CS481/CS583: Bioinformatics Algorithms

Can Alkan

EA509

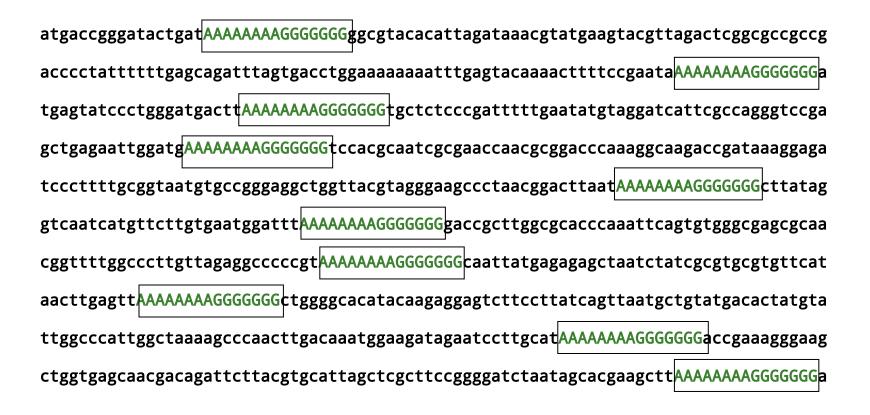
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MOTIFS

Random Sample

atgaccgggatactgataccgtatttggcctaggcgtacacattagataaacgtatgaagtacgttagactcggcgccgccg acccct attttttgag cag atttag t gacct g gaaaaaaaa ttt gag tacaaaacttttcc gaa tact g g g cataa g g tacaaaaacttttcc gaa tact g g g cataa g g tacaaaaact ttt ccgaa tact g g g cataa g g tacaaaaact ttt ccgaa tact g g g cataa g g tacaaaaact ttt ccgaa tact g g g cataa g g tacaaaaact ttt ccgaa tact g g g cataa g g tacaaaaact t t t ccgaa tact g g g cataa g g tacaaaaact t t ccgaa tact g g g cataa g g tacaaaaact t t ccgaa tact g g g cataa g g tacaaaaact t t ccgaa tact g g g cataa g g tacaaaaact t t ccgaa tact g g g cataa g g tacaaaaact t ccgaa tact g g g cataa g g tacaaaaact t ccgaa tact g g g cataa g cataa g cataa g cataa g g cataatgagtatccctgggatgacttttgggaacactatagtgctctcccgatttttgaatatgtaggatcattcgccagggtccga $\tt gctgagaattggatgaccttgtaagtgttttccacgcaatcgcgaaccaacgcggacccaaaggcaagaccgataaaggaga$ tcccttttgcggtaatgtgccgggaggctggttacgtagggaagccctaacggacttaatggcccacttagtccacttataggtcaatcatgttcttgtgaatggatttttaactgagggcatagaccgcttggcgcacccaaattcagtgtgggcgagcgcaa ctggtgagcaacgacagattcttacgtgcattagctcgcttccggggatctaatagcacgaagcttctgggtactgatagca



Where is the Implanted Motif?

at gaccgggatactgataaaaaaaagggggggggggcgtacacattagataaacgtatgaagtacgttagactcggcgccgccgtgagtatccctgggatgacttaaaaaaaagggggggtgctctcccgatttttgaatatgtaggatcattcgccagggtccga gctgagaattggatgaaaaaaagggggggtccacgcaatcgcgaaccaacgcggacccaaaggcaagaccgataaaggaga a actt gag ttaaaaaaaa gggggggct ggggcacata caa gag gag tctt ccttat cag ttaat gct gtat gacactat gtaal gag gag tag gag tag gag tag gag gag tag gag tag

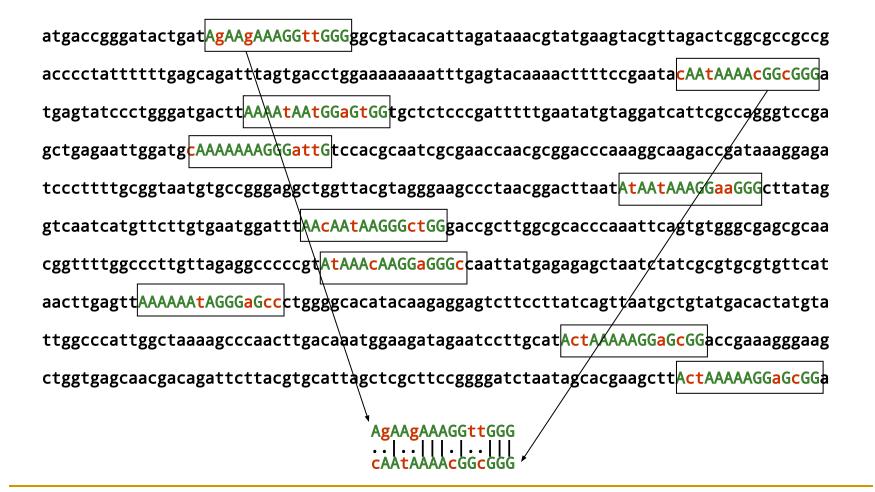
Implanting Motif **AAAAAAGGGGGGG** with Four Mutations

 $acccct attttttgag cagatttag t gacctg gaaaaaaa attt gag tacaaaacttttccgaata {\color{red} cAA} {\color{red} tAAAAA} {\color{red} cGG} {\color{$ tgagtatccctgggatgacttAAAAtAAtGGaGtGCtgctctcccgatttttgaatatgtaggatcattcgccagggtccga $tcccttttgcggtaatgtgccgggaggctggttacgtagggaagccctaacggacttaat | A {\color{red}t} A A {\color{red}t} A A A A {\color{red}G} {\color{red}G} {\color{red}aa} {\color{red}G} {\color{red}G} {\color{red}c} {\color{red}t} {\color{red}t} {\color{red}t} {\color{red}t} {\color{red}t} {\color{red}A} {\color{red}A} {\color{red}A} {\color{red}A} {\color{red}A} {\color{red}G} {\color{red}G} {\color{red}aa} {\color{red}G} {\color{red}G} {\color{red}c} {\color{red}t} {\color{red}t} {\color{red}t} {\color{red}t} {\color{red}aa} {\color{red}G} {\color{red}G} {\color{red}c} {\color{red}c} {\color{red}t} {\color{red}t} {\color{red}c} {\color{r$ cggttttggcccttgttagaggcccccgtAtAAAcAAGGaGGCcaattatgagagagctaatctatcgcgtgcgtgttcat aacttgagttAAAAAAtAGGGaGcc¢tggggcacatacaagaggagtcttccttatcagttaatgctgtatgacactatgta ctggtgagcaacgacagattcttacgtgcattagctcgcttccggggatctaatagcacgaagcttActAAAAAGGaGcGGa

Where is the Motif???

atgaccgggatactgatagaagaaaggttgggggggtacacattagataaacgtatgaagtacgttagactcggcgccgccg tgagtatccctgggatgacttaaaataatggagtggtgctctcccgatttttgaatatgtaggatcattcgccagggtccga gctgagaattggatgcaaaaaaagggattgtccacgcaatcgcgaaccaacgcggacccaaaggcaagaccgataaaggaga gtcaatcatgttcttgtgaatggatttaacaataagggctgggaccgcttggcgcacccaaattcagtgtgggcgagcgcaa aacttgagttaaaaaatagggagccctggggcacatacaagaggagtcttccttatcagttaatgctgtatgacactatgta ttggcccattggctaaaagcccaacttgacaaatggaagatagaatccttgcatactaaaaaggagcggaccgaaagggaag ctggtgagcaacgacagattcttacgtgcattagctcgcttccggggatctaatagcacgaagcttactaaaaaggagcgga

Finding (15,4) Motif



Challenge Problem

- Find a motif in a sample of
 - 20 "random" sequences (e.g., 600 chars long)
 - each sequence containing an implanted pattern of length 15,
 - each pattern appearing with 4 mismatches as (15,4)-motif.

Regulatory Regions

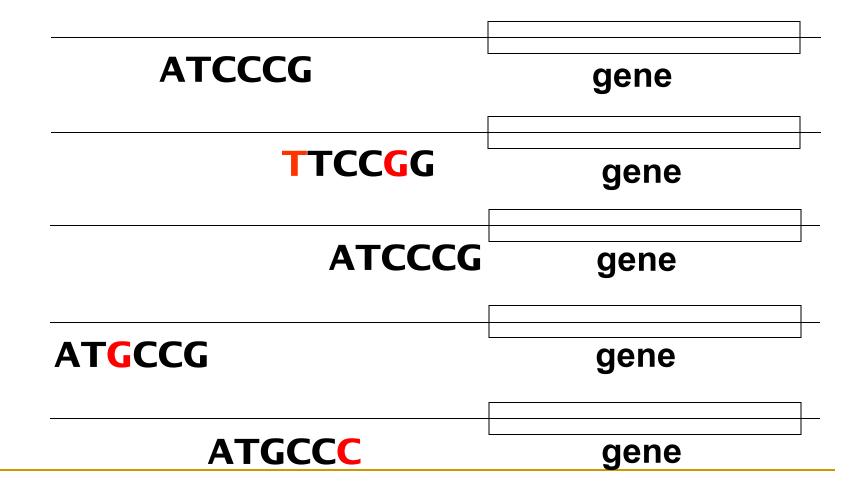
- Every gene contains a regulatory region (RR) typically stretching upstream of the transcriptional start site
- Located within the RR are the Transcription Factor
 Binding Sites (TFBS), also known as motifs, specific
 for a given transcription factor
- TFs influence gene expression by binding to a specific location in the respective gene's regulatory region -TFBS

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

Motifs and Transcriptional Start Sites



Identifying Motifs

- Genes are turned on or off by regulatory proteins
- These proteins bind to upstream regulatory regions of genes to either attract or block an RNA polymerase
- Regulatory protein (TF) binds to a short DNA sequence called a motif (TFBS)
- So finding the same motif in multiple genes' regulatory regions suggests a regulatory relationship amongst those genes

Identifying Motifs: Complications

- We do not know the motif sequence
- We do not know where it is located relative to the genes start
- Motifs can differ slightly from one gene to the next
- How to discern it from "random" motifs?

The Motif Finding Problem

Given a random sample of DNA sequences:

 Find the pattern that is implanted in each of the individual sequences, namely, the motif

Additional information:

The hidden sequence is of length 8

 The pattern is not exactly the same in each array because random point mutations may occur in the sequences

The patterns revealed with no mutations:

Consensus String

The patterns with 2 point mutations:

cctgatagacgctatctggctatccaGgtacTtaggtcctctgtgcgaatctatgcgtttccaaccat agtactggtgtacatttgatCcAtacgtacaccggcaacctgaaacaaacgctcagaaccagaagtgcaacgtTAgtgcaccctctttcttcgtggctctggccaacgagggctgatgtataagacgaaaattttagcctccgatgtaagtcatagctgtaactattacctgccacccctattacatcttacgtCcAtatacacctgttatacaacgcgtcatggcggggtatgcgttttggtcgtcgtacgctcgatcgttaCcgtacgGc

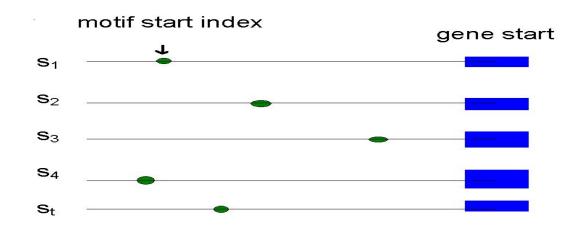
The patterns with 2 point mutations:

cctgatagacgctatctggctatccaGgtacTtaggtcctctgtgcgaatctatgcgtttccaaccat agtactggtgtacatttgatCcAtacgtacaccggcaacctgaaacaaacgctcagaaccagaagtgcaacgtTAgtgcaccctctttcttcgtggctctggccaacgagggctgatgtataagacgaaaattttagcctccgatgtaagtcatagctgtaactattacctgccacccctattacatcttacgtCcAtatacacctgttatacaacgcgtcatggcggggtatgcgttttggtcgtcgtacgctcgatcgttaCcgtacgGcctggttatacaacgcgtcatggcggggtatgcgttttggtcgtcgtacgctcgatcgttaCcgtacgGc

Can we still find the motif, now that we have 2 mutations?

Defining Motifs

- To define a motif, let's say we know where the motif starts in the sequence
- The motif start positions in their sequences can be represented as s = (s₁,s₂,s₃,...,s_t)



Motifs: Profiles and Consensus

 Line up the patterns by their start indexes

$$\mathbf{s} = (s_1, s_2, ..., s_t)$$

- A 3 0 1 0 3 1 1 0
 Profile 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
- Construct profile matrix with frequencies of each nucleotide in columns

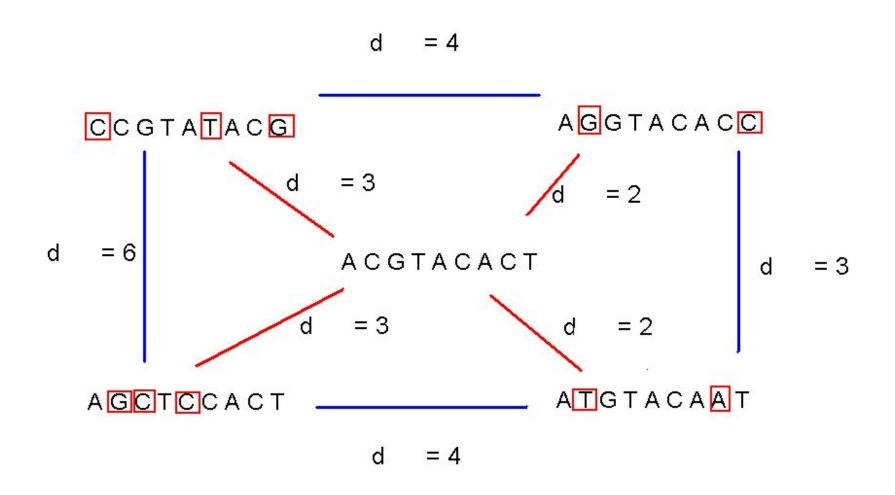
- Consensus A C G T A C G T
- Consensus nucleotide in each position has the highest score in column

Consensus

 Think of consensus as an "ancestor" motif, from which mutated motifs emerged

 The distance between a real motif and the consensus sequence is generally less than that for two real motifs

Consensus (cont'd)



Evaluating Motifs

We have a guess about the consensus sequence, but how "good" is this consensus?

 Need to introduce a scoring function to compare different guesses and choose the "best" one.

Defining Some Terms

- t number of sample DNA sequences
- n length of each DNA sequence
- DNA sample of DNA sequences (t x n array)

- I length of the motif (I-mer)
- s_i starting position of an *l*-mer in sequence i
- $s=(s_1, s_2, ..., s_t)$ array of motif's starting positions

Parameters

$$t = 8$$

$$cctgatagacgctatctggctatcc aGgtacTt aggtcctctgtgcgaatctatgcgtttccaaccat agtactggtgtacatttgat CcAtacgt acccggcaacctgaaacaaacgctcagaaccagaagtgc aaacgtTAgtgcaccctctttcttcgtggctctggccaacgagggctgatgtataagacgaaaatttt agcctccgatgtaagtcatagctgtaactattacctgccacccctattacatctt acgtCcAtataca ctgttatacaacgcgtcatggcggggtatgcgttttggtcgtcgtacgctcgatcgtta CcgtacgCc
$$n = 69$$

$$s \left\{ s_1 = 26 \quad s_2 = 21 \quad s_3 = 3 \quad s_4 = 56 \quad s_5 = 60 \right\}$$$$

Scoring Motifs

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

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

```
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 acgtacgt

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

The Motif Finding Problem

If starting positions $\mathbf{s} = (s_1, s_2, \dots s_t)$ are given, finding consensus is easy even with mutations in the sequences because we can simply construct the profile to find the motif (consensus)

But... the starting positions s are usually not given. How can we find the "best" profile matrix?

The Motif Finding Problem: Formulation

- Goal: Given a set of DNA sequences, find a set of Emers, one from each sequence, that maximizes the consensus score
- Input: A t x n matrix of DNA, and l, the length of the pattern to find
- Output: An array of t starting positions $s = (s_1, s_2, ... s_t)$ maximizing Score(s, DNA)

The Motif Finding Problem: Brute Force Solution

- Compute the scores for each possible combination of starting positions s
- The best score will determine the best profile and the consensus pattern in DNA
- The goal is to maximize Score(s, DNA) by varying the starting positions s; where:

$$s_i = [1, ..., n-l+1]$$

 $i = [1, ..., t]$

BruteForceMotifSearch

```
    BruteForceMotifSearch(DNA, t, n, ℓ)
    bestScore ← 0
    for each s=(s₁, s₂, ..., sℓ) from (1,1 ... 1)
        to (n-ℓ+1, ..., n-ℓ+1)
    if (Score(s, DNA) > bestScore)
    bestScore ← score(s, DNA)
    bestMotif ← (s₁, s₂, ..., sℓ)
    return bestMotif
```

Running Time of BruteForceMotifSearch

- Varying (n l + 1) positions in each of t sequences, we're looking at $(n l + 1)^t$ sets of starting positions
- For each set of starting positions, the scoring function makes ℓ operations, so complexity is $\ell(n \ell + 1)^t = O(\ell n^t)$
- That means that for t = 8, n = 1000, l = 10 we must perform approximately 10^{20} computations

The Median String Problem

 Given a set of t DNA sequences find a pattern that appears in all t sequences with the minimum number of mutations

This pattern will be the motif

Hamming Distance

- Hamming distance:
 - $d_H(\mathbf{v}, \mathbf{w})$ is the number of nucleotide pairs that do not match when \mathbf{v} and \mathbf{w} are aligned. For example:

$$d_{H}(AAAAAA,ACAAAC) = 2$$

Total Distance: An Example

Given v = "acgtacgt" and s

$$d_{H}(v, x) = 0$$

$$acgtacgt$$

$$cctgatagacgctatctggctatcc acgtacgt aggtcctctgtgcgaatctatgcgtttccaaccat$$

$$d_{H}(v, x) = 0$$

$$acgtacgt$$

TotalDistance(v, DNA) = 0

$$d_{H}(v, x) = 0$$

Total Distance: Example

Given v = "acgtacgt" and s

Total Distance: Definition

- □ For each DNA sequence i, compute all $d_H(v, x)$, where x is an ℓ -mer with starting position s_i $(1 \le s_i \le n \ell + 1)$
- □ Find minimum of $d_H(\mathbf{v}, \mathbf{x})$ among all ℓ -mers in sequence \mathbf{i}
- TotalDistance(v,DNA) is the sum of the minimum Hamming distances for each DNA sequence i
- □ $TotalDistance(\mathbf{v}, \mathbf{DNA}) = \min_{\mathbf{s}} d_H(\mathbf{v}, \mathbf{s})$, where \mathbf{s} is the set of starting positions $s_1, s_2, \dots s_t$

The Median String Problem: Formulation

- Goal: Given a set of DNA sequences, find a median string
- Input: A t x n matrix DNA, and l, the length of the pattern to find
- Output: A string v of l nucleotides that minimizes TotalDistance(v,DNA) over all strings of that length

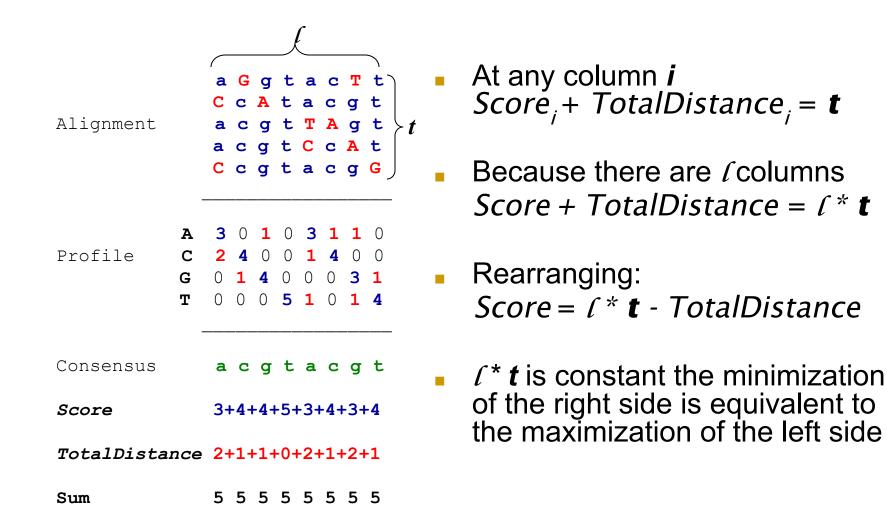
Median String Search Algorithm

- MedianStringSearch (DNA, t, n, l)
- 2. bestWord ←AAA...A
- 3. bestDistance ← ∞
- for each *E*mer s from AAA...A to TTT...T if TotalDistance(s,DNA) < bestDistance</p>
- bestDistance ← TotalDistance(s,DNA)
- 6. bestWord ← s
- 7. return bestWord

Motif Finding Problem = Median String Problem

- The Motif Finding is a maximization problem while Median String is a minimization problem
- However, the Motif Finding problem and Median String problem are computationally equivalent
- Need to show that minimizing TotalDistance is equivalent to maximizing Score

We are looking for the same thing



Motif Finding Problem vs. Median String Problem

- Why bother reformulating the Motif Finding problem into the Median String problem?
 - □ The Motif Finding Problem needs to examine all the combinations for \mathbf{s} . That is $(\mathbf{n} \ell + 1)^t$ combinations!!!
 - The Median String Problem needs to examine all 4^l combinations for v. This number is relatively smaller

STRUCTURING SEARCH

Motif Finding: Improving the Running Time

Recall the BruteForceMotifSearch:

```
1. BruteForceMotifSearch(DNA, t, n, l)

2. bestScore \leftarrow 0

3. for each \mathbf{s} = (s_1, s_2, \dots, s_t) from (1,1 . . . 1) to (\mathbf{n}-\mathbf{A}-1, . . . , \mathbf{n}-\mathbf{A}-1)

4. if (Score(\mathbf{s}, \mathbf{DNA}) > \mathbf{bestScore})

5. bestScore \leftarrow Score(\mathbf{s}, \mathbf{DNA})

6. bestMotif \leftarrow (s_1, s_2, \dots, s_t)

7. return bestMotif
```

Structuring the Search

How can we perform the line

```
for each s=(s_1, s_2, \ldots, s_r) from (1, 1, \ldots, 1) to (n-l+1, \ldots, n-l+1)?
```

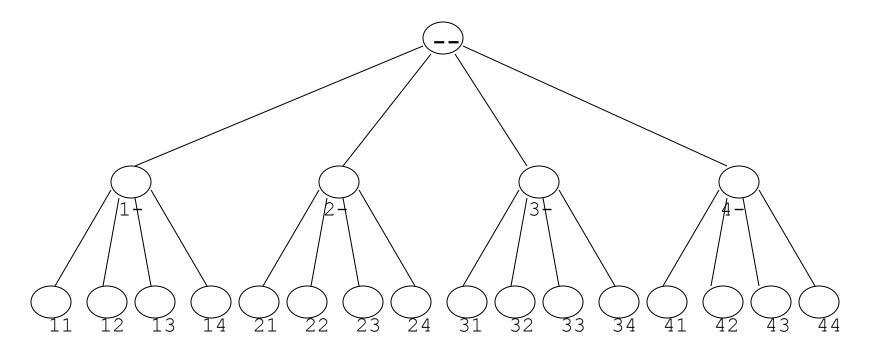
- We need a method for efficiently structuring and navigating the many possible motifs
- This is not very different than exploring all t-digit numbers

Median String: Improving the Running Time

- 1. MedianStringSearch (*DNA, t, n, l*)
- 2. bestWord ← AAA...A
- 3. bestDistance ← ∞
- for each *E*mer s from AAA...A to TTT...T
 - if TotalDistance(s,DNA) < bestDistance</pre>
- bestDistance ← TotalDistance(s,DNA)
- 6. bestWord ← s
- return bestWord

Search trees

- Assume t=2, I=2, n=5 (n-I+1=4)
 - 2 DNA sequences of length 5, look for 2-mer motif



Analyzing Search Trees

- Characteristics of the search trees:
 - The sequences are contained in its leaves
 - The parent of a node is the prefix of its children
- How can we move through the tree?

Moving through the Search Trees

- Four common moves in a search tree that we are about to explore:
 - Move to the next leaf
 - Visit all the leaves
 - Visit the next node
 - Bypass the children of a node

Visit the Next Leaf

Given a current leaf a, we need to compute the "next" leaf:

```
1. NextLeaf( \mathbf{a}, L, \mathbf{k})
2. for i \leftarrow L to 1
3. if a_i < \mathbf{k}
4. a_i \leftarrow a_i + 1
5. return a
6. a_i \leftarrow 1
7. return a
```

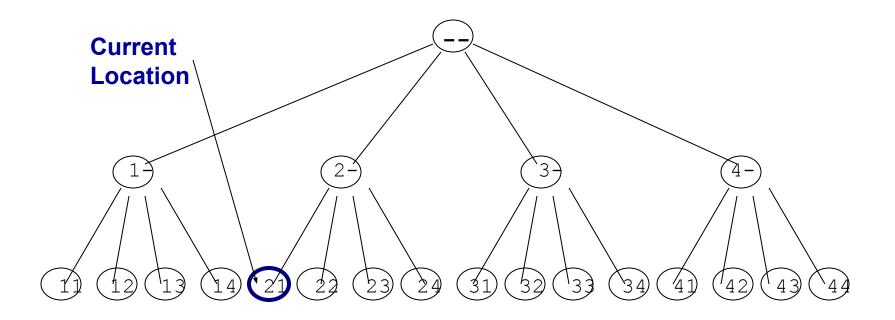
```
// a : the array of digits
// L: length of the array
// k : max digit value
```

NextLeaf (cont'd)

- The algorithm is common addition in radix k:
- Increment the least significant digit
- "Carry the one" to the next digit position when the digit is at maximal value

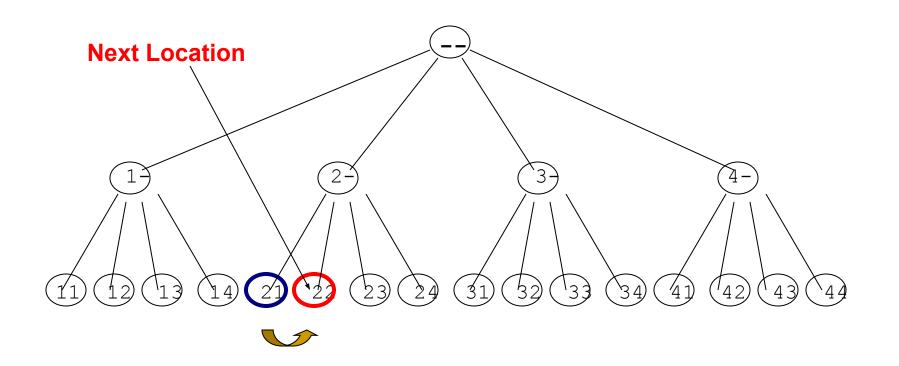
NextLeaf: Example

Moving to the next leaf:



NextLeaf: Example (cont'd)

Moving to the next leaf:



Visit All Leaves

Printing all permutations in ascending order:

```
AllLeaves(L, k) // L: length of the sequence

a \leftarrow (1,...,1) // k: max digit value

while forever // a: array of digits

output a

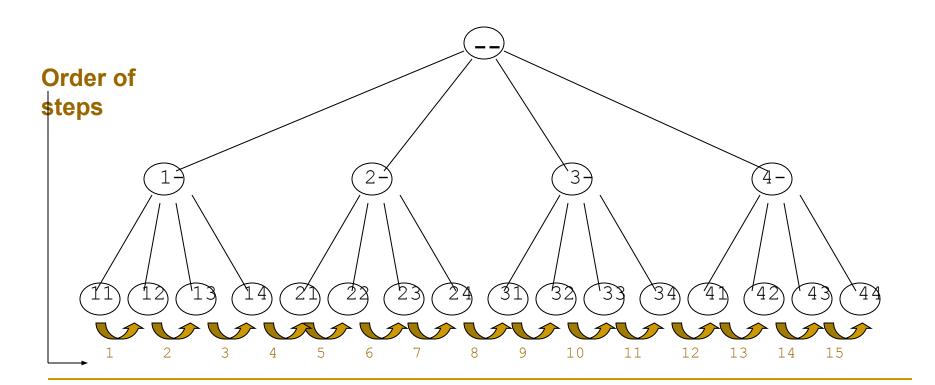
a \leftarrow NextLeaf(a, L, k)

if a = (1,...,1)

return
```

Visit All Leaves: Example

Moving through all the leaves in order:



Depth First Search

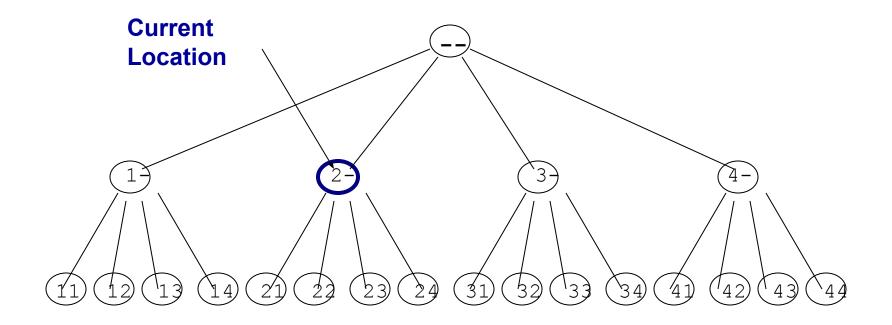
- So we can search leaves
- How about searching all vertices of the tree?
- We can do this with a depth first search

Visit the Next Vertex

```
NextVertex(\mathbf{a}, \mathbf{i}, \mathbf{l}, \mathbf{k}) // \mathbf{a}: the array of digits
        if i < L
                                      // i : prefix length
                             // L: max length
          a_{i+1} \leftarrow 1
          return (\mathbf{a}, \mathbf{i}+1) // \mathbf{k}: max digit value
       else
         for j \leftarrow l to l
            if a_j < k
              a_j \leftarrow a_j + 1
8.
              return(a,j)
9.
        return(a,0)
10.
```

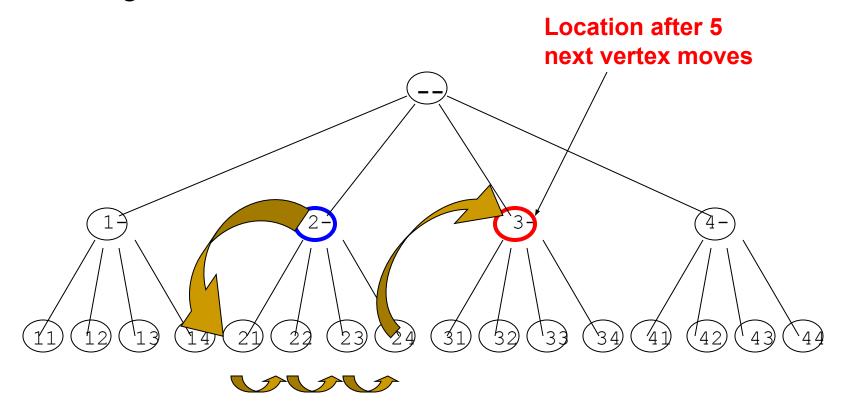
Example

Moving to the next vertex:



Example

Moving to the next vertices:



Bypass Move

 Given a prefix (internal vertex), find next vertex after skipping all its children

```
1. Bypass(\mathbf{a}, \mathbf{i}, \mathbf{k}) // \mathbf{a}: array of digits

2. for \mathbf{j} \leftarrow i to 1 // \mathbf{i}: prefix length

3. if a_j < \mathbf{k} // L: maximum length

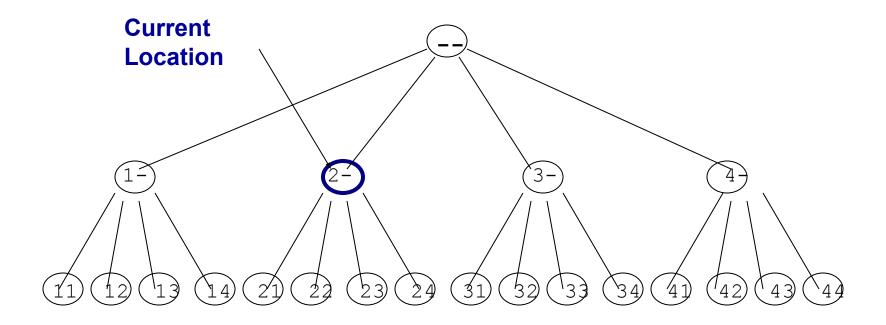
4. a_j \leftarrow a_j + 1 // \mathbf{k}: max digit value

5. return(\mathbf{a}, \mathbf{j})

6. return(\mathbf{a}, \mathbf{0})
```

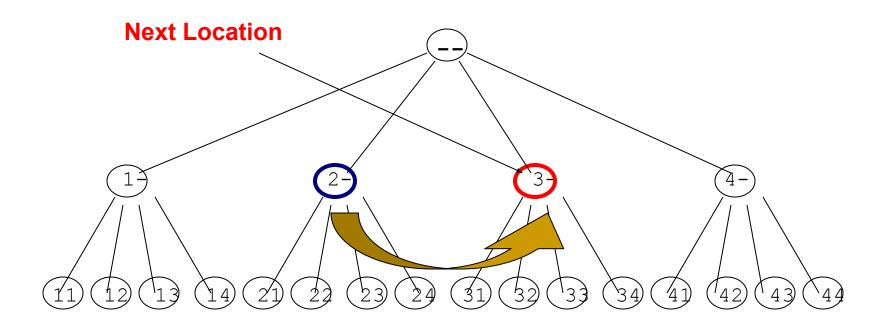
Bypass Move: Example

Bypassing the descendants of "2-":



Example

Bypassing the descendants of "2-":



Revisiting Brute Force Search

 Now that we have method for navigating the tree, let's look again at BruteForceMotifSearch

Brute Force Search Again

```
BruteForceMotifSearchAgain(DNA, t, n, l)
1.
         s \leftarrow (1, 1, ..., 1)
         bestScore \leftarrow Score(s, DNA)
         while forever
4
             \mathbf{s} \leftarrow \text{NextLeaf}(\mathbf{s}, \mathbf{t}, \mathbf{n} - \ell + 1)
5.
             if (Score(s, DNA) > bestScore)
6
                 bestScore \leftarrow Score(s, DNA)
7.
                 bestMotif \leftarrow (s_1, s_2, \ldots, s_r)
8.
         return bestMotif
```

Can We Do Better?

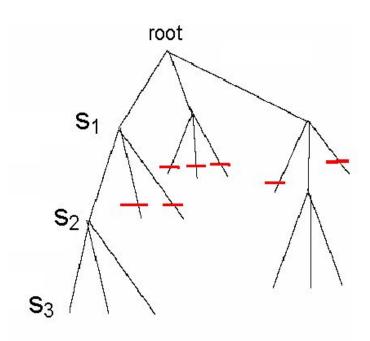
- Sets of $s=(s_1, s_2, ..., s_t)$ may have a weak profile for the first i positions $(s_1, s_2, ..., s_t)$
- Every row of alignment may add at most \(\ell\) to Score
- Optimism: if all subsequent (t-i) positions (s_{i+1}, ...s_t)

```
(t - i) * l to Score(s, i, DNA)
```

- If Score(s,i,DNA) + (t − i) * ℓ < BestScore, it makes no sense to search in vertices of the current subtree
 - Use ByPass()

Branch and Bound Algorithm for Motif Search

- Since each level of the tree goes deeper into search, discarding a prefix discards all following branches
- This saves us from looking at $(n l + 1)^{t-i}$ leaves
 - Use NextVertex() andByPass() to navigate the tree



Pseudocode for Branch and Bound Motif Search

```
BranchAndBoundMotifSearch(DNA,t,n,t)
 1.
         s \leftarrow (1,...,1)
 2.
         bestScore \leftarrow 0
 3.
         i ← 1
         while i > 0
 5.
              if i < t
 6.
                     optimisticScore \leftarrow Score(s, i, DNA) +(t - i) * \ell
 7.
                     if optimisticScore < bestScore
 8.
                       (s, i) \leftarrow Bypass(s, i, n-\ell+1)
 9.
                    else
10.
                       (s, i) \leftarrow NextVertex(s, i, n-\ell+1)
11.
              else
12.
                     if Score(s, DNA) > bestScore
13.
                        bestScore ← Score(s)
14.
                        bestMotif \leftarrow (s_1, s_2, s_3, ..., s_t)
15.
                           (\mathbf{s}, \mathbf{i}) \leftarrow \text{NextVertex}(\mathbf{s}, \mathbf{i}, \mathbf{t}, \mathbf{n} - \mathbf{i} + 1)
16.
         return bestMotif
17.
```

Structuring the Search: median string

□ For the Median String Problem we need to consider all 4^ℓ possible *E*-mers:

```
aa... aa aa... ag aa... at ... tt... tt
```

How to organize this search?

Alternative Representation of the Search Space

- Let A = 1, C = 2, G = 3, T = 4
- Then the sequences from AA...A to TT...T become:

```
11...11
11...12
11...13
11...14
•
•
•
```

 Notice that the sequences above simply list all numbers as if we were counting on base 5 without using 0 as a digit

Median String Search Improvements

- Recall the computational differences between motif search and median string search
 - □ The Motif Finding Problem needs to examine all (n-ℓ+1)^t combinations for s.
 - □ The Median String Problem needs to examine 4^{ℓ} combinations of v. This number is relatively small
- We want to use median string algorithm with the Branch and Bound trick!

Branch and Bound Applied to Median String Search

Note that if the total distance for a prefix is greater than that for the best word so far:

TotalDistance (*prefix*, *DNA*) > *BestDistance*

there is no use exploring the remaining part of the word

 We can eliminate that branch and BYPASS exploring that branch further

Bounded Median String Search

```
BranchAndBoundMedianStringSearch(DNA, t, n, l)
 1.
         s \leftarrow (1,...,1)
 2.
         bestDistance ← ∞
 3.
         i ← 1
 4.
         while i > 0
 5.
           if i < l
 6.
               prefix \leftarrow string corresponding to the first i nucleotides of s
 7.
               optimisticDistance ← TotalDistance(prefix, DNA)
 8.
                if optimisticDistance > bestDistance
 9.
                        (\mathbf{s}, \mathbf{i}) \leftarrow \text{Bypass}(\mathbf{s}, \mathbf{i}, \ell, 4)
10.
                else
11.
                        (\mathbf{s}, \mathbf{i}) \leftarrow \text{NextVertex}(\mathbf{s}, \mathbf{i}, \ell, 4)
12.
          else
13.
                word ← nucleotide string corresponding to s
14.
                if TotalDistance(s.DNA) < bestDistance
15.
                         bestDistance ← TotalDistance(word, DNA)
16.
                         bestWord ← word
17.
                (\mathbf{s}, \mathbf{i}) \leftarrow \text{NextVertex}(\mathbf{s}, \mathbf{i}, \ell, 4)
18.
         return bestWord
19.
```

previously: AAAAA, AAAAC, AAAAG,....
bestmotif = AACTA bestdist = 34
Search: ACGTA k=5
A totDist = 0
AC totDist = 23
ACG totDist = 55 X
do not search ACGT or ACGTA or ACGA or

ACGG or

Improving the Bounds

- Given an *E*-mer w, divided into two parts at point i
 - \mathbf{u} : prefix W_1, \ldots, W_i
 - \mathbf{v} : suffix $W_{i+1}, ..., W_{\ell}$
- Find minimum distance for u in a sequence
- No instances of u in the sequence have distance less than the minimum distance
- Note this doesn't tell us anything about whether u is part of any motif. We only get a minimum distance for prefix u

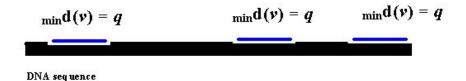
Improving the Bounds (cont'd)

Repeating the process for the suffix v gives us a minimum distance for v

Since u and v are two substrings of w, and included in motif w, we can assume that the minimum distance of u plus minimum distance of v can only be less than the minimum distance for w

Better Bounds

Searching for prefix VWe may find many instances of prefix V with a minimum distance q



Likewise for U

$$\min \mathbf{d}(u) = z \qquad \min \mathbf{d}(u) = z$$

But for U and V combined, U is not at its minimum distance location, neither is V

$$\min d(q+1 z+2)$$

But at least we know w (prefix u suffix v) cannot have distance less than $_{min}d(v) + _{min}d(u)$

Better Bounds (cont'd)

If d(prefix) + d(suffix) ≥ bestDistance:

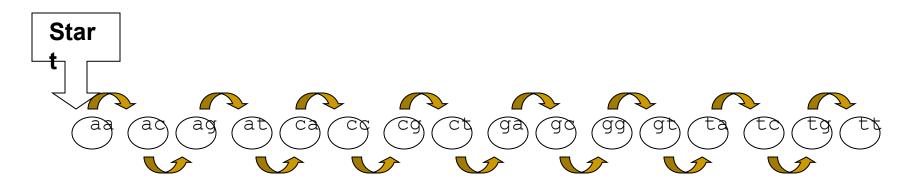
- Motif w (prefix.suffix) cannot give a better
 (lower) score than d(prefix) + d(suffix)
- In this case, we can ByPass()

Better Bounded Median String Search

```
ImprovedBranchAndBoundMedianString(DNA.t.n./)
 1.
           s = (1, 1, ..., 1)
 2.
           bestdistance = ∞
 3.
           i = 1
 4.
           while i > 0
                if i < l
 6.
                  prefix = nucleotide string corresponding to (s_1, s_2, s_3, ..., s_i)
 7.
                  optimisticPrefixDistance = TotalDistance (prefix, DNA)
 8.
                     if (optimisticPrefixDistance < bestsubstring[ i ])
9.
                        bestsubstring[ i ] = optimisticPrefixDistance
10.
                        if (\ell - i < i)
11.
                          optimisticSufxDistance = bestsubstring[\( \ell - i \) \]
12.
                        else
13.
                          optimisticSufxDistance = 0;
14.
                        if optimisticPrefixDistance + optimisticSufxDistance ≥ bestDistance
15.
                           (\mathbf{s}, i) = \text{Bypass}(\mathbf{s}, i, \ell, 4)
16.
                        else
17.
                           (\mathbf{s}, \mathbf{i}) = \text{NextVertex}(\mathbf{s}, \mathbf{i}, \ell, 4)
18.
                else
19.
                   word = nucleotide string corresponding to (s_1, s_2, s_3, ..., s_t)
20.
                  if TotalDistance( word, DNA) < bestDistance
21.
                    bestDistance = TotalDistance(word, DNA)
22.
                    bestWord = word
23.
                    (\mathbf{s}, i) = \text{NextVertex}(\mathbf{s}, i, l, 4)
24.
         return bestWord
25.
```

Linked List

Suppose ℓ = 2



 Need to visit all the predecessors of a sequence before visiting the sequence itself

Linked List (cont'd)

- Linked list is not the most efficient data structure for motif finding
- Let's try grouping the sequences by their prefixes



Search Tree

