# Comp 555 - BioAlgorithms - Spring 2020



PROBLEM SET #1 15 ON-LINE AND DUE ON 2/4/2020

Finding Hidden Patterns in DNA

#### **Initiating Transcription**

- As a precursor to transcription (the reading of DNA to construct RNAs, that eventually leading to protein synthesis) special proteins bind to the DNA, and separate it to enable its reading.
- How do these proteins know where the coding genes are in order to bind?
- Genes are relatively rare
- O(1,000,000,000) bases/genome
- O(10000) genes/genome
- O(1000) bases/gene
- Approximately 1% of DNA codes for genes (10<sup>3</sup>10<sup>4</sup>/10<sup>9</sup>)









#### **Regulatory Regions**

- RNA polymerases seek out regulatory or promoting regions located 100-1000 bp upstream from the coding region
- They work in conjunction with special proteins called transcription factors (TFs) whose presence enables gene expression
- Within these regions are the Transcription Factor Binding Sites (TFBS), special DNA sequence patterns known as motifs that are specific to a given transcription factor
- A Single TF can influence the expression of many genes. Through biological experiments one can infer, at least a subset of these affected genes.



#### **Transcription Factor Binding Sites**



- A TFBS can be located anywhere within the regulatory region.
- TFBS may vary slightly across different regulatory regions since non-essential bases could mutate
- Transcription factors are robust (they will still bind) in the presence of small sequence differences by a few bases



#### **Identifying Motifs: Complications**

- We don't know the motif sequence for every TF
- We don't know where it is located relative to a gene's start
- Moreover, motifs can differ slightly from one gene to the next
- We only know that it occurs somewhere near genes that share a TF
- How to discern a Motif's frequent similar pattern from random patterns?
- How is this problem different that finding frequent k-mers from Lecture 2?





## Let's look for an Easy Motif



1 tagtggtcttttgagtgtagatccgaagggaaagtatttccaccagttcggggtcacccagcagggcagggtgacttaat
2 cgcgactcggcgctcacagttatcgcacgtttagaccaaaacggagttagatccgaaactggagttaatcggagtcat
3 gttacttgtgagcctggttagatccgaaatataattgttggctgcatagcggagctgacatacgagtaggggaaatgcgt
4 aacatcaggctttgattaaacaatttaagcacgtagatccgaattgacctgatgacaatacggaacatgccggctccggg
5 accaccggataggctgcttattagatccgaaaggtagtatcgtaataatggctcagccatgtcaatgtgcggcattccac
6 tagatccgaatcgtgtttctccctctgtgggttaacgaggggtggaccttgtgccgaacttgtaccc
7 gaaatggttcggtgcgatatcaggccgttctcttaacttggcggtgtagatccgaacgtctctggaggggtcgtgcgcat
8 atgtatactagacattctaacgctcgcttattggcggagaccattgcccatacaagagggtcactgtgtagatccgaa
9 ttcttacacccttctttagatccgaacctgttggcgccatcttctttcgagtccttaccattgccgaaattg
10 ctacctatgtaaaacaacatctactaacgtagtccggtctttcctgatcgcctaacctaccaggtagatccgaaattcg

Problem: Given M sequences of length N find any k-mer that appears in each sequence.

How would you go about finding a 10-mer that appears in every one of these 10 strings?

#### Sneak Peek at the Answer



1 tagtggtcttttgagtgTAGATCCGAAgggaaagtatttccaccagttcggggtcacccagcagggcagggtgacttaat 2 cgcgactcggcgctcacagttatcgcacgtttagaccaaaacggagtTAGATCCGAAactggagtttaatcggagtcctt 3 gttacttgtgagcctggtTAGATCCGAAatataattgttggctgcatagcggagctgacatacgagtaggggaaatgcgt 4 aacatcaggctttgattaaacaatttaagcacgTAGATCCGAAttgacctgatgacaatacggaacatgccggctccggg 5 accaccggataggctgcttatTAGATCCGAAaggtagtatcgtaataatggctcagccatgtcaatgtgcggcattccac 6 TAGATCCGAAtcgatcgtgtttctccctctgtgggttaacgaggggtcgaccttgctcgcatgtgccgaacttgtaccc 7 gaaatggttcggtgcgatatcaggccgttctcttaacttggcggtgTAGATCCGAAcgtctctggaggggtcgtgcgta 8 atgtatactagacattctaacgctcgcttattggcggagaccattgcccactacaagagggtcactgtgTAGATCCGAA 9 ttcttacacccttcttTAGATCCGAAcctgttggcgccatcttctttcgagtccttgtacctcattgctctgatgac

 $10\ ctacctatgtaaaacaacatctactaacgtagtccggtctttcctgatctgccctaacctacaggTAGATCCGAAattcg$ 

Now that you've seen the answer, how would you find it?

#### Comp 555 - Fall 2020

#### Meet Mr Brute Force

- He's often the best starting point when approaching a problem
- He'll also serve as a straw-man when designing new approaches
- Though he's seldom elegant, he gets the job done
- Often, we can't afford to wait for him

For our current problem a brute force solution would consider every k-mer position in all strings and see if they match. Given M sequences of length N, there are:

(N−k+1)<sup>M</sup>

position combinations to consider.

How do you write M nested loops when M is a variable?





#### A Library of Helper Functions



- There's a tendancy to approach this problem with a series of nested for-loops, while the approach is valid, it doesn't generalize. It assumes a specific number of sequences.
- What we need is an iterator that generates all permutations of a sequence.
- This nested-for-loop iterator is called a Cartesian Product over sets.
- Python has a library to accomplish this

#### **Using itertools**



#### itertools: 3 loops over 2 things

#### 

#### itertools: 2 loops over 3 things

In [4]:	1 2	<pre>for number in itertools.product(range(3), repeat=2):     print(number)</pre>
	(0,	
	(0,	
	(1, (1, (1, (1, (1, (1, (1, (1, (1, (1,	
	(1, (2,	
	(2, (2,	.) 2)

#### Permutations of mixed types



In [14]:	<pre>for section in itertools.product(("I", "II", "III", "IV"), "ABC", range(1,3)):     print(section)</pre>
	('I', 'A', 1) ('I', 'A', 2)
	('I', 'B', 1) ('I', 'B', 2)
	('I', 'C', I) ('I', 'C', 2) ('II', 'A', 1)
	('II', 'A', 2) ('II', 'B', 1)
	('II', 'B', 2) ('II', 'C', 1)
	('II', 'C', 2) ('III', 'A', 1) ('III', 'A', 2)
	('III', 'B', 1) ('III', 'B', 2)
	('III', 'C', 1) ('III', 'C', 2) ('IV', 'A', 1)
	('IV', 'A', 2) ('IV', 'B', 1)
	('IV', 'B', 2) ('IV', 'C', 1)

#### **Bruteforce Exact Search**



```
In [15]:
         sequences = [
             'tagtggtcttttgagtgtagatccgaagggaaagtatttccaccagttcggggtcaccccagcagggcagggtgacttaat',
              'cgcgactcggcgctcacagttatcgcacgtttagaccaaaacggagttagatccgaaactggagtttaatcggagtcctt',
              'gttacttgtgagcctggttagatccgaaatataattgttggctgcatagcggagctgacatacgagtaggggaaatgcgt',
              'aacatcaggctttgattaaacaatttaagcacgtagatccgaattgacctgatgacaatacggaacatgccggctccggg',
              'accaccggataggctgcttattagatccgaaaggtagtatcgtaataatggctcagccatgtcaatgtgcggcattccac',
              'tagateegaategategtgttteteeetetgtgggttaaegaggggteegaeettgetegeatgtgeegaaettgtaeee',
              'gaaatggttcggtgcgatatcaggccgttctcttaacttggcggtgtagatccgaacgtctctggaggggtcgtgcgcta',
              'atgtatactagacattctaacgctcgcttattggcggagaccatttgctccactacaagaggctactgtgtagatccgaa',
             'ttcttacacccttctttagatccgaacctgttggcgccatcttcttttcgagtccttgtacctccatttgctctgatgac',
              'ctacctatgtaaaacaacatctactaacgtagtccggtctttcctgatctgccctaacctacaggtagatccgaaattcg']
         def bruteForce(dna,k):
             """Finds a *k*-mer common to all sequences from a
                list of *dna* fragments with the same length"""
             M = len(dna) # how many sequences
             N = len(dna[0]) # length of sequences
             for offset in itertools.product(range(N-k+1), repeat=M):
                 for i in range(1,len(offset)):
                     if dna[0][offset[0]:offset[0]+k] != dna[i][offset[i]:offset[i]+k]:
                         break
                 else:
                     return offset, dna[0][offset[0]:offset[0]+10]
```

#### Now let's Test and Time it



```
In [16]: M = 4
position, motif = bruteForce(sequences[0:M], 10)
print(position, motif, '\n')
for i in range(M):
    p = position[i]
    print(sequences[i][:p]+sequences[i][p:p+10].upper()+sequences[i][p+10:])
print()
%timeit bruteForce(sequences[0:M], 10)
# you can try a larger value of M, but be prepared to wait
```

(17, 47, 18, 33) tagatccgaa

tagtggtcttttgagtgTAGATCCGAAgggaaagtatttccaccagttcggggtcacccagcagggcagggtgacttaatcgcgactcggcgctcacagttatcgcacgtttagaccaaaacggagtTAGATCCGAAactggagtttaatcggagtccttgttacttgtgggcctggtTAGATCCGAAatataattgttggctgcatagcggagctgacatacgggggaaatgcgtaacatcaggctttgattaaacaatttaagcacgTAGATCCGAAttgacctgatgacaatacggaacatgccggg

6.25 s ± 143 ms per loop (mean ± std. dev. of 7 runs, 1 loop each)



Now let's consider a more realistic motif finding problem, where the binding sites do not need to match exactly.

Actually, none of the sequences have an unmodified copy of the original motif

# **Profile and Consensus**

How to find approximate string matches?

• Align candidate motifs by their start indexes

 $s = (s_1, s_2, ..., s_t)$ 

- Construct a matrix profile with the frequencies of each nucleotide in columns
- Consensus nucleotide in each position has the highest score in each column

```
a G g t a c T t
           C c A t a c g t
Alignment
            acgtTAgt
            a c g t C c A t
            CcgtacgG
                 0 3 1 1 0
Profile
           2 4 0 0 1
           0 1 4 0 0 0
                      3 1
         G
         т
           0 0 0 5 1 0 1 4
           ACGTACGT
Consensus
```



#### Consensus



- One can think of the consensus as an *ancestor* motif, from which mutated motifs emerged
- The distance between an actual motif and the consensus sequence is generally less than that for any two actual motifs
- Hamming distance is number of positions that differ between two strings



#### **Consensus Properties**



• A consensus string has a minimal hamming distance to all its source strings



#### **Scoring Motifs**



• Given  $s = (s_1, s_2, \dots, s_t)$  and DNA

$$Score(s, DNA) = \sum_{i=1}^{k} \max_{j \in A, C, G, T} count(j, i)$$

- So our approach is back to brute force
  - We consider every candidate motif in every string
  - Return the set of indices with the highest score

#### Let's try again allowing for errors



```
In [17]:
         def Score(s, DNA, k):
              .....
                  compute the consensus SCORE of a given k-mer alignment given
                  offsets into each DNA string. s = list of starting indices.
                  DNA = list of nucleotide strings. k = Target Motif length
              0.0.0
              score = 0
             for i in range(k):
                  # loop over string positions
                  cnt = dict(zip("acgt", (0, 0, 0, 0)))
                  for j, sval in enumerate(s):
                      base = DNA[j][sval+i]
                      cnt[base] += 1
                  score += max(cnt.values())
              return score
         def BruteForceMotifSearch(dna,k):
             M = len(dna)  # how many sequences
             N = len(dna[0]) # length of sequences
             bestScore = 0
              bestAlignment = []
             for offset in itertools.product(range(N-k+1), repeat=M):
                  s = Score(offset, dna, k)
                  if (s > bestScore):
                      bestAlignment = [p for p in offset]
                      bestScore = s
              print(bestAlignment, bestScore)
```

#### Test and time this one



47.4  $\mu$ s ± 5.52  $\mu$ s per loop (mean ± std. dev. of 7 runs, 10000 loops each) [17, 47, 18, 33] 36 CPU times: user 12min 57s, sys: 50.2 ms, total: 12min 57s Wall time: 12min 57s

# Running Time of BruteForceMotifSearch

- Search (N k + 1) positions in each of M sequences, by examining  $(N - k + 1)^M$  sets of starting positions
- For each set of starting positions, the scoring function makes *O*(*Mk*) operations, so the complexity is:

 $Mk(N-k+1)^M = O(MkN^M)$ 

- That means that for M = 10, N = 80, k = 10 we must perform approximately  $10^{21}$  computations
- Generously assuming  $10^9$  comps/sec it will require only  $10^{12}$  secs

$$\frac{10^{12}}{(60 \times 60 \times 24 \times 365)} > 30000 \text{ years}$$

• Want to wait?







# How conservative is this estimate?

- For the example we just did M = 4, N = 80, k = 10
- So that gives  $\approx 4.0 \times 10^9$  operations
- Using our 10<sup>9</sup> operations per second estimate, it should have taken only 4 secs.
- Instead it took closer to 700 secs, which suggests we are getting around 5.85 million operations per second.
- So, in reality it will even take longer!



# How can we find Motifs in our lifetime?



- Should we give up on Python and write in C? Assembly Language?
- Will biological insights save us this time?
- Are there other ways to find Motifs?
- Consider that if you knew what motif you were looking for, it would take only

k(N-k+1)M=O(kNM)

to find its indices in each string.

• Is that significantly better?

