

Position-specific scoring matrices (PSSM)

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	---aCtCaCA CACGTGG ACTAGC-	high
PHO84	Site D	---TTTCCA GCACGTGG GCGGA--	high
PHO81	UAS	----TTATG GCACGTG CGAATAA--	high
PHO8	Proximal	GTGATCGCT GCACGTGG CCCGA---	high
group 1	consensus	----- gCACGTGg -----	high
PHO5	UASp1	--TAAATTA GCACGTTT TCGC----	medium
PHO84	Site E	----AATAC GCACGTTT TTAATCTA	medium
group 2	consensus	----- cgCACGTTt -----	medium
Degenerate consensus		----- GCACGTKKk -----	high-med

Non-binding sites

PHO5	UASp3	--TAATTTG GCA <u>T</u> GTGCGATCTC--	No binding
PHO84	Site C	-----ACGTCC CACGTG GA ACTAT--	No binding
PHO84	Site A	-----TTTAT CACGTG <u>A</u> CACTTTTT	No binding
PHO84	Site B	-----TTAC GCACGTT <u>G</u> GTGCTG--	No binding
PHO8	Distal	---TTACCC GCACG <u>C</u> TTAATAT---	No binding

IUPAC ambiguous nucleotide code

A	A	Adenine
C	C	Cytosine
G	G	Guanine
T	T	Thymine
R	A or G	puRine
Y	C or T	pYrimidine
W	A or T	Weak hydrogen bonding
S	G or C	Strong hydrogen bonding
M	A or C	aMino group at common position
K	G or T	Keto group at common position
H	A, C or T	not G
B	G, C or T	not A
V	G, A, C	not T
D	G, A or T	not C
N	G, A, C or T	aNy

From alignments to weights

Sequence logo

Count matrix (TRANSFAC matrix F\$PHO4_01)

Residue\position	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

Tom Schneider's sequence logo

(generated with Web Logo <http://weblogo.berkeley.edu/logo.cgi>)



Frequency matrix

Pos	1	2	3	4	5	6	7	8	9	10	11	12
A	0.13	0.38	0.25	0.00	1.00	0.00	0.00	0.00	0.00	0.00	0.13	0.25
C	0.25	0.25	0.38	1.00	0.00	1.00	0.00	0.00	0.00	0.25	0.00	0.25
G	0.13	0.25	0.38	0.00	0.00	0.00	1.00	0.00	0.63	0.50	0.63	0.25
T	0.50	0.13	0.00	0.00	0.00	0.00	0.00	1.00	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	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
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_{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*
k pseudo weight (arbitrary, 1 in this case)
f'_{i,j} corrected frequency of residue *i* at position *j*

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

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)
- $n_{i,j}$ 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
- 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

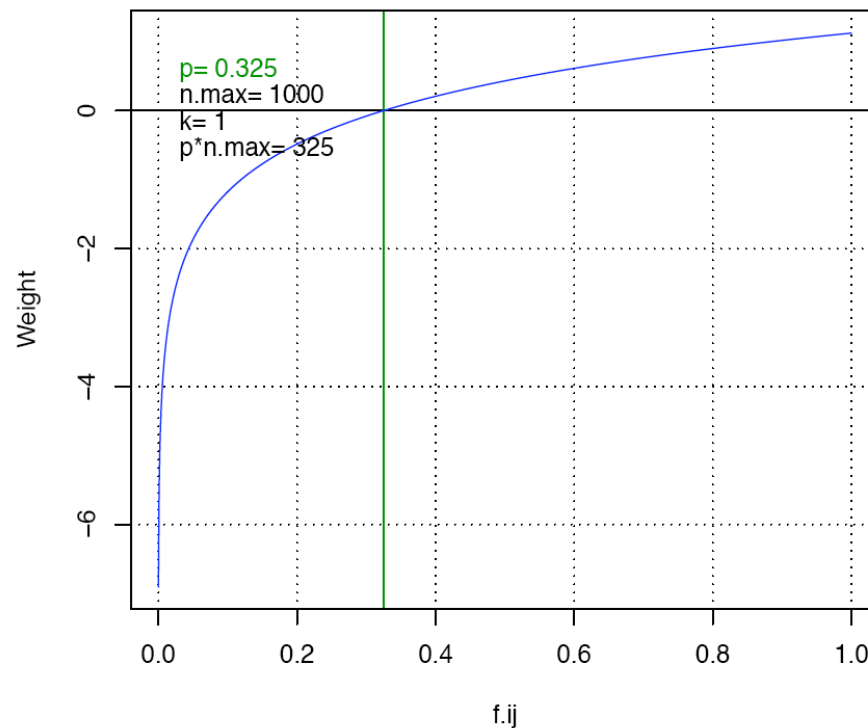
The use of a weight matrix relies on Bernoulli assumption

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

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



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

Regulatory sequence analysis

Information content

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Shannon uncertainty

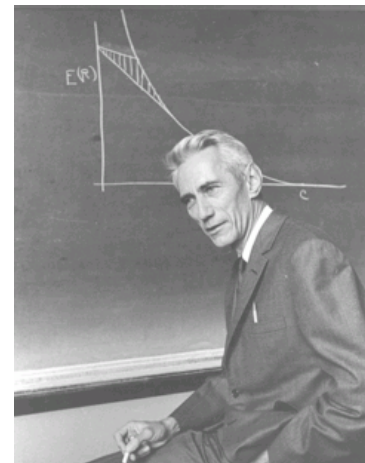
- Shannon uncertainty
 - $H_s(j)$: uncertainty of a column of a PSSM
 - H_g : uncertainty of the background (e.g. a genome)
- Special cases of uncertainty (for a 4 letter alphabet)
 - $\min(H)=0$
 - No uncertainty at all: the nucleotide is completely specified (e.g. $p=\{1,0,0,0\}$)
 - $H=1$
 - Uncertainty between two letters (e.g. $p=\{0.5,0,0,0.5\}$)
 - $\max(H) = 2$ (*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.
- R_{seq}
 - Schneider (1986) defines an **information content** based on Shannon's uncertainty.
- R_{seq}^*
 - For skewed genomes (i.e. unequal residue probabilities), Schneider recommends an alternative formula for the information content. This is the formula that is nowadays used.

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

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

$$R_{seq}(j) = H_g - H_s(j) \qquad R_{seq} = \sum_{j=1}^w R_{seq}(j)$$

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



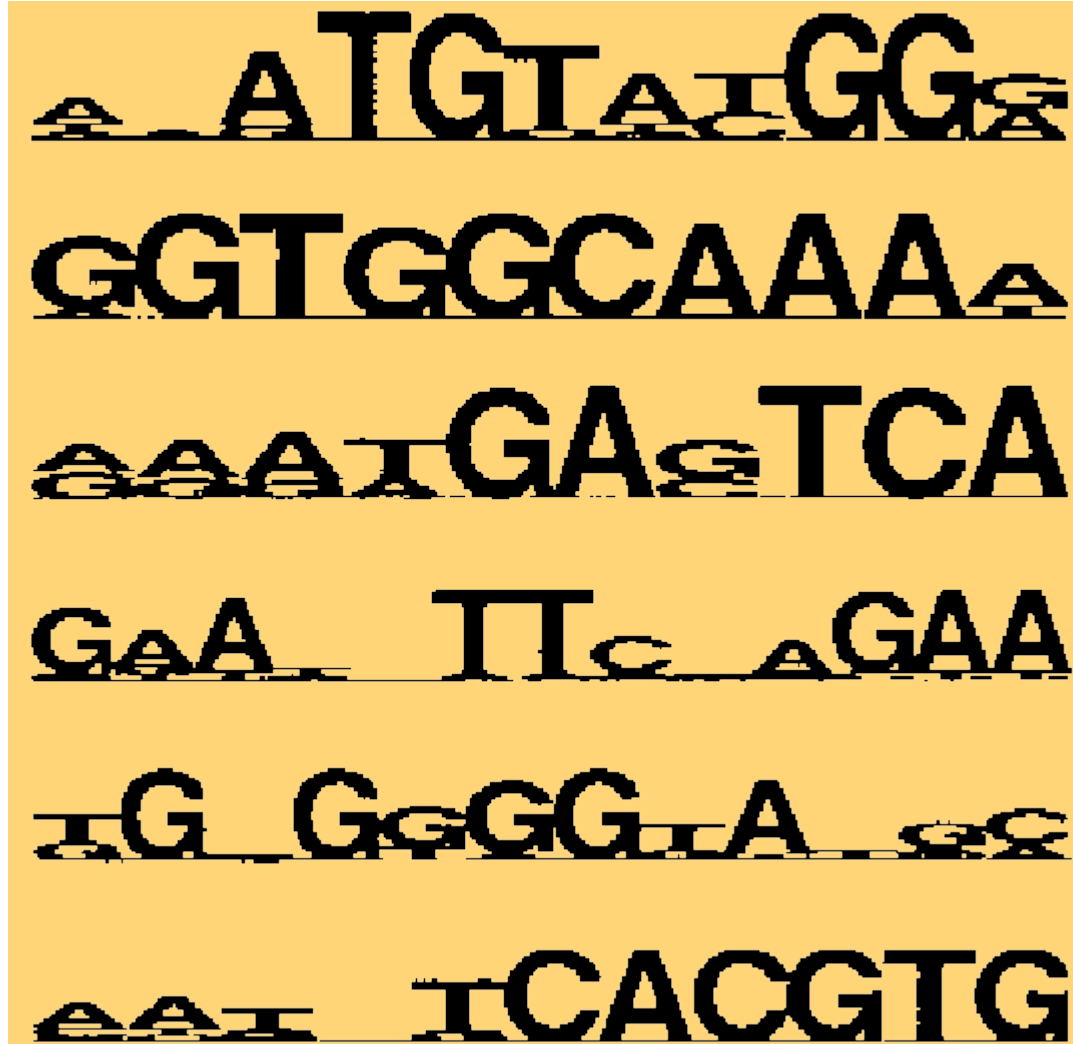
Schneider logos

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



- 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}
 - $H_s(j)$ uncertainty of column j
 - $R_{seq}(j)$ “information content” of column j (beware, this definition differs from Hertz’ information content)
 - $e(n)$ correction for small samples (pseudo-weight)
- Remarks
 - This information content does not include any correction for the prior residue probabilities (p_i)
 - This information content is expressed in bits.
- Boundaries
 - $\min(R_{seq})=0$ equiprobable residues
 - $\max(R_{seq})=2$ perfect conservation of 1 residue with a pseudo-weight of 0,
- Sequence logos can be generated from aligned sequences on the *Weblogo* server
 - <http://weblogo.berkeley.edu/>

Sequence logo



Rap1

Rpn4

Gcn4

HSE

Mig1

Cbf1

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

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

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

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

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

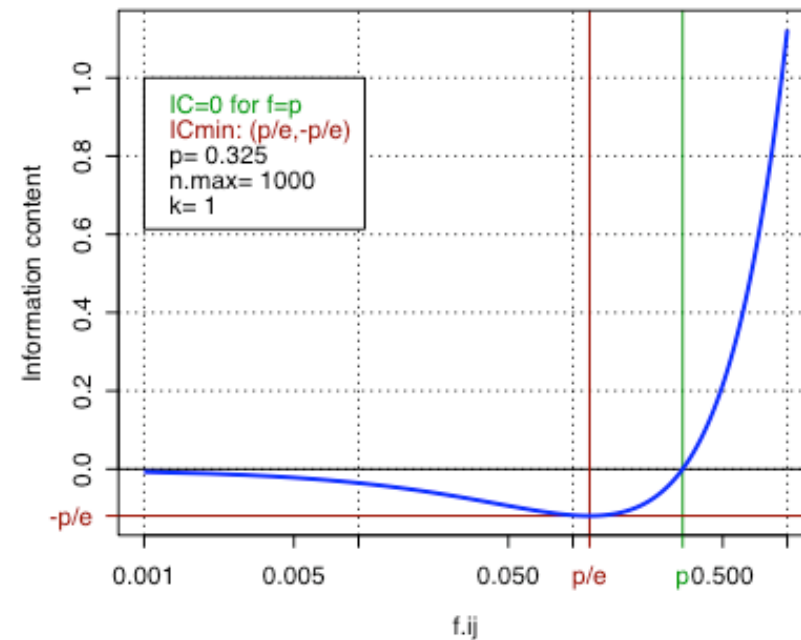
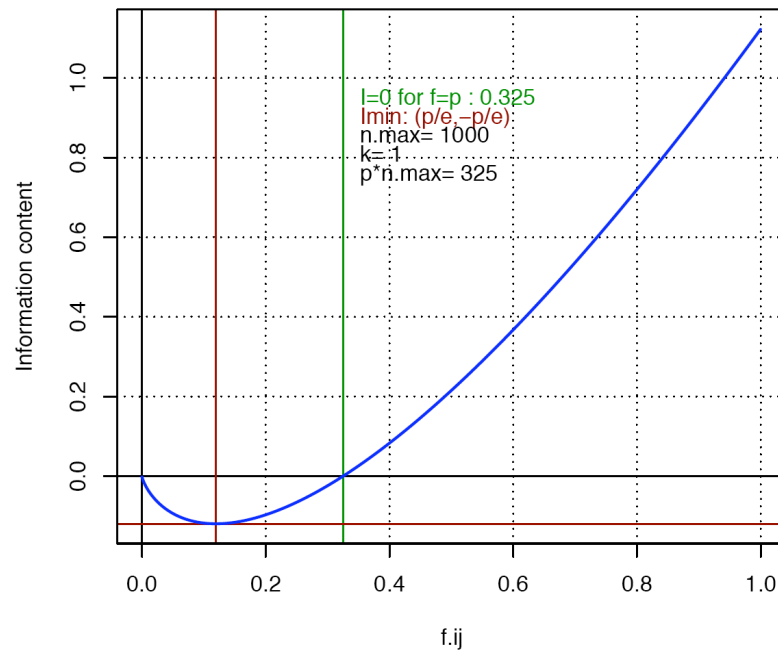
- A alphabet size (=4)
- $n_{i,j}$ occurrences of residue i at position j
- w matrix width (=12)
- p_i prior residue probability for residue i
- $f_{i,j}$ relative frequency of residue i at position j
- k pseudo weight (arbitrary, 1 in this case)
- $f'_{i,j}$ corrected frequency of residue i at position 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 15:563-577₁₃

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

- For a given cell of the matrix
 - I_{ij} is positive when $f'_{ij} > p_i$
(i.e. when residue i is more frequent at position j than expected by chance)
 - I_{ij} is negative when $f'_{ij} < p_i$
 - 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

- For a given column i of the matrix
 - The information of the column (I_j) is the sum of information of its cells.
 - I_j is always positive
 - I_j is always positive
 - I_j is 0 when the frequency of all residues equal their prior probability ($f_{ij}=p_i$)
 - 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

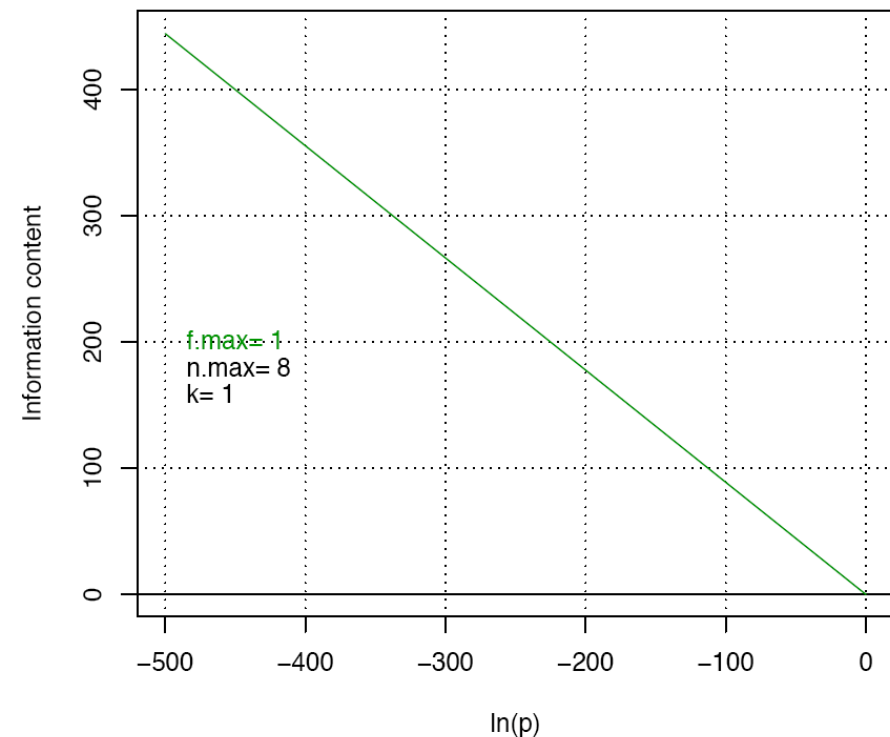
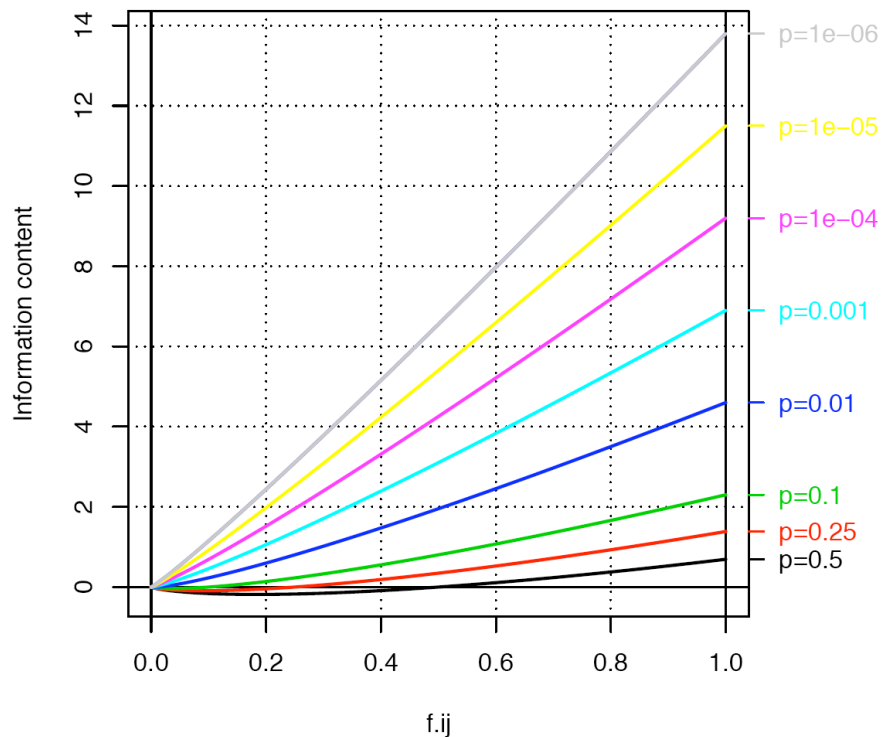
- The total information content represents the capability of the matrix to make the distinction between a binding site (represented by the matrix) and the background model.
- The information content also allows to estimate an upper limit for the expected frequency of the binding sites in random sequences.
- The pattern discovery program *consensus* (developed by Jerry Hertz) optimises the information content in order to detect over-represented motifs.
- 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

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



References - PSSM information content

■ Papers by Tom Schneider

- ❑ 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.
- ❑ Schneider, T.D. and R.M. Stephens. 1990. Sequence logos: a new way to display consensus sequences. *Nucleic Acids Res* 18: 6097-6100.
- ❑ Tom Schneider's publications online
 - <http://www.lecb.ncifcrf.gov/~toms/paper/index.html>

■ Papers by Gerald Hertz

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