

Enhanced Matrix Model for Finding Sequence Motif

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Abstract

This paper presents a probabilistic approach to locate motifs in a given set of molecular sequences. Motif is a short sequence of DNA or RNA (or amino acids) which often consists of 5- 16 nucleotides. The methodology presented here is based on probabilistic approach (Likely Hood Ratio) with three wild card locations to express motif locations and their poor, moderate and good results on the basis of matches found is observed. The proposed method allows three displacement locations (i.e. 1, 9, & 11) unlike other motif finding techniques that allow only two displacement locations.

I. INTRODUCTION & BACKGROUND

The two basic methods for computational gene prediction are *Sequence similarity search* and *Ab initio*. The sequence similarity search is based on finding similarity in gene sequences between ESTs (Expressed Sequence Tags) or other genomes and functional regions are more conserved evolutionarily than nonfunctional regions. While, *ab initio* uses gene structure as a template to detect genes and relies on two types of sequence information: signal sensors and content sensors [1]. A combinatorial optimization framework for motif finding that is flexible enough to model several variants of the problem and is not limited by the motif length [3]. The different approaches that build models to distinguish binding targets of a TF (i.e. positive sequences) from non-targets of a TF (i.e. negative sequences) [4]. The threshold for positional weight matrix compares favourably with broadly used Match method. It uses four digits [0, 1, 2, 3] of quaternary system for four nucleotides [A, C, G, T] and applied the random sampling scheme to perform the computation and sample size [5]. The versatile combinatorial optimization framework for motif finding that couples graph pruning techniques with a novel integer linear programming formulation. It combines graph theoretic and mathematical programming approach. It includes incorporating substitution matrices and phylogenetic distances [6]. The Boolean Matrices as Model for Motif Kernels [7] introduces BMA (Boolean Matrix Algorithm) which is more realistic than data models based on consensus strings and Hamming distance. Algorithm works on signal-noise ratio and selects the set of motif candidates by the minimal score of all its members. A weight matrix for a pattern of length n is defined as a matrix of numbers $W_{i,x}$ where i is in $\{1,2,\dots,n\}$ and x is in $\{A,T,G,C\}$ for DNA. The scores of the string $x_1 \dots x_n$ is given by: $W_{1,x_1} + W_{2,x_2} + \dots + W_{n,x_n}$. [2,8].

Matrix Models

In matrix, numbers containing scores for each residue or nucleotide at each position of a fixed-length motif.

There are two types of weight matrices.

A position frequency matrix (PFM) records the position-dependent frequency of each residue or nucleotide. PFMs can be experimentally determined from SELEX experiments or computationally discovered by tool such as MEME using hidden Markov models. While, position weight matrix (PWM) contains log odd weights for computing match score. A cut-off is needed to specify whether an input sequence matches the motif or not. PWMs are calculated from PFMs.

Position Weight Matrix Model:

```
Site 1 A G A T G G A T G G
Site 2 T G A T T G A T G T
Site 3 T G A T G G A T G G
Site 4 A G A T T G A T C G
Site 5 T G A T G G A T T G
Site 6 T G A T G G A T T G
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Site 7 A G A T G G A T T G

IUPAC (International Union of Pure Applied Chemistry) consensus:

W G A T G G A T N G (where W = A or T)

PWM represents frequencies of each base at each position in the motif

G	0	1.0	0	0	0.7	1.0	0	0	0.4	0.8
A	0.4	0	1.0	0	0	0	1.0	0	0	0
T	0.6	0	0	1.0	0.3	0	0	1.0	0.4	0.2
C	0	0	0	0	0	0	0	0	0.2	0

Information content IC

The least variable positions likely are important for specifying the protein-DNA interaction. Therefore high information content = low sequence variation at that position.

Information Content at position i:

$$IC_i = 2 + \sum_{b=G,A,T,C} P_b(i) * \log_2(P_b(i))$$

Where $P_b(i)$ is the probability of base b at position i
(If using \log_2 , the information is in 'bits')

G	0	1.0	0	0	0.7	1.0	0	0	0.4	0.8
A	0.4	0	1.0	0	0	0	1.0	0	0	0
T	0.6	0	0	1.0	0.3	0	0	1.0	0.4	0.2
C	0	0	0	0	0	0	0	0	0.2	0

IC	1.0	2.0	2.0	2.0	1.1	2.0	2.0	2.0	0.5	1.3	15.9 bit
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Maximum IC if P of some base is 1.0: $= 2 + [(1.0 * 0) + 0 + 0 + 0] = 2$

Minimum IC if P is 0.25 for all bases: $= 2 + [0.25 * (-2)] * 4 = 0$

Is the sequence A G A T T G A T C T a match to this matrix?

Joint probability: assuming each position is independent,

$$P(\text{motif}) = \prod_{b=G,A,T,C} P_b(i)$$

Background model: $P(G,A,T,C) = 0.25$

$$P(\text{sequence} | \text{matrix model}) = (0.4)(1.0)(1.0)(1.0)(0.3)(1.0)(1.0)(1.0)(0.2)(0.2) = 0.0048$$

$$P(\text{sequence} | \text{background model}) = (0.25)(0.25)(0.25)(0.25)(0.25)(0.25)(0.25)(0.25)(0.25)(0.25) = 6.8e-24$$

$$\text{Log-likelihood ratio LLR} = \log_2 (P(\text{sequence} | \text{matrix model}) / P(\text{sequence} | \text{background model}))$$

A measure of how different the likelihood of the sequence is, given the motif model vs. the background model.

G	0	1.0	0	0	0.7	1.0	0	0	0.4	0.8
A	0.4	0	1.0	0	0	0	1.0	0	0	0
T	0.6	0	0	1.0	0.3	0	0	1.0	0.4	0.2
C	0	0	0	0	0	0	0	0	0.2	0

Is the sequence A A A T T G A T C T a match to this matrix?

Joint probability: assuming each position is independent,

$$P(\text{motif}) = \prod_{b=G,A,T,C} P_b(i)$$

$$P(\text{sequence} | \text{matrix model}) = (0.4)(0)(1.0)(1.0)(0.3)(1.0)(1.0)(1.0)(0.2)(0.2) = 0$$

If PWM was trained on a small sample set, it might have missed some examples = *over fitting* of the matrix (ie. *too specific*)

II. PROPOSED METHODOLOGY

Formula:

The Position Weight Matrix represents frequencies of each base at each position in the motif.

TIB_i (Total Information Bit at position i)

$$TIB_i = 3 + \sum_{b=GATC} P_b(i) * \log_2(P_b(i)) \quad \dots(1)$$

Joint Probability: assuming each position is independent,

$$P_{(motif)} = \prod_{b=G,A,T,C} P_b^{(i)} \quad \dots (2)$$

Where P_b(i) is the probability of base b at position i.

The total information is on bits only when log₂ is used. [9]

Test Data I

Table: 1 Representation of Motifs (Position-Specific Weight Matrices)

	1	2	3	4	5	6	7	8	9	10	11
Site 1	A	G	A	T	G	G	A	T	G	G	C
Site 2	T	G	A	T	T	G	A	T	G	T	C
Site 3	T	G	A	T	G	G	A	T	G	G	A
Site 4	A	G	A	T	T	G	A	T	C	G	T
Site 5	T	G	A	T	G	G	A	T	T	G	T
Site 6	T	G	A	T	G	G	A	T	T	G	T
Site 7	A	G	A	T	G	G	A	T	T	G	A
	W	G	A	T	G	G	A	T	N	G	X

The above table is the combination of nucleotides. Where W (A or T), N (C or G or T) and X (C or A or T) are Wild Card locations. In the above any six locations can be fixed on the basis of the sequence of motif for which Likely Hood Ratio is to be calculated.

Table: 2 No. of occurrences of each nucleotide

G	0	7	0	0	5	7	0	0	4	6	0
A	3	0	7	0	0	0	7	0	0	0	2
T	4	0	0	7	2	0	0	7	3	1	3
C	0	0	0	0	0	0	0	0	0	0	2

The above table shows the column-wise number of occurrences of each nucleotide. Where 7 shows only one nucleotide is allowed at those locations. While, others are variable but column 5 & 10 allow only two types of nucleotides only. Here, T & G are considered.

Table:3 Finding Matches to (instances of) a PWM

G	0.00	1.00	0.00	0.00	0.71	1.00	0.00	0.00	0.57	0.86	0.00	
A	0.43	0.00	1.00	0.00	0.00	0.00	1.00	0.00	0.00	0.00	0.29	
T	0.57	0.00	0.00	1.00	0.29	0.00	0.00	1.00	0.43	0.14	0.43	
C	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.29	
*TIB=	2.01	3	3	3	2.13	3	3	3	2.01	2.41	1.44	28 Bits

The above table shows column-wise information bit and Total Information Bits (TIB). If each column is fixed (which is not possible) then the total information bit will be 33. Here loss of bits (33 – 28 = 5) i.e. the loss of 5 bits.

P(Sequence | Background Model) = Joint Probability (by formula (2)) = (0.25)¹¹ = 2.38*10⁻⁷
 P(Sequence | Matrix Model), if we take Minimum Probability of occurrence (by formula (1))

LHR = (matrix model/Background Model) = 13.16
 P(Sequence | Matrix Model), if we take Maximum Probability of occurrence (by formula (1))
 LHR = (matrix model/Background Model) = 18.16
 Seq T G A T T G A T T T A
 LHR=13.57

Further more experiments have been made on the basis of occurrences of nucleotide.

Test Data II

Table: 4 First Representations of Motifs (Position-Specific Weight Matrices)

	1	2	3	4	5	6	7	8	9	10	11
Site 1	A	G	A	T	G	G	A	T	G	G	C
Site 2	T	G	A	T	T	G	A	T	G	T	A
Site 3	T	G	A	T	G	G	A	T	G	G	A
Site 4	A	G	A	T	T	G	A	T	C	G	T
Site 5	T	G	A	T	G	G	A	T	T	G	C
Site 6	T	G	A	T	G	G	A	T	T	G	C
Site 7	A	G	A	T	G	G	A	T	T	G	A
	W	G	A	T	G	G	A	T	N	G	X

The above table is the combination of nucleotides. Where W (A or T), N (C or G or T) and X (C or A or T) are Wild Card locations. In the above any six locations can be fixed on the basis of the sequence of motif for which Likely Hood Ratio is to be calculated.

Table: 5 Finding Matches to (instances of) a PWM

G	0	7	0	0	5	7	0	0	4	6	0
A	3	0	7	0	0	0	7	0	0	0	3
T	4	0	0	7	2	0	0	7	3	1	1
C	0	0	0	0	0	0	0	0	0	0	3

The above table shows the column-wise number of occurrences of each nucleotide. Where 7 shows only one nucleotide is allowed at those locations. While, others are variable but column 5 & 10 allow only two types of nucleotides only. Here, T & G are considered.

Table: 6 No. of occurrences of each nucleotide

G	0.00	1.00	0.00	0.00	0.71	1.00	0.00	0.00	0.57	0.86	0.00	
A	0.43	0.00	1.00	0.00	0.00	0.00	1.00	0.00	0.00	0.00	0.43	
T	0.57	0.00	0.00	1.00	0.29	0.00	0.00	1.00	0.43	0.14	0.14	
C	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.43	
*TIB=	2.01	3	3	3	2.13	3	3	3	2.01	2.41	1.56	29.7 Bits

The above table shows column-wise information bit and Total Information Bits (TIB). If each column is fixed (which is not possible) then the total information bit will be 33. Here loss of bits (33 – 29.7 = 3.3) i.e. the loss of 3.3 bits.

P(Sequence | Background Model) = Joint Probability (by formula (2)) = $(0.25)^{11} = 2.38 \times 10^{-7}$
 P(Sequence | Matrix Model), if we take Minimum Probability of occurrence (by formula (1))
 LHR = (matrix model/Background Model) = 9.27
 P(Sequence | Matrix Model), if we take Maximum Probability of occurrence (by formula (1))
 LHR = (matrix model/Background Model) = 17.24

Seq. G G A T G G A T C T C
LHR=15.43

Test Data III

Table: 7 First Representation of Motifs (Position-Specific Weight Matrices)

	1	2	3	4	5	6	7	8	9	10	11
Site 1	A	G	A	T	A	G	A	T	A	A	C
Site 2	A	G	A	T	C	G	A	T	C	A	G
Site 3	T	G	A	T	G	G	A	T	T	A	A
Site 4	T	G	A	T	T	G	A	T	A	A	G
Site 5	T	G	A	T	A	G	A	T	A	G	G
Site 6	G	G	A	T	C	G	A	T	T	G	T
Site 7	G	G	A	T	T	G	A	T	G	C	A
	W	G	A	T	G	G	A	T	N	G	X

The above table is the combination of nucleotides. Where W (A or T or G), N (A or C or T or G) and X (A or C or T or G) are Wild Card locations. In the above any six locations can be fixed on the basis of the sequence of motif for which Likely Hood Ratio is to be calculated.

Table: 8. No. of occurrences of each nucleotide

G	2	7	0	0	1	7	0	0	1	2	3	7
A	2	0	7	0	2	0	7	0	3	4	2	0
T	3	0	0	7	2	0	0	7	2	0	1	0
C	0	0	0	0	2	0	0	0	1	1	1	0

The above table shows the column-wise number of occurrences of each nucleotide. Where 7 shows only one nucleotide is allowed at those locations. While, others are variable but column 5 & 10 allow all four (G, A, T & C) & three (G, A, & C) nucleotides respectively.

Table: 9 Finding Matches to (instances of) a PWM

G	0.29	1	0	0	0.14	1	0	0	0.14	0.29	0.43	
A	0.29	0	1	0	0.29	0	1	0	0.43	0.57	0.29	
T	0.43	0	0	1	0.29	0	0	1	0.29	0	0.14	
C	0	0	0	0	0.29	0	0	0	0.14	0.14	0.14	
*TIB=	1.44	3	3	3	1.05	3	3	3	1.16	1.62	1.16	24.4 Bits

The above table shows column-wise information bit and Total Information Bits (TIB). If each column is fixed (which is not possible) then the total information bit will be 33. Here loss of bits (33 – 24.4 = 8.6) i.e. the loss of 8.6 bits.

$P(\text{Sequence} | \text{Background Model}) = \text{Joint Probability (by formula (2))} = (0.25)^{11} = 2.38 \times 10^{-7}$
 $P(\text{Sequence} | \text{Matrix Model})$, if we take Minimum Probability of occurrence (by formula (1))
 $\text{LHR} = (\text{matrix model}/\text{Background Model}) = 8.87$
 $P(\text{Sequence} | \text{Matrix Model})$, if we take Maximum Probability of occurrence (by formula (1))
 $\text{LHR} = (\text{matrix model}/\text{Background Model}) = 15.19$
 Given Sequence A G A T G G A T C G A
LHR= 10.97

Test Data IV

Table: 10 First Representations of Motifs (Position-Specific Weight Matrices)

	1	2	3	4	5	6	7	8	9	10	11
Site 1	A	G	A	T	A	G	A	T	A	A	C
Site 2	T	G	A	T	C	G	A	T	C	G	G
Site 3	T	G	A	T	G	G	A	T	T	G	T
Site 4	C	G	A	T	G	G	A	T	A	C	G
Site 5	T	G	A	T	A	G	A	T	A	A	G
Site 6	G	G	A	T	C	G	A	T	T	T	A
Site 7	G	G	A	T	T	G	A	T	G	C	A
	W	G	A	T	G	G	A	T	N	G	X

The above table is the combination of nucleotides. Where W (A or C or T or G), N (A or C or T or G) and X (A or C or T or G) are Wild Card locations. In the above any six locations can be fixed on the basis of the sequence of motif for which Likely Hood Ratio is to be calculated.

Table: 11 No. of occurrences of each nucleotide

G	2	7	0	0	2	7	0	0	1	2	3
A	1	0	7	0	2	0	7	0	3	2	2
T	3	0	0	7	1	0	0	7	2	1	1
C	1	0	0	0	2	0	0	0	1	2	1

The above table shows the column-wise number of occurrences of each nucleotide. Where 7 shows only one nucleotide is allowed at those locations. While, others are variable but column 5 & 10 both allow all four (G, A, T & C) nucleotides.

Table: 12 Finding Matches to (instances of) a PWM

G	0.29	1	0	0	0.29	1	0	0	0.14	0.29	0.43	
A	0.14	0	1	0	0.29	0	1	0	0.43	0.29	0.29	
T	0.43	0	0	1	0.14	0	0	1	0.29	0.14	0.14	
C	0.14	0	0	0	0.29	0	0	0	0.14	0.29	0.14	
*TIB=	1.64	3	3	3	1.05	3	3	3	1.64	1.05	1.64	25.0 Bits

The above table shows column-wise information bit and Total Information Bits (TIB). If each column is fixed (which is not possible) then the total information bit will be 33. Here loss of bits (33 – 25.0 = 8) i.e. the loss of 8 bits.

$P(\text{Sequence} | \text{Background Model}) = \text{Joint Probability (by formula (2))}$
 $= (0.25)^{11} = 2.38 \times 10^{-7}$

$P(\text{Sequence} | \text{Matrix Model})$, if we take Minimum Probability of occurrence (by formula (1))
 $\text{LHR} = (\text{matrix model}/\text{Background Model}) = 7.82$

$P(\text{Sequence} | \text{Matrix Model})$, if we take Maximum Probability of occurrence (by formula (1))
 $\text{LHR} = (\text{matrix model}/\text{Background Model}) = 15.14$

Given Sequence G G A T G G A T C T C
 $\text{LHR} = 9.92$

ALGORITHM:

Step 1: Declare and initialize

```

char x[7][11],s[11]
int a[11],c[11],g[11],t[11]
real A[11],C[11],G[11],T[11],TI[11],AA[11],CC[11],GG[11],TT[11],lhr

```

Step 2: input sequence in x[7][11]

Step 3: count occurrences of nucleotides A,C,T,G

Step 4: calculate probabilities of occurrences column-wise

Step 5: calculation of total information bit (at log base 2)

Step 6: input the sequence of eleven nucleotides

Step 7: for non-zero probabilities

```

begin
lhr <- (mm/2.38)*10000000
initialize float LHR
LHR := (log10(lhr))
LHR := LHR/0.3010 // for the conversion at log2
end

```

Step 7: display LHR for given sequence of 11 nucleotides

Step 8: Stop

III. RESULTS

Experiments have been conducted & verified over 500 data sets and it is found that Likely Hood Ratio lies between 7.8 & 20.9. These experimental results give us some insight of motif finding problem. The three cases are generated, first, if the LLR is less than 10, it is non-considerable; second, if LLR is more than 10 and less than 15 then it is considered as motif that can be carried out for further calculations and the third, if LLR is between 15 and 20 is strongly recommended for the further calculation of motifs.

CONCLUSION

In this paper, when the motif length is of 11 nucleotides, and the probabilities of variable positions change, the LLR lies between 7.82 (min) and 20.92 (max), which are satisfied in above calculations. The loss of information bits does not affect the LHR. This method is considerable for the motif finding. The log to the base 2 is used as per the computational technique of the Information Theory. This method is position bound at locations 2 to 8 & 10 while, variable positions are 1, 9, and 11. This paper concludes for the motif finding with respect to their positions at the given motif length. This length may vary for finding motif in the researcher's interest. This calculation leads to the 3 displacement places. This is not compulsory to hold 2 to 8 & 10 as fixed while, 1, 9 and 11 as variables. The researchers can vary any three locations as per their own interest but the result will be same.

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