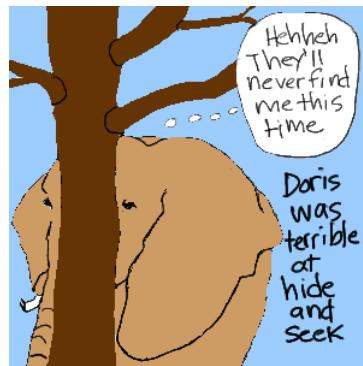


# Finding Hidden Patterns in DNA

- What makes searching for frequent subsequences hard?
  - Allowing for errors?
  - All the places they could be hiding?



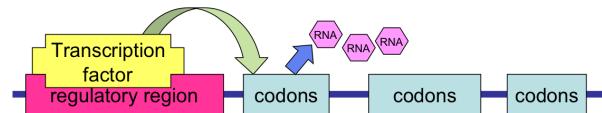
# Initiating Transcription

- 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 ( $10^3 10^4 / 10^9$ )



# 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 than finding frequent k-mers from Lecture 2?



# Let's look for an *Easy Motif*

```
1 tagtgtctttgagttagatccgaaggaaagtattccaccagttcgggtcacccagcagggcagggtgacttaat  
2 cgcgactcggcgctcacagttatcgacgcttagacaaaacggagtttagatccgaaactggagttaatcgagtcctt  
3 gttacttgtgagcctggtagatccgaaatataattttggctgcatacgaggactgacatacgagtagggaaatgcgt  
4 aacatcaggcttgattaaacaatttaaagcacgttagatccgaatttgacctgatgacaatacggaaacatgccggctccgg  
5 accaccggataggctgctttagatccgaaaggtagtatcgtaataatggctcagccatgtcaatgtgcggcatccac  
6 tagatccgaatcgatcgtttctccctgtggtaacgagggtccgaccttgcgcattgtgcgaacttgttaccc  
7 gaaatggttcggtgcgatattcggccgtctctaacttggcggttagatccgaaacgtctctggaggggtcgtcgcta  
8 atgtatactagacattctaacgctcgctttagatccgaaacctgttgcgcacatcttgcgttgcactgttagatccgaa  
9 ttcttacaccctcttagatccgaacctgttgcgcacatcttgcgttgcactgttagatccgaaattcg  
10 ctacctatgtaaaacaacatctaactaacgttagtccggctttcgatctgccctaacctacaggttagatccgaaattcg
```

**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 strings?

# Sneak Peek at the Answer

```
1 tagtgtctttgagtTAGATCCGAAgggaaagtattccaccagttcggggtcacccagcagggcagggtgacttaat  
2 cgcgactcggcgctcacagtatcgacgtagacaaaacggagtTAGATCCGAAactggagttaatcgagtcctt  
3 gttacttgtgagcctggTAGATCCGAAatataatttgtggctgcatagcggagctgacatacggatggggaaatgcgt  
4 aacatcaggcttgattaaacaatttaagcacgTAGATCCGAAttgacctgatgacaatacggAACatgcggctccgg  
5 accaccggataggctgcttatTAGATCCGAAaggtagtatcgtaataatggctcggcatgtcaatgtgcggcatccac  
6 TAGATCCGAAtcgatcggtttctccctctgtgggttaacgggggtccgaccctgctcgatgtgcggacttgtaccc  
7 gaaatggttcggtgcgatatacgccgtctcttaacttggcggtTAGATCCGAAcgtctctggaggggtcgctgccta  
8 atgtatactagacattctaacgctcgcttattggcgagaccatttgcctccactacaaggaggctactgtTAGATCCGAA  
9 ttcttacaccctcttTAGATCCGAAccctgttggcgccatctttcgactcctgtacctccatttgcctgtatgac  
10 ctacctatgtaaaacaacatctaactaactgtccggcttgcgtatctgcctaacctacaggTAGATCCGAAattcg
```

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

# 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:

$$(N - k + 1)^M$$

position combinations to consider.

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

# A Library of Helper Functions

- There's a tendency 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*

```
import itertools

for number in itertools.product("01", repeat=3):
    print ''.join(number)
```

```
000
001
010
011
100
101
110
111
```

# All permutations of items from a list

```
N = 0
for number in itertools.product(range(3), repeat=3):
    print number,
    N += 1
    if (N % 5 == 0):
        print
```

```
(0, 0, 0) (0, 0, 1) (0, 0, 2) (0, 1, 0) (0, 1, 1)
(0, 1, 2) (0, 2, 0) (0, 2, 1) (0, 2, 2) (1, 0, 0)
(1, 0, 1) (1, 0, 2) (1, 1, 0) (1, 1, 1) (1, 1, 2)
(1, 2, 0) (1, 2, 1) (1, 2, 2) (2, 0, 0) (2, 0, 1)
(2, 0, 2) (2, 1, 0) (2, 1, 1) (2, 1, 2) (2, 2, 0)
(2, 2, 1) (2, 2, 2)
```

# Permutations of mixed types

```
for section in itertools.product(("I", "II", "III", "IV"), "ABC", range(1,3)):  
    print section
```

```
('I', 'A', 1)  
('I', 'A', 2)  
('I', 'B', 1)  
('I', 'B', 2)  
('I', 'C', 1)  
('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)  
('IV', 'C', 2)
```

# Now let's try some *Brute Force* code

```
sequences = [
    'tagtggctttgagtgttagatccgaaggaaagtattccaccagttcgggtcacccagcagggcagggtaactaat',
    'cgcgactcgccgctcacagttatcgacgttagacaaaacggagttagatccaaactggagttaatcgagtcct',
    'gttacttgtgagcctggtagatccgaaataattgtggctgcatacgaggactgacatacgagtagggaaatgcgt',
    'aacatcaggcttattaaacaatttaagcactgatccgaaattgacctgatgacaatacggAACATGCCGCTCCGG',
    'accaccggataggctgttattagatccgaaaggtagtatcgatccgaaatggctcagccatgtcaatgtggcattcac',
    'tagatccgaaatcgatcggtttccctctgtggtaacggggccgacctgctcgatgtccgaaacttgttaccc',
    'gaaatgggtcggtcgatcatcggtttccctctgtggtaactggcggttagatccgaaacgtctgggggtcgctgccta',
    'atgtataactagacatttaacgctcgatccggtttccactacaagggctactgttagatccgaa',
    'ttcttacaccctttagatccgaaacctgttggccatcttgcagtccctgtacccatccatgtctgatgac',
    'ctacatatgtaaaacaacatctactaacgttagtccggctttcctgatccgaaacctacaggtagatccgaaattcg']

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 xrange(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]
```

# Test and then time it

```
M = 4
position, motif = bruteForce(sequences[0:M], 10)
print position, motif
print
for i in xrange(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

tagtgtttgagtgTAGATCCGAAggaaagtattccaccagttcgggtcaccacggcagggtgacttaat
cgcaactcgccgtcacagttatcgcacgttagacaaaacggagtTAGATCCGAAactggagttaatcgagtcctt
gttacttgtgagcctggTTAGATCCGAatataatttgtggctgcatacgggagctgacatacgagtagggaaatgcgt
aacatcaggcttattaaacaatcgacgTAGATCCGAAttgacctgtatgacaatacggaacatgccggctccggg

1 loop, best of 3: 4.74 s per loop
```

# Approximate Matching

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

```
1 taggtctttgagtTAGATCTGAAgggaaagtattccaccagttcggggtcacccagcaggcagggtgacttaat  
2 cgcgactcggcgcacagttatcgacgttagacaaaacggagtTGGATCCGAAactggagttaatcgagtcctt  
3 gttacttgtgagcctggTAGACCCGAAatataatttgtggctgcatagcggagctgacatacggatggggaaatgcgt  
4 aacatcaggcttgattaaacaatttaagcacgTAAATCCGAAttgacctgtatgacaatacggAACatgcggctccgg  
5 accaccggataggctgcttatTAGGTCCAAAaggttagtacgtataatggctcagccatgtcaatgtgcggcattccac  
6 TAGATTCGAAtcgatcggtttctccctctgtgggttaacgagggtccgacctgctcgcatgtgcgaacttgtaccc  
7 gaaatggttcggtgcgatatcaggccgtctcttaacttggcggtgCAGATCCGAAcgtctctggaggggtcgtgcgcta  
8 atgtatactagacattctaacgcgtcgctatttggcgagaccatttgccttactacaaggaggctactgtgTAGATCCGTA  
9 ttcttacaccctcttTAGATCCAAAcctgttggcgccatttttcgagtccttgcatttgcgtatgac  
10 ctacatatgaaaaacaacatctactaacgttagtccggctttcgtatctggccataacctacaggTCGATCCGAAattcg
```

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  
 $s = (s_1, s_2, \dots, s_t)$
- Construct a matrix profile with the frequencies of each nucleotide in columns
- Consensus nucleotide in each position has the highest score in each column

Alignment	a G g t a c T t C c A t a c g t a c g t T A g t a c g t C c A t C c g t a c g G
Profile	A 3 0 1 0 3 1 1 0 C 2 4 0 0 1 4 0 0 G 0 1 4 0 0 0 3 1 T 0 0 0 5 1 0 1 4
Consensus	A C G T A C G T

# 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

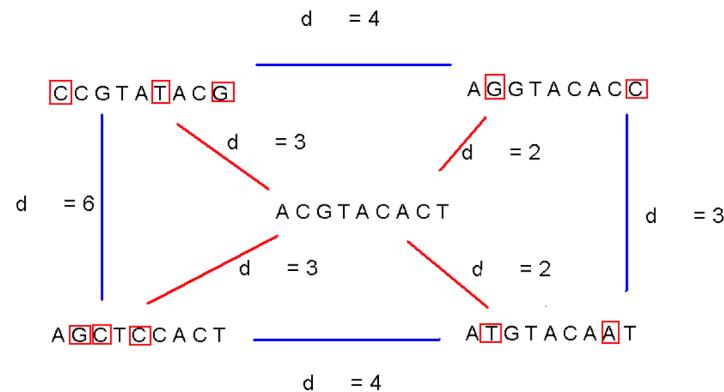
**G A G A C T C A T**  
X                    X  
**T A G A C G C A T**



A Hamming  
distance of 2

# 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

a G g t a c T t  
 C c A t a c g t  
 a c g t T A g t  
 a c g t C c A t  
 C c g t a c g G

A	3	0	1	0	3	1	1	0
C	2	4	0	0	1	4	0	0
G	0	1	4	0	0	0	3	1
T	0	0	0	5	1	0	1	4

Consensus a c g t a c g t

Score 3+4+4+5+3+4+3+4=30

# Let's try again, and handle errors this time!

```
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
    """
    score = 0
    for i in xrange(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

```
seqApprox = [
    'tagtgtctttgagtgtagatctgaaggaaaagtattccaccagttcgggtcacccagcagggcagggtgacttaat',
    'cgcgactcggcgctcacagttatcgcacgttagaccaaaacggagttggatccgaaactggagttaatcgagtcct',
    'gttacttgtgagcctggtagacccgaaataattgtggctgcatalogcgagctgacatacgagtagggaaatgcgt',
    'aacatcaggcttgattaaacaatttaagcacgtaaatccgaaattgacctgatgacaatacggaaacatgccgctccgg',
    'accacccggataggctgtttaggtccaaaaggtagtatcgtataatggctcagccatgtcaatgtgcggcattccac',
    'tagattcgaatcgatcgtttccctctgtgggtaacgaggggtccgacccgtcgcatgtgccgaacttgttaccc',
    'gaaatgttcggtgcgatatcaggccgttcttaacttggcggtcagatccgaaacgtctgtggaggggtcgctgccta',
    'atgtatactagacattctaactcgcttattggcgagaccattgctccactacaagaggctactgtgtagatccgta',
    'ttcttacaccctttagatccaaacctgttggcgccatcttcttcgactccatggctgtacccatggctgtatgac',
    'ctacctatgtaaaacaacatctactaacgttagtccggctttcctgatctgcctaaccctacaggtcgatccgaaattcg']

%timeit Score([17, 47, 18, 33, 21, 0, 46, 70, 16, 65], seqApprox, 10)
%time BruteForceMotifSearch(seqApprox[0:4], 10)
```

```
10000 loops, best of 3: 40.6 µs per loop
[17, 47, 18, 33] 36
CPU times: user 17min 22s, sys: 1.97 s, total: 17min 24s
Wall time: 17min 24s
```

# 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 complexity is

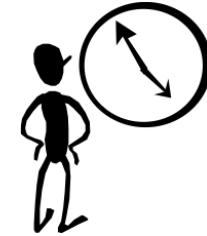
$$Mk(N - k + 1)^M = O(MkN^M)$$

- 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  
$$\frac{10^{12}}{(60*60*24*365)} > 30000 \text{ years}$$
- 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  $10^9$  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

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

- Is that significantly better?

