

*Regulatory sequence analysis*

# ***Position-specific scoring matrices (PSSM)***

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# Alignment of transcription factor binding sites

## Binding sites for the yeast Pho4p transcription factor

(Source : Oshima et al. Gene 179, 1996; 171-177)

Gene	Site Name	Sequence	Affinity
PHO5	UASp2	---aCtCaCACAGTGGACTAGC-	high
PHO84	Site D	---TTTCCAACACAGTGGGCGGA--	high
PHO81	UAS	----TTATGGCACGTGCGAATAA--	high
PHO8	Proximal	GTGATCGCTGCACGTGGCCCGA---	high
PHO5	UASp3	--TAATTTGGCATGTGCGATCTC--	low
PHO84	Site C	-----ACGTCACAGTGAACATAT--	low
PHO84	Site A	-----TTTATCACGTGACACTTTTT	low
group 1	consensus	-----gCACGTGggac-----	high-low
PHO5	UASp1	--TAAATTAGCACGTTTTCGC----	medium
PHO84	Site E	-----AATACGCACGTTTTTAATCTA	medium
PHO84	Site B	-----TTACGCACGTTGGTGCTG--	low
PHO8	Distal	---TTACCCGCACGCTTAATAT---	low
group 2	consensus	-----cgCACGTTt-----	med-low
Degenerate consensus		-----GCACGTTKKk-----	

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# ***From alignments to weights***

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## Count matrix

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

Binding site for the yeast Pho4p transcription factor  
(Source : Transfac matrix F\$PHO4\_01)

$n_{i,j}$       occurrences of residue  $i$  at position  $j$

# Frequency matrix

Pos	1	2	3	4	5	6	7	8	9	10	11	12
A	0.13	0.38	0.25	0.00	<b>1.00</b>	0.00	0.00	0.00	0.00	0.00	0.13	0.25
C	0.25	0.25	<b>0.38</b>	<b>1.00</b>	0.00	<b>1.00</b>	0.00	0.00	0.00	0.25	0.00	0.25
G	0.13	0.25	<b>0.38</b>	0.00	0.00	0.00	<b>1.00</b>	0.00	<b>0.63</b>	<b>0.50</b>	<b>0.63</b>	0.25
T	0.50	0.13	0.00	0.00	0.00	0.00	0.00	<b>1.00</b>	0.38	0.25	0.25	0.25
Sum	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00

$$f_{i,j} = \frac{n_{i,j}}{\sum_{i=1}^A n_{i,j}}$$

$A$  alphabet size (=4)

$n_{i,j}$  occurrences of residue  $i$  at position  $j$

$p_i$  prior residue probability for residue  $i$

$f_{i,j}$  relative frequency of residue  $i$  at position  $j$

# Corrected frequency matrix

Pos	1	2	3	4	5	6	7	8	9	10	11	12
A	0.15	0.37	0.26	0.04	<b>0.93</b>	0.04	0.04	0.04	0.04	0.04	0.15	0.26
C	0.24	0.24	<b>0.35</b>	<b>0.91</b>	0.02	<b>0.91</b>	0.02	0.02	0.02	0.24	0.02	0.24
G	0.13	0.24	<b>0.35</b>	0.02	0.02	0.02	<b>0.91</b>	0.02	<b>0.58</b>	<b>0.46</b>	<b>0.58</b>	0.24
T	0.48	0.15	0.04	0.04	0.04	0.04	0.04	<b>0.93</b>	0.37	0.26	0.26	0.26
Sum	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00

*1st option: identically distributed pseudo-weight*

$$f'_{i,j} = \frac{n_{i,j} + k/A}{\sum_{i=1}^A n_{i,j} + k}$$

*2nd option: pseudo-weight distributed according to residue priors*

$$f'_{i,j} = \frac{n_{i,j} + p_i k}{\sum_{i=1}^A n_{i,j} + k}$$

*A* alphabet size (=4)  
*n<sub>i,j</sub>* occurrences of residue *i* at position *j*  
*p<sub>i</sub>* prior residue probability for residue *i*  
*f<sub>i,j</sub>* relative frequency of residue *i* at position *j*  
*k* pseudo weight (arbitrary, 1 in this case)  
*f'<sub>i,j</sub>* corrected frequency of residue *i* at position *j*

Reference: Hertz (1999). Bioinformatics 15:563-577.

## Probability of a sequence segment under the matrix model

Pos	1	2	3	4	5	6	7	8	9	10	11	12
A	0.15	0.37	0.26	0.04	0.93	0.04	0.04	0.04	0.04	0.04	0.15	0.26
C	0.24	0.24	0.35	0.91	0.02	0.91	0.02	0.02	0.02	0.24	0.02	0.24
G	0.13	0.24	0.35	0.02	0.02	0.02	0.91	0.02	0.58	0.46	0.58	0.24
T	0.48	0.15	0.04	0.04	0.04	0.04	0.04	0.93	0.37	0.26	0.26	0.26

Sequence S      A    T    G    C    G    T    A    A    A    G    C    T

P(res)    0.15   0.15   0.35   0.91   0.02   0.04   0.04   0.04   0.04   0.46   0.02   0.26

P(S|M) 5.32E-13

$$P(S|M) = \prod_{j=1}^w f'_{r_j j}$$

- n Let
  - q  $M$  be a frequency matrix of width  $w$
  - q  $S = \{r_1, r_2, \dots, r_w\}$  be a sequence segment of length  $w$  (same length as the matrix)
  - q  $r_j$  is the residue found at position  $j$  of the sequence segment  $S$ .
- n The corrected frequencies  $F'_{ij}$  can be used to estimate the probability to observe residue  $i$  at position  $j$  of the motif described by the matrix
- n The probability to generate the sequence segment  $S$  under the model described by the matrix  $M$  is the product of the frequencies of residues at the corresponding columns of the matrix.

## Probability of the best sequence segment under the matrix model

Pos	1	2	3	4	5	6	7	8	9	10	11	12
A	0.15	0.37	0.26	0.04	0.93	0.04	0.04	0.04	0.04	0.04	0.15	0.26
C	0.24	0.24	0.35	0.91	0.02	0.91	0.02	0.02	0.02	0.24	0.02	0.24
G	0.13	0.24	0.35	0.02	0.02	0.02	0.91	0.02	0.58	0.46	0.58	0.24
T	<b>0.48</b>	0.15	0.04	0.04	0.04	0.04	0.04	0.93	0.37	0.26	0.26	0.26

Sequence S

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

P(res) 0.48 0.37 0.35 0.91 0.93 0.91 0.91 0.93 0.58 0.46 0.58 0.26

P(S|M) 1.59E-03

- n This segment of sequence is associated to the highest possible probability given the matrix : P(S|M)
- n Each nucleotide of the sequence corresponds to the residue with the highest probability in the corresponding column of the matrix.

$$P(S|M) = \prod_{j=1}^w f'_{r_j j}$$



## Probability of a sequence segment under the background model

Pos	Prior
A	0.325
C	0.175
G	0.175
T	0.325

Sequence S      A      T      G      C      G      T      A      A      A      G      C      T

P(res) 0.325 0.325 0.175 0.175 0.175 0.325 0.325 0.325 0.325 0.175 0.175 0.325

P(S|B) 6.29E-08

- n A background model ( $B$ ) should be defined to estimate the probability of a sequence motif outside of the motif.
- n Various possibilities can be envisaged to define the background model
  - q Bernoulli model with equiprobable residues (this should generally be avoided, because most biological sequences are biased towards some residues)
  - q Bernoulli model with residue-specific probabilities ( $p_r$ )
  - q Markov chains
- n Under a Bernoulli model, the probability of a sequence motif  $S$  is the probability of the prior frequencies of its residues  $r_j$ .

$$P(S | B) = \prod_{j=1}^w p_{r_j}$$

# Weight of a sequence segment

Pos	1	2	3	4	5	6	7	8	9	10	11	12
A	-0.79	0.13	-0.23	-2.20	1.05	-2.20	-2.20	-2.20	-2.20	-2.20	-0.79	-0.23
C	0.32	0.32	0.70	1.65	-2.20	1.65	-2.20	-2.20	-2.20	0.32	-2.20	0.32
G	-0.29	0.32	0.70	-2.20	-2.20	-2.20	1.65	-2.20	1.19	0.97	1.19	0.32
T	0.39	-0.79	-2.20	-2.20	-2.20	-2.20	-2.20	1.05	0.13	-0.23	-0.23	-0.23
residue r	A	T	G	C	G	T	A	A	A	G	C	T
W(r)	-0.79	-0.79	0.70	1.65	-2.20	-2.20	-2.20	-2.20	-2.20	0.97	-2.20	-0.23
Weight	-11.67 =SUM[W(r)]											

$$W_S = \ln \left( \frac{P(S|M)}{P(S|B)} \right) = \ln \left( \frac{\prod_{j=1}^w f'_{r_j j}}{\prod_{j=1}^w p_{r_j}} \right) = \ln \left( \prod_{j=1}^w \frac{f'_{r_j j}}{p_{r_j}} \right) = \sum_{j=1}^w \ln \left( \frac{f'_{r_j j}}{p_{r_j}} \right) = \sum_{j=1}^w W_{r_j j}$$

$W_S$	weight of sequence segment $S$
$P(S M)$	probability of the sequence segment, given the matrix
$P(S B)$	probability of the sequence segment, given the background
$j$	position within the segment and within the matrix
$r_j$	residue at position $j$ of the sequence segment
$p_{r_j}$	prior probability of residue $r_j$
$f'_{r_j j}$	probability of residue $r_j$ at position $j$ of the matrix

- n The **weight** of a sequence segment is defined as the log-ratio between
  - q  $P(S|M)$ , the sequence probability under the model described by the PSSM, and
  - q  $P(S|B)$ , the sequence probability under the background model.
- n The weight represents the likelihood that this segment is an occurrence of the motif rather than being issued from the background model.
- n The weight matrix  $W_{ij}$  allows to easily calculate segment weights.

# Weight matrix (Bernoulli model)

Prior	Pos	1	2	3	4	5	6	7	8	9	10	11	12
0.325	A	-0.79	0.13	-0.23	-2.20	1.05	-2.20	-2.20	-2.20	-2.20	-2.20	-0.79	-0.23
0.175	C	0.32	0.32	0.70	1.65	-2.20	1.65	-2.20	-2.20	-2.20	0.32	-2.20	0.32
0.175	G	-0.29	0.32	0.70	-2.20	-2.20	-2.20	1.65	-2.20	1.19	0.97	1.19	0.32
0.325	T	0.39	-0.79	-2.20	-2.20	-2.20	-2.20	-2.20	1.05	0.13	-0.23	-0.23	-0.23
1.000	Sum	-0.37	-0.02	-1.02	-4.94	-5.55	-4.94	-4.94	-5.55	-3.08	-1.13	-2.03	0.19

$$f'_{i,j} = \frac{n_{i,j} + p_i k}{\sum_{r=1}^A n_{r,j} + k}$$

$$W_{i,j} = \ln \left( \frac{f'_{i,j}}{p_i} \right)$$

- A* alphabet size (=4)  
*n<sub>i,j</sub>* occurrences of residue *i* at position *j*  
*p<sub>i</sub>* prior residue probability for residue *i*  
*f<sub>i,j</sub>* relative frequency of residue *i* at position *j*  
*k* pseudo weight (arbitrary, 1 in this case)  
*f'<sub>i,j</sub>* corrected frequency of residue *i* at position *j*  
*W<sub>i,j</sub>* weight of residue *i* at position *j*

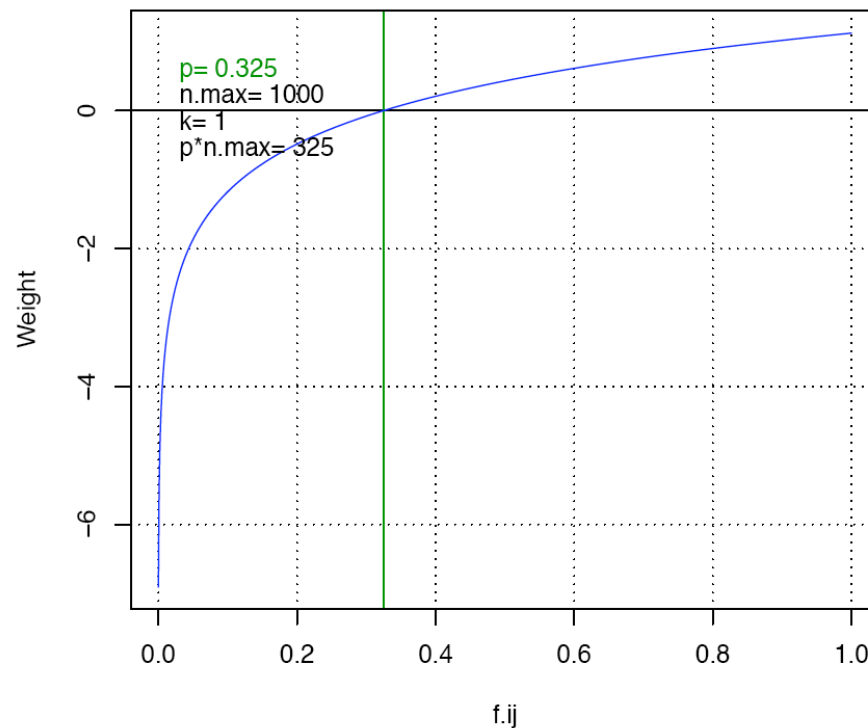
## Bernoulli assumption

If we assume, for the background model, an independent succession of nucleotides (Bernoulli model), the weight  $W_S$  of a sequence segment *S* is simply the sum of weights of the nucleotides at successive positions of the matrix ( $W_{i,j}$ ).

In this case, it is convenient to convert the PSSM into a weight matrix, which can then be used to assign a score to each position of a given sequence.

# Properties of the weight function

$$W_{i,j} = \ln\left(\frac{f'_{i,j}}{p_i}\right) \quad f'_{i,j} = \frac{n_{i,j} + p_i k}{\sum_{i=1}^A n_{i,j} + k} \quad \sum_{i=1}^A f'_{i,j} = 1$$



- n The weight is
  - q *positive* when  $f'_{i,j} > p_i$   
(*favourable* positions for the binding of the transcription factor)
  - q *negative* when  $f'_{i,j} < p_i$   
(*unfavourable* positions)

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# ***Information content***

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# Shannon uncertainty

- n Shannon uncertainty
  - q  $H_s(j)$ : uncertainty of a column of a PSSM
  - q  $H_g$ : uncertainty of the background (e.g. a genome)
- n Properties of the uncertainty (for a 4 letter alphabet)
  - q  $\min(H)=0$ 
    - No uncertainty at all: the nucleotide is completely specified (e.g.  $p=\{1,0,0,0\}$ )
  - q  $H=1$ 
    - Uncertainty between two letters (e.g.  $p=\{0.5,0,0,0.5\}$ )
  - q  $\max(H) = 2$  (Complete uncertainty)
    - One bit of information is required to specify the choice between each alternative (e.g.  $p=\{0.25,0.25,0.25,0.25\}$ ).
    - Two bits are required to specify a letter in a 4-letter alphabet.
- n Schneider (1986) defines an information content  $R_{seq}^*$  based on Shannon's uncertainty.

$$H_s(j) = - \sum_{i=1}^A f_{i,j} \log_2(f_{i,j})$$

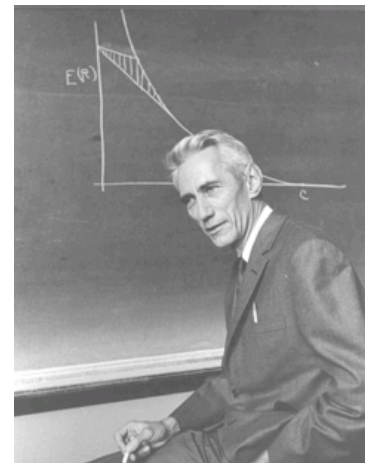
$$H_g = - \sum_{i=1}^A p_i \log_2(p_i)$$

$$R_{seq}(j) = H_g - H_s(j)$$

$$R_{seq} = \sum_{j=1}^w R_{seq}(j)$$

$$R_{seq}^*(j) = \sum_{i=1}^A f_{i,j} \log_2\left(\frac{f_{i,j}}{p_i}\right)$$

$$R_{seq}^* = \sum_{j=1}^w R_{seq}^*(j)$$



*Adapted from Schneider (1986)*

# Information content

Prior	Pos	1	2	3	4	5	6	7	8	9	10	11	12
0.325	A	-0.12	0.05	-0.06	-0.08	0.97	-0.08	-0.08	-0.08	-0.08	-0.08	-0.12	-0.06
0.175	C	0.08	0.08	0.25	1.50	-0.04	1.50	-0.04	-0.04	-0.04	0.08	-0.04	0.08
0.175	G	-0.04	0.08	0.25	-0.04	-0.04	-0.04	1.50	-0.04	0.68	0.45	0.68	0.08
0.325	T	0.19	-0.12	-0.08	-0.08	-0.08	-0.08	-0.08	0.97	0.05	-0.06	-0.06	-0.06
1.000	Sum	0.11	0.09	0.36	1.29	0.80	1.29	1.29	0.80	0.61	0.39	0.47	0.04

$$f'_{i,j} = \frac{n_{i,j} + p_i k}{\sum_{i=1}^A n_{i,j} + k}$$

$$I_{i,j} = f'_{i,j} \ln \left( \frac{f'_{i,j}}{p_i} \right)$$

$$I_j = \sum_{i=1}^A I_{i,j}$$

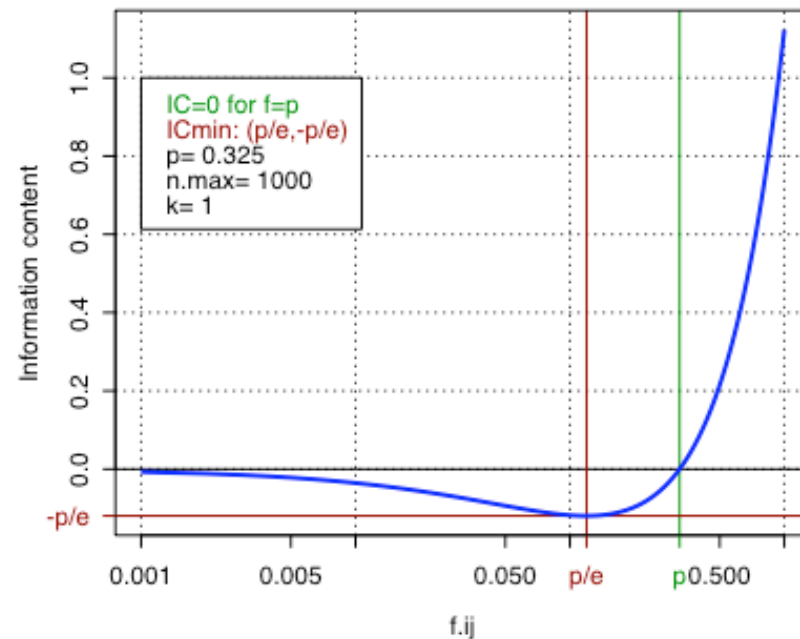
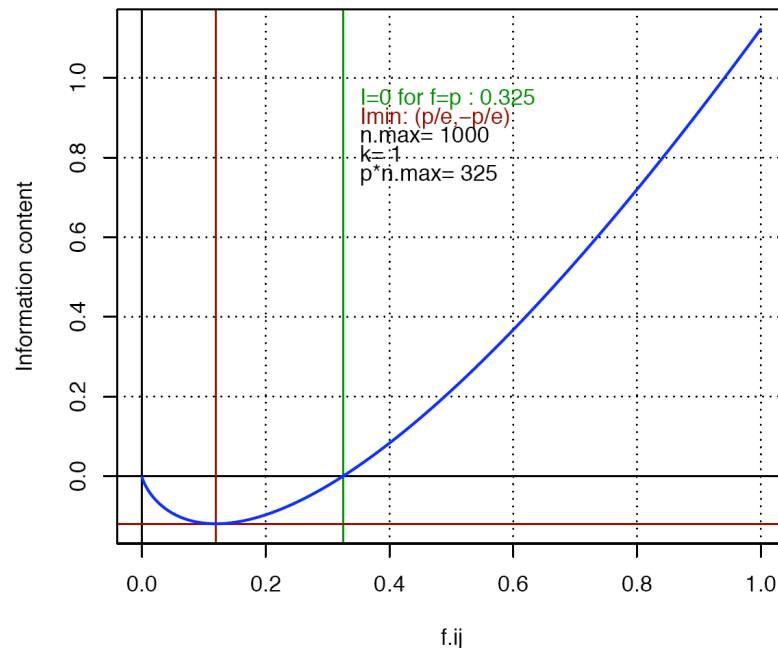
$$I_{matrix} = \sum_{j=1}^w \sum_{i=1}^A I_{i,j}$$

- $A$  alphabet size (=4)
- $n_{i,j}$  occurrences of residue  $i$  at position  $j$
- $w$  matrix width (=12)
- $p_i$  prior residue probability for residue  $i$
- $f_{i,j}$  relative frequency of residue  $i$  at position  $j$
- $k$  pseudo weight (arbitrary, 1 in this case)
- $f'_{i,j}$  corrected frequency of residue  $i$  at position  $j$
- $W_{i,j}$  weight of residue  $i$  at position  $j$
- $I_{i,j}$  information of residue  $i$  at position  $j$

Reference: Hertz (1999).  
Bioinformatics 15:563-577.

# Information content $I_{ij}$ of a cell of the matrix

- n For a given cell of the matrix
  - q  $I_{ij}$  is positive when  $f'_{ij} > p_i$   
(i.e. when residue  $i$  is more frequent at position  $j$  than expected by chance)
  - q  $I_{ij}$  is negative when  $f'_{ij} < p_i$
  - q  $I_{ij}$  tends towards 0 when  $f'_{ij} \rightarrow 0$  (because  $\lim_{x \rightarrow 0} x \ln(x) = 0$ )





# Information content of a column of the matrix

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- n For a given column  $i$  of the matrix
- q The information of the column ( $I_j$ ) is the sum of information of its cells.
  - q  $I_j$  is always positive
  - q  $I_j$  is always positive
  - q  $I_j$  is 0 when the frequency of all residues equal their prior probability ( $f_{ij}=p_i$ )
  - q  $I_j$  is maximal when
    - the residue  $i_m$  with the lowest prior probability has a frequency of 1 (all other residues have a frequency of 0)
    - and the pseudo-weight is 0

$$I_j = \sum_{i=1}^A I_{i,j} = \sum_{i=1}^A f'_{i,j} \ln \left( \frac{f'_{i,j}}{p_i} \right)$$

$$i_m = \arg \min_i (p_i) \quad k = 0$$
$$\max(I_j) = 1 * \ln \left( \frac{1}{p_i} \right) = -\ln(p_i)$$

# Information content of the matrix

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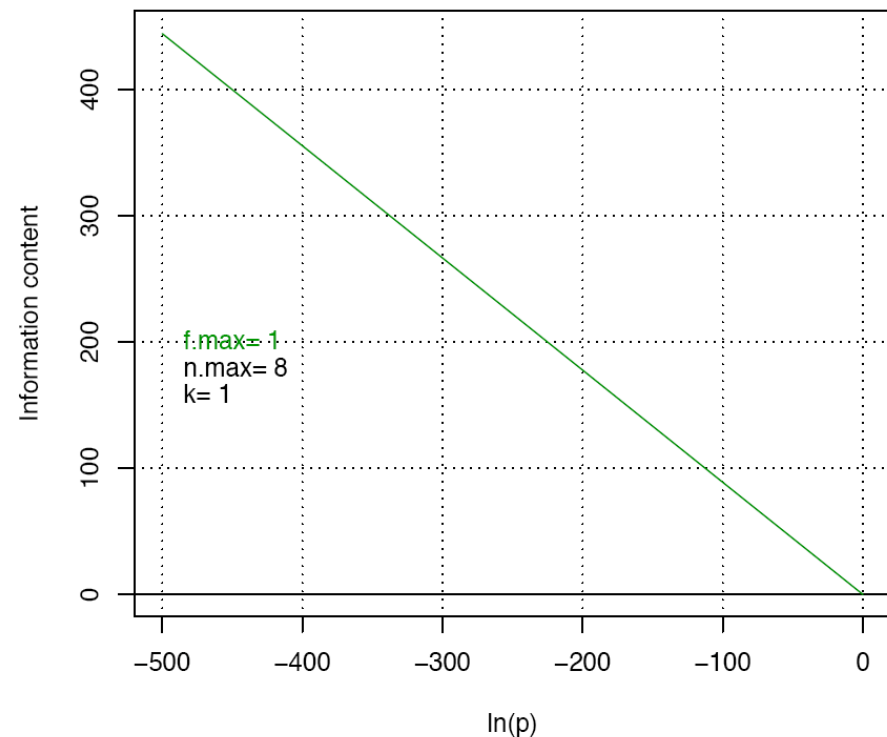
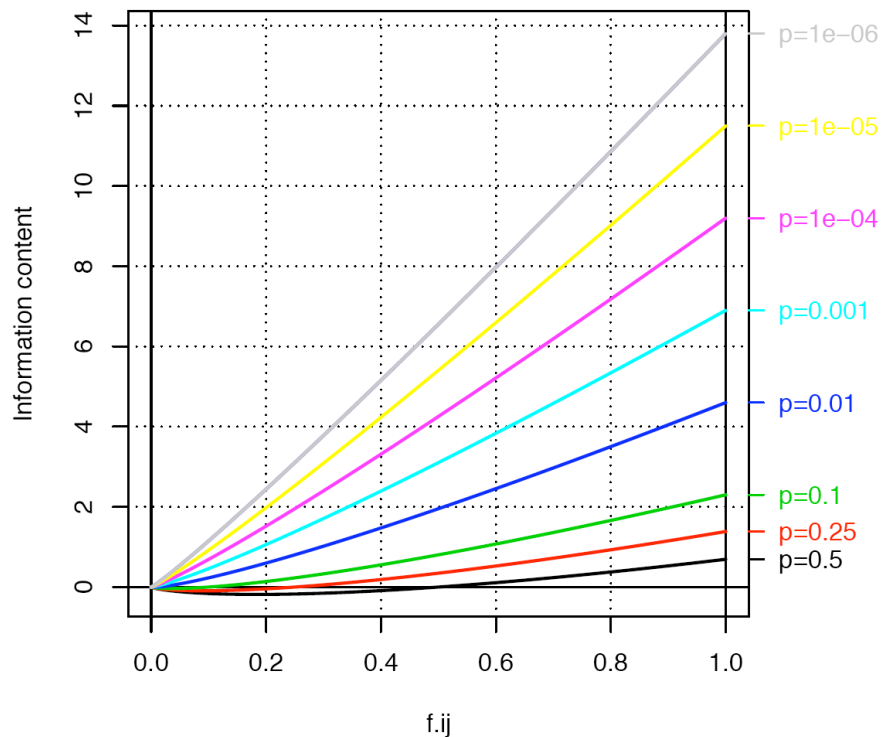
- n The total information content represents the capability of the matrix to make the distinction between a binding site (represented by the matrix) and the background model.
- n The information content also allows to estimate an upper limit for the expected frequency of the binding sites in random sequences.
- n The pattern discovery program *consensus* (developed by Jerry Hertz) optimises the information content in order to detect over-represented motifs.
- n Note that this is not the case of all pattern discovery programs: the gibbs sampler algorithm optimizes a log-likelihood.

$$I_{matrix} = \sum_{j=1}^w \sum_{i=1}^A I_{i,j}$$

$$P(site) \leq e^{-I_{matrix}}$$

# Information content: effect of prior probabilities

- n The upper bound of  $I_j$  increases when  $p_i$  decreases
  - q  $I_j \rightarrow \text{Inf}$  when  $p_i \rightarrow 0$
- n The information content, as defined by Gerald Hertz, has thus no upper bound.



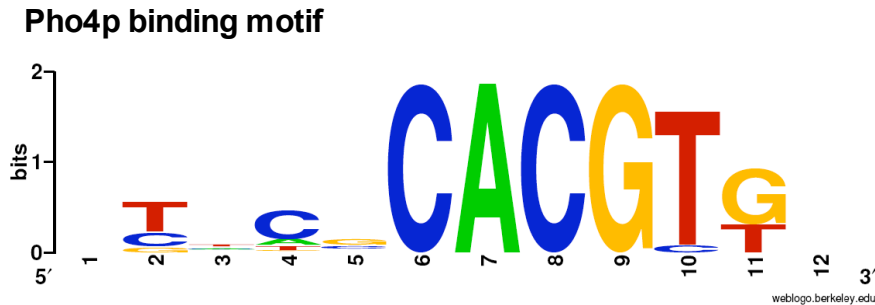
*Regulatory sequence analysis*

# ***Sequence logos***

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# Schneider logos

$$H_s(j) = - \sum_{i=1}^A f_{ij} \log_2(f_{ij})$$
$$R_{seq}(j) = 2 - H_s(j) + e(n)$$
$$h_{ij} = f_{ij} R_{seq}(j)$$



- n Schneider (1990) proposes a graphical representation based on his previous entropy (H) for representing the importance of each residue at each position of an alignment. He provides a new formula for  $R_{seq}$ 
  - q  $H_s(j)$  uncertainty of column  $j$
  - q  $R_{seq}(j)$  information content of column  $j$
  - q  $e(n)$  correction for small samples (pseudo-weight)
- n Remarks
  - q This information content does not include any correction for the prior residue probabilities ( $p_i$ )
  - q This information content is expressed in bits.
- n Boundaries
  - q  $\min(R_{seq})=0$  equiprobable residues
  - q  $\max(R_{seq})=2$  perfect conservation of 1 residue, all the others are forbidden
- n Sequence logos can be generated from aligned sequences on the *Weblogo* server
  - q <http://weblogo.berkeley.edu/>

## Sequence logo

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A T G T A T G G

Rap1

G G T G G C A A A A

Rpn4

A A A T G A G T C A

Gcn4

G A A T T C A G A A

HSE

T G G G G G T A G G

Mig1

A A T T C A C G T G

Cbf1

# References - PSSM information content

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## n Papers by Tom Schneider

- q Schneider, T.D., G.D. Stormo, L. Gold, and A. Ehrenfeucht. 1986. Information content of binding sites on nucleotide sequences. J Mol Biol 188: 415-431.
- q Schneider, T.D. and R.M. Stephens. 1990. Sequence logos: a new way to display consensus sequences. Nucleic Acids Res 18: 6097-6100.
- q Tom Schneider's publications online
  - <http://www.lecb.ncifcrf.gov/~toms/paper/index.html>

## n Papers by Gerald Hertz

- q Hertz, G.Z. and G.D. Stormo. 1999. Identifying DNA and protein patterns with statistically significant alignments of multiple sequences. Bioinformatics 15: 563-577.