

CAP 5510: Introduction to Bioinformatics

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Pattern Discovery

Patterns

- ❑ Nature **stumbles** upon **recipes** to accomplish tasks.
- ❑ With high probability, such recipes are reused.
- ❑ This causes the recipe to be conserved through evolution.
- ❑ Such recipes give rise to **patterns**.

Why Pattern Discovery?

- Modern Biomedical Research
 - Generates a “ton of data”.
 - Use analytical tools to find patterns in data.
- Pattern Discovery facilitates this process!
 - Pattern Discovery in *sequences*
 - Pattern Discovery in *structures*
 - Pattern Discovery in *quantitative data*
- Patterns help to *detect* members of a class
- Patterns help to *characterize* classes

Sequence Patterns: Examples

- Protein active sites and functional domains
 - For e.g., Zinc-finger motifs & Helix-turn-helix motifs
- Protein family signatures
- Signals in DNA e.g., protein binding sites
- MicroRNA and Anti-sense RNA

Example 1: Protein Motifs

□ DNA-binding motifs

● **Helix-turn-Helix**

□ Motifs in Cys₂His₂-Zinc-binding proteins

Example: Zinc Finger Motif

...**Y****K****C**₃**G****L****C**₆**E****R****S****F****V****E****K****S****A****L****S****R****H**₁₉**O****R****V****H**₂₃**K****N**...

□ Motifs in proteins that bind to [4Fe-4S]-complex

Example: Ferredoxin subfamily

...**C****x****x****C****x****x****C****x****x****x****C****P**...

How to Represent Patterns

- Consensus sequence
- Alignments
- LOGO format
- Frequency Matrices
- Weight Matrices (Profiles, PSSMs, PWMs)

Pattern Representations

□ Consensus sequences

[Pribnow, 1975]

TACGAT

TATAAT

TATAAT

GATACT

TATGAT

TATGTT

TATAAT Consensus

TATRNT Consensus
w/ IUPAC

TATAAT Multi-level
G CGC Consensus
T

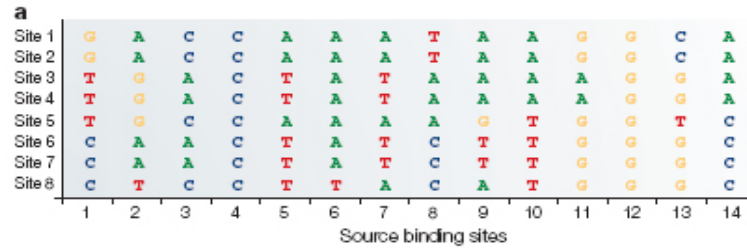
Needs Alignment

Pattern Representations

- Consensus sequences
- Weight Matrices (Profiles, PSSMs)
 - Frequency Counts
 - Relative Frequency Measures
 - Normalized Measures
 - Log-transformed Measures
 - Information content
 - "Logo" technique
 - HMMs

Pattern Representation: Weight Matrix

Alignment



Consensus



Frequencies

c Position frequency matrix (PFM)

	1	2	3	4	5	6	7	8	9	10	11	12	13	14
A	0	4	4	0	3	7	4	3	5	4	2	0	0	4
C	3	0	4	8	0	0	0	3	0	0	0	0	2	4
G	2	3	0	0	0	0	0	0	1	0	6	8	5	0
T	3	1	0	0	5	1	4	2	2	4	0	0	1	0

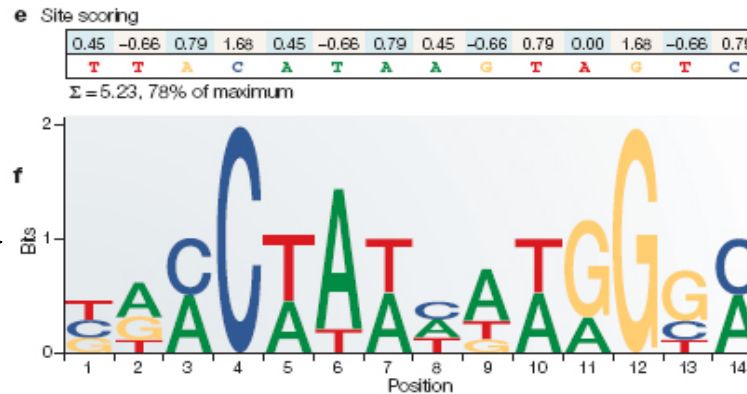
Scoring a sequence against a profile

Profile/
PSSM/PWM

d Position weight matrix (PWM)

A	-1.93	0.79	0.79	-1.93	0.45	1.50	0.79	0.45	1.07	0.79	0.00	-1.93	-1.93	0.79
C	0.45	-1.93	0.79	1.68	-1.93	-1.93	-1.93	0.45	-1.93	-1.93	-1.93	-1.93	-1.93	0.00
G	0.00	0.45	-1.93	-1.93	-1.93	-1.93	-1.93	-1.93	0.66	-1.93	1.30	1.68	1.07	-1.93
T	0.15	0.66	-1.93	-1.93	1.07	0.66	0.79	0.00	0.00	0.79	-1.93	-1.93	-0.66	-1.93

Visualizing a profile



[Wasserman, Sandelin, Nat Genet, 2004]

Formulae

□ Prob of char **b** in position **i**: $p(b, i) = \frac{f_{b,i}}{N}$

Frequency
Sequences

□ Corrected prob: $P(b, i) = \frac{f_{b,i} + s(b)}{N + \sum_{a \in \mathbf{A}} s(a)}$

PseudoCount

□ Weight matrix entry: $W_{b,i} = \log_2 \frac{P(b, i)}{BP(b)}$

Background Frequency

□ Information content of position of **i**:

$$D_i = 2 + \sum_b P(b, i) \log_2 P(b, i)$$

[Wasserman, Sandelin,
Nat Genet, 2004]

Statistical Evaluation Fundamentals

- Probability of finding a sequence w in some position of a DNA/protein sequence (assuming independence at each position)

$$\Pr(w) = \prod_{i=1}^m \Pr(w_i)$$

- $\Pr(w_i) = \text{BP}(b)$ [Background Frequency]

Statistical Evaluation

□ **Z-score** of a motif with a certain frequency: →

$$z(w) = \frac{Obs(w) - Exp(w)}{\sqrt{Var(w)}}$$

□ **Information Content** or Relative Entropy of an alignment or profile: →

$$IC(M) = \sum_{i=1}^4 \sum_{j=1}^m m_{i,j} \log \frac{m_{i,j}}{b_i}$$

□ **Maximum a Posteriori (MAP) Score**: →

$$MAP(M) = - \sum_{i=1}^4 \sum_{j=1}^m n_{i,j} \log \frac{m_{i,j}}{b_i}$$

□ **Model Vs Background Score**: →

$$L(w) = \frac{\Pr(w | M)}{\Pr(w | Bg)} = \prod_{j=1}^m \frac{m_{i,j}}{b_i}$$

Pattern Discovery in Protein Sequences

Motifs are combinations of secondary structures in proteins with a specific **structure** and a specific **function**. They are also called **super-secondary structures**.

Examples: Helix-Turn-Helix, Zinc-finger, Homeobox domain, Hairpin-beta motif, Calcium-binding motif, Beta-alpha-beta motif, Coiled-coil motifs.

Several motifs may combine to form **domains**.

- Serine proteinase domain, Kringle domain, calcium-binding domain, homeobox domain.

Motif Detection

Profile Method

● If many examples of the motif are known, then

➤ **Training**: build a **Profile** and compute a **threshold**

➤ **Testing**: **score** against profile

Combinatorial Pattern Discovery Methods

Gibbs Sampling

Expectation Method

HMM

How to evaluate these methods?

- Calculate TP, FP, TN, FN
- Compute **sensitivity** fraction of known sites predicted, **specificity**, and more.
 - **Sensitivity** = $TP / (TP + FN)$
 - **Specificity** = $TN / (TN + FN)$
 - **Positive Predictive Value** = $TP / (TP + FP)$
 - **Performance Coefficient** = $TP / (TP + FN + FP)$
 - **Correlation Coefficient** =

Supervised Pattern Discovery

□ Input: Alignment of known motifs, and
Query sequence

Output: Is the query sequence a motif?

● Profile Method [Gribskov et al., 1996]

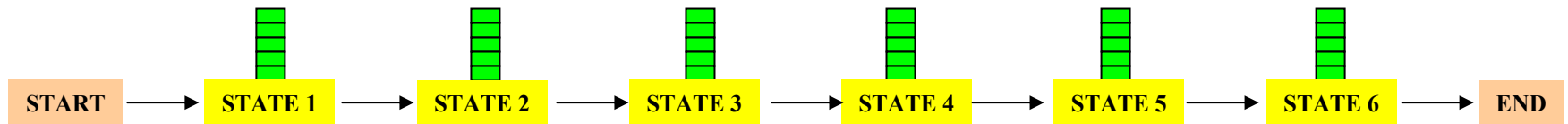
- Build a **profile** from the alignment and **score** query sequence against the profile to decide if it "fits the profile".
- Need to pick a **threshold** score.

● Enumerative/Combinatorial Methods

Profile HMMs

PROFILE METHOD, [M. Gribskov et al., '90]

Location in Seq.	Sequence						Protein Name
	1	2	3	4	5	6	
14	G	V	S	A	S	A	Ka RbtR
32	G	V	S	E	M	T	Ec DeoR
33	G	V	S	P	G	T	Ec RpoD
76	G	A	G	I	A	T	Ec TrpR
178	G	C	S	R	E	T	Ec CAP
205	C	L	S	P	S	R	Ec AraC
210	C	L	S	P	S	R	St AraC
13	G	V	N	K	E	T	Br MerR



Combinatorial Method: GYM

Pattern Generation:

Aligned Motif
Examples

Pattern Generator

Motif Detection:

New Protein
Sequence

Motif Detector

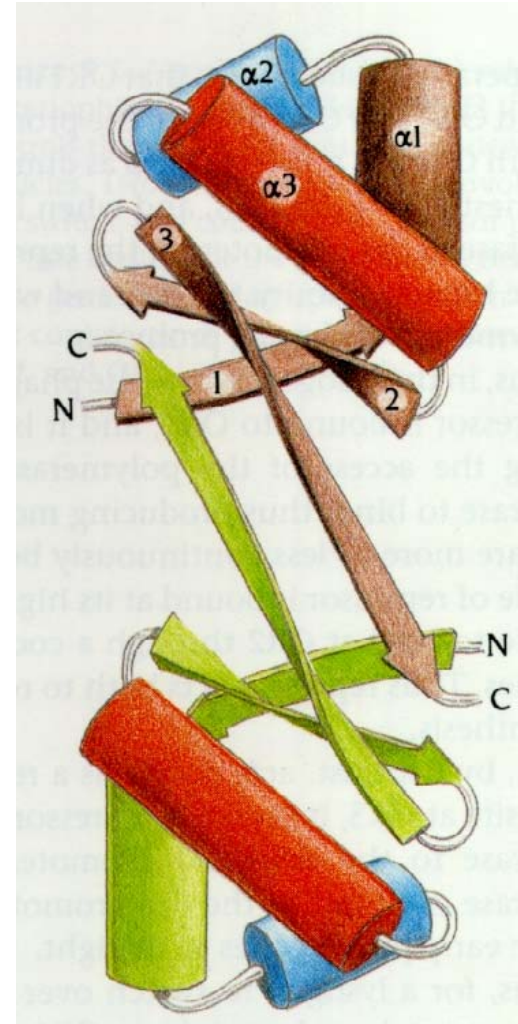
Detection
Results

Pattern
Dictionary

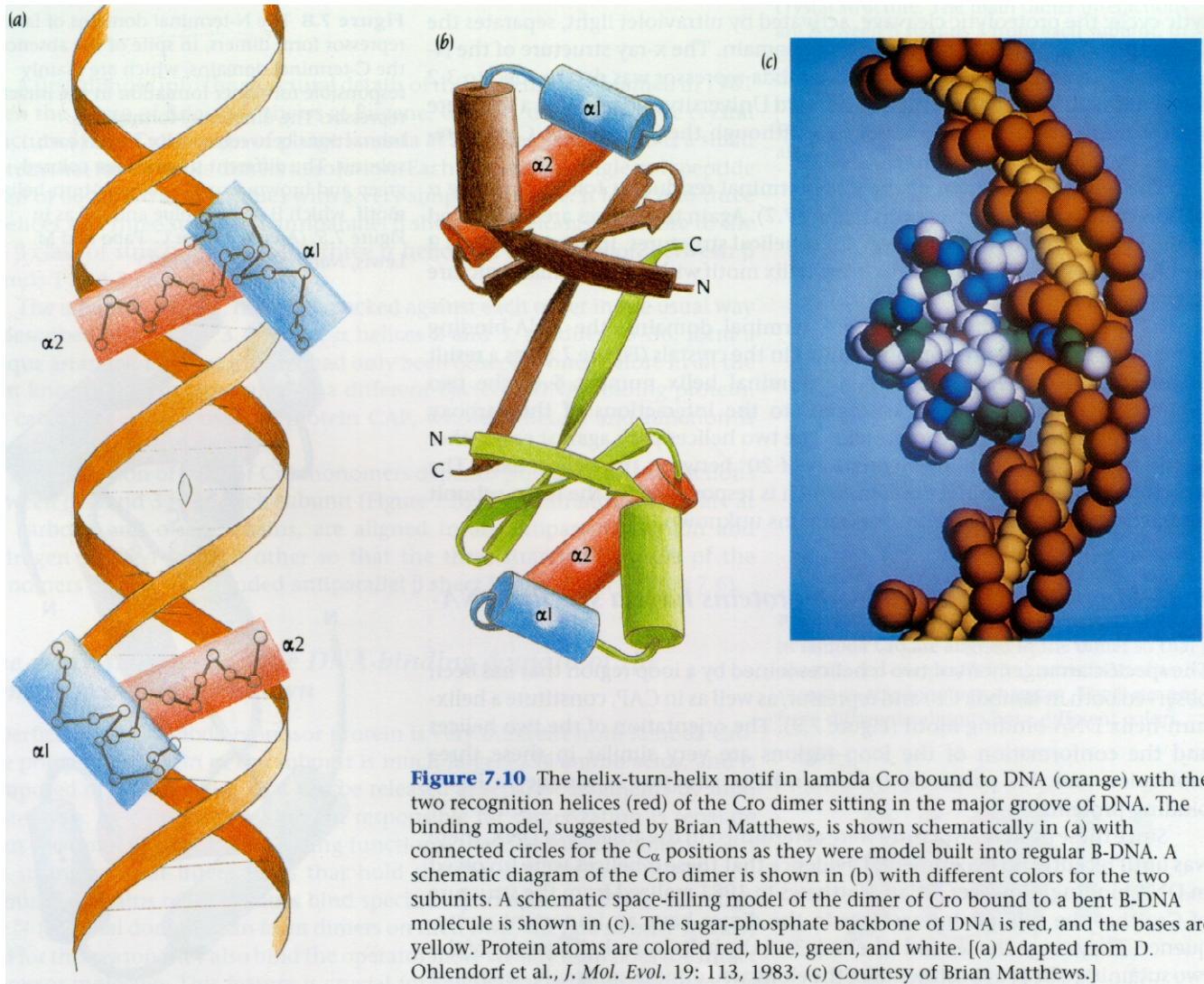
[Narasimhan, Bu, Wang, Xu, Yang, Mathee, J Comput Biol, 2002]

Helix-Turn-Helix Motifs

- Structure
 - 3-helix complex
 - Length: 22 amino acids
 - Turn angle
- Function
 - Gene regulation by binding to DNA



DNA Binding at HTH Motif



HTH Motifs: Examples

<i>Loc</i>	<i>Protein Name</i>	<i>Helix 2</i>									<i>Turn</i>				<i>Helix 3</i>								
		-1	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20
14	Cro	F	G	Q	E	K	T	A	K	D	L	G	V	Y	Q	S	A	I	N	K	A	I	H
16	434 Cro	M	T	Q	T	E	L	A	T	K	A	G	V	K	Q	Q	S	I	Q	L	I	E	A
11	P22 Cro	G	T	Q	R	A	V	A	K	A	L	G	I	S	D	A	A	V	S	Q	W	K	E
31	Rep	L	S	Q	E	S	V	A	D	K	M	G	M	G	Q	S	G	V	G	A	L	F	N
16	434 Rep	L	N	Q	A	E	L	A	Q	K	V	G	T	T	Q	Q	S	I	E	Q	L	E	N
19	P22 Rep	I	R	Q	A	A	L	G	K	M	V	G	V	S	N	V	A	I	S	Q	W	E	R
24	CII	L	G	T	E	K	T	A	E	A	V	G	V	D	K	S	Q	I	S	R	W	K	R
4	LacR	V	T	L	Y	D	V	A	E	Y	A	G	V	S	Y	Q	T	V	S	R	V	V	N
167	CAP	I	T	R	Q	E	I	G	Q	I	V	G	C	S	R	E	T	V	G	R	I	L	K
66	TrpR	M	S	Q	R	E	L	K	N	E	L	G	A	G	I	A	T	I	T	R	G	S	N
22	BlaA Pv	L	N	F	T	K	A	A	L	E	L	Y	V	T	Q	G	A	V	S	Q	Q	V	R
23	TrpI Ps	N	S	V	S	Q	A	A	E	Q	L	H	V	T	H	G	A	V	S	R	Q	L	K

Combinatorial Method: GYM

- **Combinations of residues** in specific locations (may not be contiguous) contribute towards stabilizing a structure.
- Some **reinforcing** combinations are relatively rare.
- GYM algorithm is inspired by the APriori algorithm [**Agrawal et al., 1996**]

[Narasimhan, Bu, Wang, Xu, Yang, Mathee, J Comput Biol, 2002]

Patterns

Loc	Protein Name	Helix 2									Turn				Helix 3								
		-1	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20
14	Cro	F	G	Q	E	K	T	A	K	D	L	G	V	Y	Q	S	A	I	N	K	A	I	H
16	434 Cro	M	T	Q	T	E	L	A	T	K	A	G	V	K	Q	Q	S	I	Q	L	I	E	A
11	P22 Cro	G	T	Q	R	A	V	A	K	A	L	G	I	S	D	A	A	V	S	Q	W	K	E
31	Rep	L	S	Q	E	S	V	A	D	K	M	G	M	G	Q	S	G	V	G	A	L	F	N
16	434 Rep	L	N	Q	A	E	L	A	Q	K	V	G	T	T	Q	Q	S	I	E	Q	L	E	N
19	P22 Rep	I	R	Q	A	A	L	G	K	M	V	G	V	S	N	V	A	I	S	Q	W	E	R
24	CII	L	G	T	E	K	T	A	E	A	V	G	V	D	K	S	Q	I	S	R	W	K	R
4	LacR	V	T	L	Y	D	V	A	E	Y	A	G	V	S	Y	Q	T	V	S	R	V	V	N
167	CAP	I	T	R	Q	E	I	G	Q	I	V	G	C	S	R	E	T	V	G	R	I	L	K
66	TrpR	M	S	Q	R	E	L	K	N	E	L	G	A	G	I	A	T	I	T	R	G	S	N
22	BlaA Pv	L	N	F	T	K	A	A	L	E	L	Y	V	T	Q	G	A	V	S	Q	Q	V	R
23	TrpI Ps	N	S	V	S	Q	A	A	E	Q	L	H	V	T	H	G	A	V	S	R	Q	L	K

- Q1 G9 N20
- A5 G9 V10 I15

Pattern Mining Algorithm

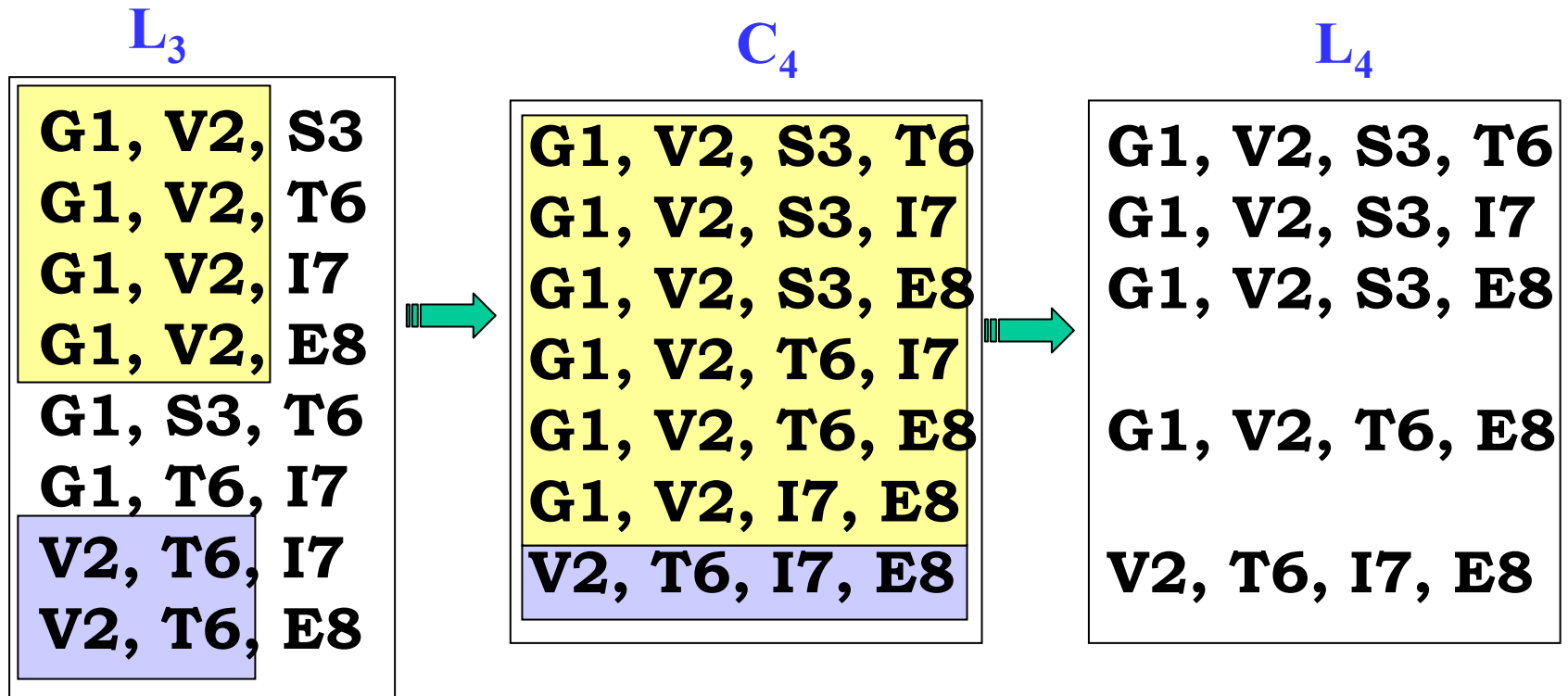
Algorithm **Pattern-Mining**

Input: Motif length m , support threshold T ,
list of aligned motifs M .

Output: Dictionary L of frequent patterns.

1. $L_1 :=$ All frequent patterns of length 1
2. **for** $i = 2$ **to** m **do**
3. $C_i :=$ **Candidates**(L_{i-1})
4. $L_i :=$ Frequent candidates from C_i
5. **if** ($|L_i| \leq 1$) **then**
6. **return** L as the union of all L_j , $j \leq i$.

Candidates Function



Motif Detection Algorithm

Algorithm **Motif-Detection**

Input : Motif length m ,
threshold score T ,
pattern dictionary L ,
and input protein sequence $P[1..n]$.

Output : Detected motif(s).

1. **for** each location i **do**
2. $S := \text{MatchScore}(P[i..i+m-1], L)$.
3. **if** $(S > T)$ **then**
4. Report it as a possible motif

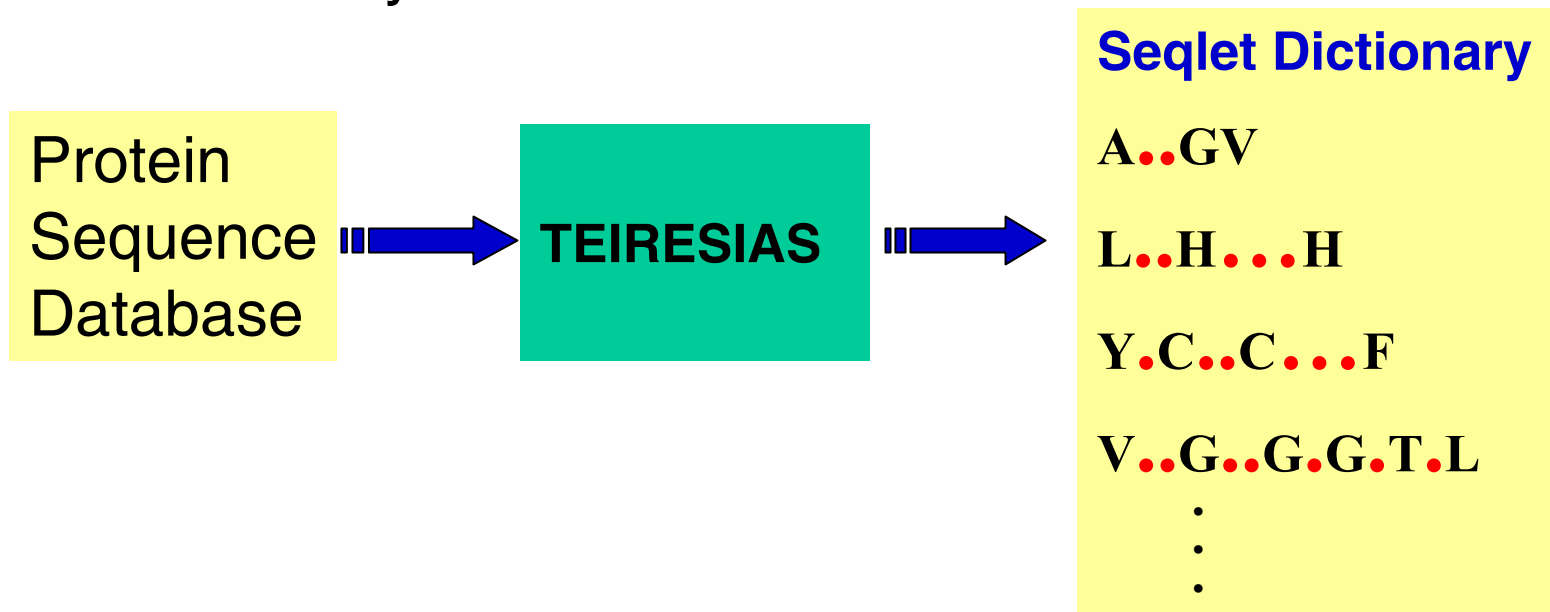
Experimental Results: GYM 2.0

<i>Motif</i>	<i>Protein Family</i>	<i>Number Tested</i>	<i>GYM = DE Agree</i>	<i>Number Annotated</i>	<i>GYM = Annot.</i>
<i>HTH Motif (22)</i>	Master	88	88 (100 %)	13	13
	Sigma	314	284 + 23 (98 %)	96	82
	Negates	93	86 (92 %)	0	0
	LysR	130	127 (98 %)	95	93
	AraC	68	57 (84 %)	41	34
	Rreg	116	99 (85 %)	57	46
	Total	675	653 + 23 (94 %)	289	255 (88 %)

Unaligned Pattern Discovery

TEIRESIAS:

The algorithm is similar to that used in GYM for aligned Pattern discovery.



Rigoutsos & Floratos, Bioinformatics, '98

TEIRESIAS: Key Features

- ❑ Starts with a set of seed patterns (Enumeration step)
- ❑ Convolution operator applied to all pairs of patterns:
$$A..GV.S \oplus V.S.GR = A..GV.S.GR$$
- ❑ Order of Evaluation carefully chosen so that long patterns get longer first
- ❑ Finds all maximal patterns.
- ❑ Combinatorial explosion avoided by generating only relevant maximal patterns.

Rigoutsos & Floratos, Bioinformatics, '98

SPLASH

- ❑ Structural Pattern Localization Analysis by Sequential Histogram (**SPLASH**)
- ❑ Not limited to fixed alphabet size
- ❑ Patterns are modeled by a homology metric and thus allow mismatches
- ❑ Early pruning of inconsistent seed patterns, leading to increased efficiency.
- ❑ Easily parallelized with availability of extra resources.

Califano, Bioinformatics, '00; **Califano** et al., J Comput Biol, '00

Precomputed Sequence Patterns

- ❑ PROSITE
- ❑ BLOCKS and PRINTS
- ❑ eMOTIF
- ❑ SPAT
- ❑ PRODOM
- ❑ Pfam

Motif Detection Tools

- PROSITE (Database of protein families & domains)
 - Try [PDOC00040](#). Also Try [PS00041](#)
- PRINTS [Sample Output](#)
- BLOCKS (multiply aligned ungapped segments for highly conserved regions of proteins; automatically created) [Sample Output](#)
- Pfam (Protein families database of alignments & HMMs)
 - Multiple Alignment, domain architectures, species distribution, links: [Try](#)
- MoST
- PROBE
- ProDom
- DIP

Protein Information Sites

❑ **SwissPROT & GenBank**

❑ **InterPRO** is a database of protein families, domains and functional sites in which identifiable features found in known proteins can be applied to unknown protein sequences. [See sample.](#)

❑ **PIR** [Sample Protein page](#)

Modular Nature of Proteins

- Proteins are collections of “modular” domains. For example,

Coagulation Factor XII



PLAT

Domain Architecture Tools

□ CDART

- Protein [AAH24495](#); [Domain Architecture](#);
- It's [domain relatives](#);
- Multiple [alignment](#) for 2nd domain

□ SMART

Predicting Specialized Structures

- ❑ COILS - Predicts coiled coil motifs
- ❑ TMPred - predicts transmembrane regions
- ❑ SignalP - predicts signal peptides
- ❑ SEG - predicts nonglobular regions

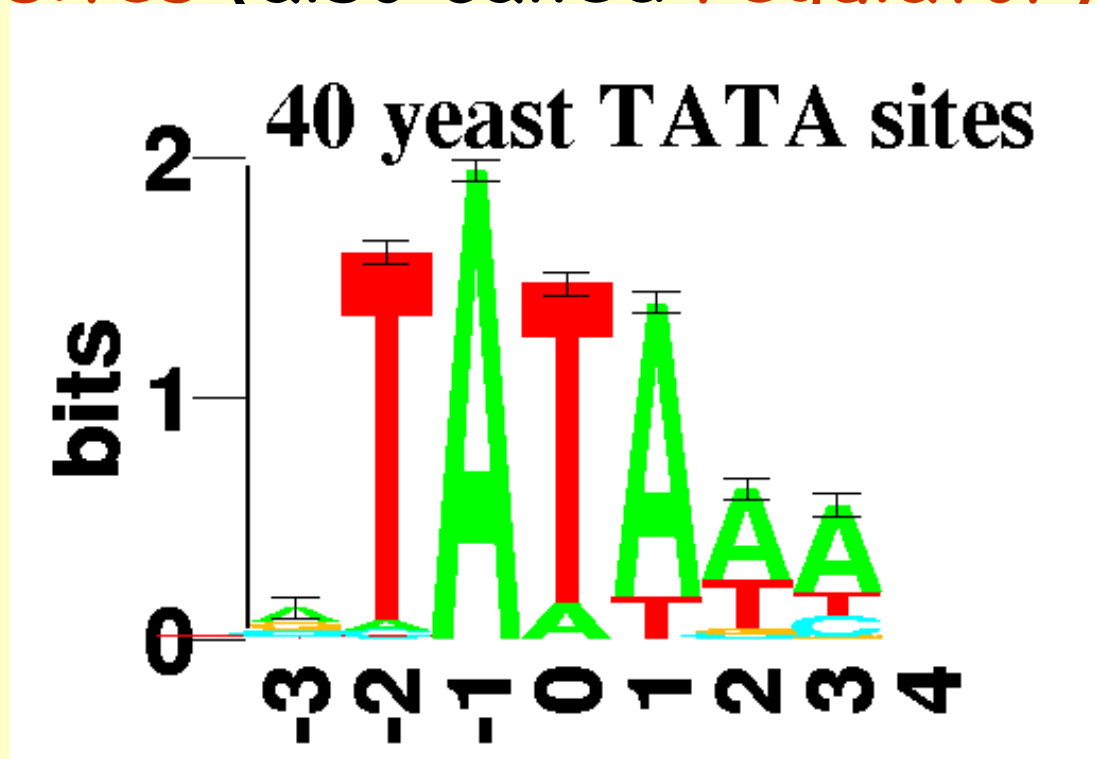
Patterns in DNA Sequences

- Signals in DNA sequence control events
 - Start and end of genes
 - Start and end of introns
 - Transcription factor binding sites (regulatory elements)
 - Ribosome binding sites
- Detection of these patterns are useful for
 - Understanding gene structure
 - Understanding gene regulation

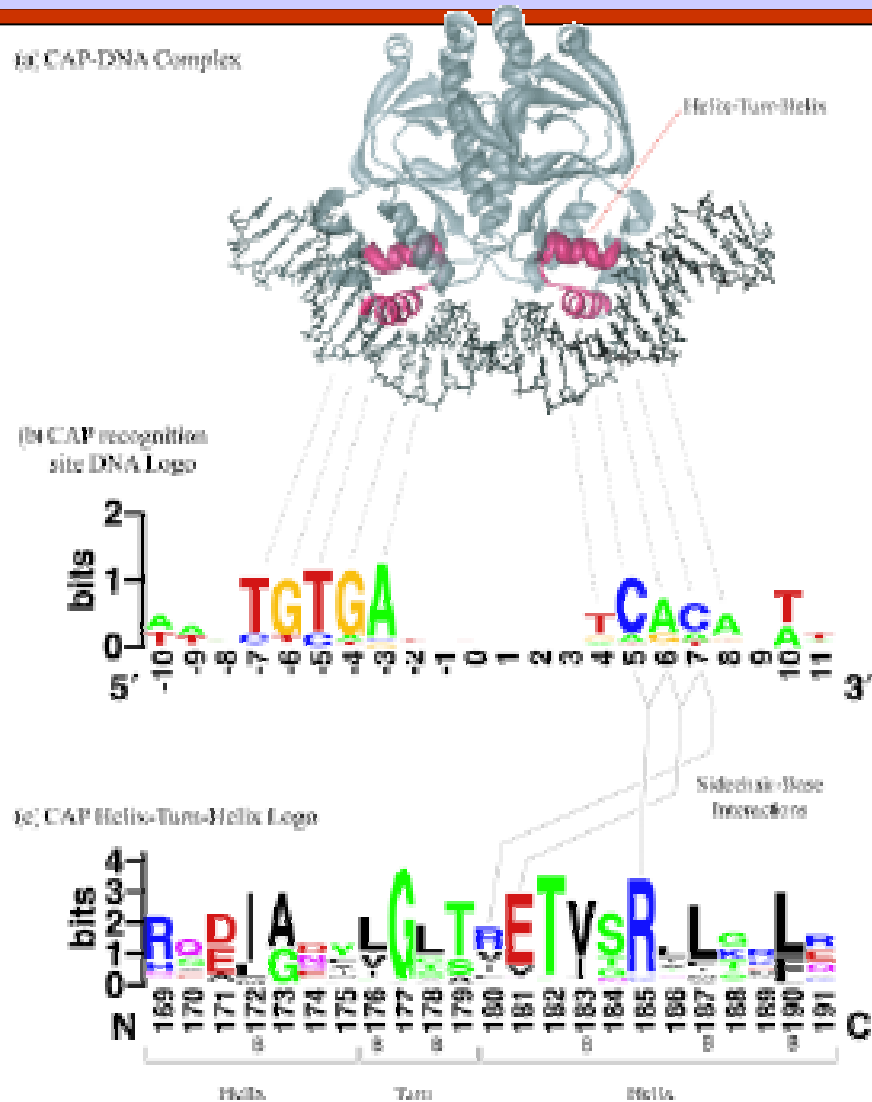
Motifs in DNA Sequences

□ Given a collection of DNA sequences of promoter regions, locate the transcription factor binding sites (also called regulatory elements)

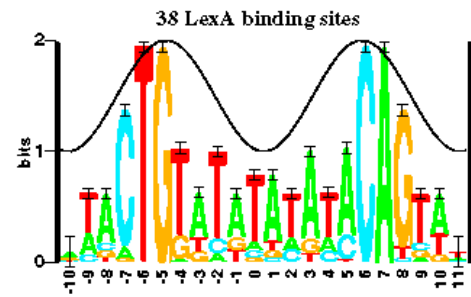
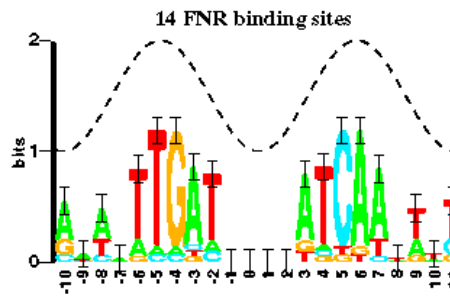
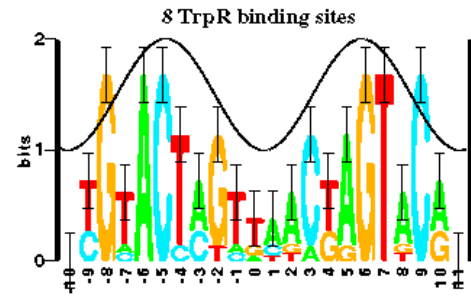
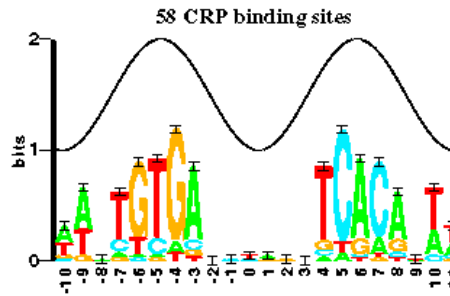
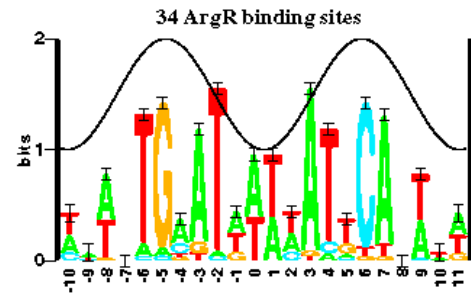
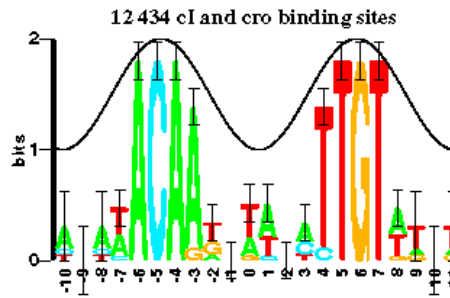
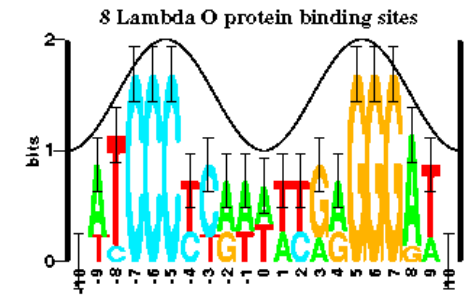
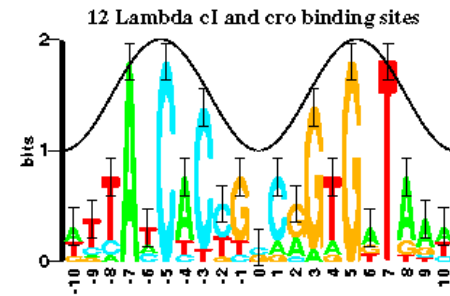
● Example:



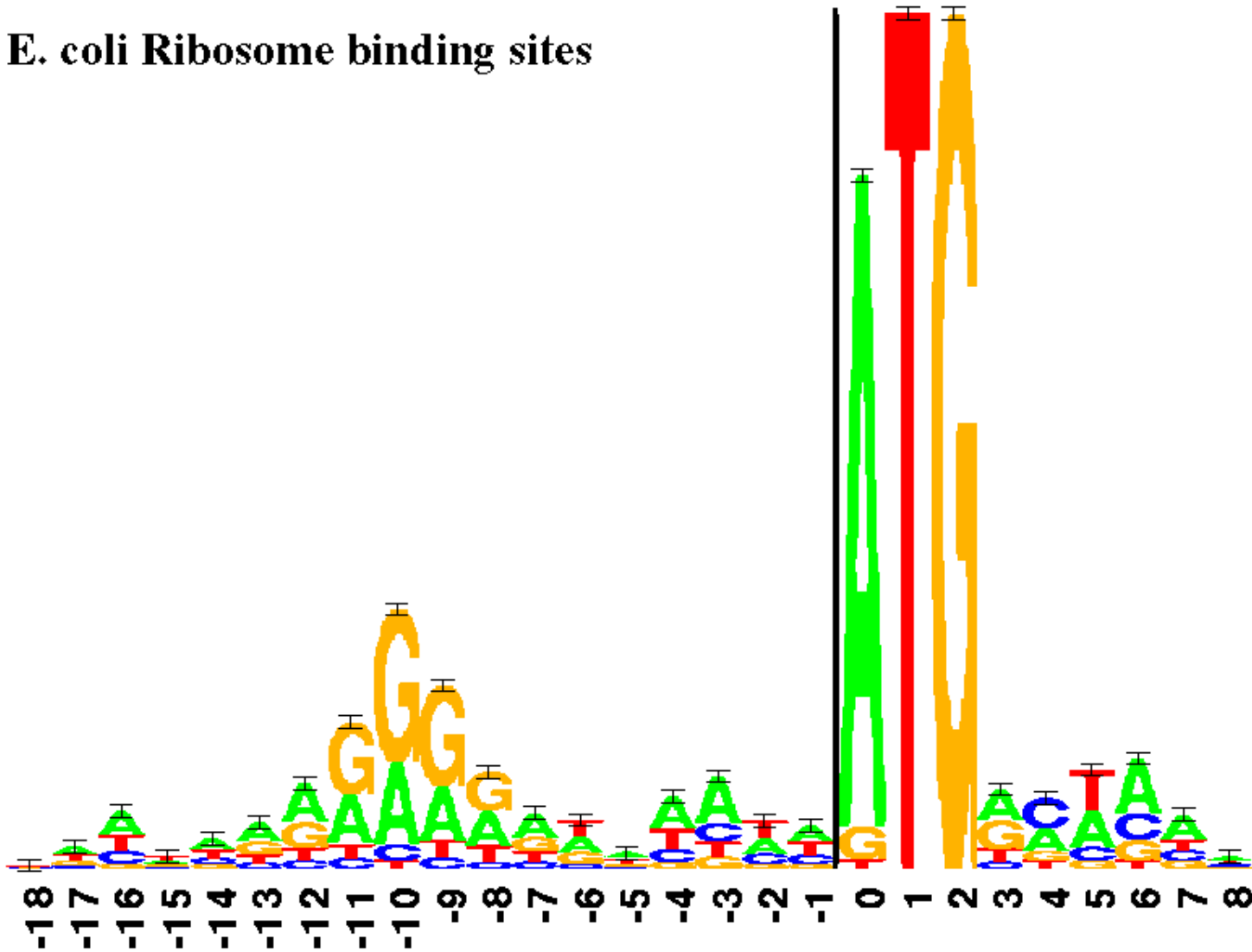
Motifs



More Motifs in *E. Coli* DNA Sequences



E. coli Ribosome binding sites



Motifs in DNA Sequences

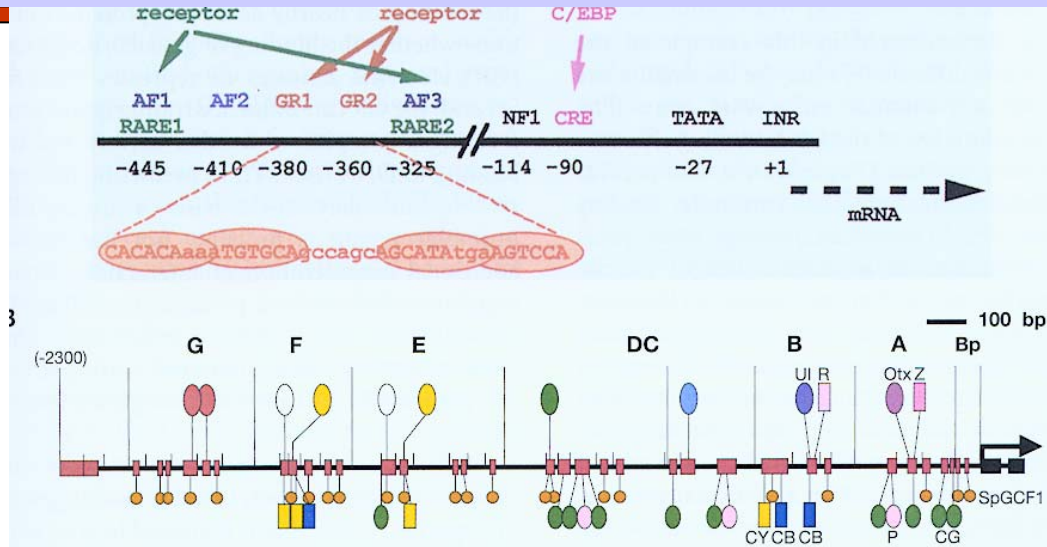


FIGURE 9.13. Regulatory elements of two promoters. (A) The rat *pepCK* gene. The relative positions of the TF-binding sites are illustrated (Yamada et al. 1999). The glucocorticoid response unit (GRU) includes three accessory factor-binding sites (AF1, AF2, and AF3), two glucocorticoid response elements (GR1 and GR2), and a cAMP response element (CRE). A dimer of glucocorticoid receptors bound to each GR element is depicted. The retinoic response unit (RAU) includes two retinoic acid response elements (RARE1 and RARE2) that coincide with the AF1 and AF3, respectively (Sugiyama et al. 1998). The sequences of the two GR sites and the binding of the receptor to these sites are shown. These sites deviate from the consensus sites and depend on their activity on accessory proteins bound to other sites in the GRU. This dependence on accessory proteins is reduced if a more consensus-like (canonical) GR element comprising the sequence 'TGT'TCT' is present. The CRE that binds factor C/EBP is also shown. (B) The 2300-bp promoter of the developmentally regulated gene *endo16* of the sea urchin (Bolouri and Davidson 2002). Different colors indicate different binding sites for distinct proteins and proteins shown above the line bind at unique locations, below the line at several locations. The regions A–G are functional modules that determine the expression of the gene in a particular tissue at a particular time of development and may either serve to induce transcription of the gene as a necessary developmental step (A, B, and G) or repress transcription (C–F) in tissues when it is not appropriate. (Reprinted, with permission, from Bolouri and Davidson 2002 [©2002 Elsevier].)

Motif Detection (TFBMs)

- See evaluation by Tompa et al.
 - [bio.cs.washington.edu/assessment]
- Gibbs Sampling Methods: AlignACE, GLAM, SeSiMCMC, MotifSampler
- Weight Matrix Methods: ANN-Spec, Consensus,
- EM: Improbizer, MEME
- Combinatorial & Misc.: MITRA, oligo/dyad, QuickScore, Weeder, YMF

Gibbs Sampling for Motif Detection

