End of Semester Logistics

Last day of class: Thursday Dec 5th

PS7/711-4: due Saturday, Dec 7, 11:59pm

Review session: Sunday, Dec 8th, 3pm to 7pm, WH 5407

Problem Set 7 review

Your questions!

Final exam: Monday, Dec 9th, 1pm-4pm

- Cummulative, emphasis last 3rd of the semester
- Closed book, 4 pages of notes

End of Semester Logistics ...

Homework

- Problem Set 7 due at 11:59pm on Sat 12/7
- Late homework receives a zero score once the solution sets have been posted.
- In calculating your final score, your lowest homework score will be dropped provided that all assignments have been submitted by the last day of classes.

End of Semester Logistics ...

Problem set 7

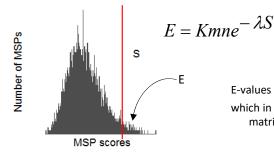
- Run 6 Blast searches with different parameter settings
- Record some results in Tables (excel worksheet)
- Interpret in terms of Blast heuristics and Karlin Altschul stats

Recommendations:

- Run all six searches in one session
- Record results immediately
- Interpret results at your leisure

BLAST (Karlin-Altschul) Statistics

E = Expected number of matches with score at least S under the null model

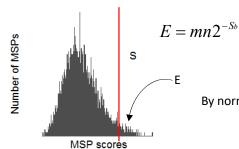


E-values depend on K and λ, which in turn depend on the scoring matrix, *S*[*i,j*].

<u>Maximal Segment Pair (MSP)</u>: an ungapped local alignment that cannot be improved by making it bigger or smaller.

BLAST (Karlin-Altschul) Statistics

E = Expected number of matches with score at least S under the null model



By normalizing the raw alignment scores

$$S_b = \frac{\lambda S - \ln K}{\ln 2}$$

we obtain a "bit score" S_b and an expression for E that is independent of K and λ

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E ~ the number of potential starting points for a high-scoring local alignment.

 $\simeq m \times n$, the number of cells in the alignment matrix

 $E = mn^{2^{-Sb}}$

Not quite!

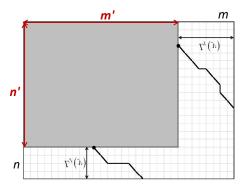
An alignment that starts too close to the end of s_1 or s_2 will not achieve a score of at least S_b

An alignment must start within the gray box to accrue a score of at least S_7 before reaching the end of the sequence.

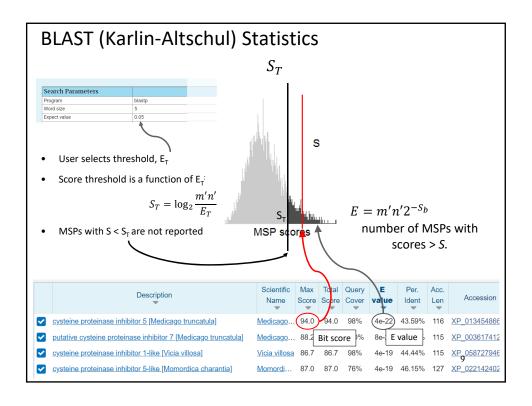
$$E = m'n'2^{-S_b}$$

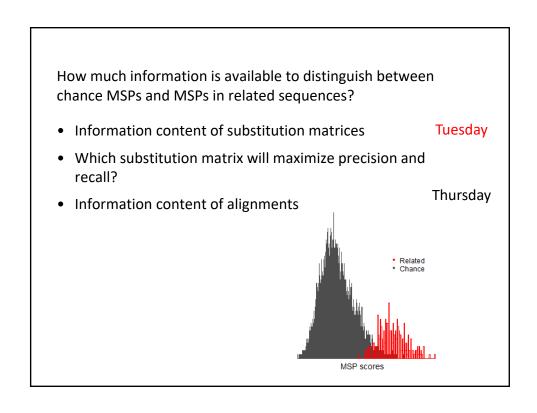
where the m' and n' are the "effective" lengths that reflect this edge correction

Blast software estimates the effective lengths automatically



New finite-size correction for local alignment score distributions. Park, Sheetlin, Ma, Madden, Spouge* *BMC Research Notes* 2012, 5:286 doi:10.1186/1756-0500-5-286

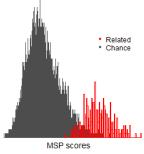




How much information is available to distinguish between chance MSPs and MSPs in related sequences?

False postives and false negatives depend on the overlap of the distributions of

- Chance MSPs
- MSPs in related sequences



What factors influence this overlap?

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A warm-up thought experiment:



Alternate Hypothesis: Coin is biased

•
$$pr(Heads|H_A) = q$$
, $pr(Tails|H_A) = (1 - q)$,
where $q \neq 0.5$

Null Hypothesis: : Coin is fair

•
$$pr(Heads|H_0) = p$$
, $pr(Tails|H_0) = (1 - p)$,
where $p = 0.5$

How many coin tosses are required to decide if the coin is biased?

- If $q \gg 0.5$, then a short series of coin tosses is sufficient
- If $q \approx 0.5$ (e.g., q = 0.5001), then we require a much longer series of coin tosses is sufficient to convince us that $p(H) \neq 0.5$.

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Relative Entropy

Given an event space $\mathcal{E} = \{E_1 E_2 \cdots E_N\}$ and probability distributions, P and Q, defined on \mathcal{E} :

$$Q = pr\{\mathcal{E}|H_A\} = \{q_1q_2 \cdots q_N\}$$
$$P = pr\{\mathcal{E}|H_0\} = \{p_1p_2 \cdots p_N\}$$

the relative entropy or Kullback-Leibler Divergence

$$\mathcal{H} = \sum_{\mathcal{E}} q_i \log_2 \frac{q_i}{p_i}$$

is the expected information provided by each observation to discriminate in favor of hypothesis H_A against hypothesis H_A , when H_A is true.

Note: the KL Divergence is not symmetric and therefore not a distance.

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Relative Entropy – coin toss example

Given an event space $\mathcal{E} = \{\textit{Heads}, \textit{Tails}\}\$ and probability distributions, P and Q, defined on \mathcal{E} :

$$Q = pr\{\mathcal{E}|H_A\} = \{q, 1 - q\}, q \neq 0.5$$
$$P = pr\{\mathcal{E}|H_0\} = \{0.5, 0.5\}$$

the relative entropy or Kullback-Leibler Divergence

$$\mathcal{H} = \sum_{\{H.T\}} q_i \log_2 \frac{q_i}{p_i} = q \log_2 \left(\frac{q}{0.5} \right) + (1 - q) \log_2 \left(\frac{1 - q}{0.5} \right)$$

is the expected information provided by each coin toss to discriminate in favor of hypothesis H_A (bias) against hypothesis H_0 (fair) when H_A is true.

Note: the KL Divergence is not symmetric and therefore not a distance.

Relative Entropy for coin tosses

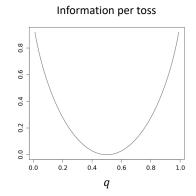
$$\sum_{\{H.T\}} q_i \log_2 \frac{q_i}{p_i} = q \log_2 \left(\frac{q}{0.5} \right) + (1-q) \log_2 \left(\frac{1-q}{0.5} \right)$$

Alternate hypothesis (H_A) :

- probability of Heads: $q \neq 0.5$
- probability of Tails: 1 q

Null hypothesis (H_0) :

- probability of Heads: 0.5
- probability of Tails: 0.5



Ungapped local alignments

T Q S S R A A K R Y S V C S L ...

- Alternate hypothesis: Sequences are related at N PAMs divergence. Amino acids x and y are aligned with frequency, q^N_{xy}
- Null hypothesis: Sequences are unrelated.
 Amino acids x and y are aligned with background frequencies, p_xp_y

How many aligned sites are required to decide if the sequences are related or share chance similarity?

Relative Entropy – ungapped local alignments

T Q S S R A A K R Y S V C S L ..
| + | + + + | | | + | + | |
E O A S O S A K R W S T. A G T. ...

- Alternate hypothesis: Sequences are related at N PAMs divergence $(q^N_{_{XY}})$
- Null hypothesis: Sequences are unrelated $(p_x p_y)$

The relative entropy

$$\mathcal{H} = \sum_{\{xy\}} q_{xy}^N \log_2 \frac{q_{xy}^N}{p_x p_y}$$

$$= \sum_{\{xy\}} q_{xy}^N S^N[x,y]$$

gives the number of bits per position available tzo distinguish chance MSPs from MSPs in related sequences with *N* PAMs of divergence.

The relative entropy of a substitution matrix is given in bits per position and can be calculated from S^N using the equation

$$\mathcal{H}^N = \sum_{x,y} q_{xy}^N S^N[x,y]$$

BLO	BLOSUM		PAM	Sequence	
	bits/site		bits/site	identity	
		20	2.95	83%	
		30	2.57		
		60	2.00	63%	
		70	1.60		
90	1.18	100	1.18	43%	
80	0.99	120	0.98	38%	
60	0.66	160	0.70	30%	
50	0.52	200	0.51	25%	
45	0.38	250	0.36	20%	

Since PAM 30 has 2.57 bits of information per site, should you always score your alignments with PAM30?

P/	ΔM	Seq Id		
30	2.57			
100	1.18	43 %		
120	0.98	38%		
160	0.70	30 %		
200	0.51	25%		
250	0.36	20 %		

NO!

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Target frequencies q_{xy}^N

Empirical "target" frequencies:

frequency of x aligned with y in sequences related to the query

Query
$$\mathbf{x} \dots \mathbf{x} \dots \mathbf{y} \dots \mathbf{x} \dots \mathbf{x} \dots$$

Theoretical "target" frequencies:

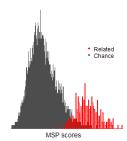
frequency of x aligned with y used to construct the matrix

$$S^{N}[x,y] = \log_2 \underbrace{q_{xy}^{N}}_{p_x p_y}$$

The highest alignment scores are obtained when the matrix target frequencies match the empirical target frequencies.

Scoring an alignment with a matrix that does not match the target frequencies characteristic of the query and sequences related to it, will result in lower MSP scores in related matches

If the matrix does not match the target frequencies, the related (red) distribution will move to the left, increasing the overlap.



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The average score (in bits) per alignment position when using a PAM *Y* matrix to compare sequences in fact separated by n PAMs

(Calculated by simulation)

PAM matrix	Actual PAM distance n								
	40	80	120	160	200	240	280	320	
	40	2.26	1.31	0.62	0.10	-0.30	-0.61	-0.86	-1.06
	80	2-14	1.44	0.92	0.53	0.23	-0.02	-0.21	-0.37
Υ	120	1.93	1.39	0.98	0.67	0.42	0.22	0.06	-0.07
•	160	1.71	1.28	0.95	0.70	0.50	0.33	0.20	0.09
	200	1.51	1.16	0.90	0.68	0.51	0.38	0.26	0.17
	240	1.32	1.05	0.82	0.65	0.51	0.39	0.29	0.21
	280	1.17	0.94	0.75	0.60	0.48	0.38	0.30	0.23
	320	1.03	0.84	0.68	0.56	0.46	0.37	0.30	0.24

Maxima highlighted in yellow

Best discrimination between related and chance MSPs :

Matrix divergence ~ Family divergence

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Altschul SF, J. Mol. Biol., 219, 555-565 (1991)

The average matrix	•	, ,	per aligi		•		_		
		(Ca	Iculated b	y simul	ation)	·			
DANA		Actual PAM distance n							
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						□ = Ef	= Efficiency ≥ 94%		
F#: ai a		Score w	ith PAM	Υ					
Efficie	iicy = –	Score with PAM n							
								31	
				А	ltschul SF.	J. Mol. Biol.	. 219. 555-	565 (1993	