

Modern Homology Search

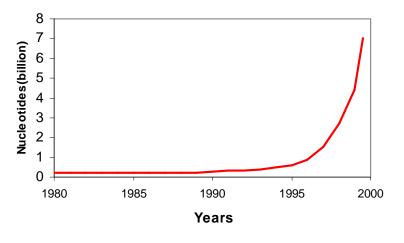
Ming Li

Canada Research Chair in Bioinformatics
University of Waterloo

Joint work with Bin Ma, John Tromp Joint work: X.F. Cui, T. Vinar, D. Shasha I will present three simple ideas

A gigantic gold mine

The trend of genetic data growth



30 billion

in year 2005

- 400 Eukaryote genome projects underway
- GenBank doubles every 18 months
- n Comparative genomics Ł all-against-all search



What is homology search

- Given two DNA sequences, find all local similar regions, using "edit distance" (match=1, mismatch=-1, gapopen=-5, gapext=-1).
- Example. Input:
 - E. coli genome: 5 million base pairs
 - h. H. influenza genome: 1.8 million base pairs

Output: all local alignments.

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Comparing to internet search

- n Internet search
 - Size limit: 5 billion people x homepage size
 - Supercomputing power used: ½ million CPU-hours/day
 - Query frequency: Google --- 112 million/day
 - Query type: exact keyword search --- easy to do

Homology search

- Size limit: 5 billion people x 3 billion basepairs + millions of species x billion bases
- 10% (?) of world's supercomputing power
- Query frequency: NCBI BLAST -- 150,000/day, 15% increase/month
- Query type: approximate search --- topics today



Tremendous Cost

- Bioinformatics Companies living on BLAST:
 - Paracel (Celera)
 - _n TimeLogic
 - TurboGenomics (TurboWorx)
- NSF, NIH, pharmaceuticals proudly support many supercomputing centers for homology search
- However: hardware become obsolete in 2-3 years. Software solution is indispensable.



Old Homology Search

- Too slow (dynamic programming)
- Too lossy (BLAST) and in fact still too slow
- No specific. Hundreds of unrelated answers.



- Dynamic programming (1970-1980)
 - Human vs mouse genomes: 10⁴ CPU-years
- BLAST, FASTA heuristics (1980-1990)
 - Human vs mouse genomes: 19 CPU-years
 - BLAST paper was referenced 100000 times
- n PatternHunter
 - h Human vs mouse genomes: 20 CPU-days
- PatternHunter II: dynamic programming sensitivity, BLAST speed.



Model Homology Search

- ~100% sensitivity, approaching to dynamic programming.
- ~100% specificity: return only the correct match, not hundreds of junk alignments
- Still at higher (than BLAST) speed.



BLAST Algorithm & Example

- Find seeded matches of 11 base pairs
- Extend each match to right and left, until the scores drop too much, to form an alignment
- Report all local alignments



BLAST Dilemma:

- If you want to speed up, have to use a longer seed. However, we now face a dilemma:
 - increasing seed size speeds up, but loses sensitivity;
 - decreasing seed size gains sensitivity, but loses speed.
- How do we increase sensitivity & speed simultaneously? For 20 years, many tried: suffix tree, better programming ...



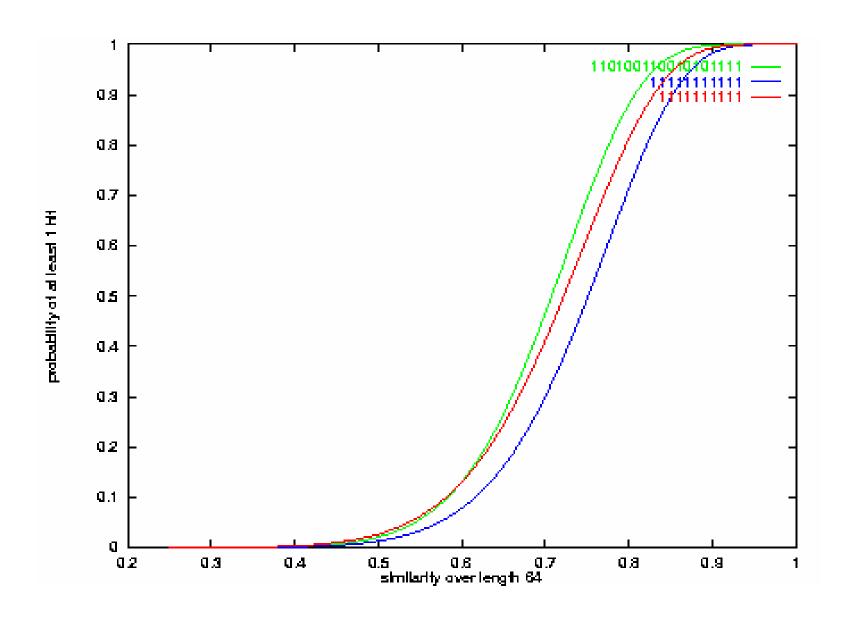
- A simple (but profound) idea
- A simpler (but working) idea
- n Another simple idea



New Idea: Spaced Seed

- Spaced Seed: nonconsecutive matches and optimize match positions.
- Represent BLAST seed by 11111111111
- Spaced seed: 111*1**1**11*111
 - 1 means a required match
 - * means "don't care" position
- This seemingly simple change makes a huge difference: significantly increases hit to homologous region while reducing bad hits.

Sensitivity: PH weight 11 seed vs BLAST 11 & 10



Formalize

Given i.i.d. sequence (homology region) with Pr(1)=p and Pr(0)=1-p for each bit:

Mhich seed is more likely to hit this region:

_n BLAST seed: 11111111111

Spaced seed: 111*1**1**11*111

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Expect Less, Get More

Lemma: The expected number of hits of a weight W length M seed model within a length L region with homology level p is (L-M+1)pW

Proof. E(#hits) =
$$\sum_{i=1...L-M+1} p^{W}$$

- Example: In a region of length 64 with p=0.7
 - Pr(BLAST seed hits)=0.3 E(# of hits by BLAST seed)=1.07
 - Pr(optimal spaced seed hits)=0.466, 50% more E(# of hits by spaced seed)=0.93, 14% less

Why Is Spaced Seed Better?

A wrong, but intuitive, proof: seed s, interval I, similarity p $E(\#hits) = Pr(s \ hits) \ E(\#hits \mid s \ hits)$

Thus:

```
Pr(s hits) = Lp^w / E(\#hits | s hits)
```

For optimized spaced seed, E(#hits | s hits)

| 111*1**1**11*111 | Non overlap | Prob |
|------------------|-------------|-------|
| 111*1**1**11*111 | 6 | p^6 |
| 111*1**1*111*111 | 6 | p^6 |
| 111*1**1*11*111 | 6 | p^6 |
| 111*1**1*111 | 7 | p^7 |

.

- For spaced seed: the divisor is $1+p^6+p^6+p^7+...$
- ⁿ For BLAST seed: the divisor is bigger: $1+p+p^2+p^3+...$



Complexity of finding the optimal spaced seed (Li, Ma, Zhang, SODA)

Theorem 1. Given a seed and it is NP-hard to find its sensitivity, even in a uniform region.

Theorem 2. The sensitivity of a given seed can be efficiently approximated with arbitrary accuracy, with high probability.



Computing Spaced Seeds

(Keich, Li, Ma, Tromp, Discrete Appl. Math)

Let *f(i,b)* be the probability that seed *s* hits the length *i* prefix of *R* that ends with *b*.

Thus, if s matches b, then

$$f(i,b)=1,$$

otherwise we have the recursive relationship:

$$f(i,b) = (1-p)f(i-1,0b') + pf(i-1,1b')$$

where b' is b deleting the last bit.

Then the probability of *s* hitting *R* is

$$\Sigma_{|b|=M}$$
 Prob(b) $f(L-M,b)$



Related Literature

- Random or multiple spaced q-grams were used in the following work:
 - FLASH by Califano & Rigoutsos
 - Multiple filtration by Pevzner & Waterman
 - LSH of Buhler
 - Praparata et al on probe design
- Optimizing & further work
 - _n Buhler-Keich-Sun
 - Brejova-Bronw-Vinar
 - Choi-Zhang
 - Tsur, Farach-Colton, Landau, Sahinalp
 - Over 100 research papers.



PatternHunter

(Ma, Tromp, Li: *Bioinformatics*, 18:3, 2002, 440-445)

- PH used optimal spaced seeds, novel usage of data structures: red-black tree, queues, stacks, hashtables, new gapped alignment algorithm.
- Written in Java.
- Used in Mouse Genome Consortium (*Nature*, Dec. 5, 2002), as well as in hundreds of institutions and industry.

Comparison with BLAST



On Pentium III 700MH, 1GB

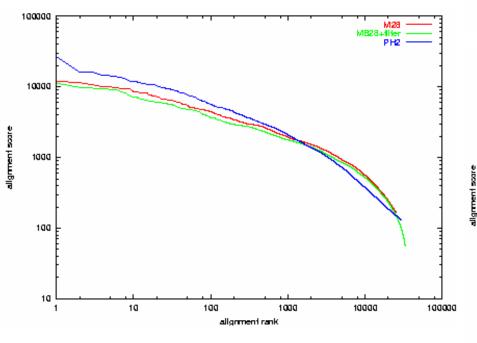
| | BLAST | PatternHunter |
|----------------------------|-------------|---------------|
| E.coli vs H.inf | <i>716s</i> | 14s/68M |
| Arabidopsis 2 vs 4 | | 498s/280M |
| Human 21 vs 22 | | 5250s/417M |
| Human(3G) vs Mouse(x3=9G)* | 19 years | 20 days |

- n All with filter off and identical parameters
- 16M reads of Mouse genome against Human genome for MIT Whitehead. Best BLAST program takes 19 years at the same sensitivity

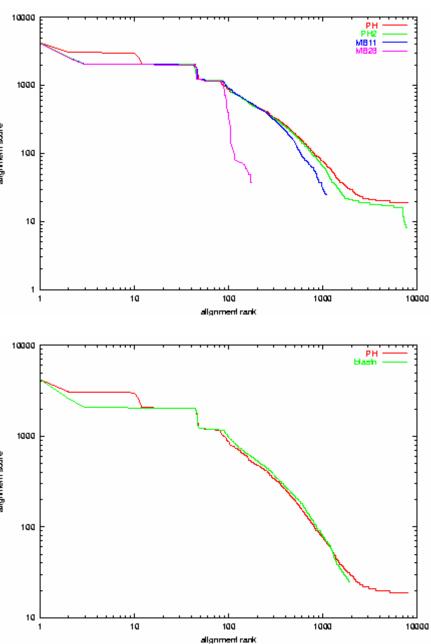
Quality Comparison:

x-axis: alignment rank y-axis: alignment score

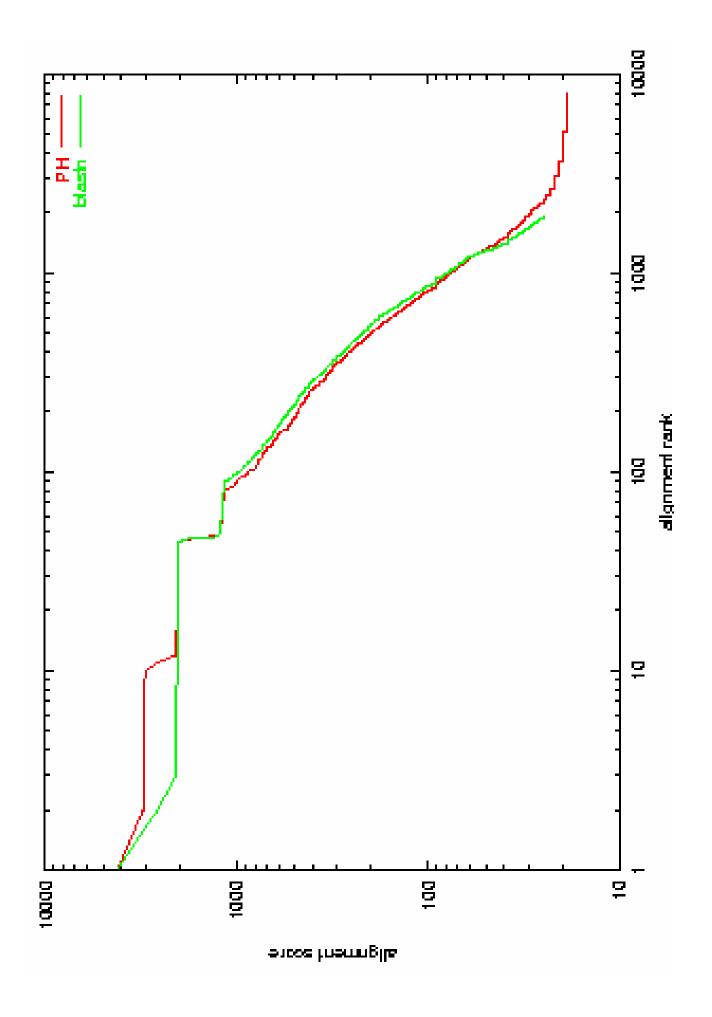
both axes in logarithmic scale

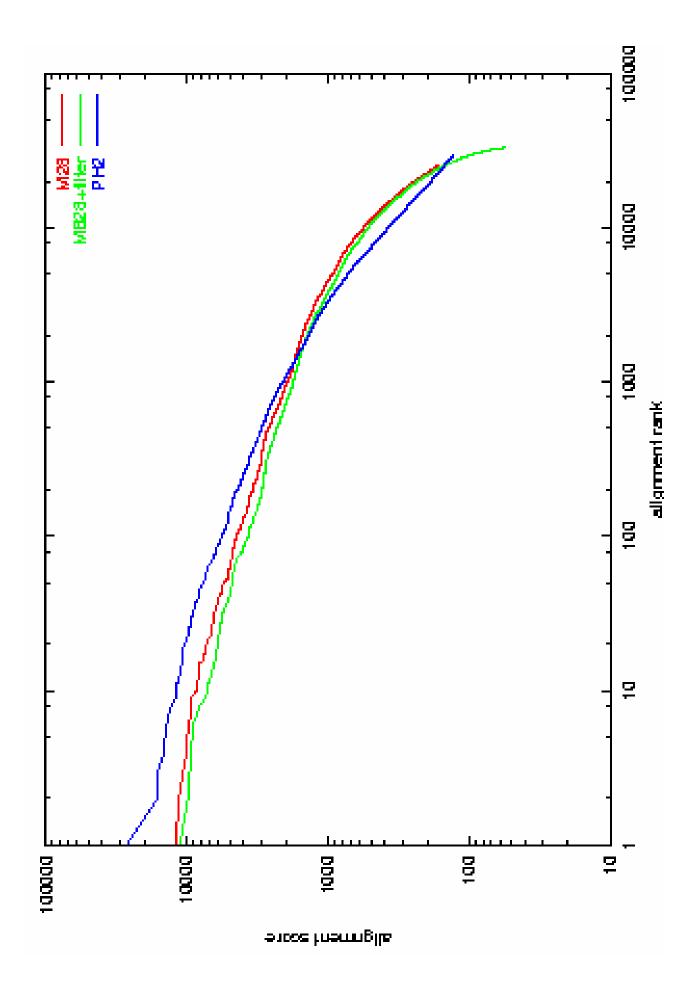


A. thaliana chr 2 vs 4

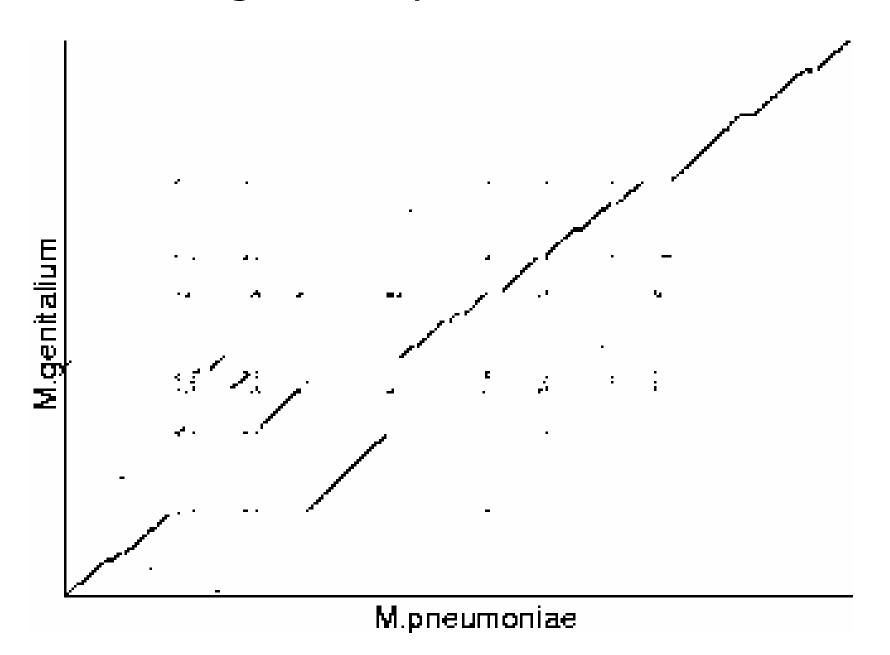


E. Coli vs H. influenza





Genome Alignment by PatternHunter (4 seconds)



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PattternHunter II:

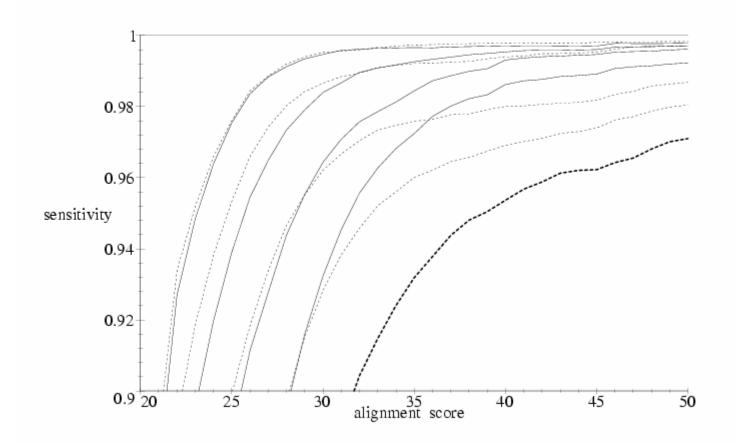


-- Smith-Waterman Sensitivity, BLAST Speed

(Li, Ma, Kisman, Tromp, J. Bioinfo Comput. Biol. 2004)

- The biggest problem for BLAST was low sensitivity (and low speed). Massive parallel machines are built to do S-W exhaustive dynamic programming.
- Spaced seeds give PH a *unique* opportunity of using several optimal seeds to achieve optimal sensitivity, this was not possible by BLAST technology.
- We have designed PH II, with multiple optimal seeds.
- PH II approaches Smith-Waterman sensitivity, and 3000 times faster.
- Experiment: 29715 mouse EST, 4407 human EST.

Sensitivity Comparison with Smith-Waterman (at 100%) The thick dashed curve is the sensitivity of BLAST, seed weight 11. From low to high, the solid curves are the sensitivity of PH II using 1, 2, 4, 8 weight 11 coding region seeds, and the thin dashed curves are the sensitivity 1, 2, 4, 8 weight 11 general purpose seeds, respectively





Speed Comparison with Smith-Waterman

- Smith-Waterman (SSearch): 20 CPU-days.
- PatternHunter II with 4 seeds: 475 CPU-seconds. 3638 times faster than Smith-Waterman dynamic programming at the same sensitivity.

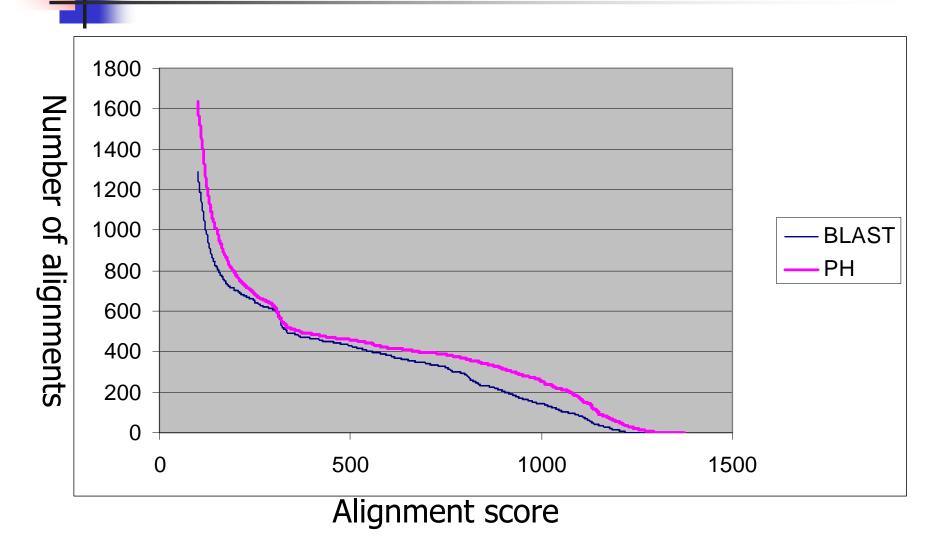


Translated PatternHunter

- Has all the functionalities of
 - Blastp
 - tBlastx with gapped alignments
 - tBlastn, Blastx with gapped alignments
- More sensitive and faster new algorithm replacing 6-frame translation

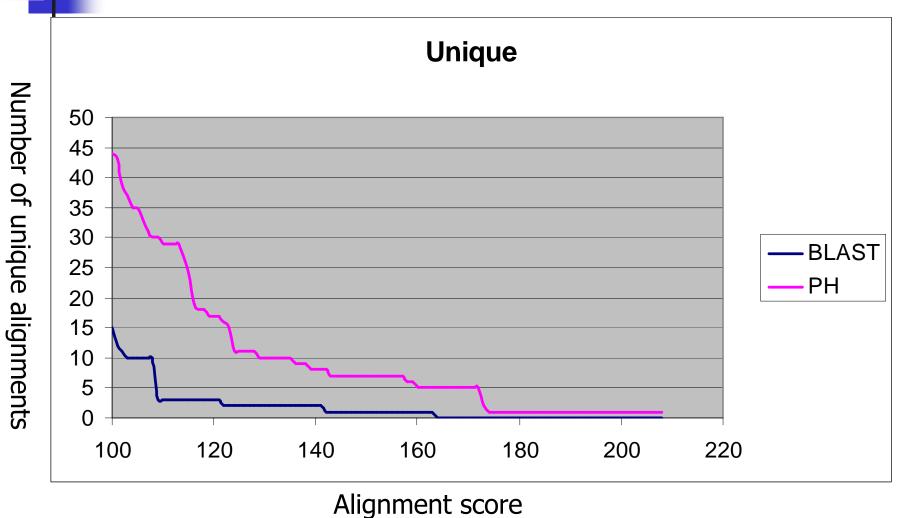
Alignment comparison: tBLASTx vs tPH

tPH: 253 seconds tBLASTx: 807 seconds



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Unique Alignments: tBLASTx vs tPH





Old field, new trend

Research trend

- Over 30 papers on spaced seeds have appeared since our original paper, in 2 years.
- Many more have used PH in their work.
- Most modern alignment programs (including BLAST) have now adopted spaced seeds
- Spaced seeds are serving thousands of users/day

PatternHunter direct users

- Pharmaceutical/biotech firms.
- Mouse/Rat Genome Consortiums, Nature, Dec. 5, 2002.
- Hundreds of academic institutions.

Running PH

Available at: www.BioinformaticsSolutions.com

Java –Xmx512m –jar ph.jar –i query.fna –j subject.fna –o out.txt

- -Xmx512m --- for large files
- -j missing: query.fna self-comparison
- -db: multiple sequence input, 0,1,2,3 (no, query, subject, both)
- -W: seed weight
- -G: open gap penalty (default 5)
- -E: gap extension (default 1)
- -q: mismatch penalty (default 1)
- -r: reward for match (default 1)
- -model: specify model in binary
- -H: hits before extension
- -P: show progress
- -multi 4: use 4 seeds

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Meaningful Match?

- n Given a gene sequence, BLAST or PH simply returns a bunch of meaningless alignments.
- Can we return a complete gene match?
- Idea: Combine PH with ExonHunter (Brejova, Brown, Li, Vinar, ISMB'2005): take the ab initio genefinder (HMM) trained for the database genome, further train/bias it with the query gene model (its splice sites etc). Use PH to find possible hot regions and use this HMM to do extension, deciding on introns/exons.

Example:

Given a human gene [GI:35560], want a homologous gene in mouse genome [GI:293767]



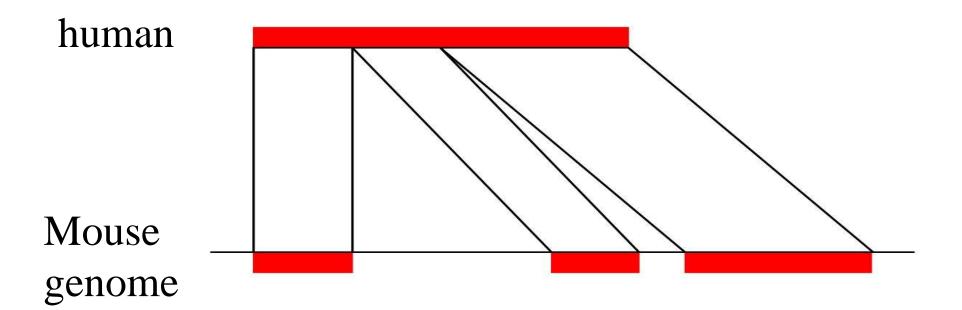
BLAST Result

- 249 alignments are returned
- only 3 alignments are relevant
- Exons / Splice sites are not detected



New gPH results

Fully correct homologous gene-match is returned. Just one alignment!





Best ideas are simple ones.

The most difficult theoretical studies are those that actually work.

Open questions:

- Polynomial time probabilistic algorithm for finding (near) optimal seed, multiple seeds.
- gPH for distant species, via 3D modeling?



Acknowledgement

- PH is joint work with Bin Ma and John Tromp
- PH II is joint work with Ma, Kisman, and Tromp
- Some joint theoretical work with Ma, Keich, Tromp, Xu, and Brown.
- gPH is joint work with X.F. Cui, D. Shasha, T. Vinar.
- Financial support: Bioinformatics Solutions Inc, NSERC, Killam Fellowship, Steacie Fellowship, CRC chair program.



Conclusion – continued

- Another simple idea applied to data mining: from irreversibly computing 1 bit requires 1kT energy (von Neumann, Landauer), we derived shared information d(x,y) between x,y, to classify
 - Species & genomes, Li et al, in Bioinformatics, 2001
 - ⁿ Chain letters, Bennett, Li, Ma, *Scientific American*, 2003
 - Languages, Benedeto, Caglioti, Loreto, Phy. Rev. Let. '02
 - Music, Cilibrasi, Vitanyi, de Wolf, New Scientist, 2003
 - Time series/anomaly detection, Keogh, Lonardi, Ratanamahatana, KDD'04. They compared d(x,y) with 51 methods/measures from SIGKDD, SIGMOD, ICDM, ICDE, SSDB, VLDB, PKDD, PAKDD and concluded our method the simplest & best --- Keogh tutorial ICDM'04.

PH 2-hit sensitivity vs BLAST 11, 12 1-hit

