# ✓ LAB 4

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**Problem 1**: Find the distance between a pattern and a set of strings.

 $\textbf{Given} \hbox{:} \ A \ \mathsf{DNA} \ \mathsf{string} \ \mathsf{Pattern} \ \mathsf{and} \ \mathsf{a} \ \mathsf{collection} \ \mathsf{of} \ \mathsf{DNA} \ \mathsf{strings} \ \mathsf{Dna}.$ 

Return: DistanceBetweenPatternAndStrings(Pattern, Dna).

### **Sample Dataset**

AAA

TTACCTTAAC GATATCTGTC ACGGCGTTCG CCCTAAAGAG CGTCAGAGGT

### Sample Output

5

```
def HammingDistance(Pattern,pattern):
    count=0
    for i in range(len(pattern)):
        if(pattern[i]!=Pattern[i]):
            count+=1
    return count
```

```
Pattern='AAA'

dna=['TTACCTTAAC','GATATCTGTC','ACGGCGTTCG','CCCTAAAGAG','CGTCAGAGGT']

print("The distance is",DistanceBetweenPatternandStrings(Pattern,dna))
```

The distance is 5

## Problem 2:

Implement Brute Force Motif Search for a set of DNA strings.

Given a collection of strings Dna and an integer d, a k-mer is a **(k,d)-motif** if it appears in every string from Dna with at most d mismatches. The following algorithm finds (k,d)-motifs.

## **Implanted Motif Problem**

Implement MotifEnumeration (shown above) to find all (k, d)-motifs in a collection of strings.

**Given**: Integers k and d, followed by a collection of strings Dna.

Return: All (k, d)-motifs in Dna.

# **Sample Dataset**

3 1

ATTTGGC

**TGCCTTA** 

CGGTATC

GAAAATT

## **Sample Output**

ATA ATT GTT TTT

```
def _isMotif(Dna, motif, d):
    for string in Dna:
        found = False
        for i in range(len(string) - len(motif) + 1):
            mismatches = 0
            for j in range(len(motif)):
                if string[i + j] != motif[j]:
                   mismatches += 1
            if mismatches <= d:</pre>
                found = True
                break
        if not found:
           return False
    return True
def _MotifEnumeration(Dna, k, d, i, current_motif, motifs):
    if len(current_motif) == k:
        \# Check if current motif is a valid (k, d)-motif
        if _isMotif(Dna, current_motif, d):
            motifs.add("".join(current_motif))
        return
    \ensuremath{\text{\# Try}} all possible nucleotides at the current position
    for nucleotide in "ACGT":
        _MotifEnumeration(Dna, k, d, i + 1, current_motif + [nucleotide], motifs)
def MotifEnumeration(Dna, k, d):
   motifs = set()
    n = len(Dna[0])
    for i in range(n - k + 1):
        _MotifEnumeration(Dna, k, d, i, [], motifs)
    return list(motifs)
dna=['ATTTGGC','TGCCTTA','CGGTATC','GAAAATT']
k=3
d=1
motifs=MotifEnumeration(dna,k,d)
print("The motifs are")
for i in motifs:
    print(i)
     The motifs are
     ATA
     ATT
     GTT
     TTT
```

### Problem 3:

Given a set of 't' DNA Strings, display a Motif Matrix and calculate the corresponding Count matrix and Profile matrix. Use the profile matrix to form the Consensus string.

Dataset: Use NF-xB binding sites and form consensus "TCGGGGATTTCC"

image-2.png

```
def count_matrix(motifs):
   count = {'A': [], 'C': [], 'G': [], 'T': []}
   k = len(motifs[0])
   for nucleotide in ['A', 'C', 'G', 'T']:
       count[nucleotide] = [0] * k
   for i in range(k):
       for motif in motifs:
           count[motif[i]][i] += 1
   return count
def profile_matrix(motifs):
   counts = count_matrix(motifs)
   profile = {}
   k = len(motifs[0])
   for nucleotide in ['A', 'C', 'G', 'T']:
       profile[nucleotide] = [count / len(motifs) for count in counts[nucleotide]]
   return profile
```

```
def consensus_string(profile):
    consensus = ''
    k = len(profile['A'])
    for i in range(k):
       max_freq = -1
        cons_nucleotide = ''
        for nucleotide in ['A', 'C', 'G', 'T']:
            if profile[nucleotide][i] > max_freq:
                max_freq = profile[nucleotide][i]
                cons nucleotide = nucleotide
        consensus += cons_nucleotide
    return consensus
motifs = ["TCGGGGGTTTTT",
        "CCGGTGACTTAC",
        "ACGGGGATTTTC",
        "TTGGGGACTTTT"
        "AAGGGGACTTCC",
        "TTGGGGACTTCC",
        "TCGGGGATTCAT"
        "TCGGGGATTCCT",
        "TAGGGGAACTAC"
        "TCGGGTATAACC"]
t = len(motifs)
count_matrix_motifs = count_matrix(motifs)
profile_matrix_motifs = profile_matrix(motifs)
consensus = consensus_string(profile_matrix_motifs)
print("Motif Matrix:")
for motif in motifs:
   print(motif)
print("\nCount Matrix:")
for nucleotide in ['A', 'C', 'G', 'T']:
   print(f"{nucleotide}: {' '.join(map(str, count_matrix_motifs[nucleotide]))}")
print("\nProfile Matrix:")
for nucleotide in ['A', 'C', 'G', 'T']:
   print(f"{nucleotide}: {' '.join(map(str, profile_matrix_motifs[nucleotide]))}")
print("\nConsensus String:", consensus)
     Motif Matrix:
     TCGGGGGTTTTT
     CCGGTGACTTAC
     ACGGGGATTTTC
     TTGGGGACTTTT
     AAGGGGACTTCC
     TTGGGGACTTCC
     TCGGGGATTCAT
     TCGGGGATTCCT
     TAGGGGAACTAC
     TCGGGTATAACC
     Count Matrix:
     A: 2 2 0 0 0 0 9 1 1 1 3 0
     C: 160000041246
     G: 0 0 10 10 9 9 1 0 0 0 0
     T: 7 2 0 0 1 1 0 5 8 7 3 4
     Profile Matrix:
     A: 0.2 0.2 0.0 0.0 0.0 0.0 0.9 0.1 0.1 0.1 0.3 0.0
     C: 0.1 0.6 0.0 0.0 0.0 0.0 0.0 0.4 0.1 0.2 0.4 0.6
     G: 0.0 0.0 1.0 1.0 0.9 0.9 0.1 0.0 0.0 0.0 0.0 0.0
     T: 0.7 0.2 0.0 0.0 0.1 0.1 0.0 0.5 0.8 0.7 0.3 0.4
```

### Problem 4:

Given a profile matrix Profile, we can evaluate the probability of every k-mer in a string Text and find a **Profile-most probable** k-mer in Text, i.e., a k-mer that was most likely to have been generated by Profile among all k-mers in Text.

For example, ACGGGGATTACC is the Profile-most probable 12-mer in GGTACGGGGATTACCT. Indeed, every other 12-mer in this string has probability 0.

In general, if there are multiple Profile-most probable k-mers in Text, then we select the first such k-mer occurring in Text.

Profile-most Probable k-mer Problem Find a Profile-most probable k-mer in a string.

Given: A string Text, an integer k, and a 4 × k matrix Profile.

Consensus String: TCGGGGATTTCC

Return: A Profile-most probable k-mer in Text. (If multiple answers exist, you may return any one.)

# Sample Dataset

ACCTGTTTATTGCCTAAGTTCCGAACAAACCCAATATAGCCCGAGGGCCT

```
5
```

0.2 0.2 0.3 0.2 0.3

0.4 0.3 0.1 0.5 0.1

0.3 0.3 0.5 0.2 0.4

0.1 0.2 0.1 0.1 0.2

### Sample Output

CCGAG

```
def profile_most_probable_kmer(text, k, profile):
    max_prob = -1
   most_probable_kmer = ""
   for i in range(len(text) - k + 1):
       kmer = text[i:i+k]
       prob = 1
       for j in range(k):
           nucleotide = kmer[j]
           if nucleotide == 'A':
               prob *= profile[0][j]
           elif nucleotide == 'C':
              prob *= profile[1][j]
            elif nucleotide == 'G':
               prob *= profile[2][j]
            elif nucleotide == 'T':
               prob *= profile[3][j]
       if prob > max_prob:
           max_prob = prob
           most_probable_kmer = kmer
   return most_probable_kmer
```

```
text = "ACCTGTTTATTGCCTAAGTTCCGAACCAATATAGCCCGAGGGCCT"
k = 5
profile = [
    [0.2, 0.2, 0.3, 0.2, 0.3],
    [0.4, 0.3, 0.1, 0.5, 0.1],
    [0.3, 0.3, 0.5, 0.2, 0.4],
    [0.1, 0.2, 0.1, 0.1, 0.2]]
print(profile_most_probable_kmer(text, k, profile))
```

CCGAG

```
Start coding or generate with AI.

Start coding or generate with AI.
```