotide in the current column over the length of the column. Pseudocounts were also incorporated for each match and delete state. Insert states do not need pseudocounts since they have balanced emission probabilities, in which we will just add an equal probability to each nucleotide. Finally, all the different types of transition and emission probabilities were added to the HMM builder, including start and end probabilities. Start states and end states are needed because we could start or end with an insertion or a match. Because the start and end states also need to emit something, 'S' and 'E' emissions were added with emission probability 1.

To calculate the log-Viterbi scores, delta was first calculated. Delta represents a matrix of the best Viterbi scores among any path ending at some state at some timestep. Viterbi scores are probabilities of the most likely state path, though they do not directly reveal the state path on their own. To keep track of the most likely state transitions, we used another matrix called psi. The transition probabilities were calculated from any state to the current state and the emission probabilities were calculated in its current state with the current observation. Delta was then updated to capture the highest score obtained from each timestep. Once delta was fully computed, we called the maximum score among all states at the last timestep the Viterbi score of

the whole sequence. If needed, the actual optimal state path could be computed by traversing backwards through ψ , starting with the state that maximized Δ at the last timestep.

III. Results

NW_012150019.1:1286304-1289063	CTTATCAGGTTTGTGCGCTGTCTGCCCTGGTACGCC---TGGTTCTCTCTCTCCATTTCAT	1945
NW_024100919.1:131130662-131133756	GCCAGCCCTG-GCCTGCGCTCTTAGCCCTGAGGTGCTCGTTTTTTTTTCTCCATTTCAT	2253
NW_018508801.1:631021-634098	GCCGTCCCTG-GCCTGCATCTTAGCCCTGAGGTGCTGCTGTTTT---TTTCTCTCCATTTCAT	1926
NW_016820117.1:442773-445553	GCCTGTCTG-GCCTGCGCTCTTAGCCCTGAGTTGTCTGGTTTT---TGCTCTCCATTTCAT	1942
NW_012003394.1:351660-354591	GCCTGTCTG-GCCTGCGCTCTTAGCCCTGAGTTGTCCGGTTTT---TCTCTCTCCATTTCAT	1942
NC_041757.1:13793450-137937146	GCCTGTCTG-ACCTGCGCTCTTAGCCCTGAGTTGTCCGGTTTT---TCTCTCTCCATTTCAT	1942
NW_012011989.1:775591-778797	GCCGTCTG-GCCTGCGCTCTTAGCCCTGAGTTGTCCGGTTTT---TCTCTCTCCATTTCAT	1942
NC_033665.1:1462712-1465617	GGATGGAGGAAGAGTGGGTGAATGGAGAGGAAACCAGA-CGTAAATCAGACGTGTAC	1008
NC_048386.1:35786290-35788191	TGTTGAACGCATGGAGAGTGAATACACAGATGAATGGAGAAAAAA-GCAGACACCTCAG	813
NW_022437128.1:419840-423140	TGTTGAATGCAGGAGGGTGAATACCGAGATGAATGGAGAAAAAA-AGCAGACCTCAG	1400
NW_036909.1:31862568-31865345	TGTTGAATGCCGTGAGGGTGAATACGCAGATGAATGGAGAGAGAAAACCAGACACCTCAG	887
NG_007462.1:4993-7764	TGTTGAATGCCGTGAGGGTGAATACACAGATGAATGGAGAGAGAAAACCAGACACCTCAG	880
NC_036885.1:30973796-30977728	CGTTGAATGCCGTGAGGGTGAATACACAGATGAATGGAGAGAGAAAACCAGACACCTCAG	2041
NC_048245.1:31236650-31239423	CGTTGAATGCCGTGAGGGTGAATACACAGATGAATGGAGAGAGAAAACCAGACACCTCAG	882
NW_044402.1:52604482-52607247	TGTTGAATGCATGGAGGGTGAATACGCAGATGAGTGGAGAGAGAAAACCAGACACCTCAG	874
NW_022611662.1:30018474-30021238	TGTTGAATGCATGGAGGGTGAATACGCAGATGAGTGGAGAGAGAAAACCAGACACCTCAG	874
NW_022681455.1:31615357-31618475	TGTTGAATGCATGGAGGGTGAATACACAGATGAATGGAGAGACAAAACCAGACAACTCAG	1211
NC_045438.1:136321804-136324643	TGTTGAATGCATGGAGGGTGAATACACAGATGAATGGAGACAGCAAACCAGACAACTCAG	925
NC_044552.1:25134100-25136912	TGTTGAATTCATAGAGGGTGAATACACAGATGAATGGAGAGACAAAACCAGACAACTCAG	902
NW_023666044.1:31477411-31480675	TGTTGAATGCATGGAGGGTGAATACACAGATGAATGGAGAGAGAAAACCAGACAACTCAG	1354

Training was also performed on a conserved and variable 250 base pairs region from 1561bp to 1801bp for the related set as shown in Figure 2.

NW_003573558.1:1525860-1529761	AAATAGAGGGGGCCC-GATTCA-TGCGGA--G-AGGCAGGG--GCTTCATATCTCAGGA	1535
NW_011515227.1:483309-486096	AAATAGAGGGAACCTG-GCCCTG-CTAGGG--GTGGGGGTGG--GCCCTCCATC--TCAG	408
NC_037350.1:27716170-27718943	AAACAGAGGGAGTTG-GCCAG--TGGGG--TTGGGGCTGG--GCTTCCCACC--TCAG	407
NC_040271.1:29552630-29555397	AAACAGAGGGAGTTG-GCCAG--TGGGG--TTGGGGCTGG--GCTTCCCACC--TCAG	407
NT_176404.1:41351534-41354077	AAATAGAGGGGCTTG-GCCCTGTGGGGAG--GGAGGGTTGG--GCTCTATAGCTCAGGG	397
NC_000083.7:35418343-35420983	AAATAGAGGGGGG---CTGGCTCTGTGA--GGAAGGCTGT--GCATTGCACCTCAGGG	405
NW_003614548.1:582968-584758	-----	0
NC_013680.1:20399432-20402011	CCGGCTGAGCTGCCA-CCTGTTGCTCCT---TTGAGCGTGATTCCCCCATGCTAATCCTC	792
NW_006804941.1:169600-171296	GTAGATAAGCAGTCTAGTT---ATTCTCTCTTTAGAGGTGACTTGCTATAATTTTAATCC	675
NC_040906.2:104407235-104409853	CTAGATACGTA-----GCCAGCTGTTTCTATCTAGGGGCGACTTGCTCTGATTCTAATTC	983
NC_051816.1:1219807-1221672	AGTCTGAGCTGCATAAGCTGTTTCTCCT---ATAGGGGTGACTTGCTCTGATGCTAAACC	700
NC_048222.1:26734884-26737746	TACATAAAGCAGCCTGGCTGTTTCTCAT---TTAGGGGTGACTTGCTCTGTTGCTAAACC	856
NW_022631180.1:1350442-1353155	TACAGAAGCAGCC-TAGCTGTTT-CTCA---TTTGGGGTGACTTGCTCTGATGCTAAAC	892
NC_030830.1:22245930-22248693	CTAGAGAAGCAGCCA-GCAGTTTCTCCTTC---AGGGGTGACTTGCTCTAACACTCATCC	925
NC_037546.1:25245470-25248257	CTAGAGAAGCAGCCA-GCAGTTTCTCCTTC---AGGGGTGACTTGCTCTAACACTCATCC	953
NW_011493987.1:129540-132367	CTAGAGAAGCAGCCA-GCAGTTTCTCCTTC---AGGGGTGACTTGCTCTAACACTCATCC	993
NW_005394817.1:170906-173674	CTAGAGAAGCAGCCA-GCAGTTTCTCCTTC---AGGGGTGACTTGCTCTAACACTCATCC	934
NC_047043.1:23230434-23233200	CTAGAGAAGCAGCCG-GCTGTTTCTCCTTC---AGGGGTGACTTGCTTTGATACTAATCC	914
NW_022098071.1:4290720-4293500	CTAGAGAAGCAGCCG-GCTGTTTCTCCTTC---AGGGGTGACTTGCTTTGATACTAATCC	934
NC_018727.3:32583572-32585339	CTGCATAAACACCTAGCTGTTTCTCGGTT---AGGGGTAACTTGTTCTGATGCTAAACC	720
NC_010449.5:23699635-23702393	CTAGAGAAGCAGGTG-GCTGTTTCTCCTTCAGAGGGGACTTATTCAAATCTAATTAATCC	935
NC_009163.3:32223398-32226182	CTAGATAAGCAGCCTGGCTGTTTCTCCTGT---AGGGATGACTTGCTCTGATGCTAATCC	929
NC_051355.1:3622011-3624629	CTGGCAAAGAGCGGG-GAGGCTTCTCCTC---TTTGTGGTGAGT-CTGCTACTAACCTAC	813
NW_024404946.1:35623427-35626009	CTGAATAAGCAGCCA-GCTGATTCTC-----CTCTGGGTGAT-TCCTCAGTACT-AAA	863
NW_017871006.1:263552-265756	CTATAAAGCATGGT-TATCTATCTT-----TTGGGGG-TGAT-TCCTCAGATGCTGACC	852

Figure 2: A section of the conserved and variable region from the Basic set MSA alignment from Clustal Omega.

After the MSA alignment was performed, the cladograms and phylogenetic trees were obtained from our code and compared against Clustal Omega. Figure 3A and 3B show the cladogram and phylogenetic tree for the basic set and Figure 4A and 4B show the cladogram and phylogenetic tree for the related set all from Clustal Omega.

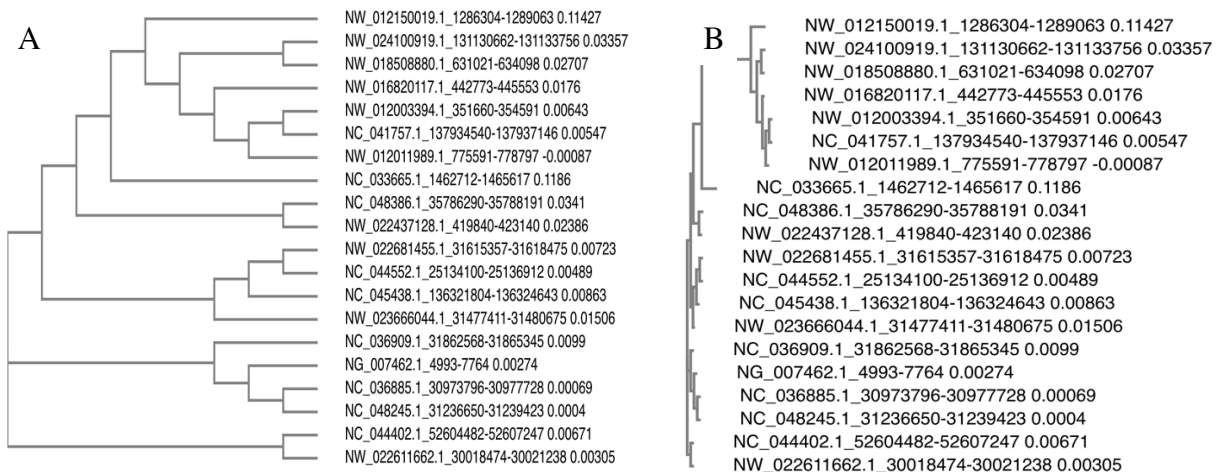


Figure 3: A. Cladogram of the Basic Set displayed from Clustal Omega. B. Phylogenetic tree of the Basic Set from Clustal Omega.

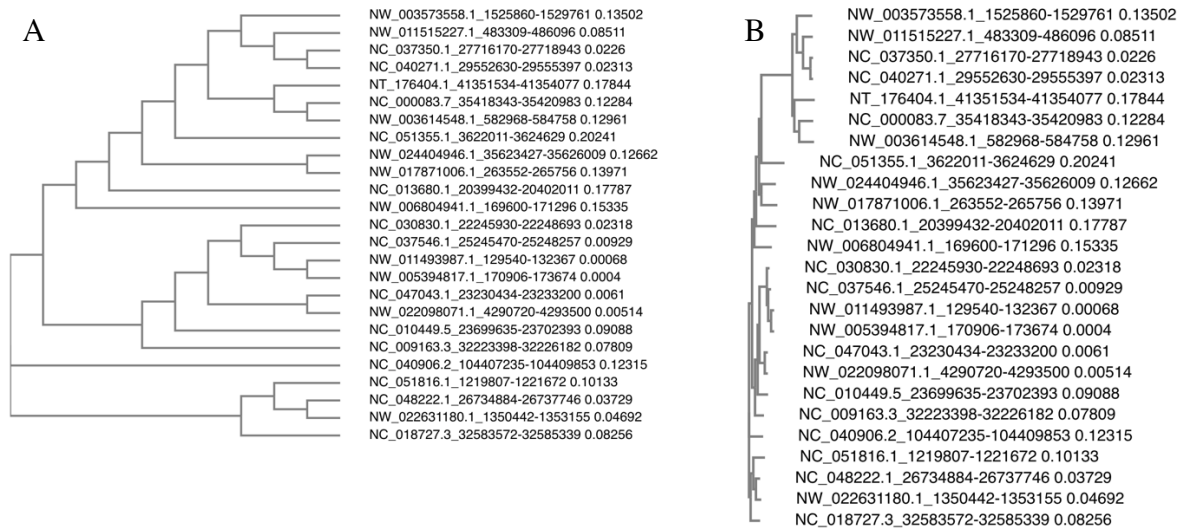


Figure 4: A. Cladogram of the Related Set displayed from Clustal Omega. B. Phylogenetic tree of the Related Set from Clustal Omega.

Viterbi Scores were calculated from our Viterbi algorithm. The log-Viterbi scores were computed for the remaining 20% sequences for both the basic set and related set and compared against both the trained 80%, or 20, sequences of the basic set and the trained 80% of sequences of the related set. Their values are displayed below in Table 1 and Table 2.

Test Sequence from Basic Set	Accession Number	Log-Viterbi Scores: From Trained Basic Set on 20 sequences	Log-Viterbi Scores: From Trained Related Set on 20 sequences
Gelada	NC_037671	-756.85	-609.25
Small-Eared Galago	NW_003852441	-798.31	-616.43
Angola Colobus	NW_012118154	-780.07	-563.01
Drill	NW_012104920	-812.35	-596.18
Crab-Eating Macaque	NC_022287	-799.43	-505.25

Table 1: Log-Viterbi scores computed from the basic test set trained from 2521bp to 2761bp

Test Sequence from Related Set	Accession Number	Log-Viterbi Scores: From Trained Basic Set on 20 sequences	Log-Viterbi Scores: From Trained Related Set on 20 sequences
Giant Panda	NC_048222	-799.94	-563.95
N.American River Otter	NW_022631180	-inf	-575.65
Angola Colobus	NW_022098071	-759.77	-579.29
Beluga Whale	NW_005394817	-758.69	-602.94
W. European Hedgehog	NW_006804941	-inf	-inf

Table 2: Log-Viterbi scores computed from the related test set trained from 1561bp to 1801bp.

IV. Discussion

After aligning the 25 human and primate TNF sequences for the basic set using multiple sequence alignment, it was found that there were many indel regions in the beginning of the alignment. Only after the alignment reached 60 base pairs did we finally begin to see some alignment between the first 7 species. However, some of the other species in the basic set continued to not show alignment for hundreds of base pairs. This observation could be explained by the fact that not all primates are that closely related genetically. One example of species that do show close alignment are humans and chimpanzees. This would make sense as it had been found that humans are most closely related to chimpanzees and pygmy chimpanzees. The high volume of indels could also be attributed to the observation that not all the sequence lengths are the same. As we traverse through the base pairs, we do begin to see more conserved domains that contain some variable regions. These regions would be good candidates for the HMM to train on.

Similarly, we also saw that there were many regions of indels in the beginning of the alignment in the related set. The more conserved regions occurred later in the alignment, indicating that these conservations could be related to the functional property of TNF. The related set's alignment results were not surprising since having so many indels could be due to how these mammals could be vastly different genetically.

From the phylogenetic tree results of the basic set, it was found that there are many closely related primates. For instance, the branch length between chimpanzees and pygmy chimpanzees was among one of the shorter branch lengths. As shown in Figure 5A, the next branch length was that of human, indicating that humans are closely related to both these chimpanzees. In the phylogenetic tree results of the related set, it was found that mammals of similar taxa were closely related to each other. As shown in Figure 5B, the branch length

between cattle and sheep was among one of the shorter branch lengths. These results correlate to our current understanding of the cattle family, which contains sheep.

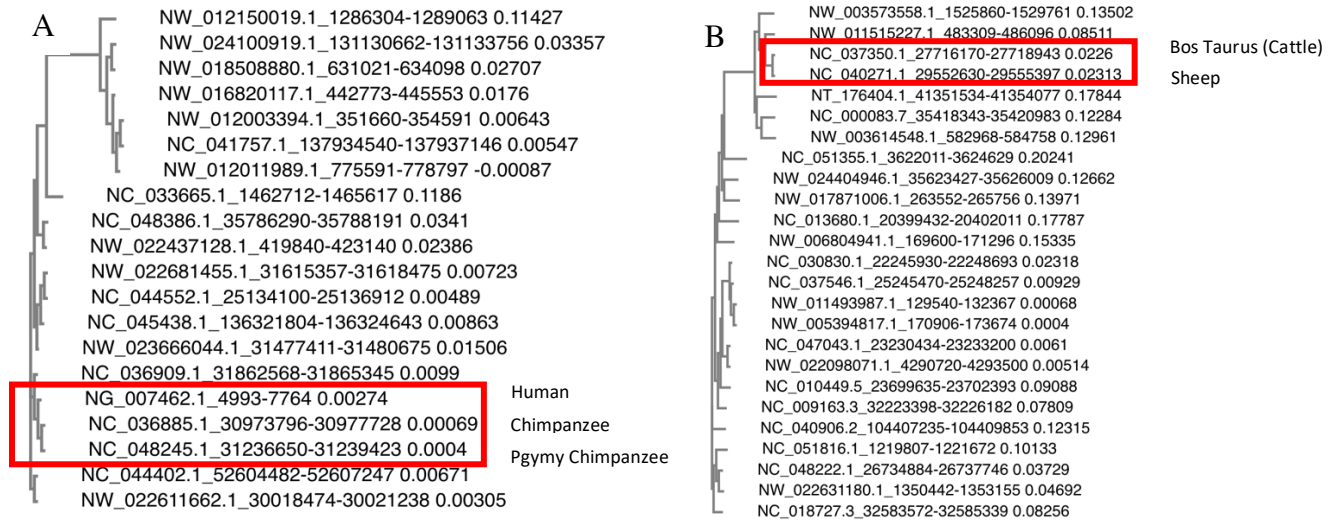


Figure 5: Comparison of phylogenetic trees. A: Phylogenetic tree of the basic set. B: Phylogenetic tree of the related set.

After analyzing the MSA and the phylogenetic tree results, the sequences were then used to train an HMM for scoring other sequences. Comparing the log-Viterbi scores could help us determine the similarity between the species among the primate clade and mammals from the other taxa. The higher the log-Viterbi score, the more similar the sequences in the test set are to the training set. Likewise, the lower the log-Viterbi score, the less similar the sequences in the test set are to the training set. Because the primates are in the same clade and therefore are phenotypically and genetically similar, it was expected that the basic training set would produce higher log-Viterbi scores to the basic test set. Since primates are not genetically similar to other mammals in the related set, it was expected that the basic training set would produce lower log-Viterbi scores to the related test set. However, the opposite result was observed in Table 1, in that the trained basic set produced lower log-Viterbi scores for the basic test set compared with the related test set. This result could be due to the regions selected for training and testing.

Because the subset chosen to train was somewhat arbitrarily chosen, the quality of the alignment could be highly dependent on the regions used. Although the chosen region appeared to have contained a good combination of conserved domains and variable region, perhaps it still wasn't conserved enough. There could be higher variability in the sequences than expected, and thus led to higher log-Viterbi scores.

We would expect that a TNF sequence in the related set would be more closely aligned with sequences in the related set than in the primate set. Hence, the related test set would produce higher log-Viterbi scores to the related training set. Likewise, the related test set would produce lower log-Viterbi scores to the basic training set. The results observed in Table 2 follow the hypothesis. The related set that was used to train an HMM on the related test set show better alignment compared with the basic set. However, there were a couple of scores that had a result of “-inf”. This could be due to the sequence having a zero probability of occurring, either because of a length mismatch or possibly an oversight in our code. These results indicated that mammals in the related set are still more closely aligned with each other than the primates in the basic set.

The results of this research were somewhat inconclusive, neither strongly supporting nor refuting the hypothesis. This is because the MSA supports the hypothesis, but the HMM does not. More analysis would need to be done. If there was more time to improve the findings of this project, multiple regions of the alignment could be tested to determine which region is the best for training the HMM model. Overall, the log-Viterbi scores were still very low, which could be a result of the chosen region's poor alignment quality. Since the TNF sequences ranged upward of 2000 bp, it was also difficult to train the HMM model well. Training on the whole gene, rather than a single section could also produce interesting results. Perhaps, it would also have been

more fruitful to find a gene that had a shorter sequence first before testing it on a much longer sequence or look at multiple genes to find any similarity between them.

References

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Appendix

Species Name	Length of TNF Sequence	Accession Number	Start Seq, End Seq
<u>Human</u>	2772 bp	NG_007462	4993, 7764
<u>Rhesus Monkey</u>	2607 bp	NC_041757	137934540, 137937146
<u>Chimpanzee</u>	3933 bp	NC_036885	30973796, 30977728
<u>Sumatran Orangutan</u>	2778 bp	NC_036909	31862568, 31865345
<u>White-Tufted-Ear Marmoset</u>	1902 bp	NC_048386	35786290, 35788191
<u>Northern White-Cheeked Gibbon</u>	2766 bp	NC_044402	52604482, 52607247
<u>Sooty Mangabey</u>	2932 bp	NW_012003394	351660, 354591
<u>Francois's Langur</u>	3119 bp	NW_022681455	31615357, 31618475
<u>Tufted Capuchin</u>	3301 bp	NW_022437128	419840, 423140
<u>Blank Snub-Nosed Monkey</u>	2781 bp	NW_016820117	442773, 445553
<u>Gray Mouse Lemur</u>	2906 bp	NC_033665	1462712, 1465617
<u>Coquerel's Sifaka</u>	2760 bp	NW_012150019	1286304, 1289063
<u>Ma's Night Monkey</u>	3078 bp	NW_018508880	631021, 634098
<u>Pig-Tailed Macaque</u>	3207 bp	NW_012011989	775591, 778797
<u>Golden Snub-Nosed Monkey</u>	2813 bp	NC_044552	25134100, 25136912
<u>Bolivian Squirrel Monkey</u>	3095 bp	NW_024100919	131130662, 131133756
<u>Silvery Gibbon</u>	2765 bp	NW_022611662	30018474, 30021238
<u>Pgymy Chimpanzee (Bonobo)</u>	2774 bp	NC_048245	31236650, 31239423
<u>Green Monkey</u>	3265 bp	NW_023666044	31477411, 31480675
<u>Ugandan Red Colobus</u>	2840 bp	NC_045438	136321804, 136324643
<u>Gelada</u>	2767 bp	NC_037671	34037747, 34040513
<u>Small-Earged Galago</u>	3509 bp	NW_003852441	11768637, 11772145
<u>Angola Colobus</u>	3434 bp	NW_012118154	3616485, 3619918
<u>Drill</u>	3436 bp	NW_012104920	1288916, 1292351
<u>Crab-Eating Macaque</u>	3669 bp	NC_022287	7627176, 7630844

Table 3: Primate species names for the basic set and their accession numbers

Species Name	Length of TNF Sequence	Accession Number	Start Seq, End Seq
<u>Mouse</u>	2641 bp	NC_000083	35418343, 35420983
<u>Norway Rat</u>	2619 bp	NC_051355	3622011, 3624629
<u>Pale Spear-Nosed Bat</u>	2619 bp	NC_040906	104407235, 104409853
<u>Bos Taurus</u> (Cattle)	2774 bp	NC_037350	27716170, 27718943
<u>Sus scrofa</u> (pig)	2759 bp	NC_010449	23699635, 23702393
<u>Canis Lupis</u> (dog)	1866 bp	NC_051816	1219807, 1221672
<u>Rabbit</u>	2580 bp	NC_013680	20399432, 20402011
<u>Sheep</u>	2768 bp	NC_040271	29552630, 29555397
<u>Horse</u>	2785 bp	NC_009163	32223398, 32226182
<u>Bison</u>	2828 bp	NW_011493987	129540, 132367
<u>Felis catus</u> (Cat)	1768 bp	NC_018727	32583572, 32585339
<u>Water Buffalo</u>	2788 bp	NC_037546	25245470, 25248257
<u>Common Bottlenose Dolphin</u>	2767 bp	C_047043	23230434, 23233200
<u>Guinea Pig</u>	2544 bp	NT_176404	41351534, 41354077
<u>Bactrian Camel</u>	2788 bp	NW_011515227	483309, 486096
<u>Goat</u>	2764 bp	NC_030830	22245930, 22248693
<u>Chinese Hamster</u>	1791 bp	NW_003614548	582968, 584758
<u>African Savanna Elephant</u>	3902 bp	NW_003573558	1525860, 1529761
<u>Squirrel</u>	2583 bp	NW_024404946	35623427, 35626009
<u>Beaver</u>	2205 bp	NW_017871006	263552, 265756
<u>Giant Panda</u>	2863 bp	NC_048222	26734884, 26737746
<u>N.A. River Otter</u>	2714 bp	NW_022631180	1350442, 1353155
<u>Beluga Whale</u>	2781 bp	NW_022098071	4290720, 4293500
<u>Wild Yak</u>	2769 bp	NW_005394817	170906, 173674
<u>W. Euro. Hedgehog</u>	1697 bp	NW_006804941	169600, 171296

Table 4: Mammals species names for the related set and their accession numbers

Set Name	Fasta File Link Shared on Google Drive
Basic Set	https://drive.google.com/file/d/1AYFgjGRFfyuf6LcVzfrEkxot3I5H7Ta/view?usp=sharing
Related Set	https://drive.google.com/file/d/1Iww5MEPz1jmd0IC0Fu_zeWY2_exH_YsL/view?usp=sharing

Table 5: Fasta files containing MSA alignment for both the basic set and the related set.