

Graduate Computational Genomics

02-710 / 10-810 & MSCBIO2070

Introduction to motif finding

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Lecture #10, February 15, 2007

Reading: handouts & papers



Outline

- The problem
- Motif representation
- Motif discovery
 - A greedy approach
 - Expectation-Maximization
 - Gibbs sampling
 - Self-organizing maps



The problem



- **Motif** is a pattern that nature has preserved in biological sequences, usually for a reason.
- Motifs can be viewed as degenerate sequences (strings).
- Motifs are typically detected either by examining **one gene in multiple species** or by examining **many genes in one species**.

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Why find motifs?

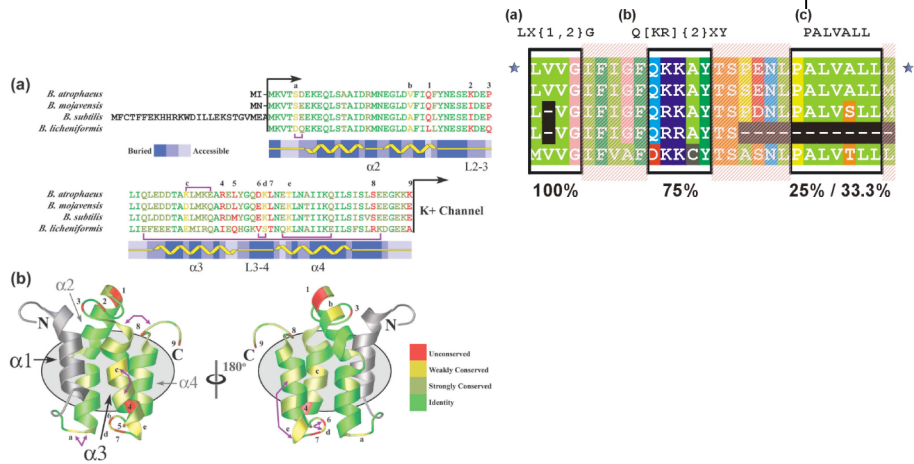


- In proteins - may be an important component
 - Find similarities to known proteins
 - Find important areas of a protein family
- In DNA - may be a *regulatory element*
 - Discover how the gene expression is regulated

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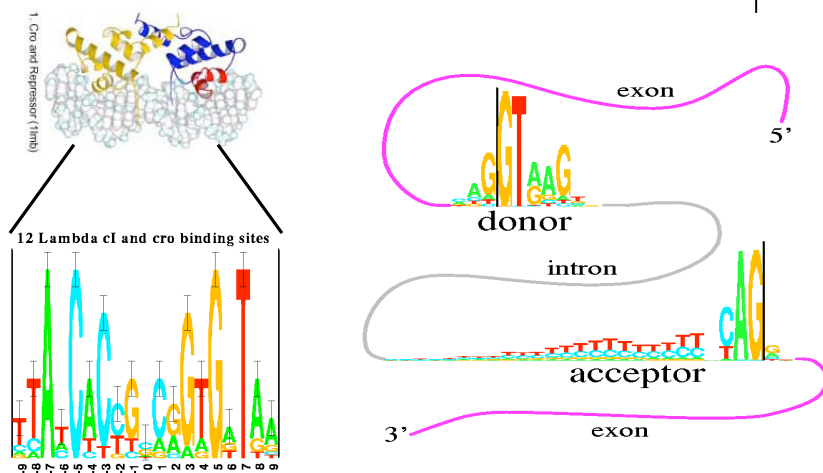
Protein motifs



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DNA motifs



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Pattern representation



- Majority rule
- String representation
 - Consensus sequence
 - Regular expressions

C-26	AGGATATT
C-57	AGGATATT
C-25	CATATTTT
7	AGAGTTTT
67	AGCATTTT
	AGNATTTT

C - X(2,4) - **C** - X(3) - [LIVMFYWC] - X(8) - **H** - X(3,5) - **H**

ISP1:

KKFA- **C** - PE - **C** - PKR - **F** - MRSDHLSK - **H** - IKT - **H** - QNKK

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Pattern representation



- Majority rule
- String representation
 - Consensus sequence
 - Regular expressions
- Probabilistic modeling
 - Weight matrices (PSSM models)
 - ✓ HMMs

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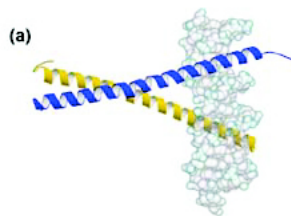
Transcription factors

- *Function:* they recognise a specific DNA target sequence (typically in the proximity of the gene) and facilitate transcription initiation and/or repression.
- Preferential DNA-binding.
- Tolerant in (some) base substitutions at the most preferred target sequence.
- Moderate change in affinity between “good” targets.
- TF binding patterns are generally more difficult to detect than protein patterns. (*why?*)

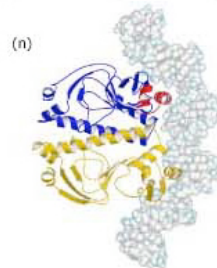
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Transcription factors (cntd)



21. Leucine zipper (2dgc)



14. Catabolic gene activator (2cgp)

Source: Luscombe et al. (2000) *Genome Biol.* 1(1):REVIEWS001

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Consensus representation

Sequences:

```

A G G A T A T T
A G G A T A T T
C A T A T T T T
A G A G T T T T
A G C A T T T T
    
```

Putative
new targets:

```

A G A G T T G T
G G A G T T T T
    
```

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Position Specific Scoring Matrices (PSSM)

A	9	214	63	142	118	8
C	22	7	26	31	52	13
G	18	2	29	38	29	5
T	193	19	124	31	43	216

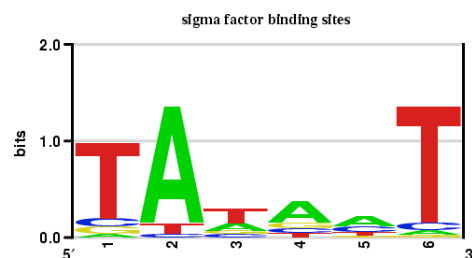
Counts from 242 known σ^{70} sites

A	.04	.88	.26	.59	.49	.03
C	.09	.03	.11	.13	.21	.05
G	.07	.01	.12	.16	.12	.02
T	.80	.08	.51	.13	.18	.89

Relative frequencies $f_b(i)$

A	-38	19	1	12	10	-48
C	-15	-38	-8	-10	-3	-32
G	-13	-48	-6	-7	-10	-40
T	17	-32	8	-9	-6	19

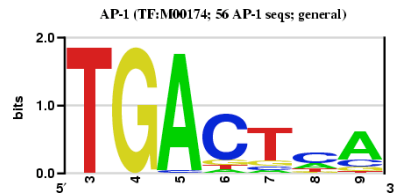
$10 \times \log_2 f_b(i)/f_{REF}$



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Sequence LOGO



- A motif is interesting if it differs from the background. So...

$$RH(model) = \sum_{x=1}^L \sum_{b=A}^T f(b,x) \cdot \ln \frac{f(b,x)}{f_{REF}(x)}$$

Relative Entropy
(Information Content)

$$Height_{col}(x) = 2 + \underbrace{\sum_{b=A}^T f(b,x) \cdot \log_2 f(b,x)}_{\text{Shannon's entropy (H)}}$$

not a proper metric!!

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Characteristics of Regulatory Motifs

```

ATATATAA TT T
CTGATAA CAG
GTGA TCA
AGGCG AGC
AA AA TAA
TTTAA TAA
GAAACG TTGCG
AA TTA T A
TTTAA TAA
GGGACGAG
AAAAATTT
A GA AAAA
TATATA TT
AAA AA AAAA
TTT AA AA
T T T T A AAA
ATATAT AT A
ATATAA TT
    
```

- Tiny
- Highly Variable
- ~Constant Size
 - Because a constant-size transcription factor binds
- Often repeated
- Low-complexity-ish

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A little more about PSSMs

- The PSSM model is a convenient representation of a [transcription factor binding preferences](#).
- Scoring a sequence against a PSSM model calculates the log-likelihood ratio of the sequence coming from the model vs. the background model.
- PSSM models assume [position independency](#), meaning that the nucleotide distributions between two positions are independent.
- PSSM scores are related to [interaction energy](#) *via* the Boltzmann distribution.

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1.- PSSM scoring

	C	T	A	T	A	A	T	C
A	-38	19	1	12	10	-48		
C	-15	-38	-8	-10	-3	-32		
G	-13	-48	-6	-7	-10	-40		
T	17	-32	8	-9	-6	19		

-93

C	T	A	T	A	A	T	C
A	-38	19	1	12	10	-48	
C	-15	-38	-8	-10	-3	-32	
G	-13	-48	-6	-7	-10	-40	
T	17	-32	8	-9	-6	19	

+85

C	T	A	T	A	A	T	C
A	-38	19	1	12	10	-48	
C	-15	-38	-8	-10	-3	-32	
G	-13	-48	-6	-7	-10	-40	
T	17	-32	8	-9	-6	19	

-95

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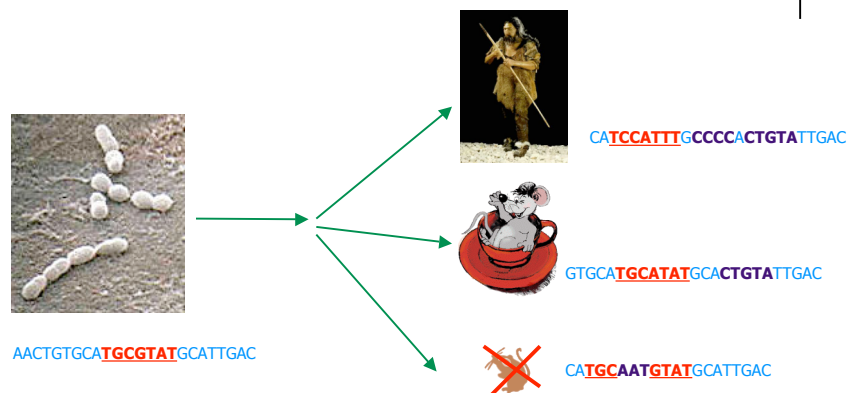
More on PSSM scoring

- PSSM scanning can be used to identify sites that match the profile in new promoters.
- Setting up a threshold is somewhat arbitrary.
 - Controlling FDR
 - Controlling false positives
- Even if the threshold is perfect, still some sequences will turn up as “positive” (why?)
- **Solution:** ask nature’s help!

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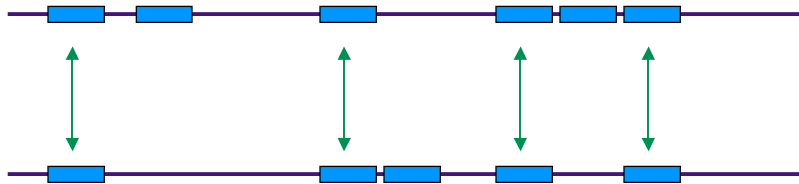
Evolution of regulatory regions



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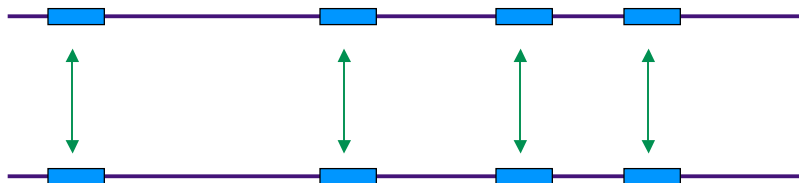
Phylogenetic footprinting



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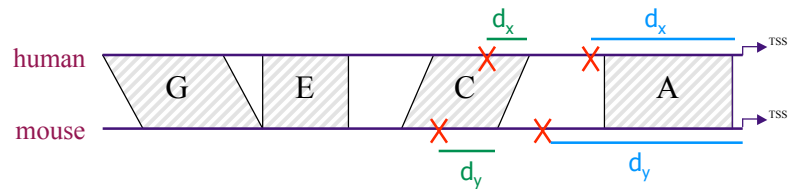
Phylogenetic footprinting



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FOOTER scoring



$$PF_D = P(d_{XY} \leq d_1) = \frac{1}{D} + \sum_{k=1}^{d_1} \frac{2 \cdot (D-k)}{D^2}$$

$$PF_S = P((S+T) \leq (s+t) \mid M_1, M_2)$$

$$PF = -w_D \cdot \ln(PF_D) - w_S \cdot \ln(PF_S)$$

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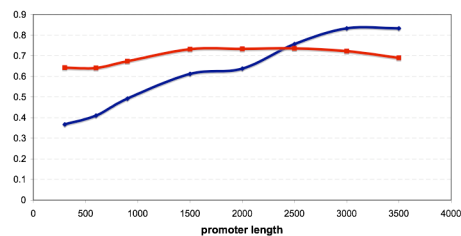
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FOOTER performance

$$S_N = \frac{TP}{TP + FN}$$

$$S_P = \frac{TP}{TP + FP}$$

Sensitivity and Specificity



	Footer	<i>ConSite</i> (def)	<i>ConSite</i> (adj)	<i>rVista</i> v1.0
# of sites	72	49	49	69
TP	60	23	34	54
FP	23	15	28	189
S_N	83.3%	46.9%	69.4%	78.2%
S_P	72.3%	60.5%	54.8%	22.2%

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PSSM to Binding Energy

- Assuming equilibrium...

$$P(d|p) = \frac{P_{REF}(d) \cdot e^{-E(p,d)/k_B T}}{\sum_{x_i} P_{REF}(x_i) \cdot e^{-E(p,x_i)/k_B T}} \Rightarrow \frac{P(d|p)}{P_{REF}(d)} = \frac{e^{-E(p,d)/k_B T}}{Z} \Rightarrow$$

$$\Rightarrow \ln \frac{P(D|P)}{P_{REF}(D)} = -E(p,d)/k_B T - c$$



Ludwig
Boltzmann
(1844-1906)

- The PSSM log-likelihood score is equal to the binding energy (in $k_B T$ units) minus the average background energy.
- The absolute energy is not important.
 - Try adding/subtracting a number from all energy values

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Acknowledgements

Some of the slides used in this lecture are adapted or modified slides from lectures of:

- Serafim Batzoglou, Stanford University
- Eric Xing, Carnegie-Mellon University

Theory and examples from the following:

- <http://www-lmmb.ncifcrf.gov/~toms/sequencelogo.html>

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