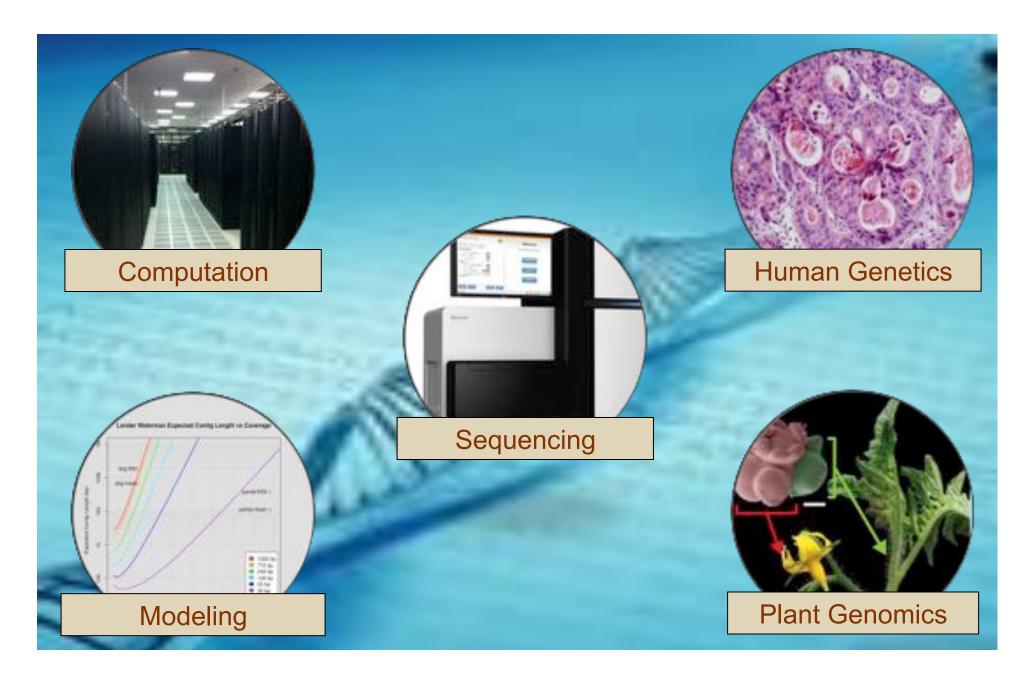
# Sequence Alignment & Computational Thinking Michael Schatz

Sept 23, 2013 SBU Introduction to Physical and Quantitative Biology



## Schatz Lab Overview



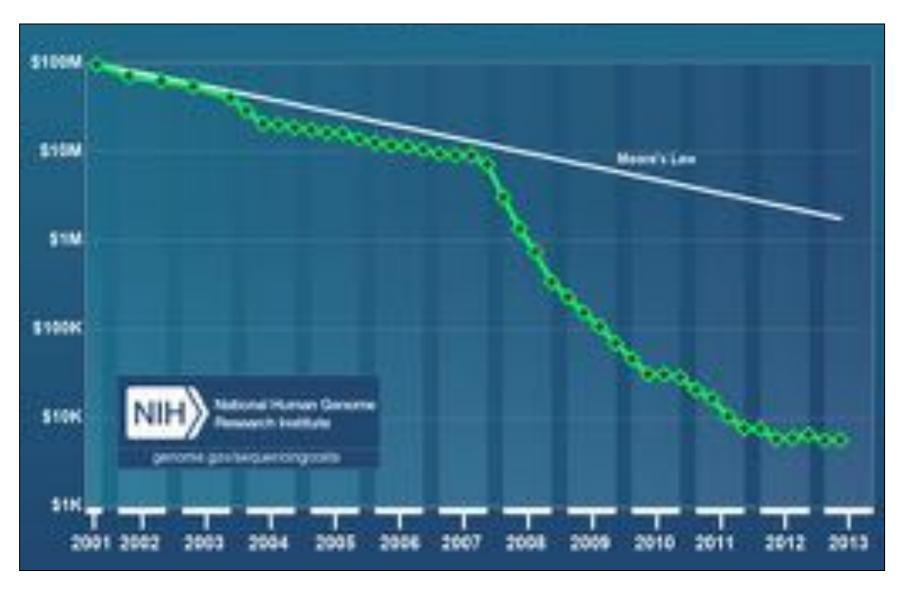


## Outline

- I. Rise of DNA Sequencing
- 2. Sequence Alignment Basics

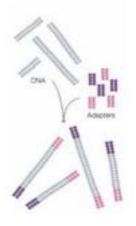
- 3. Understanding Bowtie
- 4. Genetics of Autism

## Cost per Genome

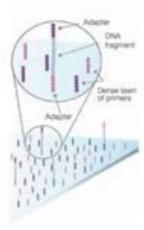


http://www.genome.gov/sequencingcosts/

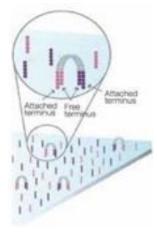
# Illumina Sequencing by Synthesis



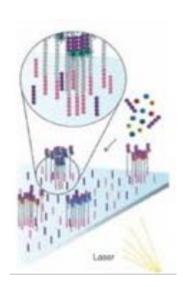
1. Prepare



2. Attach



3. Amplify



4. Image













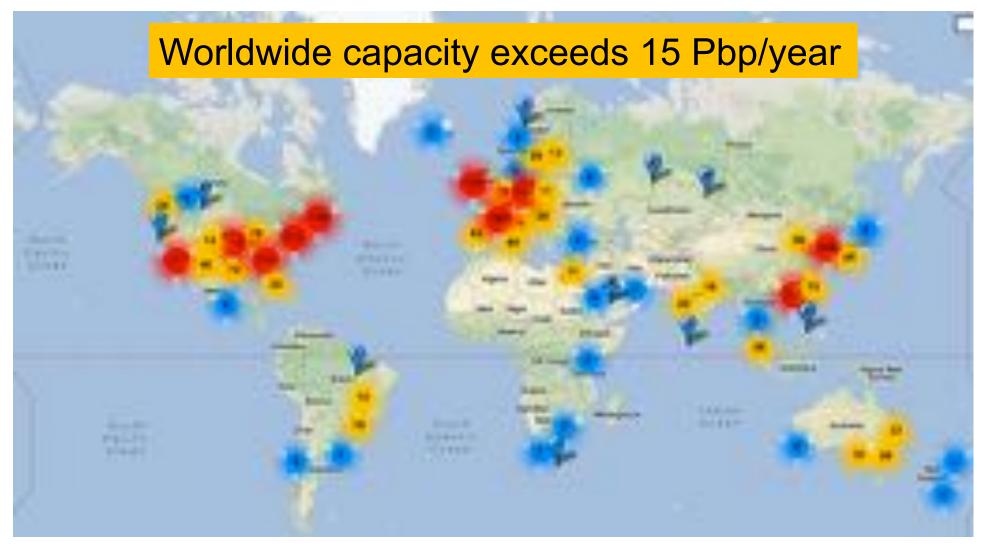
5. Basecall

## Inside the NY Genome Center

Sequencing Capacity: 16 HiSeq 2500 @ 600 Gbp / 11 day = 872 Gbp / day



# Sequencing Centers



Next Generation Genomics: World Map of High-throughput Sequencers http://omicsmaps.com

Milestones in Molecular Biology

#### There is tremendous interest to sequence:

- What is your genome sequence?
- How does your genome compare to my genome?
- Where are the genes and how active are they?
- How does gene activity change during development?
- How does splicing change during development?
- How does methylation change during development?
- How does chromatin change during development?
- How does is your genome folded in the cell?
- Where do proteins bind and regulate genes?
- What virus and microbes are living inside you?
- How has the disease mutated your genome?
- What drugs should we give you?







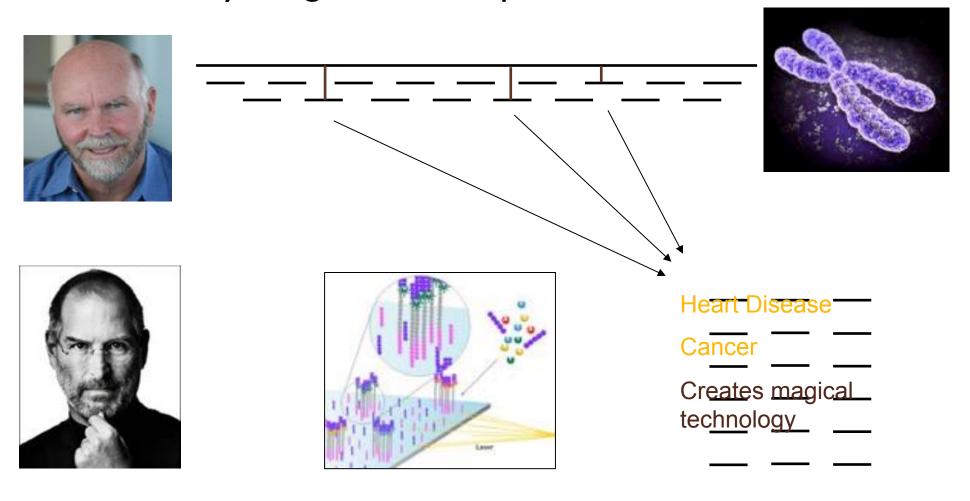
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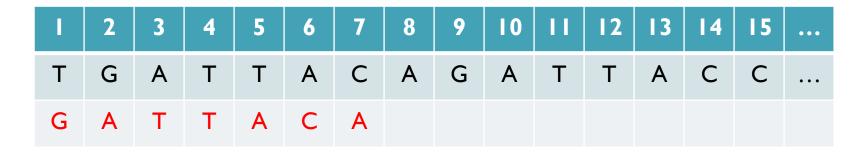
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#### Personal Genomics

How does your genome compare to the reference?



- Where is GATTACA in the human genome?
- Strategy I: Brute Force



No match at offset I

- Where is GATTACA in the human genome?
- Strategy I: Brute Force

I	2	3	4	5	6	7	8	9	10	П	12	13	14	15	•••
Т	G	Α	Т	Т	Α	С	Α	G	Α	Т	Т	Α	С	С	• • •
	G	Α	Т	Т	Α	С	Α								

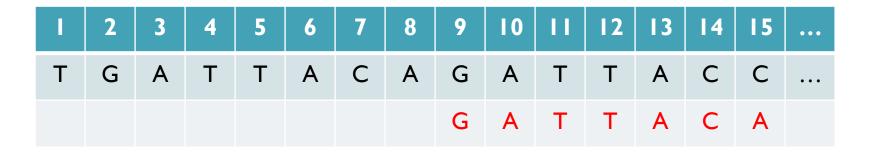
Match at offset 2

- Where is GATTACA in the human genome?
- Strategy I: Brute Force

1	2	3	4	5	6	7	8	9	10	Ш	12	13	14	15	•••
Т	G	Α	Т	Т	Α	С	Α	G	Α	Т	Т	Α	С	С	•••
		G	Α	Т	Т	Α	С	Α	•••						

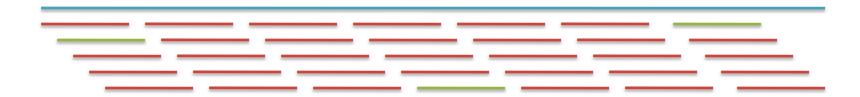
No match at offset 3...

- Where is GATTACA in the human genome?
- Strategy I: Brute Force



No match at offset 9 <- Checking each possible position takes time

#### Brute Force Analysis



- Brute Force:
  - At every possible offset in the genome:
    - Do all of the characters of the query match?
- Analysis
  - Simple, easy to understand

<ul><li>Genome length = n</li></ul>	[3B]
— Query length = m	[7]
<ul><li>Comparisons: (n-m+1) * m</li></ul>	[21B]

Overall runtime: O(nm)

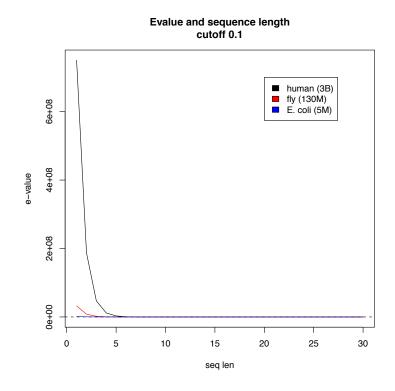
[How long would it take if we double the genome size, read length?] [How long would it take if we double both?]

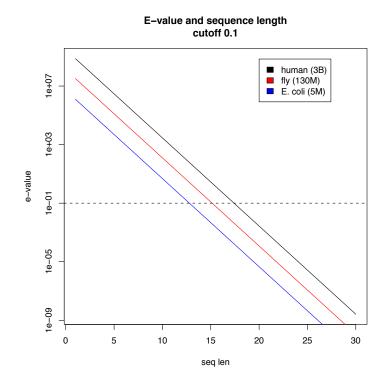
## **Expected Occurrences**

The expected number of occurrences (e-value) of a given sequence in a genome depends on the length of the genome and inversely on the length of the sequence

- I in 4 bases are G, I in 16 positions are GA, I in 64 positions are GAT, ...
- I in 16,384 should be GATTACA
- $E=n/(4^{m})$

[183,105 expected occurrences] [How long do the reads need to be for a significant match?]





#### **Brute Force Reflections**

#### Why check every position?

- GATTACA can't possibly start at position 15

[WHY?]

1	2	3	4	5	6	7	8	9	10	Ш	12	13	14	15	•••
Т	G	Α	Т	Т	Α	С	Α	G	Α	Т	Т	Α	С	С	•••
								G	Α	Т	Т	Α	С	Α	

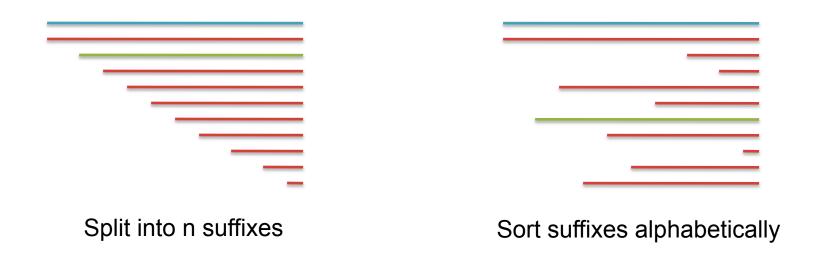
Improve runtime to O(n + m)

[3B + 7]

- If we double both, it just takes twice as long
- Knuth-Morris-Pratt, 1977
- Boyer-Moyer, 1977, 1991
- For one-off scans, this is the best we can do (optimal performance)
  - We have to read every character of the genome, and every character of the query
  - For short queries, runtime is dominated by the length of the genome

## Suffix Arrays: Searching the Phone Book

- What if we need to check many queries?
  - We don't need to check every page of the phone book to find 'Schatz'
  - Sorting alphabetically lets us immediately skip 96% (25/26) of the book without any loss in accuracy
- Sorting the genome: Suffix Array (Manber & Myers, 1991)
  - Sort every suffix of the genome



[Challenge Question: How else could we split the genome?]

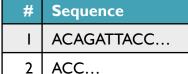
- Strategy 2: Binary search
  - Compare to the middle, refine as higher or lower
- Searching for GATTACA
  - Lo = I; Hi = I5;



#	Sequence	Pos
Ι	ACAGATTACC	6
2	ACC	13
3	AGATTACC	8
4	ATTACAGATTACC	3
5	ATTACC	10
6	C	15
7	CAGATTACC	7
8	CC	14
9	GATTACAGATTACC	2
10	GATTACC	9
П	TACAGATTACC	5
12	TACC	12
13	TGATTACAGATTACC	I
14	TTACAGATTACC	4
15	TTACC	П

Hi

- Strategy 2: Binary search
  - Compare to the middle, refine as higher or lower
- Searching for GATTACA
  - Lo = I; Hi = I5; Mid = (I+I5)/2 = 8
  - Middle = Suffix[8] = CC



Hi

Lo

l	ACAGATTACC	6
2	ACC	13
3	AGATTACC	8
4	ATTACAGATTACC	3
5	ATTACC	10
6	C	15
7	CAGATTACC	7
8	CC	14
9	GATTACAGATTACC	2
10	GATTACC	9
П	TACAGATTACC	5
12	TACC	12
12	TACC TGATTACAGATTACC	12 I

Pos

- Strategy 2: Binary search
  - Compare to the middle, refine as higher or lower
- Searching for GATTACA
  - Lo = I; Hi = I5; Mid = (I+I5)/2 = 8
  - Middle = Suffix[8] = CC => Higher: Lo = Mid + I



#	Sequence	Pos
	ACAGATTACC	6
2	ACC	13
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  - Lo = 9; Hi = 15;

#	Sequence	Pos
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10	GATTACC	9
П	TACAGATTACC	5
12	TACC	12
13	TGATTACAGATTACC	I
14	TTACAGATTACC	4
15	TTACC	П

Lo

Ηį

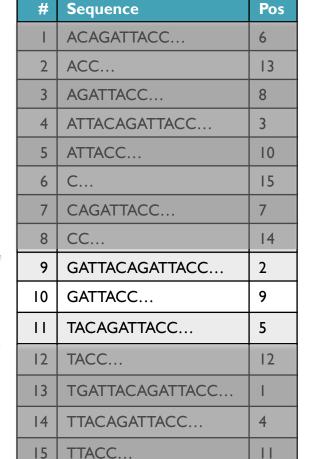
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  - Compare to the middle, refine as higher or lower
- Searching for GATTACA
  - Lo = I; Hi = I5; Mid = (I+I5)/2 = 8
  - Middle = Suffix[8] = CC=> Higher: Lo = Mid + I
  - Lo = 9; Hi = 15; Mid = (9+15)/2 = 12
  - Middle = Suffix[12] = TACC

#	Sequence	Pos
I	ACAGATTACC	6
2	ACC	13
3	AGATTACC	8
4	ATTACAGATTACC	3
5	ATTACC	10
6	C	15
7	CAGATTACC	7
8	CC	14
9	GATTACAGATTACC	2
10	GATTACC	9
П	TACAGATTACC	5
12	TACC	12
13	TGATTACAGATTACC	I
14	TTACAGATTACC	4
15	TTACC	11

Hi

Lo

- Strategy 2: Binary search
  - Compare to the middle, refine as higher or lower
- Searching for GATTACA
  - Lo = I; Hi = 15; Mid = (1+15)/2 = 8
  - Middle = Suffix[8] = CC=> Higher: Lo = Mid + I
  - Lo = 9; Hi = 15; Mid = (9+15)/2 = 12
  - Middle = Suffix[12] = TACC=> Lower: Hi = Mid I
  - Lo = 9; Hi = 11;







- Strategy 2: Binary search
  - Compare to the middle, refine as higher or lower
- Searching for GATTACA
  - Lo = I; Hi = 15; Mid = (1+15)/2 = 8
  - Middle = Suffix[8] = CC=> Higher: Lo = Mid + I
  - Lo = 9; Hi = 15; Mid = (9+15)/2 = 12
  - Middle = Suffix[12] = TACC=> Lower: Hi = Mid I
  - Lo = 9; Hi = 11; Mid = (9+11)/2 = 10
  - Middle = Suffix[10] = GATTACC

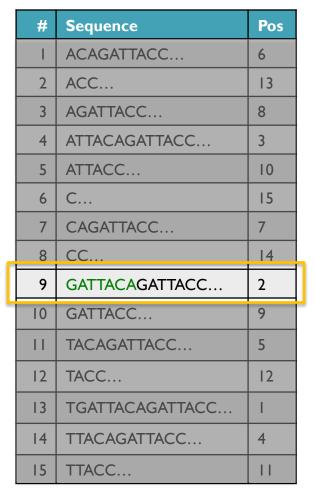
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  - Middle = Suffix[8] = CC=> Higher: Lo = Mid + I
  - Lo = 9; Hi = 15; Mid = (9+15)/2 = 12
  - Middle = Suffix[12] = TACC=> Lower: Hi = Mid I
  - Lo = 9; Hi = 11; Mid = (9+11)/2 = 10
  - Middle = Suffix[10] = GATTACC=> Lower: Hi = Mid I
  - Lo = 9; Hi = 9;



#	Sequence	Pos
I	ACAGATTACC	6
2	ACC	13
3	AGATTACC	8
4	ATTACAGATTACC	3
5	ATTACC	10
6	C	15
7	CAGATTACC	7
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11	TACAGATTACC	5
12	TACC	12
13	TGATTACAGATTACC	1
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- Strategy 2: Binary search
  - Compare to the middle, refine as higher or lower
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  - Lo = I; Hi = 15; Mid = (1+15)/2 = 8
  - Middle = Suffix[8] = CC
     => Higher: Lo = Mid + I
  - Lo = 9; Hi = 15; Mid = (9+15)/2 = 12
  - Middle = Suffix[12] = TACC=> Lower: Hi = Mid I
  - Lo = 9; Hi = 11; Mid = (9+11)/2 = 10
  - Middle = Suffix[10] = GATTACC=> Lower: Hi = Mid I
  - Lo = 9; Hi = 9; Mid = (9+9)/2 = 9
  - Middle = Suffix[9] = GATTACA...=> Match at position 2!





#### Binary Search Analysis

Binary Search

```
Initialize search range to entire list

mid = (hi+lo)/2; middle = suffix[mid]

if query matches middle: done

else if query < middle: pick low range

else if query > middle: pick hi range

Repeat until done or empty range
```

[WHEN?]

- Analysis
  - More complicated method
  - How many times do we repeat?
    - How many times can it cut the range in half?
    - Find smallest x such that:  $n/(2^x) \le 1$ ;  $x = \lg_2(n)$

[32]

- Total Runtime: O(m lg n)
  - More complicated, but much faster!
  - Looking up a query loops 32 times instead of 3B

[How long does it take to search 6B or 24B nucleotides?]





## Outline

- I. Rise of DNA Sequencing
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- 3. Understanding Bowtie
- 4. Genetics of Autism





# Fast gapped-read alignment with Bowtie 2

Ben Langmead and Steven Salzberg (2012) Nature Methods. 9, 357–359

## In-exact alignment

- Where is GATTACA approximately in the human genome?
  - And how do we efficiently find them?
- It depends...
  - Define 'approximately'
    - Hamming Distance, Edit distance, or Sequence Similarity
    - Ungapped vs Gapped vs Affine Gaps
    - Global vs Local
    - All positions or the single 'best'?
  - Efficiency depends on the data characteristics & goals
    - Smith-Waterman: Exhaustive search for optimal alignments
    - BLAST: Hash-table based homology searches
    - Bowtie: BWT alignment for short read mapping

• Where is GATTACA approximately in the human genome?

1	2	3	4	5	6	7	8	9	10	Ш	12	13	14	15	•••
Т	G	Α	Т	Т	Α	С	Α	G	Α	Т	Т	Α	С	С	•••
G	Α	Т	Т	Α	С	Α									

Match Score: 1/7

• Where is GATTACA approximately in the human genome?

1	2	3	4	5	6	7	8	9	10	Ш	12	13	14	15	•••
Т	G	Α	Т	Т	Α	С	Α	G	Α	Т	Т	Α	С	С	•••
	G	Α	Т	Т	Α	С	Α								

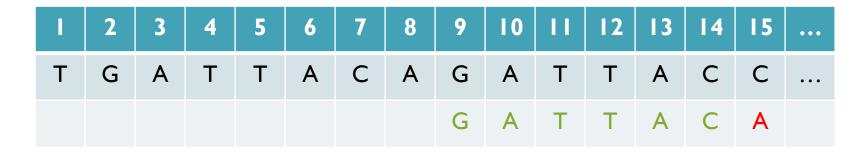
Match Score: 7/7

• Where is GATTACA approximately in the human genome?

1	2	3	4	5	6	7	8	9	10	П	12	13	14	15	•••
Т	G	Α	Т	Т	Α	С	Α	G	Α	Т	Т	Α	С	С	• • •
		G	A	Т	Т	Α	С	Α	•••						

Match Score: 1/7

Where is GATTACA approximately in the human genome?



Match Score: 6/7 <- We may be very interested in these imperfect matches Especially if there are no perfect end-to-end matches

# Similarity metrics

#### Hamming distance

Count the number of substitutions to transform one string into another

GATTACA	GATTTTTACA
GATCACA	GATTACA
1	6

#### • Edit distance

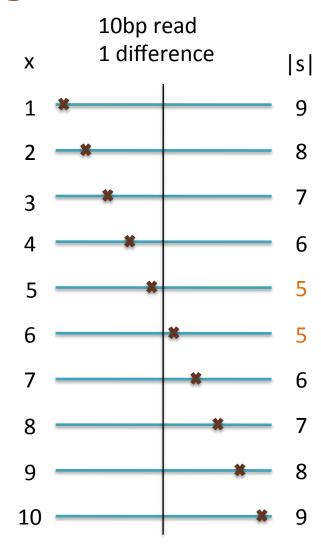
 The minimum number of substitutions, insertions, or deletions to transform one string into another

GATTACA	GATTTTTACA
	xxx
GATCACA	GATTACA
1	3

## Seed-and-Extend Alignment

Theorem: An alignment of a sequence of length m with at most k differences must contain an exact match at least s=m/(k+1) bp long (Baeza-Yates and Perleberg, 1996)

- Proof: Pigeonhole principle
  - I pigeon can't fill 2 holes
- Seed-and-extend search
  - Use an index to rapidly find short exact alignments to seed longer in-exact alignments
    - BLAST, MUMmer, Bowtie, BWA, SOAP, ...
  - Specificity of the depends on seed length
    - Guaranteed sensitivity for k differences
    - Also finds some (but not all) lower quality alignments <- heuristic</li>



## Algorithm Overview

1. Split read into segments

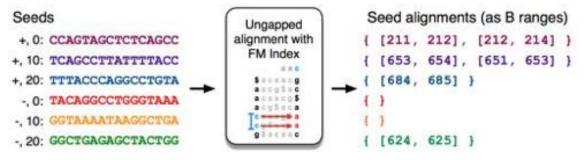
Read Read (reverse complement)

CCAGTAGCTCTCAGCCTTATTTTACCCAGGCCTGTA TACAGGCCTGGGTAAAATAAGGCTGAGAGCTACTGG

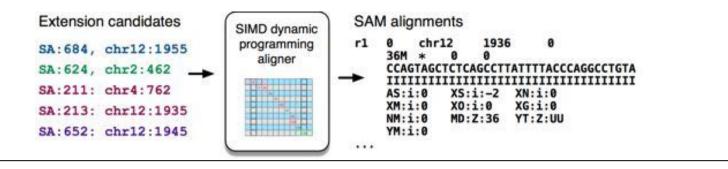
Policy: extract 16 nt seed every 10 nt

Seeds
+, 0: CCAGTAGCTCTCAGCC
+, 0: TACAGGCCTGGTAAA
+, 10: TCAGCCTTATTTTACC
+, 20: TTTACCCAGGCCTGTA
-, 20: GGCTGAGAGCTACTGG

2. Lookup each segment and prioritize



3. Evaluate end-to-end match





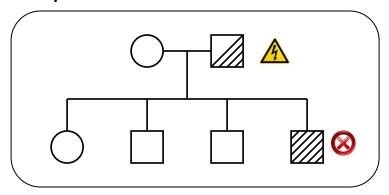
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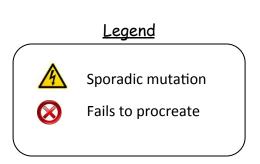
### Unified Model of Autism

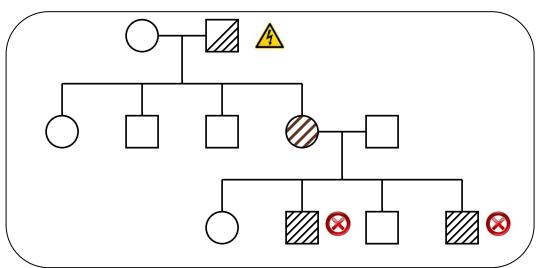
Sporadic Autism: 1 in 100



**Prediction**: De novo mutations of high penetrance contributes to autism, especially in low risk families with no history of autism.

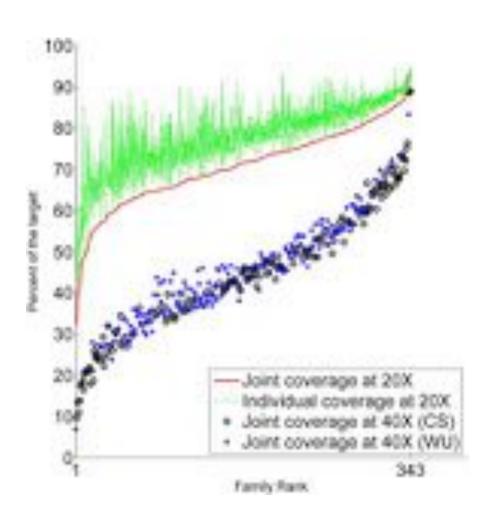
Familial Autism: 90% concordance in twins





A unified genetic theory for sporadic and inherited autism Zhao et al. (2007) PNAS. 104(31)12831-12836.

## Exome-Capture and Sequencing



Sequencing of 343 families from the Simons Simplex Collection

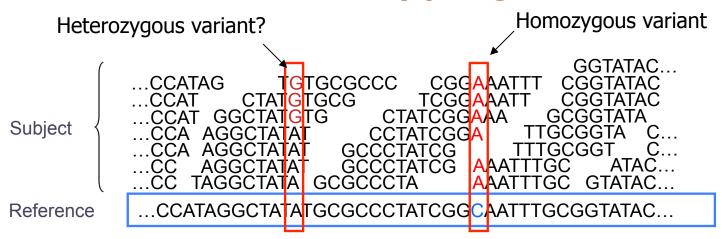
- Parents plus one child with autism and one non-autistic sibling
- Enriched for higher-functioning individuals

Families prepared and captured together to minimize batch effects

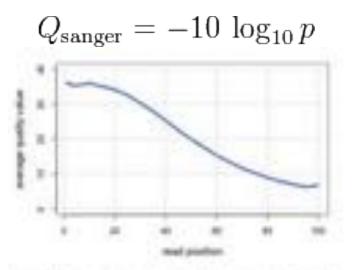
- Exome-capture performed with NimbleGen SeqCap EZ Exome v2.0 targeting 36 Mb of the genome.
- ~80% of the target at >20x coverage with ~93bp reads

De novo gene disruptions in children on the autism spectrum lossifov et al. (2012) Neuron. 74:2 285-299

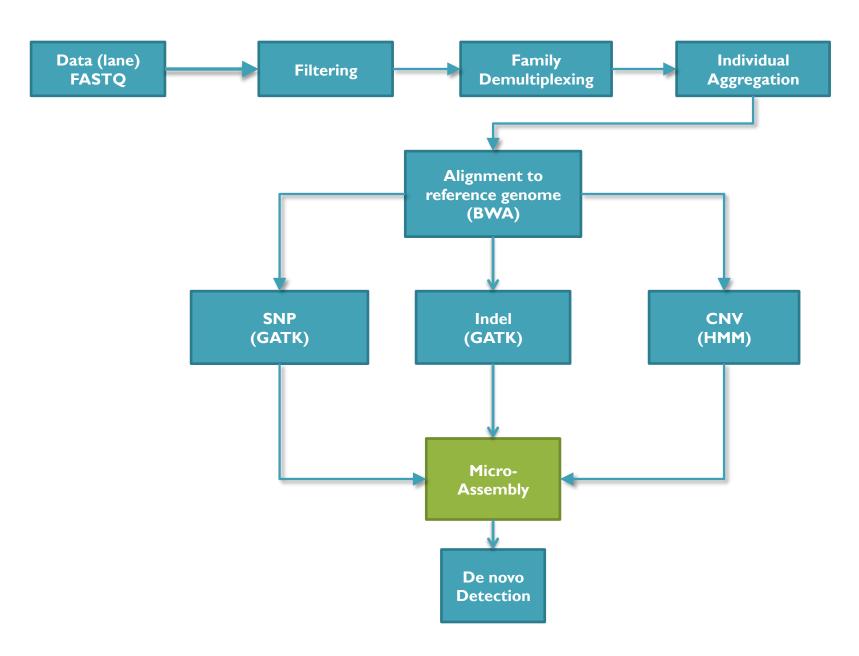
### Genotyping



- Sequencing instruments make mistakes
  - Quality of read decreases over the read length
- A single read differing from the reference is probably just an error, but it becomes more likely to be real as we see it multiple times
  - Often framed as a Bayesian problem of more likely to be a real variant or chance occurrence of N errors
  - Accuracy improves with deeper coverage



## **Exome Sequencing Pipeline**



### Scalpel: Haplotype Microassembly

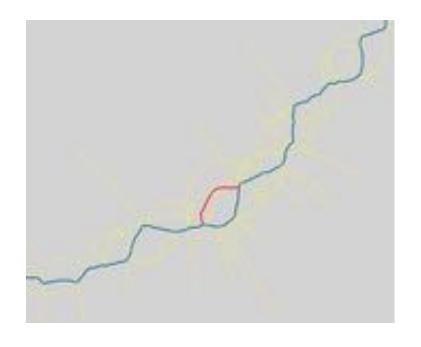
G. Narzisi, J. O'Rawe, I. Iossifov, Y. Lee, Z. Wang, G. Lyon, M. Wigler, and M. C. Schatz

DNA sequence **micro-assembly** pipeline for accurate detection and validation of *de novo* mutations (SNPs, indels) within exome-capture data.



#### **Features**

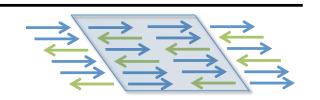
- I. Combine mapping and assembly
- 2. Exhaustive search of haplotypes
- De novo mutations



NRXN1 de novo SNP (auSSC12501 chr2:50724605)

### Scalpel Pipeline

Extract reads mapping within the exon including (1) well-mapped reads, (2) soft-clipped reads, and (3) anchored pairs



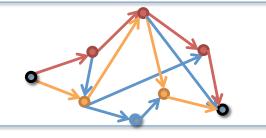


Decompose reads into overlapping *k*-mers and construct de Bruijn graph from the reads





Find end-to-end haplotype paths spanning the region





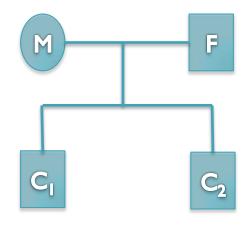
Align assembled sequences to reference to detect mutations



### De novo mutation discovery and validation

**Concept:** Identify mutations not present in parents.

Challenge: Sequencing errors in the child or low coverage in parents lead to false positive de novos



```
Father: ...TCAGAACAGCTGGATGAGATCTTAGCCAACTACCAGGAGATTGTCTTTGCCCGGA...

Mother: ...TCAGAACAGCTGGATGAGATCTTAGCCAACTACCAGGAGATTGTCTTTGCCCGGA...

Sib: ...TCAGAACAGCTGGATGAGATCTTAGCCAACTACCAGGAGATTGTCTTTGCCCGGA...

Aut(1): ...TCAGAACAGCTGGATGAGATCTTAGCCAACTACCAGGAGATTGTCTTTGCCCGGA...

Aut(2): ...TCAGAACAGCTGGATGAGATCTTAGCCAACTACCAGGAGATTGTCTTTGCCCGGA...
```

6bp heterozygous deletion at chr13:25280526 ATP12A

### De novo Genetics of Autism

- In 343 family quads so far, we see significant enrichment in de novo *likely gene killers* in the autistic kids
  - Overall rate basically 1:1 (432:396)
  - 2:1 enrichment in nonsense mutations
  - 2:1 enrichment in frameshift indels
  - 4:1 enrichment in splice-site mutations
  - Most de novo originate in the paternal line in an age-dependent manner (56:18 of the mutations that we could determine)
- Observe strong overlap with the 842 genes known to be associated with fragile X protein FMPR
  - Related to neuron development and synaptic plasticity

De novo gene disruptions in children on the autism spectrum lossifov et al. (2012) Neuron. 74:2 285-299

### Computational Biology

"Computer science is no more about computers than astronomy is about telescopes." Edsger Dijkstra

- Computer Science = Science of Computation
  - Solving problems, designing & building systems
  - Computers are very, very dumb, but we can instruct them
    - Build complex systems out of simple components
    - They will perfectly execute instructions forever
- CompBio = Thinking Computationally about Biology
  - Processing: Make more powerful instruments, analyze results
  - Designing & Understanding: protocols, procedures, systems

"Think Harder & Compute Less"

Dan Gusfield

Recommended: CSE 549 - Introduction to Computational Biology



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# Questions?

http://schatzlab.cshl.edu

