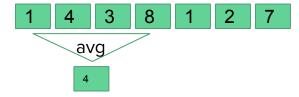
Scientific Programming Practical 10

Introduction

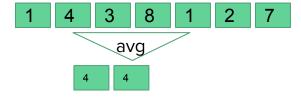
6. Implement a function movingAvg(A, n) where A is a numpy one dimensional array and n is the window size on which computing the average, that outputs the moving average over a numpy one dimensional array. Es. A = [1,2,3,4,5] movingAvg(A,2) = [1.5, 2.5, 3.5, 4.5], while movingAvg(A,3) = [2,3,4] without using for loops. Hint: use cumsum and clever slicing. Assess the smoothing effect of the moving average by creating a numpy array containing a sinusoidal wave with some additional noise (i.e. use np.sin and np.random.rand) and testing several values of the window size n.

Example (win:4):



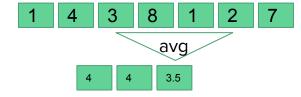
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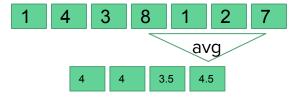
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Example (win:4):

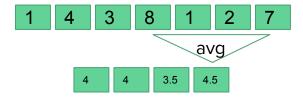


Cumulative sum:

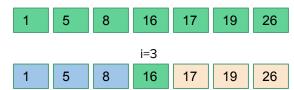


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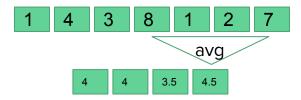
Cumulative sum:



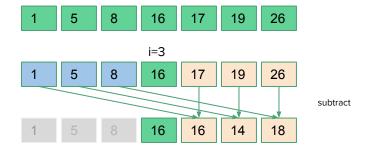
Clever bit: when we move to the right with the window we need to disregard (i.e. subtract) the blue elements one after the other

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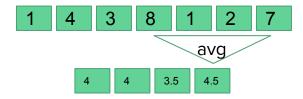
Cumulative sum:



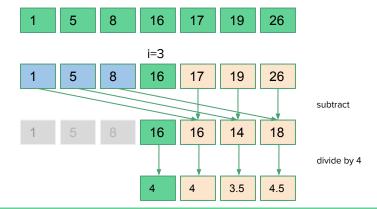
Clever bit: when we move to the right with the window we need to disregard (i.e. subtract) the blue elements one after the other Let's subtract them after the starting point (i=3)

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Example (win:4):



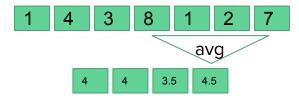
Cumulative sum:



Clever bit: when we move to the right with the window we need to disregard (i.e. subtract) the blue elements one after the other Let's subtract them after the starting point (i=3) Finally, let's compute the mean value (i.e. divide by 4)

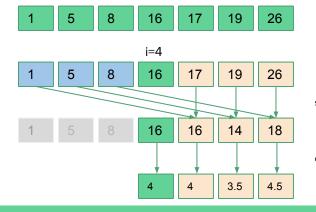
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Example:

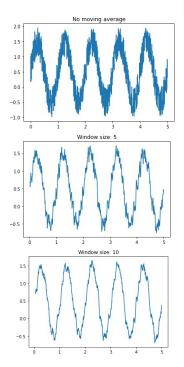


of the window size n.

Cumulative sum:



```
def movingAvg(v, n):
          """computes the moving average of n values in the array v"""
          cs = np.cumsum(v)
         #print(cs)
         #print("\t",cs[n:])
         #print("\t", cs[:-n])
          cs[n:] = cs[n:] - cs[:-n]
         #print("CS:", cs)
         return cs[n - 1:] / n
         X = np.arange(0, 5, 0.005)
         \#B = np.random(100)
         B = np.sin(2*np.pi*X)
         #Let's add random numbers uniformly distributed in [0,1)
         B += np.random.random sample(1000)
         \#B += np.random.rand(1000)
         C = movingAvg(B, 5)
         D = movingAvg(B, 10)
         E = movingAvg(B, 50)
         #X = np.arange(0, B.shape[0])
         plt.plot(X,B)
         plt.title("No moving average")
         plt.show()
         plt.close()
         plt.plot(np.arange(0, C.shape[0]),C)
         plt.title("Window size: 5")
         plt.show()
subtract
         plt.close()
         plt.title("Window size: 10")
         plt.plot(np.arange(0, D.shape[0]),D)
         plt.show()
divide by 4 plt.close()
         plt.title("Window size: 50")
         plt.plot(np.arange(0, E.shape[0]), E)
         plt.show()
```



```
6. Implement a function <a href="movingAvg(A,n">movingAvg(A,n</a>) where A is a numpy one dimensional array and n is the window size on which computing the average, that outputs the moving average over a numpy one dimensional array. Es. A = [1,2,3,4,5] <a href="movingAvg(A,2">movingAvg(A,2)</a> = [1.5, 2.5, 3.5, 4.5] while <a href="movingAvg(A,3">movingAvg(A,2)</a> = [1.5, 2.5, 3.5, 4.5] while <a href="movingAvg(A,3">movingAvg(A,2)</a> = [1.5, 2.5, 3.5, 4.5] while <a href="movingAvg(A,3)">movingAvg(A,2)</a> = [1.5, 2.5, 3.5, 4.5] while <a href="movingAvg(A,3)">movingAvg(A,2)</a> = [1.5, 2.5, 3.5, 4.5] while <a href="movingAvg(A,2)">movingAvg(A,2)</a> = [1.5, 2.5, 3.5, 4.5] while <a href="movingAvg(A,2)">movingAvg(A,2)</a>
```

Window size: 50

1.25

1.00

0.50

0.25

0.00

-0.25

```
def movingAvg(v, n):
    """computes the moving average of n values in the array v"""
    cs = np.cumsum(v)
    #print(cs)
    #print("\t", cs[n:])
    #print("\t", cs[:-n])
    cs[n:] = cs[n:] - cs[:-n]
    #print("CS:", cs)
    return cs[n - 1:] / n
```

```
X = np.arange(0, 5, 0.005)
\#B = np.random(100)
B = np.sin(2*np.pi*X)
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B += np.random.random sample(1000)
\#B += np.random.rand(1000)
C = movingAvg(B, 5)
D = movingAvg(B, 10)
E = movingAvg(B, 50)
\#X = np.arange(0, B.shape[0])
plt.plot(X.B)
plt.title("No moving average")
plt.show()
plt.close()
plt.plot(X[4:],C) # X has 1000- 5 -1 elements
plt.title("Window size: 5")
plt.show()
plt.close()
plt.title("Window size: 10")
plt.plot(X[9:],D) # X has 1000- 9 -1 elements
plt.show()
plt.close()
plt.title("Window size: 50")
plt.plot(X[49:],E) # X has 1000- 49 -1 elements
plt.show()
```

Biopython



FROM Biopython's website:

The Biopython Project is an international association of developers of freely available **Python tools for computational molecular biology**.

The goal of Biopython is to make it as easy as possible to use **Python for bioinformatics** by creating high-quality, reusable modules and classes.

Biopython



Biopython:

- 1. Provides tools to **parse several common bioinformatics formats** (e.g. FASTA, FASTQ, BLAST, PDB, Clustalw, Genbank,..).
- 2. Provides an **interface towards biological data repositories** (e.g. NCBI, Expasy, Swiss-Prot,..)
- 3. Provides an **interface towards some bioinformatic tools** (e.g. clustalw, MUSCLE, BLAST,...)
- 4. **Implements some tools** like pairwise alignment **and data structures** to deal with biological data.

More material at:

Seq objects are more powerful than strings to deal with sequences and are defined in the module **Bio.Seq**.

They have two information:

1. SEQUENCE

2 AI PHARET

(optional, but useful to check things)

defined in Bio.Alphabet.IUPAC

Some options:

```
Bio.Alphabet.generic_dna,
Bio.Alphabet.generic_protein,
Bio.Alphabet.ThreeLetterProtein,
Bio.Alphabet.generic_alphabet, ...
```

They are **immutable objects**. The mutable version is **MutableSeq**.

```
from Bio. Seg import Seg
from Bio.Alphabet import IUPAC
#No alphabet specified
s = Seg("GATTACATAATA")
dna seq = Seq("GATTATACGTAC", IUPAC.unambiguous dna)
print("5:", s)
print("S's alphabet:", s.alphabet)
print("dna seq:", dna seq)
print("dna seg's alphabet:", dna seg.alphabet)
my prot = Seq("MGNAAAAKKGSEQE", IUPAC.protein)
print("my prot:", my prot)
print("my prot's alphabet:", my prot.alphabet)
S: GATTACATAATA
S's alphabet: Alphabet()
dna seg: GATTATACGTAC
dna seq's alphabet: IUPACUnambiguousDNA()
my prot: MGNAAAAKKGSEQE
```

my prot's alphabet: IUPACProtein()

Seq objects behave like strings, but the consistency of the alphabet is checked too.

For example we cannot concatenate a unambiguous_dna with a IUPAC.protein sequence.

```
my mess = dna seq + my prot
S: GATTACATAATA
S's alphabet: Alphabet()
dna seg: GATTATACGTAC
dna seg's alphabet: IUPACUnambiguousDNA()
my prot: MGNAAAAKKGSEQE
my prot's alphabet: IUPACProtein()
                                          Traceback (most recent call last)
TypeError
<ipython-input-20-4c1f6d65a691> in <module>()
     14 print("my prot's alphabet:", my prot.alphabet)
     15
---> 16 my mess = dna seq + my prot
/usr/local/lib/python3.5/dist-packages/Bio/Seq.py in add (self, other)
    296
                        raise TypeError(
    297
                            "Incompatible alphabets {0!r} and {1!r}".format(
--> 298
                                self.alphabet, other.alphabet))
    299
                   # They should be the same sequence type (or one of them is generic)
    300
                    a = Alphabet. consensus alphabet([self.alphabet, other.alphabet])
TypeError: Incompatible alphabets IUPACUnambiguousDNA() and IUPACProtein()
```

Seq objects behave like strings, but the consistency of the alphabet is checked too

For example we cannot concatenate a **unambiguous_dna** with a **IUPAC.protein** sequence.

```
from Bio.Seq import Seq
from Bio.Alphabet import generic_alphabet

dna_seq = Seq("GATTATACGTAC", IUPAC.unambiguous_dna)
my_prot = Seq("MGNAAAAKKGSEQE", IUPAC.protein)

my_prot.alphabet = generic_alphabet

#Does it really make sense though?!?
print(dna_seq + my_prot)
```

GATTATACGTACMGNAAAAKKGSE0E

Seq objects behave like strings, but the consistency of the alphabet is checked too.

We can loop through the elements of the sequence and perform slicing...

```
from Bio. Seq import Seq
from Bio.Alphabet import IUPAC
dna seq = Seq("GATTATACGTACGGCTA", IUPAC.unambiguous dna)
for base in dna seq:
    print(base, end = " ")
print("")
sub seq = dna seq[4:10]
print(sub seq)
#Let's reverse the string:
print("Reversed: ", dna seq[::-1])
#from Seg to string:
dna str = str(dna seq)
print("As string:", dna str)
print(type(dna str))
GATTATACGTACGGCTA
ATACGT
Reversed: ATCGGCATGCATATTAG
As string: GATTATACGTACGGCTA
<class 'str'>
```

Biopython provides several methods working on Seq objects (remember Seq are immutable!)

3



5' AUGGCCAUUGUAAUGGGCCGCUGAAAGGGUGCCCGAUAG Single stranded messenger RNA General methods (return int and Seq objects):

Seq.count(s) : counts the number of times s appears in the sequence;
Seq.upper() : makes the sequence of the object Seq in upper case
Seq.lower() : makes the sequence of the object Seq in lower case

Only for DNA/RNA (return **Seq** objects):

Seq.complement() to complement the sequence
Seq.reverse_complement() to reverse complement the sequence.
Seq.transcribe() transcribes the DNA into mRNA
Seq.back_transcribe() back transcribes mRNA into DNA
Seq.translate() translates mRNA or DNA into proteins

Other functions are in **SeqUtils**(ex. use from Bio.SeqUtils import molecular_weight):

```
SeqUtils.GC(Seq) computes GC content
SeqUtils.molecular_weight(Seq) computes the molecular weight of the seq
....
```

Check out: http://biopython.org/DIST/docs/api/

Biopython provides several methods working on Seq objects (remember Seq are immutable!)

Up to first stop:

MAIVMGRWKGAR*

MAIVMGR

ATGGCCATTGTAATGGGCCGCTGAAAGGGTGCCCGATAG
AUGGCCAUUGUAAUGGGCCGCUGAAAGGGUGCCCGAUAG

IUPACUnambiguousRNA()

... and back
ATGGCCATTGTAATGGGCCGCTGAAAGGGTGCCCGATAG

Translation to protein:
MAIVMGR*KGAR*

Mitocondrial translation: (TGA is W!)

```
from Bio. Seg import Seg
from Bio.Alphabet import IUPAC
coding dna = Seg("ATGGCCATTGTAATGGGCCGCTGAAAGGGTGCCCGATAG",
                 IUPAC.unambiguous dna)
print(coding dna)
mrna = coding dna.transcribe()
print(mrna)
print("")
print(mrna.alphabet)
print("")
print("... and back")
print(mrna.back transcribe())
print("")
print("Translation to protein:")
prot = mrna.translate()
print(prot)
print("")
print("Up to first stop:")
print(mrna.translate(to stop = True))
print("")
print("Mitocondrial translation: (TGA is W!)")
mit prot = mrna.translate(table=2)
print(mit prot)
```

Sequence annotations

The **SeqRecord** object is used to store annotations associated to sequences. They might provide:

SeqRecord.seq: the sequence (the Seq object)
 SeqRecord.id: the identifier of the sequence, typically an accession number
 SeqRecord.name: a "common" name or identifier sometimes identical to the accession number
 SeqRecord.description: a human readable description of the sequence
 SeqRecord.letter_annotations: a per letter annotation using a restricted dictionary (e.g. quality)
 SeqRecord.annotations: a dictionary of unstructured annotation (e.g. organism, publications,...)
 SeqRecord.features: a list of SeqFeature objects with more structured information (e.g. genes pos).
 SeqRecord.dbxrefs: a list of database cross references.

Sequence annotations

Read a fasta file NC005816.fna containing the whole sequence for Yersinia pestis biovar Microtus str. 91001 plasmid pPCP1 and retrieve some information about the sequence.

← → C A https://www.ncbi.nlm.nih.gov/nuccore/NC 005816

>gi|45478711|ref|NC_005816.1| Yersinia pestis biovar Microtus str. 91001 plasmid pPCP1, complete sequence

S NCBI Resources ☑ How To ☑ Nucleotide Nucleotide ▼ NC005816 Advanced Learn more about upcoming changes to the Nucleotide, EST, and GSS databases GenBank -Send to: Yersinia pestis biovar Microtus str. 91001 plasmid pPCP1, complete sequence NCBI Reference Sequence: NC 005816.1 FASTA Graphics Go to: 🗸 LOCUS NC 005816 9609 bp DNA circular CON 11-JAN-2018 DEFINITION Yersinia pestis biovar Microtus str. 91001 plasmid pPCP1, complete sequence. ACCESSION NC 005816 VERSION NC 005816.1 BioProject: PRJNA224116 BioSample: SAMN02602970 Assembly: GCF 000007885.1 KEYWORDS RefSeq. SOURCE Yersinia pestis biovar Microtus str. 91001 Yersinia pestis biovar Microtus str. 91001 Bacteria; Proteobacteria; Gammaproteobacteria; Enterobacterales; Yersiniaceae; Yersinia. REFERENCE 1 (bases 1 to 9609) Zhou, D., Tong, Z., Song, Y., Han, Y., Pei, D., Pang, X., Zhai, J., Li, M., Cui, B., Qi, Z., Jin, L., Dai, R., Du, Z., Wang, J., Guo, Z., Wang, J., Huang, P. and Yang, R. Genetics of metabolic variations between Yersinia pestis biovars

https://www.ncbi.nlm.nih.gov

Sequence annotations

Read a fasta file NC005816.fna containing the whole sequence for Yersinia pestis biovar Microtus str. 91001 plasmid pPCP1 and retrieve some information about the sequence.

```
ID: gi|45478711|ref|NC 005816.1|
Name: gi|45478711|ref|NC 005816.1|
Description: gi|45478711|ref|NC 005816.1| Yersinia
pestis biovar Microtus str. 91001 plasmid pPCP1,
complete sequence
Number of features: 0
Seg('TGTAACGAACGGTGCAATAGTGATCCACACCCAACGCCTGAAATCAGAT
CCAGG...CTG', SingleLetterAlphabet())
Sequence [first 30 bases]:
TGTAACGAACGGTGCAATAGTGATCCACAC
The id:
gi|45478711|ref|NC 005816.1|
The description:
gi|45478711|ref|NC 005816.1| Yersinia pestis biovar
Microtus str. 91001 plasmid pPCP1, complete sequence
The record is a: <class 'Bio.SeqRecord.SeqRecord'>
```

```
from Bio import SeqIO
record =
SegIO.read("file samples/NC 005816.fna",
"fasta")
print(record)
print("")
print("Sequence [first 30 bases]:")
print(record.seq[0:30])
print("")
print("The id:")
print(record.id)
print("")
print("The description:")
print(record.description)
print("")
print("The record is a: ", type(record))
```

SeqIO.parse

The Bio.SeqIO module aims to provide a simple way to work with several different sequence file formats

The method Bio.SeqIO.parse is used to parse some sequence data into a SeqRecord iterator. In particular, the basic syntax is:

```
SeqRecordIterator = Bio.SeqIO.parse(filename, file_format)
```

where filename is typically an open handle to a file and file_format is a lower case string describing the file format. Possible options include fasta, fastq-illumina, abi, ace, clustal... all the

Note that Bio.Seq10.parse returns an iterator, therefore it is possible to manually fetch one SeqRecord after the other with the next(iterator) method.

Formats available: https://biopython.org/wiki/SeqIO

WARNING: When dealing with very large FASTA or FASTQ files, the overhead of working with all these objects can make scripts too slow. In this case SimpleFastaParser and FastqGeneralIterator parsers might be better as they return just a tuple of strings for each record.

SegIO

Example: Let's get the first 3 entries of the .fasta file contigs82.fasta printing off the length of the sequence and the first 50 bases of each sequence followed by "...".

```
In [12]:
         from Bio import SeqIO
         seqIterator = SeqIO.parse("file samples/contigs82.fasta", "fasta")
         labels = ["1st", "2nd", "3rd"]
         for l in labels:
            segRec = next(segIterator)
            print(l, "entry:")
            print(seqRec.id, " has size ", len(seqRec.seq))
            print(seqRec.seq[:50]+"...")
            print("")
        1st entry:
        MDC020656.85 has size 2802
        GAGGGGTTTAGTTCCTCATACTCGCAAAGCAAAGATACATAAATTTAGAA...
        2nd entry:
        MDC001115.177 has size 3118
        TGAATGGTGAAAATTAGCCAGAAGATCTTCTCCACACATGACATATGCAT...
         3rd entry:
        MDC013284.379 has size 5173
```

SeqIO

With SimpleFastaParser...

```
labels = ["1st", "2nd", "3rd"]
with open("file samples/contigs82.fasta") as cont handle:
   for l in labels:
       ID, seg = next(SimpleFastaParser(cont handle))
       print(l, "entry:")
       print(ID, " has size ", len(seq))
       print(seq[:50]+"...")
       print("")
1st entry:
MDC020656.85 has size 2802
GAGGGGTTTAGTTCCTCATACTCGCAAAGCAAAGATACATAAATTTAGAA...
2nd entry:
MDC013284.379 has size 5173
3rd entry:
MDC018185.241 has size 23761
AAAACGAGGAAAATCCATCTTGATGAACAGGAGATGCGGAGGAAAAAAAT...
```

SeqIO

The module Bio. SeqIO also has three different ways to allow random access to elements:

- Bio.SeqIO.to_dict(file_handle/iterator) : builds a dictionary of all the SeqRecords keeping them in memory and allowing modifications to the records. This potentially uses a lot of memory but is very fast;
- Bio.SeqIO.index(filename,file_type) : builds a sort of read-only dictionary, parses the elements into SeqRecords on demand (i.e. it returns an iterator!). This method is slower, but more memory efficient;
- Bio.SeqIO.index_db(indexName.idx, filenames, file_format): builds a read-only dictionary, but stores ids and offsets on a SQLite3 database. It is slower but uses less memory.

SeqIO.write

The module Bio.SeqIO provides also a way to write sequence records to files in various formats (like fasta, fastq, genbank, pfam...)

SeqRecords can be written out to files by using

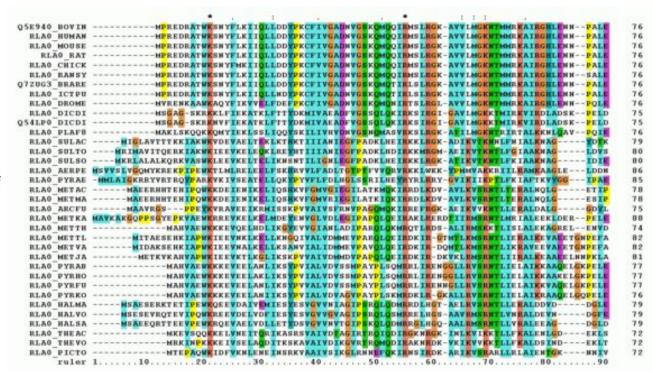
```
N = Bio.SeqIO.write(records,out_filename, file_format)
```

where **records** is a list of the SeqRecords to write, **out_filename** is the string with the filename to write and **file_format** is the format of the file to write. **N** is the number of sequences written.

WARNING: If you write a file that is already present, SeqIO.write will just rewrite it without telling you.

Multiple sequence alignment

Multiple Sequence
Alignments are a
collection of multiple
sequences which have
been aligned together —
usually with the insertion of
gap characters, and
addition of leading or
trailing gaps — such that all
the sequence strings are
the same length.



In Biopython, each row is a SeqRecord object and alignments are stored in an object MultipleSeqAlignment

Parsing MSAs: The basic syntax of the two functions: **AlignIO**

```
Bio.AlignIO.parse(file handle, alignment format)
Bio.AlignIO.read(file handle, alignment format)
```

where file handle is the handler to the opened file, while the alignment format is a lower case string with the alignment format (e.g. fasta, clustal, stockholm, mauve, phylip,...).

The function Bio.AlignIO.parse() returns an iterator of MultipleSegAlignment objects that is a collection of SegRecords.

Each SegRecord contains several information like the **ID**, **Name**, **Description**, Number of features, start, end and sequence.

In the frequent case that we have to deal with a single multiple alignment we will have to use the Bio.AlignIO.read() function.

```
from Bio import AlignIO
alignments = AlignIO.read("file samples/PF02171 seed.sth", "stockholm")
for align in alignments:
    start = align.annotations["start"]
   end = align.annotations["end"]
    seq = align.seq
    desc = align.description
    dbref = ",".join([x for x in align.dbxrefs])
   print("{} S:{} E:{}".format(desc, start, end))
    if(len(dbref) > 0):
        print(dbref)
    print("{}".format(seq))
    print("")
AG01 SCHP0/500-799 S:500 E:799
YLFFILDK-NSPEP-YGSIKRVCNTMLGVPSQCAISKHILQS------KPQYCANLGMKINVKVGGIN-CSLIPKSNP----L
AG06 ARATH/541-851 S:541 E:851
FILCILPERKTSDI-YGPWKKICLTEEGIHTOCICPIKI-----SDOYLTNVLLKINSKLGGIN-SLLGIEYSYNIPLI
AG04 ARATH/577-885 S:577 E:885
FILCVLPDKKNSDL-YGPWKKKNLTEFGIVTOCMAPTROPND------OYLTNLLLKINAKLGGLN-SMLSVERTPAFTVI
TAG76 CAEEL/660-966 S:660 E:966
CIIVVLQS-KNSDI-YMTVKEQSDIVHGIMSQCVLMKNVSRP-----TPATCANIVLKLNMKMGGIN--SRIVADKITNKYL
```

Writing and converting MSAs

Biopython provides a function Bio.AlignIO.write() to write alignments to file

and

Bio.AlignIO.convert() to convert one format into the other (provided that all information needed for the second format is available)

```
N = Bio.AlignIO.write(alignments,outfile,file format)
where alignments are a MultipleSeqAlignment object with the alignments to write to the output
file with name outfile that has format file format (a low case string with the file format). N is
the number of entries written to the file.
 Ex.
 my alignments = [align1, align2, align3]
 N = AlignIO.write(my alignments, "file samples/my malign.phy", "phylip")
  Bio.AlignIO.convert(input file, input file format, output file, output file format)
 basically by passing the input file name and format and output file name and format.
 Ex:
  Bio.AlignIO.convert("PF05371 seed.sth", "stockholm", "PF05371 seed.aln", "clustal")
```

Example: Convert the seed alignment of the <u>Piwi (PF02171) family</u> stored in the pfam (stockholm) format <u>PF02171_seed.sth</u> into phylip format. Print some stats on the data.

```
N. of seq: 16
Len of seq: 395
1 multiple alignments converted to phylip
```

STOCKHOLM 1.0

#=GS AG01 SCHP0/500-799

#=GS AG06 ARATH/541-851

AC 074957.1

AC 048771.2

```
from Bio import AlignIO
alignments = AlignIO.read("file samples/PF02171 seed.sth", "stockholm")
out = AlignIO.convert("file samples/PF02171 seed.sth",
                        "stockholm".
                        "file samples/PF05371 seed.aln",
                        "clustal")
print("N. of seg: {}\nLen of seg: {}".format(
                                                      len(alignments),
                                                      len(alignments[0])))
print("{} multiple alignments converted to phylip".format(out))
       CLUSTAL X (1.81) multiple sequence alignment
       AGO1 SCHPO/500-799
                                     YLFFILDK-NSPEP-YGSIKRVCNTMLGVPSOCAISKHILOS------
       AG06 ARATH/541-851
                                     FILCILPERKTSDI-YGPWKKICLTEEGIHTOCICPIKI------
       AGO4 ARATH/577-885
                                     FILCVLPDKKNSDL-YGPWKKKNLTEFGIVTOCMAPTROPND------
```

```
#=GS AGO4 ARATH/577-885
                          AC Q9ZVD5.2
#=GS TAG76 CAEEL/660-966 AC P34681.2
#=GS 016720 CAEEL/566-867 AC 016720.2
#=GS 062275_CAEEL/594-924 AC 062275.1
#=GS YQ53 CAEEL/650-977
#=GS NRDE3 CAEEL/673-1001 AC 021691.1
#=GS 017567 CAEEL/397-708 AC 017567.1
#=GS AUB_DROME/555-852
#=GS PIWI_DROME/538-829
#=GS PIWL1_HUMAN/555-847 AC Q96J94.1
#=GS PIWI ARCFU/110-406
                          AC 028951.1
#=GS PIWI ARCFU/110-406
                          DR PDB: 2W42 B: 110-406:
#=GS PIWI ARCFU/110-406
                          DR PDB: 1YTU B: 110-406:
#=GS PIWI ARCFU/110-406
                          DR PDB; 2BGG B; 110-406;
#=GS PIWI ARCFU/110-406
                          DR PDB; 1W9H A; 110-406;
#=GS PIWI_ARCFU/110-406
                          DR PDB; 2BGG A; 110-406;
#=GS PIWI_ARCFU/110-406
                          DR PDB; 1YTU A; 110-406;
#=GS PIWI ARCFU/110-406
                          DR PDB; 2W42 A; 110-406;
#=GS Y1321 METJA/426-699
                          AC 058717.1
#=GS 067434 AQUAE/419-694
                          AC 067434.1
#=GS 067434 AQUAE/419-694 DR PDB; 1YVU A; 419-694;
#=GS 067434 AQUAE/419-694 DR PDB; 2F8S A; 419-694;
#=GS 067434 AQUAE/419-694 DR PDB; 2F8T A; 419-694;
#=GS 067434 AOUAE/419-694 DR PDB: 2F8S B: 419-694:
#=GS 067434 AOUAE/419-694 DR PDB: 2NUB A: 419-694:
#=GS 067434 AOUAE/419-694 DR PDB: 2F8T B: 419-694:
#=GS AG010_ARATH/625-946 AC Q9XGW1.1
AG01 SCHP0/500-799
YLFFILDK.NSPEP.YGSIKRVCNTMLGVPSQCAISKHILQS.......KPQYCANLGMKINVKVGGIN.CSLIPKSNP....LGNVPTL......ILGGDVYHPG\
AGO6 ARATH/541-851
FILCILPERKTSDI.YGPWKKICLTEEGIHTQCICPIKI......SDQYLTNVLLKINSKLGGIN.SLLGIEYSYNIPLINKIPTL.....ILGMDVSHGPF
AGO4 ARATH/577-885
FILCVLPDKKNSDL.YGPWKKKNLTEFGIVTQCMAPTRQPND......QYLTNLLLKINAKLGGLN.SMLSVERTPAFTVISKVPTI......ILGMDVSHGSF
TAG76 CAEEL/660-966
CIIVVLQS.KNSDI.YMTVKEQSDIVHGIMSQCVLMKNVSRP......TPATCANIVLKLNMKMGGIN..SRIVADKITNKYLVDQPTM......VVGIDVTHPT(
016720 CAEEL/566-867
LIVVVLPG..KTPI.YAEVKRVGDTVLGIATQCVQAKNAIRT.......TPQTLSNLCLKMNVKLGGVN.SILLPNVRPR...IFNEPVI.......FLGCDITHPA/
TEVELITO.DSITT.LHORYKMIEKDTKMIVODMKLSKALSV..IN...AGKRLTLENVINKTNVKLGGSN..YVEVDAKKOL.....DSHL.......IIGVGISAPPA
```

```
TAG76 CAEEL/660-966
                                    CIIVVLOS-KNSDI-YMTVKEOSDIVHGIMSOCVLMKNVSRP------
016720 CAEEL/566-867
                                    LIVVVLPG--KTPI-YAEVKRVGDTVLGIATOCVOAKNAIRT------
062275 CAEEL/594-924
                                    TFVFIITD-DSITT-LHQRYKMIEKDTKMIVQDMKLSKALSV--IN---A
Y053 CAEEL/650-977
                                    DILVGIAR-EKKPD-VHDILKYFEESIGLOTIOLCOOTVDKMMGG----O
NRDE3 CAEEL/673-1001
                                    TIVFGIIA-EKRPD-MHDILKYFEEKLGOOTIQISSETADKFMRD----H
017567 CAEEL/397-708
                                    MLVVMLAD-DNKTR-YDSLKKYLCVECPIPNQCVNLRTLAGKSKDGGENK
AUB DROME/555-852
                                    IVMVVMRS-PNEEK-YSCIKKRTCVDRPVPSQVVTLKVIAPRQQKP---T
PIWI DROME/538-829
                                    LILCLVPN-DNAER-YSSIKKRGYVDRAVPTOVVTLKTTKNRSL-----
PIWL1 HUMAN/555-847
                                    IVVCLLSS-NRKDK-YDAIKKYLCTDCPTPSQCVVARTLGKQQT-----
PIWI ARCFU/110-406
                                    GIMLVLPE-YNTPL-YYKLKSYLINS--IPSOFMRYDILSNRNL-----
Y1321 METJA/426-699
                                    CFALIIGKEKYKDNDYYEILKKQLFDLKIISQNILWENWRKDDK-----
                                    LVIVFLEEYPKVDP-YKSFLLYDFVKRELLKKMIPSQVILNRTLKN---E
067434 AQUAE/419-694
AG010_ARATH/625-946
                                    LLLAILPD-NNGSL-YGDLKRICETELGLISOCCLTKHVFKI------
AG01 SCHP0/500-799
                                    -KPOYCANLGMKINVKVGGIN-CSLIPKSNP----LGNVPTL-----
```

Manipulating/writing MSA

It is possible to slice alignments using the operator applied on a SeqRecord.

Think about it as a matrix

```
1. SeqRecord[i,j] returns the jth character of alignment i as a string;
```

- 2. SeqRecord[:,j] returns all the jth characters of the multiple alignment as a string;
- SeqRecord[:,i:j] returns a MultipleSeqAlignment with the sub-alignments going for i to j
 (excluded)
- 4. SeqRecord[a:b,i:j] similar to 3. but for alignments going from a to b (excluded) only

```
YLFFILDK-NSPEP-YGSIKLVPPVYYAHLVSNLARYODV
FILCILPERKTSDI-YGPWKIVAPVRYAHLAAAOVAOFTK
FILCVLPDKKNSDL-YGPWKVVAPICYAHLAAAOLGTFMK
CIIVVLOS-KNSDI-YMTVKIPTPVYYADLVATRARCHVK
LIVVVLPG--KTPI-YAEVKIPAPAYYAHLVAFRARYHLV
TEVELITD-DSITT-LHORYLPTPLYVANEYAKRGRNLWN
DILVGIAR-EKKPD-VHDILVPDVLYAAENLAKRGRNNYK
TIVFGIIA-EKRPD-MHDILIPNVSYAAONLAKRGHNNYK
MLVVMLAD-DNKTR-YDSLKVPAPCOYAHKLAFLTAOSLH
IVMVVMRS-PNEEK-YSCIKVPAVCHYAHKLAFLVAESIN
LILCLVPN-DNAER-YSSIKVPAVCOYAKKLATLVGTNLH
IVVCLLSS-NRKDK-YDAIKVPAPCOYAHKLAFLVGOSIH
GIMLVLPE-YNTPL-YYKLKLPVTVNYPKLVAGIIANVNR
CFALIIGKEKYKDNDYYEILIPAPIHYADKFVKALGKNWK
LVIVFLEEYPKVDP-YKSFLLPATVHYSDKITKLMLRGIE
LLLAILPD-NNGSL-YGDLKIVPPAYYAHLAAFRARFYLE
```

```
align[0,0] is Y
align[2,1] is I
align[:,0] is YFFCLTDTMILIGCLL

align[:,0:3] gets first 3 rows (SeqRecords)
YLFFILDK-N...
FILCILPERK...
FILCVLPDK...

align[0:3,0:3] first 3 cols of first 3 rows (SeqRecords):
YLF
FIL
FIL
```

Pairwise alignment

Biopython has its own module to make pairwise alignment. It provides two algorithms: <u>Smith-Waterman</u> for local alignment and <u>Needleman-Wunsch</u> for global alignment. These methods are implemented in two Biopython functions of the <u>Bio.pairwise2</u> module:

```
pairwise2.align.globalxx()
pairwise2.align.localxx()
```

```
aligns = pairwise2.align.globalxx(seq1,seq2)
aligns = pairwise2.align.localxx(seq1,seq2)
```

where seq1 and seq2 are two str objects. These methods return a list of alignments (at least one) that have the same **optimal score**. Each alignment is represented as tuples with the following 5 elements in order:

- 1. The alignment of the first sequence;
- 2. The alignment of the second sequence;
- The alignment score;
- 4. The start of the alignment (for global alignments this is always 0);
- 5. The end of the alignment (for global alignments this is always the length of the alignment).

```
Example:
alignments = pairwise2.align.globalxx("ACCGTTATATAGGCCA", "ACGTACTAGTATAGGCCA")
for i in range(len(alignments)):
    print(alignments[i])

('ACCGT--TA-TATAGGCCA', 'A-CGTACTAGTATAGGCCA', 15.0, 0, 19)
('ACCGT--TA-TATAGGCCA', 'AC-GTACTAGTATAGGCCA', 15.0, 0, 19)
```

Pairwise alignment

OPTIONS FOR MATCHES/MISMATCHES AND GAP OPENS/EXTENSIONS

```
pairwise2.align.globalxx pairwise2.align.globalmx pairwise2.align.globalms pairwise2.align.globalmd pairwise2.align.globalxd pairwise2.align.globalxx pairwise2.align.localxx pairwise2.align.localmx pairwise2.align.localms pairwise2.align.localmd pairwise2.align.localxd pairwise2.align.localxs
```

Match parameters can be:

- x : means that a match scores 1 a mismatch 0;
- m: the match and mismatch score are passed as additional params after the sequence (es. aligns = pairwise2.align.globalmx(seq1,seq2, 1, -1) to set 1 as match score and -1 as mismatch penalty.

Gap parameters can be:

- x : gap penalty is 0;
- s : same gap open and gap extend penalties for the 2 sequences (passed as additional paramatter seqs).
- d: different gap open and gap extend penalties for the 2 seqs (additional params after the seqs).

The first letter is the score for a match the second letter is the penalty for a gap

Pairwise alignment

```
('ACCGT--TA-TATAGGCCA', 'A-CGTACTAGTATAGGCCA', 15.0, 0, 19)
('ACCGT--TA-TATAGGCCA', 'AC-GTACTAGTATAGGCCA', 15.0, 0, 19)
Looping through aligns
ACCGT -- TA-TATAGGCCA
A - CGTACTAGTATAGGCCA
Score: 15.0, Start: 0, End: 19
ACCGT - - TA - TATAGGCCA
AC-GTACTAGTATAGGCCA
Score: 15.0, Start: 0, End: 19
Match: 1, Mismatch: -1, Gap open: -0.5, Gap extend: -0.2
ACCGT - - TA - TATAGGCCA
A-CGTACTAGTATAGGCCA
Score: 13.3, Start: 0, End: 19
ACCGT -- TA-TATAGGCCA
AC-GTACTAGTATAGGCCA
Score: 13.3, Start: 0, End: 19
```

```
from Bio import pairwise2
from Bio import SeqIO
alignments = pairwise2.align.globalxx("ACCGTTATATAGGCCA",
                                       "ACGTACTAGTATAGGCCA")
for i in range(len(alignments)):
    print(alignments[i])
print("")
print("Looping through aligns")
for align in alignments:
        print(align[0])
        print(align[1])
        print("Score: {}, Start: {}, End: {}".format(align[2],
                                                      align[3],
                                                      align[4]))
        print("")
alignments = pairwise2.align.globalms("ACCGTTATATAGGCCA",
                                       "ACGTACTAGTATAGGCCA".
                                       1.-1.-0.5.-0.2
print("")
print("Match: 1, Mismatch: -1, Gap open: -0.5, Gap extend: -0.2")
for align in alignments:
    print(align[θ])
    print(align[1])
    print("Score: {}, Start: {}, End: {}".format(align[2],
                                                  align[3],
                                                  align[4]))
    print("")
```

http://biopython.org



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Biopython

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Introduction

Biopython is a set of freely available tools for biological computation written in Python by an international team of developers.

It is a distributed collaborative effort to develop Python libraries and applications which address the needs of current and future work in bioinformatics. The source code is made available under the Biopython License, which is extremely liberal and compatible with almost every license in the world.

We are a member project of the Open Bioinformatics Foundation (OBF), who take care of our domain name and hosting for our mailing list etc. The OBF used to host our developement repository, issue tracker and website but these are now on GitHub.

This wiki will help you download and install Biopython, and start using the libraries and tools.

Get Started	Get help	Contribute
Download Biopython	Tutorial (PDF)	What's being worked on
Installation help (PDF)	Documentation on this wiki	Developing on Github
	Cookbook (working examples)	Google Summer of Code
	Discuss and ask questions	Report bugs (older issues)

The latest release is Biopython 1.70, released on 10 July 2017.

http://biopython.org/DIST/docs/api/

Check:

Seq SegRecord MultipleSegAlignment

Table of Contents Everything Modules Bio.Affv Bio.Affy.CelFile Bio.Align Bio.Align.AlignInfo Bio Align Applications Bio, Align, Applications, Clustal Omega Bio.Align.Applications. Clustalw Bio.Align.Applications. Dialign Bio.Align.Applications. MSAProbs Bio.Align.Applications. Mafft Everything All Classes Bio.Affv.CelFile.ParserError Bio.Affv.CelFile.Record Bio.Align.AlignInfo.PSSM Bio.Align.AlignInfo.SummaryInfo Bio.Align.Applications. ClustalOmega.ClustalOmegaCom Bio.Align.Applications. Clustalw.ClustalwCommandline Bio.Align.Applications. Dialign.DialignCommandline Bio.Align.Applications. MSAProbs.MSAProbsCommandl Bio.Align.Applications. Mafft.MafftCommandline Bio.Align.Applications. Muscle.MuscleCommandline Bio.Align.Applications. Prank.PrankCommandline Bio.Align.Applications. Probcons.ProbconsCommandline Bio.Align.Applications. TCoffee.TCoffeeCommandline Bio.Align.MultipleSeqAlignment Bio.AlignIO.ClustalIO.ClustalIterator Bio.AlignIO.ClustalIO.ClustalWriter Bio.AlignIO.EmbossIO.EmbossIterator Bio.AlignIO.EmbossIO.EmbossWriter Bio.AlignIO.Interfaces.AlignmentIterator Bio.AlignIO.Interfaces.AlignmentWriter Bio.AlignIO.Interfaces.SequentialAlignmentWriter Bio AlignIO MafIO MafIndex Bio AlignIO MafIO MafWriter Bio.AlignIO.MauveIO.MauveIterator Bio.AlignIO.MauveIO.MauveWriter Bio.AlignIO.NexusIO.NexusWriter Bio.AlignIO.PhylipIO.PhylipIterator Bio.AlignIO.PhylipIO.PhylipWriter Bio.AlignIO.PhylipIO.RelaxedPhylipIterator Bio.AlignIO.PhylipIO.RelaxedPhylipWriter Bio.AlignIO.PhylipIO.SequentialPhylipIterator Bio.AlignIO.PhylipIO.SequentialPhylipWriter

Trees Indices Help Module Hierarchy . Bio: Collection of modules for dealing with biological data in Python. o Bio.Affy: Deal with Affymetrix related data such as cel files. Bio.Align: Code for dealing with sequence alignments. Bio.AlignIO: Multiple sequence alignment input/output as alignment objects. Bio.AlignIO.Interfaces: AlignIO support module (not for general use). Bio.AlignIO.MafIO: Bio.AlignIO support for the "maf" multiple alignment format. o Bio.Alphabet: Alphabets used in Seq objects etc to declare sequence type and letters. o Bio.Blast: Code for dealing with BLAST programs and output. Bio.Blast.ParseBlastTable: A parser for the NCBI blastpap version 2.2.5 output format. Currently only supports the '-m 9' option. (table w/ annotations). Returns a BlastTableRec instance Bio.CAPS: Cleaved amplified polymorphic sequence (CAPS) markers. o Bio.Cluster: Cluster Analysis. Bio.Cluster.cluster: C Clustering Library Bio.Compass: Code to deal with COMPASS output, a program for profile/profile comparison. Bio.Data: Collections of various bits of useful biological data.

[Module Hierarchy | Class Hierarchy]

- - Bio.Affv.CelFile: Reading information from Affymetrix CEL files version 3 and 4.
 - - Bio.Align.AlignInfo: Extract information from alignment objects.
 - Bio.Align.Applications: Alignment command line tool wrappers.
 - Bio.Align.Applications, ClustalOmega: Command line wrapper for the multiple alignment program Clustal Omega.
 - Bio.Align.Applications. Clustalw: Command line wrapper for the multiple alignment program Clustal W.
 - Bio.Align.Applications. Dialign: Command line wrapper for the multiple alignment program DIALIGN2-2.
 - Bio.Align.Applications. MSAProbs: Command line wrapper for the multiple sequence alignment program MSAProbs.

 - Bio.Align.Applications. Mafft: Command line wrapper for the multiple alignment programme MAFFT.
 - Bio.Align.Applications. Muscle: Command line wrapper for the multiple alianment program MUSCLE.
 - Bio.Align.Applications. Prank: Command line wrapper for the multiple alignment program PRANK.
 - Bio.Align.Applications. Probcons: Command line wrapper for the multiple alignment program PROBCONS.
 - Bio.Align.Applications, TCoffee: Command line wrapper for the multiple alignment program TCOFFEE.
 - - Bio.AlignIO.ClustalIO: Bio.AlignIO support for "clustal" output from CLUSTAL W and other tools.
 - Bio.AlignIO.EmbossIO: Bio.AlignIO support for "emboss" alignment output from EMBOSS tools.
 - Bio.AlignIO.FastaIO: Bio.AlignIO support for "fasta-m10" output from Bill Pearson's FASTA tools.

 - Bio.AlignIO.MauveIO: Bio.AlignIO support for "xmfa" output from Mauve/ProgressiveMauve.
 - Bio.AlignIO.NexusIO: Bio.AlignIO support for the "nexus" file format.
 - Bio.AlignIO.PhylipIO: AlignIO support for "phylip" format from Joe Felsenstein's PHYLIP tools.
 - Bio.AlignIO.StockholmIO: Bio.AlignIO support for "stockholm" format (used in the PFAM database).
 - - Bio.Alphabet.IUPAC: Standard nucleotide and protein alphabets defined by IUPAC.
 - Bio.Alphabet.Reduced: Reduced alphabets which lump together several amino-acids into one letter.
 - Bio.Application: General mechanisms to access applications in Biopython.

 - Bio.Blast.Applications: Definitions for interacting with BLAST related applications.
 - Bio.Blast.NCBIStandalone: Code for calling standalone BLAST and parsing plain text output (DEPRECATED).
 - Bio.Blast.NCBIWWW: Code to invoke the NCBI BLAST server over the internet.
 - Bio.Blast.NCBIXML: Code to work with the BLAST XML output.

 - . Bio.Blast.Record: Record classes to hold BLAST output.

 - o Bio.Crystal: Represent the NDB Atlas structure (a minimal subset of PDB format).

 - - Bio.Data.CodonTable: Codon tables based on those from the NCBI.
 - Bio.Data.IUPACData: Information about the IUPAC alphabets.
 - Bio.Data.SCOPData: Additional protein alphabets used in the SCOP database and PDB files.
 - Bio.DocSQL: Bio.DocSQL: easy access to DB API databases (DEPRECATED).

Installing biopython

```
In windows installing Biopython should be as easy as opening the command prompt as admininstrator (typing cmd and then right clicking on the link choosing run as admininstrator) and then pip3 install biopython.

In linux sudo pip3 install biopython will install biopython for python3 up to python3.5. On python 3.6, the command is: python3.6 -m pip install biopython.
```

http://qcbsciprolab2019.readthedocs.io/en/latest/practical10.html

Exercises

- Write a python function that reads a genebank file given in input and prints off the following information:
 - 1. Identifier, name and description;
 - 2. The first 100 characters of the sequence;
 - 3. Number of external references (dbxrefs) and ids of the external refs.
 - 4. The name of the organism (hint: check the annotations dictionary at the key "organism")
 - Retrieve and print all (if any) associated publications (hint: annotation dictionary, key:"references")
 - Retrieve and print all the locations of "CDS" features of the sequence (hint: check the features)

Hint: go back and check the details of the SeqRecord object.

Test the program downloading some files from genebank like this

Show/Hide Solution

- 2. Write a python program that loads a pfam file (stockholm format .sth) and reports for each record of the alignment:
 - 1. the id of the entry
 - 2. the start and end points
 - 3. the number of gaps and the % of gaps on the total length of the alignment
 - 4. the number of external database references (dbxrefs), and the first 3 external references comma separated (hint: use join).

Print these information to the screen. Finally, write this information in a tab separated file (.tsv) having the following format: #ID\tstart\tend\tnum gaps\tpercentage gaps\tdbxrefs .