Python Assignment

```
In [1]: from Bio import SeqIO
    from Bio.Data import CodonTable
    import pandas as pd
    from Bio.Seq import Seq
    import matplotlib as plt
    import matplotlib.pyplot as plt
    plt.style.use('ggplot')
```

```
In [3]: | pip install biopython
```

```
Requirement already satisfied: biopython in c:\users\shazid\anaconda3\lib\si te-packages (1.78)
Requirement already satisfied: numpy in c:\users\shazid\anaconda3\lib\site-p ackages (from biopython) (1.21.5)
Note: you may need to restart the kernel to use updated packages.
```

1

```
In [4]: def get_sequences_from_file(fasta_fn):
    sequence_data_dict = {}
    for record in SeqIO.parse(fasta_fn, "fasta"):
        description = record.description.split()
        species_name = description[1] + " " + description[2]
        sequence_data_dict[species_name] = record.seq
    return(sequence_data_dict)
```

Here, def is defining the function get_sequences_from_file with the fasta file as an argument.

This is a Python function called <code>get_sequences_from_file</code> that takes a string fasta_fn as its input parameter. The purpose of this function is to read a fasta file and store the sequence data in a dictionary format.

First, an empty dictionary called sequence data dict is created.

The function uses the SeqIO.parse function from the Biopython library to read the fasta file. This function takes two parameters, the name of the fasta file (fasta_fn in this case) and the format of the file ("fasta"). It returns an iterator that produces SeqRecord objects for each sequence in the fasta file.

The function then loops through each SeqRecord object produced by SeqIO.parse. For each record, it splits the description string into a list of words using the split function. The second and third elements of this list are concatenated to form the species_name variable.

Finally, the sequence_data_dict dictionary is updated with a new key-value pair, where the key is the species_name variable and the value is the actual DNA sequence, accessed using the seq attribute of the SeqRecord object.

Once all sequences in the fasta file have been processed, the function returns the sequence_data_dict dictionary containing all the DNA sequences from the file, with species

2

In [5]: mito_table = CodonTable.unambiguous_dna_by_name["Vertebrate Mitochondrial"]
 print(mito_table) #to look at mito_table contents

Table 2 Vertebrate Mitochondrial, SGC1

	T	С	Α	G	
T T T	TTT F	TCT S TCC S TCA S TCG S	TAT Y TAC Y TAA Stop TAG Stop		T C A G
С	CTT L	CCT P	CAT H	CGT R	Т
С	CTC L	CCC P	CAC H	CGC R	С
C	CTA L	CCA P	CAA Q	CGA R	Α
C	CTG L	CCG P	CAG Q	CGG R	G
	++				
Α	ATT I(s)	ACT T	AAT N	AGT S	Т
Α	ATC I(s)		AAC N	AGC S	C
Α	\mid ATA M(s) \mid	ACA T	aaa k	AGA Stop	Α
Α	ATG M(s)	ACG T	AAG K	AGG Stop	G
G	GTT V	GCT A	GAT D	GGT G	Т
G	GTC V	GCC A	GAC D	GGC G	С
G	GTA V	GCA A	GAA E	GGA G	Α
G	GTG V(s)	GCG A	GAG E	GGG G	G
	+				

The "Vertebrate Mitochondrial" codon table is one of the predefined tables available in Biopython. It is a special codon table used by mitochondrial DNA in vertebrates, which has some differences from the standard codon table used by nuclear DNA in those same organisms.

```
In [6]:

def translate_function(string_nucleotides):
    mito_table = CodonTable.unambiguous_dna_by_name["Vertebrate Mitochondrial
    aa_seq_string = ''
    if len(string_nucleotides)%3 == 0:
        for i in range(0, len(string_nucleotides), 3):
            codon = string_nucleotides[i:i + 3]
    if codon in ["TAA", "TAG", "AGA", "AGG"]:
            return aa_seq_string
    aa_seq_string+= mito_table[codon]
    return aa_seq_string
```

This is a Python function called translate_function which takes a string of nucleotides as input and returns a string of amino acids.

The function first accesses the Vertebrate Mitochondrial codon table from the CodonTable module provided by Biopython. This table maps each possible codon (triplet of nucleotides) to the corresponding amino acid.

Next, the function checks if the length of the input nucleotide string is divisible by 3 (the length of a codon). If not, the function does not attempt to translate the string and returns an empty string.

If the string length is divisible by 3, the function proceeds to iterate over the string, processing one codon at a time. For each codon, the function checks if it is a stop codon (TAA, TAG, AGA, AGG). If it is, the function returns the current amino acid sequence string without including the stop codon.

If the current codon is not a stop codon, the function looks up the corresponding amino acid for that codon in the mito_table using the codon as the key. The function then appends the amino acid to the current amino acid sequence string.

Finally, the function returns the completed amino acid sequence string.

This code is collected from https://www.geeksforgeeks.org/dna-protein-python-3/ (https://github.com/florist-notes/Data-Analysis/blob/master/data-analysis%20project%20-%203%20-%20DNA_Translation/dna_seq.py_(https://github.com/florist-notes/Data-Analysis/blob/master/data-analysis%20project%20-%203%20-%20DNA_Translation/dna_seq.py_))

3

This is a Python function called translate which takes a string of nucleotides as input and returns a string of amino acids.

The function uses Biopython's Seq object to translate the input nucleotide sequence into a protein sequence. The translate method of the Seq object takes several optional arguments. In this case, the table argument is set to 2 in the library which is "Vertebrate Mitochondrial", which corresponds to the standard genetic code used by most organisms, and the to_stop argument is set to True, which means translation will stop at the first in-frame stop codon.

The resulting protein sequence is returned as a Seq object, so the function converts it to a string using the str function before returning it.

Note that the translate method will automatically handle any necessary adjustments for frameshifts or incomplete codons at the end of the input sequence.

4

This is a Python function called compute_molecular_weight which takes a string of amino acids as input and returns the molecular weight of the protein.

The function uses Biopython's ProteinAnalysis module to analyze the input amino acid sequence. The ProteinAnalysis class provides various methods to compute different properties of a protein sequence, such as molecular weight, isoelectric point, and secondary structure composition.

In this case, the ProteinAnalysis object is initialized with the input amino acid sequence aa_seq . Then, the molecular_weight method is called on the aa_analysis object to calculate the molecular weight of the protein sequence.

Finally, the calculated molecular weight is returned as the output of the function.

```
In [24]: def gc_content(DNA_string):
    G_count = DNA_string.count("G")
    C_count = DNA_string.count("C")
    DNAstring_length = len(DNA_string)
    GC_content = (C_count+G_count)/DNAstring_length
    return GC_content
```

This is a Python function called gc_content which takes a string of DNA nucleotides as input and returns the GC content of the DNA sequence.

The function first counts the number of occurrences of the nucleotides G and C in the input DNA sequence using the count method. These counts are stored in the variables G_count and C_count .

The function also calculates the length of the input DNA sequence by using the len function and storing the result in the variable DNAstring_length.

The GC content is then calculated by dividing the sum of G and C counts by the total length of the DNA sequence. The result is stored in the variable GC_content.

Finally, the GC content is returned as the output of the function.

```
In [26]: cytb_seqs = get_sequences_from_file("penguins_cytb.fasta")
In [27]: penguins_df = pd.read_csv("penguins_mass.csv")
    species_list = list(penguins_df.species)
```

In [13]: cytb_seqs #checking if it was correctly imported

'Aptenodytes patagonicus': Seq('ATGGCCCCAAACCTCCGAAAATCCCATCCTCTAAAAAATAA TTAATAACTCC...TAA'),

'Eudyptes chrysocome': Seq('ATGGCCCCCAACCTCCGAAAATCCCACCCCCTCCTAAAAAACAATCAA TAACTCC...TAA'),

'Eudyptes chrysolophus': Seq('ATGGCCCCCAACCTCCGAAAATCCCACCCCCTCCTAAAAAACAATC AATAACTCC...TAA'),

'Eudyptula minor': Seq('ATGGCCCCCAACCTCCGAAAATCTCACCCCCTCCTAAAAATAATCAACAACTCT...TAA'),

'Pygoscelis adeliae': Seq('ATGGCCCCCAACCTCCGAAAATCCCACCCTCTCCTAAAAATAATTAAC AACTCC...TAA'),

'Pygoscelis antarctica': Seq('ATGGCCCCCAACCTCCGAAAATCCCACCCTCTCCTAAAAATAATCAACAACTCC...TAG'),

'Pygoscelis papua': Seq('ATGGCCCCCAACCTTCGAAAAATCCCACCCTCTCCTAAAAAATAATCAACAA ATCC...TAG'),

'Spheniscus demersus': Seq('ATGGCCCCCAACCTCCGAAAATCCCACCCTCTCCTAAAAACAATCAA CAACTCC...TAA'),

'Spheniscus humboldti': Seq('ATGGCCCCCAACCTCCGAAAATCCCACCCTCTCCTAAAAACAATCA ACAACTCC...TAA'),

'Spheniscus magellanicus': Seq('ATGGCCCCCAACCTCCGAAAATCCCACCCTCTCCTAAAAAACAA TCAACAACTCC...TAA')}

In [14]: penguins_df #checking the csv file information

Out[14]:

	species	mass
0	Aptenodytes forsteri	28.00
1	Aptenodytes patagonicus	13.40
2	Eudyptes chrysocome	2.80
3	Eudyptes chrysolophus	4.50
4	Eudyptes sclateri	4.25
5	Eudyptula minor	1.60
6	Pygoscelis adeliae	4.60
7	Pygoscelis antarctica	4.10
8	Pygoscelis papua	6.10
9	Spheniscus demersus	3.20
10	Spheniscus humboldti	4.75
11	Spheniscus magellanicus	3.40

Out[16]:

	species	mass	Molecular Weight
0	Aptenodytes forsteri	28.00	NaN
1	Aptenodytes patagonicus	13.40	NaN
2	Eudyptes chrysocome	2.80	NaN
3	Eudyptes chrysolophus	4.50	NaN
4	Eudyptes sclateri	4.25	NaN
5	Eudyptula minor	1.60	NaN
6	Pygoscelis adeliae	4.60	NaN
7	Pygoscelis antarctica	4.10	NaN
8	Pygoscelis papua	6.10	NaN
9	Spheniscus demersus	3.20	NaN
10	Spheniscus humboldti	4.75	NaN
11	Spheniscus magellanicus	3.40	NaN

In [17]: penguins_df["GC content"]= penguins_df.shape [0]*["NaN"] #adding "GC content"
penguins_df

Out[17]:

	species	mass	Molecular Weight	GC content
0	Aptenodytes forsteri	28.00	NaN	NaN
1	Aptenodytes patagonicus	13.40	NaN	NaN
2	Eudyptes chrysocome	2.80	NaN	NaN
3	Eudyptes chrysolophus	4.50	NaN	NaN
4	Eudyptes sclateri	4.25	NaN	NaN
5	Eudyptula minor	1.60	NaN	NaN
6	Pygoscelis adeliae	4.60	NaN	NaN
7	Pygoscelis antarctica	4.10	NaN	NaN
8	Pygoscelis papua	6.10	NaN	NaN
9	Spheniscus demersus	3.20	NaN	NaN
10	Spheniscus humboldti	4.75	NaN	NaN
11	Spheniscus magellanicus	3.40	NaN	NaN

Here, two new columns named "Molecular Weight" and "GC content" have been added to a pandas DataFrame called penguins_df. The columns were initialized with NaN values.





	species	mass	Molecular Weight	GC content
0	Aptenodytes forsteri	28.00	42459.6021	48.381452
1	Aptenodytes patagonicus	13.40	42563.7067	49.693788
2	Eudyptes chrysocome	2.80	42475.5753	51.181102
3	Eudyptes chrysolophus	4.50	42445.5493	50.918635
4	Eudyptes sclateri	4.25	42475.5753	50.831146
5	Eudyptula minor	1.60	42491.6408	49.256343
6	Pygoscelis adeliae	4.60	42458.6140	49.081365
7	Pygoscelis antarctica	4.10	42404.5423	47.769029
8	Pygoscelis papua	6.10	42595.8759	47.156605
9	Spheniscus demersus	3.20	42431.5490	48.293963
10	Spheniscus humboldti	4.75	42399.5520	49.256343
11	Spheniscus magellanicus	3.40	42459.6021	48.206474

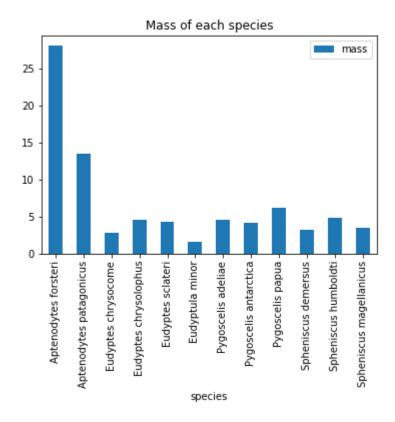
The first two lines of the code create empty lists called Molecular_w_list and GC_content_list that will be used to store the computed molecular weights and GC contents of the DNA sequences.

The code then loops over the items in the cytb_seqs dictionary using a for loop. In each iteration of the loop, the DNA sequence value is translated into an amino acid sequence using the translate function from Part 3, and the molecular weight of the resulting amino acid sequence is computed using the compute_molecular_weight function from Part 4. The GC content of the DNA sequence is also computed and multiplied by 100 to obtain the percentage value.

The computed molecular weight and GC content values are then appended to the Molecular_w_list and GC_content_list lists, respectively, using the append method.

Finally, the "Molecular Weight" and "GC content" columns of the penguins_df DataFrame are updated with the computed molecular weight and GC content values, respectively, using the assignment operator =. The updated DataFrame is then printed to the console using the print function.

```
In [19]: %matplotlib inline
In [77]: penguins_df.plot(kind='bar',x='species',y='mass', title='Mass of each species
Out[77]: <AxesSubplot:title={'center':'Mass of each species'}, xlabel='species'>
```



Q1: What is the smallest penguin species? Ans: Eudyptula minor is the smallest penguin by mass Q2: What is the geographical range of this species? Ans: Geographical range: Coastlines of southern Australia and New Zealand

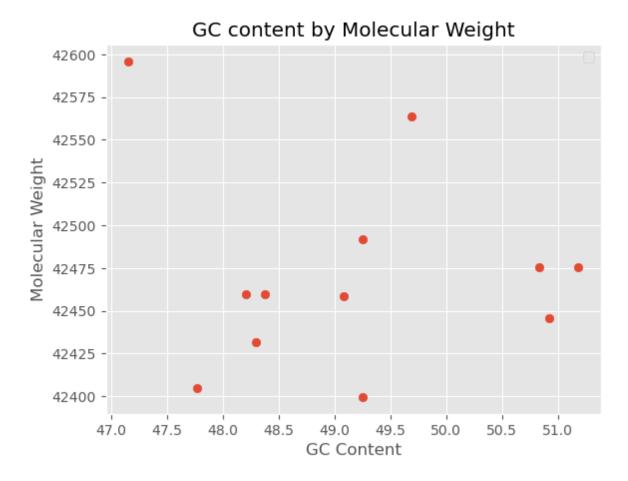
Source: https://www.wildlife.vic.gov.au/ data/assets/pdf file/0023/91391/Little-Penguin.pdf (https://www.wildlife.vic.gov.au/ data/assets/pdf file/0023/91391/Little-Penguin.pdf)

```
In [20]: fig, ax =plt.subplots()  #plotting a scatter plot
my_scatter_plot =ax.scatter(
    penguins_df["GC content"],  #GC content from penguins dataframe
    penguins_df["Molecular Weight"]  #Molecular Weight from penguins dataframe
)

plt.title('GC content by Molecular Weight')
plt.xlabel('GC Content')
plt.ylabel('Molecular Weight')
plt.legend () #Added a title, labeled x and y axes
```

No artists with labels found to put in legend. Note that artists whose labe 1 start with an underscore are ignored when legend() is called with no argum ent.

Out[20]: <matplotlib.legend.Legend at 0x2744ad95d60>



```
In [29]: penguins_df.to_csv('penguins_mass_cytb.csv', index=False) #saving the new data
In []:
```