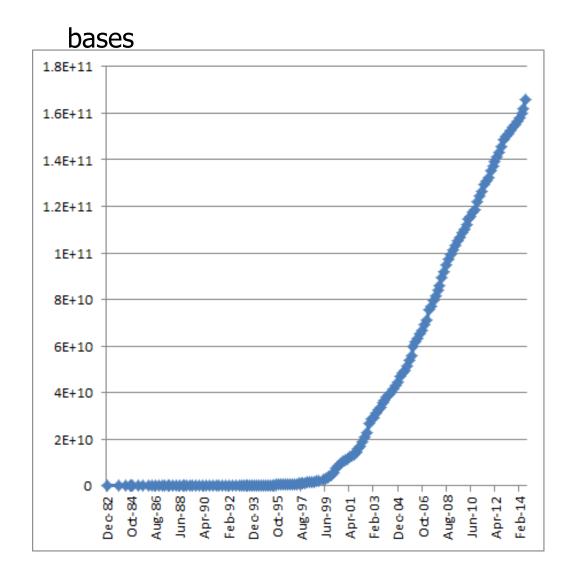
# CS3225v: Combinatorial Methods in Computation Biology Searching biological database

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#### Biological databases

- Biological data increases rapidly.
- Searching methods also need to scale-up to the large datasets



#### Problem definition

- Consider a database D of genomic sequences (or protein sequences)
- Given a query string Q,
  - we look for a string S in D which is the closest match to the query string Q
  - There are two meanings for closest match:
    - S and Q has a semi-global alignment (forgive the spaces on the two ends of Q)
    - S and Q have a local alignment

# Measurement of the goodness of a search algorithm

#### Sensitivity

- Ability to detect "true positive".
- Sensitivity can be measured as the probability of finding the match given the query and the database sequence has only x% similarity.

#### Specificity

- Ability to reject "false positive"
- Specificity is related to the efficiency of the algorithm.
- A good search algorithm should be both sensitive and specific

# Different approaches

- Exhaustive approach
  - Smith-Waterman Algorithm
- Heuristic methods
  - BLAST and BLAT
  - PatternHunter
- Filter and refine approaches
  - LSH

- Note: many approaches are local alignment!
- There are other searching algorithms. We don't have enough time to cover them.

#### Smith-Waterman Algorithm

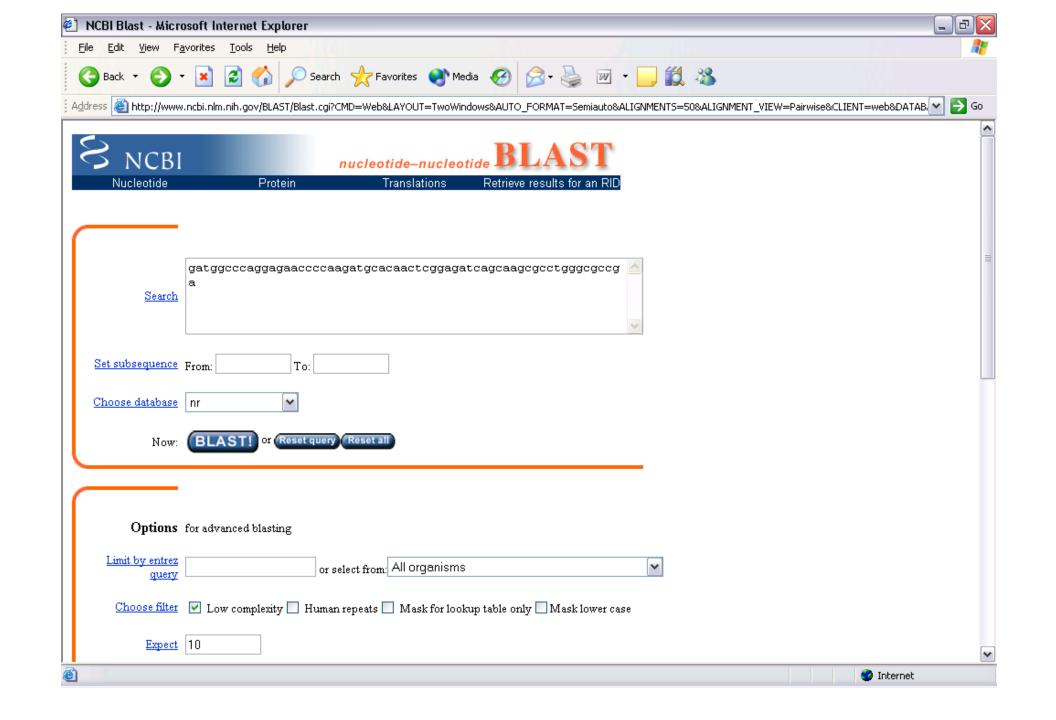
- Input:
  - the database D (total length: n) and
  - the query Q (length: m)
- Output: all closest matches (based on local alignment)

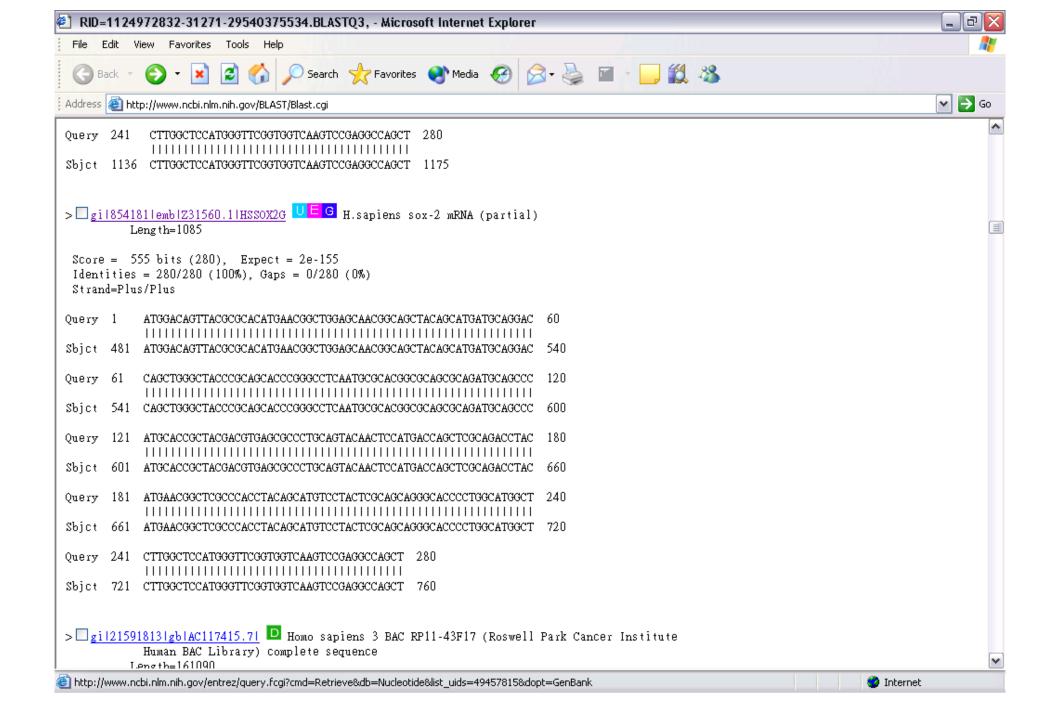
#### **Algorithm**

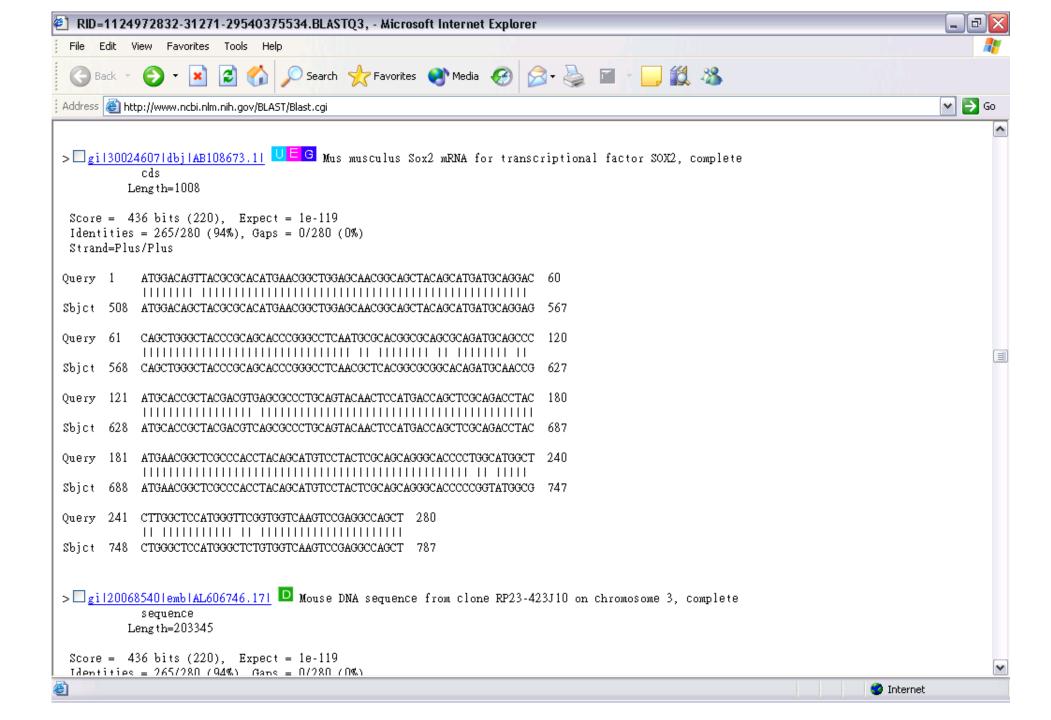
- For every sequences S in the database,
  - Use Smith-Waterman algorithm to compute the best local alignment between S and Q
- Return all alignments with the best score
- Time: O(nm)
- This is a brute force algorithm. So, it is the most sensitive algorithm.

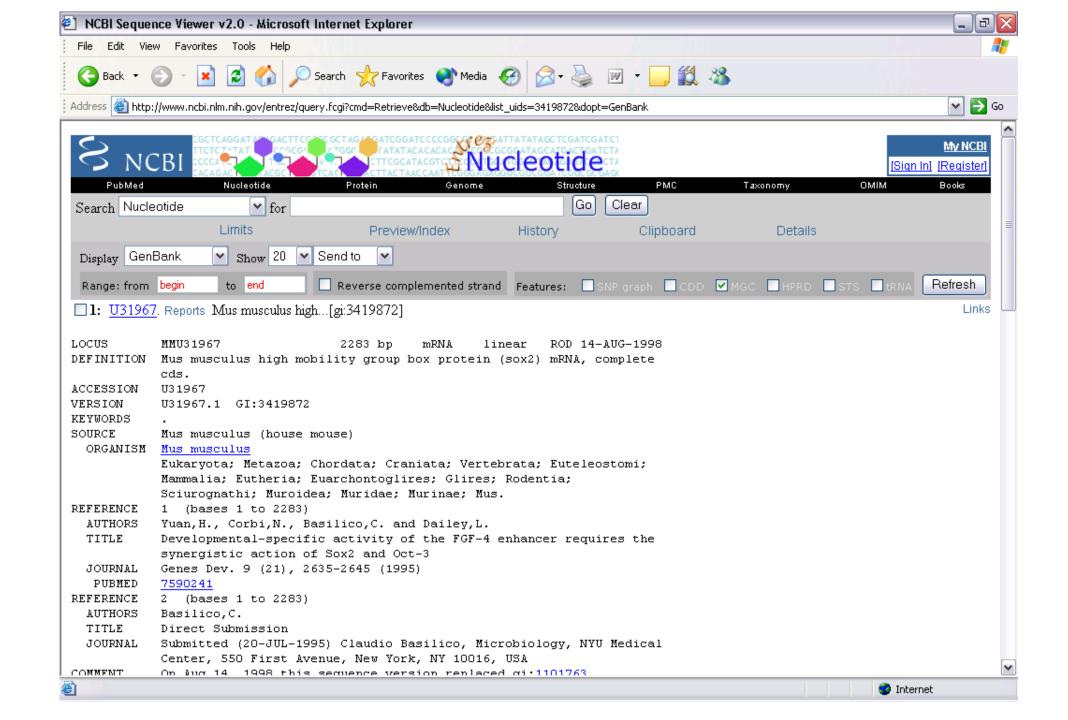
#### What is BLAST?

- BLAST = Basic Local Alignment Search Tool
- Input:
  - A database D of sequences
  - A sequence s
- Aim of BLAST:
  - Compare s against all sequences in D faster based on heuristics.
- Disadvantage of BLAST:
  - To be fast, it scarifies the accuracy. Thus, less sensitive









### History of BLAST

- 1990: Birth of BLAST1
  - It is very fast and dedicate to the search of local similarities without gaps
  - Altschul et al, Basic local alignment search tool. J. Mol. Biol., 215(3):403-410, 1990.
  - The most highly cited paper in 1990 and the third most highly cited paper in 1983-2002.
- 1996-1997: Birth of BLAST2
  - BLAST2 allows insertion of gaps
  - BLAST2 have two versions. Developed by two groups of authors independently
    - 1997: NCBI-BLAST2 (National Center for Biotechnology Information)
    - 1996: WU-BLAST2 (Washington University)

#### BLAST1

- A heuristic method which searches for local similarity without gap
- It can be divided into four steps:
  - Step 1: Query preprocessing
  - Step 2: Scan the database for hits
  - Step 3: Extension of hits

# Step 1: Query preprocessing

• For every position p of the query, insert the w-tuple (w=11 default) at position p into the hash table.

Q=TCATCATG

w-tuple	positions
ATCA	3
CATC	2
CATG	5
TCAT	1, 4

#### Step 2: Generation of hits

- Scan every sequence in the database DB.
  - For each position q in the sequence, if there is an exact match between the w-tuple at position q and a w-tuple in the query, a hit is made.
- A hit is characterized by the positions in both query and DB sequences.

>seq1 CCGCTCATGATGATCA

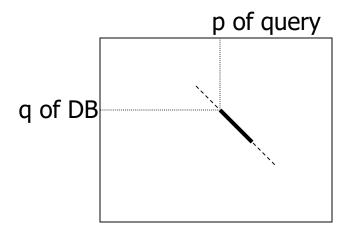
#### The list of hits:

- (5 of DB, 1 of query)
- (5 of DB, 4 of query)
- (13 of DB, 3 of query)

W-tuple	positions
ATCA	3
CATC	2
CATG	5
TCAT	1, 4

# Step 3: Extension of hits (I)

- For every hit, extend it in both directions, without gap.
- The extension is stopped as soon as the score decreases by more than X(parameter of the program) from the highest value reached so far.



# Step 3: Extension of hits (II)

- If the extended segment pair has score better than or equal to S(parameter of the program), it is called an HSP (High scoring segment pair). Then, they will be reported.
- For every sequence in the database, the best scoring HSP is called the MSP (Maximal segment pair).

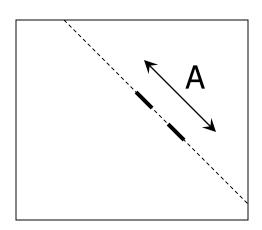
#### NCBI-Blast2

Allows local alignment with gaps.

- The first 2 steps are the same as BLAST1.
- Two major differences:
  - Two-hits requirement (implemented for protein)
  - Gapped extension

#### Step 3: Two-hits requirement

- To extend a hit, we require that there is another hit on the same diagonal within a distance smaller than A
- By default, A=40
- Note: Two-hits requirement is implemented for protein sequences (not DNA).

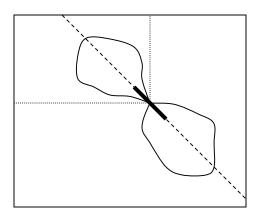


# Step 4: Gapped extension (I)

- For hits satisfying the two-hits requirement, extend them similar to Step 3 of BLAST1
- Among the generated HSP, we perform gapped extension for those with score > some threshold

# Step 4: Gapped extension (II)

- Gapped extension is a modified Smith-Waterman algorithm
  - Explore the dynamic programming starting from the middle of the hit
  - When the alignment score drops off by more than  $X_g$ , stop



#### BLAST1 vs. NCBI-BLAST2

- BLAST1 spends 90% of its time on extension
- For NCBI-BLAST2, due to the two-hits requirement, the number of extensions is reduced.
  - NCBI-BLAST2 is about 3 times faster than BLAST1.

# BLAST program options

Program	Query	Database	Alignment type		
blastn	DNA	DNA	Search DNA query sequence in DNA database		
blastp	Protein	Protein	Search protein query sequence in protein database		
blastx	DNA	Protein	Convert DNA query sequence into protein sequences in all 6 reading frames. Search these translated proteins in protein database		
tblastn	Protein	DNA	Search protein query sequence again protein sequences generated from the 6 reading frames of the DNA sequences in the DNA database		
tblastx	DNA	DNA	Convert DNA query sequence into protein sequences in all 6 reading frames. Search these translated protein query sequence again protein sequences generated from the 6 reading frames of the DNA sequences in the DNA database		

# Statistics for local alignment

- A local alignment without gaps consists simply of a pair of equal length segments.
- BLAST finds the local alignments whose score cannot be improved by extension. Such local alignments are called high-scoring segment pairs or HSPs.
- To determine the significant of the local alignments, BLAST gives E-value and bit score. Below, we give a brief discussion on them.
- Assumption: We required the expected score for aligning a random pair of residues/bases to be negative.
  - Otherwise, the longer the alignment, the higher is the score independent of whether the segments aligned are related or not.

#### Raw Score for BLAST

• Raw score = 8\*2 - 3 - (5+2\*3) = 2.

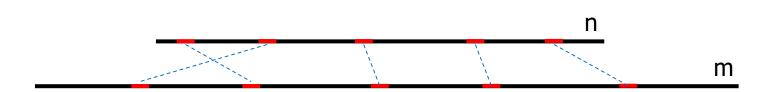
$$g(q)=5+2q$$

	A	С	G	Т
A	2	-3	-3	-3
С	-3	2	-3	-3
G	-3	-3	2	-3
Т	-3	-3	-2	2

**BLAST Matrix** 

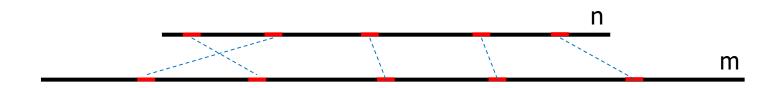
#### E-value

- E-value is the expected number of alignments having raw score > S totally at random.
- Let m and n be the lengths of the query sequence and the database sequence.
- Intuition:
  - Double the length of either sequence will double the expected number of HSPs. (i.e.  $E \propto nm$ )
  - Double the score S will exponentially reduce the expected number of HSPs. (i.e.  $E \propto e^{-\lambda S}$ )



# E-value (II)

- Mathematically, when both m and n are sufficiently long,
  - the expected number E of HSPs with score at least S follows the extreme distribution (Gumbel distribution). We have
    - $E=Kmne^{-\lambda S}$  for some parameters K and  $\lambda$  which depends on the scoring matrix  $\delta$  and the expected frequencies of the residues/bases.
- Hence, when E-value is small, the HSP is significant.



# E-value (III)

- For more information on estimating K and  $\lambda$ , please read
  - http://www.ncbi.nlm.nih.gov/BLAST/tutorial/Altschul-3.html
  - http://oreilly.com/catalog/blast/chapter/ch04.pdf

#### Bit score

- The raw score S of an alignment depends on the scoring system.
- Without knowing the scoring system, the raw score is meaningless.
- The bit score is defined to normalize the raw score, which is defined as follows.

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

- Note that  $E = Kmne^{-\lambda S}$ . By definition of S',  $E = mn2^{-S'}$ .
- Hence, when S' is big, the HSP is significant.

#### P-value

- The number of random HSPs with score ≥ S follows a Poisson distribution.
- Pr(exactly x HSPs with score  $\geq$  S) =  $\frac{e^{-E}E^x}{x!}$ 
  - where  $E = Kmne^{-\lambda S}$  is the E-score
- Hence, p-value = Pr(at least 1 HSPs with score  $\geq$  S) =  $1 e^{-E}$ .
- Note:
  - when E increases, p-value is approaching 1.
  - When E=3, p-value is  $1-e^{-3} = 0.95$ .
  - When E=10, p-value is  $1-e^{-10} = 0.99995$
  - when E<0.01, 1-e<sup>-E</sup>≈E.
- Hence, in BLAST, p-value is not shown since we expect p-value and E-value are approximately the same when E<0.01 while p-value is almost 1 when E>10.

### Local alignment with gaps

- There is no solid theoretical foundation for local alignment with gaps.
- Moreover, experimental results suggested that the theory for ungapped local alignment can be applied to the gapped local alignment as well.

# Completeness of BLAST (I)

- BLAST is the most popular solution for finding local alignments. It is well-known that BLAST is heuristics and it will miss solution.
- We would like to check how many good alignments are missed by BLAST.
- We extracted 2000 mRNA sequences from each of the 4 different species. We aligned them on human genome. Then, we checked how many significant alignments are missed by BLAST.

# Completeness of BLAST (II)

	Chimpanzee	Mouse	Chicken	Zebrafish	All 4 species
E-value (≤)	Missing %	Missing %	Missing %	Missing %	Missing %
$1.0 \times 10^{-16}$	0.00	0.03	0.05	0.06	0.01
$1.0 \times 10^{-15}$	0.00	0.03	0.05	0.06	0.02
$1.0 \times 10^{-14}$	0.00	0.04	0.06	0.06	0.02
$1.0 \times 10^{-13}$	0.00	0.03	0.07	0.14	0.02
$1.0 \times 10^{-12}$	0.01	0.04	0.10	0.17	0.03
$1.0 \times 10^{-11}$	0.02	0.05	0.11	0.28	0.05
$1.0 \times 10^{-10}$	0.02	0.07	0.13	0.39	0.06
$1.0 \times 10^{-9}$	0.03	0.09	0.16	0.60	0.08
$1.0 \times 10^{-8}$	0.05	0.11	0.25	0.77	0.12
$1.0 \times 10^{-7}$	0.10	0.19	0.31	0.81	0.18
$1.0 \times 10^{-6}$	0.17	0.31	0.45	1.08	0.28
$1.0 \times 10^{-5}$	0.32	0.47	0.70	1.45	0.45
$1.0 \times 10^{-4}$	0.57	0.88	0.99	1.81	0.75
$1.0 \times 10^{-3}$	0.99	1.36	1.25	2.25	1.17
$1.0 \times 10^{-2}$	1.69	2.11	1.68	2.61	1.84
$1.0 \times 10^{-1}$	2.70	2.97	2.33	2.86	2.76

- 2000 queries for each species.
  - BLAST only missed 0.06% of those 8000 queries (with E-value smaller than 1.0x10<sup>-10</sup>).
  - In conclusion, BLAST is accurate enough in most cases, yet the few alignments missed could be critical for biological research.

#### Variation of BLAST

- MegaBLAST
- BLAT
- PatternHunter
- PSI-BLAST

#### MegaBLAST

- Only for DNA
- For DNA, in BLAST, w = 11 by default.
- To improve efficiency, MegaBLAST uses longer w-tuples (by default, w=28).
- The cost is the reduction in sensitivity.

#### **BLAT**

- Only for DNA.
- By default, BLAT uses w=11 and two-hit.

- BLAT is very fast.
  - The main trick is to index the database and put the index in the main memory
  - Note that BLAT is less sensitive than BLAST, but more sensitive than MegaBLAST.

### Main trick of BLAT

- BLAST cannot build index of human genome since it it big.
- BLAT's index stores the positions of non-overlapping w-tuples in memory.

### Database = ACTTGTACTTGTA

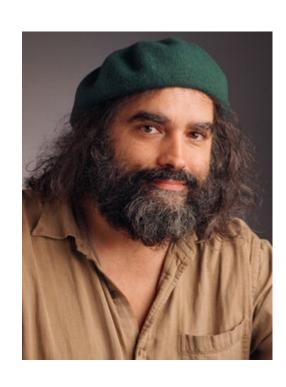
### Index of all w-mers

w-mer	positions
ACTT	1, 7, 13
CTTG	2, 8, 14
GTAC	5, 11
TACT	6, 12
TGTA	4, 10
TTGT	3, 9, 15
TGTA	16
•	

### Index of w-mers at positions iw+1

w-mer	positions
ACTT	1, 13
GTAC	5
TTGT	9

### About the inventor: Jim Kent



- Education: University of California, Santa Cruz
- Awards: Overton Prize, Benjamin Franklin Award

### PatternHunter

- PatternHunter can only apply to DNA
- PatternHunter is similar to BLAST. Moreover, it uses gapped w-tuple.
  - For w=11, they use 111010010100110111
  - Example,

```
111010010100110111
ACTCCGATATGCGGTAAC
| | | | - | - - | - | | - | | |
ACTTCACTGTGAGGCAAC
```

 They found that gapped w-tuple can increase the sensitivity while increase the efficiency.

## Advantage of gapped w-tuple (I)

- Increase sensitivity
  - Gapped w-tuples are more independent.
  - Examples:
    - Two adjacent ungapped 11-tuples share 10 symbols

```
• 1111111111 1/4 chances to have 2nd hit 11111111111 next to the 1st hit
```

Two adjacent gapped 11-tuples share 5 symbols

• If the w-tuples are more independent, the probability of having at least one hit in a homologous region is higher.

# Advantage of gapped w-tuple (II)

- Reduce the number of hits.
  - For the same query length (says, 64),
    - It covers by 54 ungapped 11-tuples
    - It covers by 47 gapped 11-tuples
  - So, the number of hits is smaller.
- Thus, the efficiency is increased!

### PatternHunter I

Ma et al., *Bioinformatics* 18:440-445, 2002

Proposition. The expected number of hits of a weight-W length-M model within a length-L region of similarity p is  $(L-M+1)*p^W$ 

Proof.

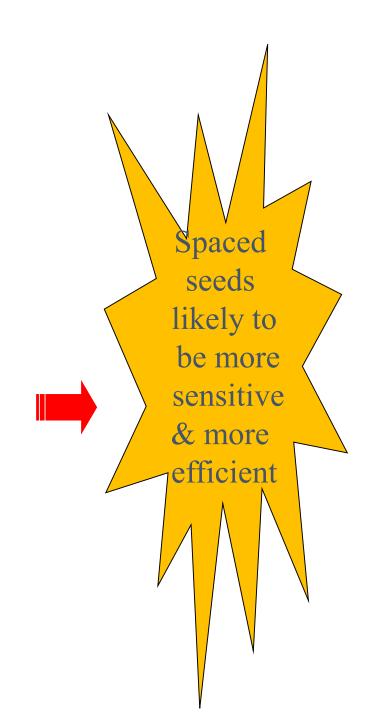
For any fixed position, the prob of a hit is  $p^{W}$ .

There are L - M + 1 candidate positions.

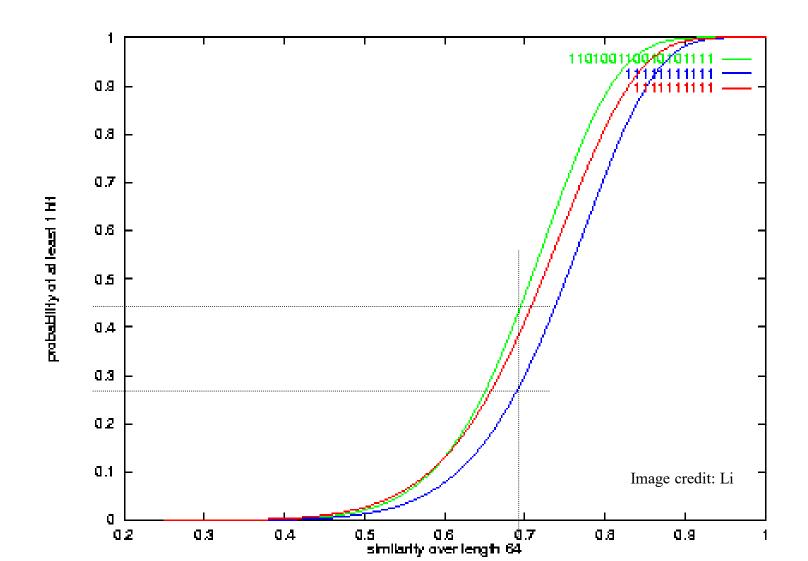
The proposition follows.

# **Implication**

- For L = 1017
  - BLAST seed expects  $(1017 11 + 1) * p^{11} = 1007 * p^{11}$  hits
  - But  $\sim 1/4$  of these overlap each other. So likely to have only  $\sim 750$  \*  $p^{11}$  distinct hits
  - Our example spaced seed expects (1017 18 + 1) \*  $p^{11} = 1000 * p^{11}$  hits
  - But only  $1/4^6$  of these overlap each other. So likely to have  $\sim 1000 * p^{11}$  distinct hits



# Sensitivity of PatternHunter I



### More for PatternHunter

- To further improve the efficiency,
  - PatternHunter uses a variety of advanced data structures including priority queues, a variation of red-black tree, queues, hash tables.
  - PatternHunter also uses a new method of sequence alignment.
- To further improve the accuracy,
  - PatternHunter II suggested to use multiple gapped seeds.
  - They show that the accuracy can approach smith-waterman algorithm while the speed 3000 times faster than smith-waterman.
- PatternHunter II is both faster and sensitive than BLAST, MegaBLAST.

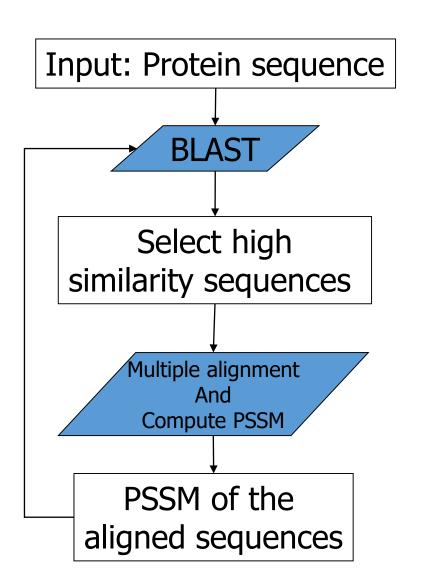
## About the Inventor: Ming Li

- Ming Li
  - Canada Research Chair Professor of Bioinformatics, University Professor, Univ of Waterloo
  - Fellow, Royal Society of Canada. Fellow, ACM.
     Fellow, IEEE.



## PSI-BLAST (Position Specific Iterated BLAST)

- PSI-BLAST is an implementation of BLAST for finding protein families. It allows us to detect distant homology.
- Input: a protein sequence
  - Using BLAST, we get a set of sequences that align with the query protein with E-score below a threshold, 0.01 (by default).
  - Align the selected sequences
  - Generate a PSSM profile from the multiple alignment
  - Iterate until no significant alignment found,
    - Using a modified BLAST, search the database with the PSSM profile.
    - Align the selected sequences
    - Generate a PSSM from the multiple alignment
- This version automatically combines statistically significant alignments produced by BLAST into a position-specific score matrix.
- It is much more sensitive to weak but biologically relevant sequence similarities

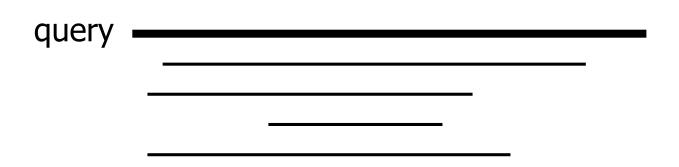


Find a set of sequences similar to the query

• Using BLAST 2.0, we get a set of sequences that align with the query protein with E-score below a threshold, 0.01 (by default).

# Multiple sequence alignment of the selected sequences

- Using the query sequence as the template, we aligned the selected sequences.
- All gap characters inserted into the query sequence are ignored.
- Note:
  - the length of the alignment is the same as the query sequence.
  - Some columns of the multiple sequence alignment may include nothing except the query.



### Generate a PSSM profile from the alignment

- Given the multiple alignment of length n,
  - We generate the position-specific score matrix (PSSM) profile, which is a 20xn matrix.
  - For each column and each residue a in the profile, we generate a log-odds score  $log(O_{ia}/P_a)$ .
    - where O<sub>ia</sub> is the observed frequency of residue a at position i and P<sub>a</sub> is the expected frequency respectively of the residue a.
- Since number of sequences may be small, data-dependent pseudo frequency is added to O<sub>ia</sub>.

## Find a set of sequences similar to the PSSM profile

- We apply a modified BLAST to the PSSM profile.
  - Basically, when we compare a position of the PSSM and a residue in the database, we use the corresponding log-odds score in that position.

Repeat until we satisfy.

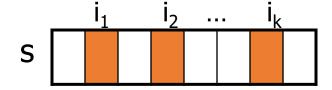
# Locality-Sensitive Hashing (LSH)

#### **LSH-ALL-PAIRS**

- Input: biosequence database D
- Aim: find pairs of w-mers that differ by at most d substitutions (ungapped local alignment) in a collection of biosequences D.

## Locality-sensitive hash function

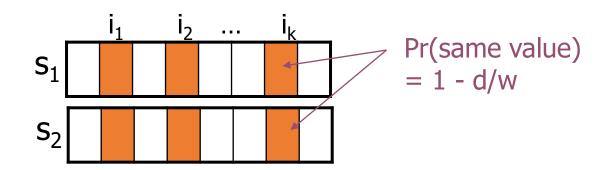
- Consider an w-mers s,
  - choose k indices i<sub>1</sub>, i<sub>2</sub>, ..., i<sub>k</sub> uniformly from the set {1, 2, ..., w}
  - Define  $\pi(s) = (s[i_1], s[i_2], ..., s[i_k])$ . This function is called the locality-sensitive hash function



# Property of locality-sensitive hash function (I)

- Consider two w-mers s<sub>1</sub> and s<sub>2</sub>,
  - the more similar are they, the higher probability that  $\pi(s_1) = \pi(s_2)$ .
- More precisely, if the hamming distance of  $s_1$  and  $s_2 = d$ ,

• 
$$Pr[\pi(s_1) = \pi(s_2)]$$
 =  $\prod_{j=1,...,k} Pr[s_1[i_j] = s_2[i_j]]$   
=  $(1 - d/w)^k$ 



# Property of locality-sensitive hash function (II)

- Hence, s<sub>1</sub> and s<sub>2</sub> are similar if
  - $\pi(s_1) = \pi(s_2)$
- However, we may have false positive and false negative
  - False positive:  $s_1$  and  $s_2$  are dissimilar but  $\pi(s_1) = \pi(s_2)$ .
    - False positive can be distinguished from true positive by computing hamming distance between s<sub>1</sub> and s<sub>2</sub>
  - False negative:  $s_1$  and  $s_2$  are similar but  $\pi(s_1) \neq \pi(s_2)$ .
    - We cannot detect false negative.
    - We can only reduce the number of false negative by repeating the test using different  $\pi$ () functions

### LSH-ALL-PAIRS

### Algorithm:

- 1. Generate m random locality-sensitive hash functions  $\pi_1()$ ,  $\pi_2()$ , ...,  $\pi_m()$ .
- 2. For every w-mer s in the database, compute  $\pi_i(s)$  for  $1 \le j \le m$ .
- 3. For every pair of w-mers s and t such that  $\pi_i(s) = \pi_i(t)$  for some j,
  - If hamming distance(s, t) < d, report (s, t)-pair.</li>

### Conclusion

- This lecture presents some database searching methods.
- In fact, there are many other methods. For examples:
  - CAFÉ, FLASH, RAMdb, FD, suffix tree, suffix array, compressed suffix array

### More information

- The list of database used by blast
  - ftp://ftp.ncbi.nlm.nih.gov/blast/db/