Exhaustive Search: Motif Finding

Motif Discovery and Median String Detection

Informally. A regulatory motif is a nucleotide sequence widespread in DNA and conjectured to have some biological significance. Motif discovery in-silico usually involves spotting nucleotide sequences of certain lengths that tend to repeat in DNA significantly more often than the probability theory would dictate, if the occurrences were purely random. While finding repeated substrings in a string can be done using suffix trees (see previous handouts), motifs often vary from one another by a small number of differences. Figure 1 shows an example of a motif AATGCTGA as expressed in six different occurrences, each of which differs from the motif and other occurrences in at most two positions.

Informally, the motif finding problem is the problem of discovery of related patterns of a specified length in a collection of DNA fragment. We formalize this definition below.

Alignment Matrix. Let $D = (D_1, \ldots, D_t)$ be a collection of DNA strings in nucleotide alphabet $\{A, C, G, T\}$. Let $|D_i| = n$ and let l < n be the length of the desired regulatory motif. Let $s = (s_1, \ldots, s_t)$, $1 \le s_i \le n - l + 1$ be a list of starting positions in strings from D. An alignment matrix A(s) is a set of l-mers (strings of length l) from $D_1, \ldots D_t$ starting at positions s_1, \ldots, s_t respectively.

Example. Figure 2 shows an alignment matrix of six DNA strings for a collection of positions s = (4, 3, 5, 4, 3, 6).

Motif:	A	A	Τ	G	С	Τ	G	A
Sequence 1:	Α	Τ	Т	G	С	Τ	G	Α
Sequence 2:	A	A	T	\mathbf{C}	\mathbf{C}	T	G	A
Sequence 3:	A	Α	Τ	G	\mathbf{C}	<u>A</u>	G	Α
Sequence 4:	$\underline{\mathrm{T}}$	Α	T	G	\mathbf{C}	\mathbf{T}	G	Α
Sequence 5:	A	Α	T	G	\mathbf{C}	\mathbf{T}	$\underline{\mathbf{C}}$	Α
Sequence 6:	Α	Α	Τ	G	$\underline{\mathbf{G}}$	Τ	$\underline{\mathbf{C}}$	Α

Figure 1: Motif AATGCTGA as expressed in six different occurrences.

		A(s)	
D_1 :	ATC	ATTGCTGA	CCGTAT
D_2 :	GC	AATCCTGA	TGTCCGT
D_3 :	AGGCT	AATGCAGA	TATG
D_4 :	TTAC	TATGCTGA	GTAAT
D_5 :	TAT	AATGCTCA	TCATCC
D_6 :	TGCAG	AATGGTCA	TTAT

Figure 2: Alignment matrix.

D_1 :	ATC	Α	Τ	Τ	G	С	Τ	G	Α	CCGTAT
D_2 :	GC	Α	A	\mathbf{T}	\mathbf{C}	\mathbf{C}	Τ	G	A	TGTCCGT
D_3 :	AGGCT	Α	A	${\rm T}$	G	\mathbf{C}	A	G	A	TATG
D_4 :	TTAC	Т	A	${\rm T}$	G	\mathbf{C}	Τ	G	A	GTAAT
D_5 :	TAT	Α	A	${\rm T}$	G	\mathbf{C}	Τ	\mathbf{C}	A	TCATCC
D_6 :	TGCAG	Α	A	\mathbf{T}	G	G	\mathbf{T}	\mathbf{C}	A	TTAT
	A:	5	5	0	0	0	1	0	6	
P(s):	C:	0	0	0	1	5	0	2	0	
	G:	0	0	0	5	1	0	4	0	
	T:	1	1	6	0	0	5	0	0	
		Α	Α	Τ	G	С	Τ	G	A	

Figure 3: Profile matrix and the consensus string.

Profile matrix. Given a collection of DNA strings $D = (D_1, ..., D_t)$ and a list of positions $s = (s_1, ..., s_t)$, a profile matrix P(s) is a function

$$P: \{1, \dots, l\} \times \{A, C, T, G\} \rightarrow \{1, \dots, l\},\$$

defined as follows:

P(s)(i, X) = number of occurrences of letter X at position i in A(s).

Example. The profile matrix for the alignment matrix from Figure 2 is shown in Figure 3.

Consensus sequence. The consensus sequence for an alignment matrix A(s) is a string formed out of the nucleotides, most frequently occurring in each position.

Example. The consensus sequence for the alignment matrix from Figure 2 is shown at the bottom of Figure 3.

Consensus score. Given an alignment matrix A(s) its consensus score (or the consensus score of its starting position s) is defined as

$$Score(s) = \sum_{j=1}^{t} \max_{X \in \{A,C,G,T\}} P(s)(i,X)$$

Example. The *consensus score* of the alignment matrix A(s) from Figure 2 is:

$$Score(s) = 5 + 5 + 6 + 5 + 5 + 5 + 4 + 6 = 41.$$

Note. When all strings in A(s) are in full agreement, the consensus score will reach its maximum $Score(s) = t \cdot l$. The smallest possible consensus score is $Score(s) = \frac{t \cdot l}{4}$, which is achieved when P(s)(i,A) = P(s)(i,C) = P(s)(i,G) = P(s)(i,T) for all i = 1, ... l.

As such, the maximum possible score for an alignment of six strings of length 8 is $Score(s) = 6 \cdot 8 = 48$. A score of 41 shows a high degree of agreement on the consensus string among the six strings from the collection D.

Motif Finding Problem. The formal definition of the *motif finding problem* is:

Given a collection of DNA strings $D = (D_1, ..., D_t)$, each of length n, and an integer number $0 < l \le n$, find the list of starting positions (alignment) $s = (s_1, ..., s_t)$ which maximizes the consensus score Score(s) for all l-mer alignment matrices.

Median String Problem

Median String problem is equivalent to the motif finding problem.

Hamming Distance. Let $S = s_1 \dots s_l$ and $T = t_1 \dots t_l$ be two strings of length l in the same alphabet Σ . **Hamming distance** between S and T, denoted $d_H(S,T)$ is total number of positions i in which $s_i \neq t_i$.

For example, for S = ATTCGAT and T = ATGGCAT, $d_H(S, T) = 3$.

Definition. Let $D = (D_1, ..., D_t)$ be a collection of DNA strings in the $\{A, C, G, T\}$ alphabet and l be an integer such that $0 < l \le n = |D_1| = ... = |D_t|$. Let $s = (s_1, ..., s_t)$ be a list of starting positions in $D_1, ..., D_t$ respectively, and let us denote the set of all lists of starting positions as S. Let A(s) be the alignment matrix for s. Let $v = v_1 ... v_l$ be a string in the alphabet $\{A, C, G, T\}$.

The **Hamming distance** between v and A(s), denoted $d_H(v,s)$ is defined as

$$d_H(v,s) = \sum_{j=1}^t d_H(v, D_j[s_j : s_{j+l}]).$$

The total distance between a string v and the collection D is defined as

$$TotalDistance(v, D) = \min_{s \in \mathcal{S}} d_H(v, s).$$

A string $v = v_1 \dots v_l$ is a **median** of the collection $D = (D_1, \dots D_t)$ if $TotalDistance(v, D) \leq TotalDistance(v', D)$ for any other string v' of length l.

Median Finding Problem. Given a collection of DNA strings $D = (D_1, \ldots, D_t)$, each of length n, and an integer $0 < l \le n$, find the median string of length l for D.

Note. The median finding problem and the motif finding problem are equivalent because the *consensus string* from the motif finding problem minimizes the Hamming distance between a string and a string collection.

Algorithm Ideas

Observations.

- For the motif finding problem, we need to iterate over all possibe starting sequences $s = (s_1, \ldots, s_t)$. There are $(n l + 1)^t$ such sequences, and the optimal solution can, potentially be found at any of them.
- For the median string finding problem, we essentially need to find (or estimate) TotalDistance(v, D) for **every** l-mer string in the $\{A, C, G, T\}$ alphabet. There are 4^t such strings.
- In either case we need to enumerate a space of possible solutions that are representable as sequence of known length.
- In either case, a naïve algorithm is to set up the enumeration, and then:
 - For each $s = (s_1, \ldots, s_t)$, compute Score(s), and find the consensus string.
 - or, for each *l*-mer string $v = v_1 \dots v_l$, compute TotalDistance(v, D).
- Computation of TotalDistance(v, D) is straightforward: for each string $D_i \in \{D_1, \ldots, D_t\}$, find the position s_i that minimizes $d_H(v, s_i)$. Take $s = (s_1, \ldots, s_t)$ computed this way, and add up $d_H(v, s_i)$ to obtain TotalDistance(v, D).

So, solving either problem boils down to correctly and, if possible, efficiently enumerating the space of all possible solutions.

Search Trees

Definition. Consider a set S of all sequences $s = (s_1, \ldots, s_t)$, where $s_i \in \{1, \ldots, K\}$. A search tree over S is a tree of depth t+1 constructed as follows.

- Each non-leaf node in the tree has exactly K children.
- Each node in the tree has a label. The root has an empty label.
- The K children of the root have labels $1, 2, \ldots K$.
- Given a node with a label s', its K children will have labels s'1, s'2, ..., s'K.

Proposition. The leaf nodes in a search tree over S enumerate every single sequence in S.

Search trees for motif finding. For motif finding, K = n - l + 1, and the leaf labels will have the form $s = (s_1, \ldots, s_t)$, where t is the number of the DNA strings in D.

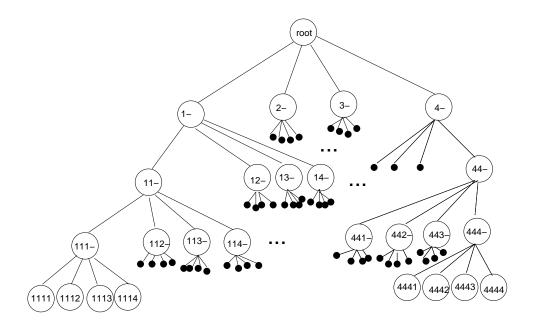


Figure 4: An example of a search tree.

Search trees for median string finding. For median string finding problem, we map A=1, C=2, G=3, T=4, set K=4 and t=l. The labels will be l-mers in the alphabet $\{1,2,3,4\}$, which uniquely map to l-mers in the nucleotide alphabet.

E.g., 1324 is AGCT, 11443 is AATTG and so on.

Example. Figure ?? shows how a search tree for a set of 4-mers in the $\{A, C, G, T\}$ alphabet would look.

Traversing the leaves of a search tree. Two simple algorithms NextLeaf() and AllLeaves() traverse the leaves of a search tree.

```
ALGORITHM NextLeaf(a[1..t],t,K)
                                                                    ALGORITHM AllLeaves(t,K)
             // a: current leaf; t: length of
                                                                    begin
sequence; K: size of alphabet
                                                                       a = (1, \dots, 1);
  \mathbf{for} \mathrel{\mathsf{i}}= L \; \mathbf{to} \; 1 \quad \mathbf{do}
                                                                       repeat
    if a[i] < K then
                                                                        print a;
      a[i] \leftarrow a[i] + 1;
                                                                         a \leftarrow \mathsf{NextLeaf}(a, t, K);
      return a;
                                                                      until a = (1, ..., 1);
    end if
                                                                    end
    a[i] \leftarrow 1;
  end for
end
```

References

[1] S.S. Skiena, W.D. Smith, P. Lemke (1990) Reconstructing Sets From Interpoint Distances. In *Proc. Sixth Annual Symposium on Computational*

Geometry, pp. 332-339, Berkeley, CA, June 1990.