

# CS 68: BIOINFORMATICS

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Swarthmore College  
Spring 2018

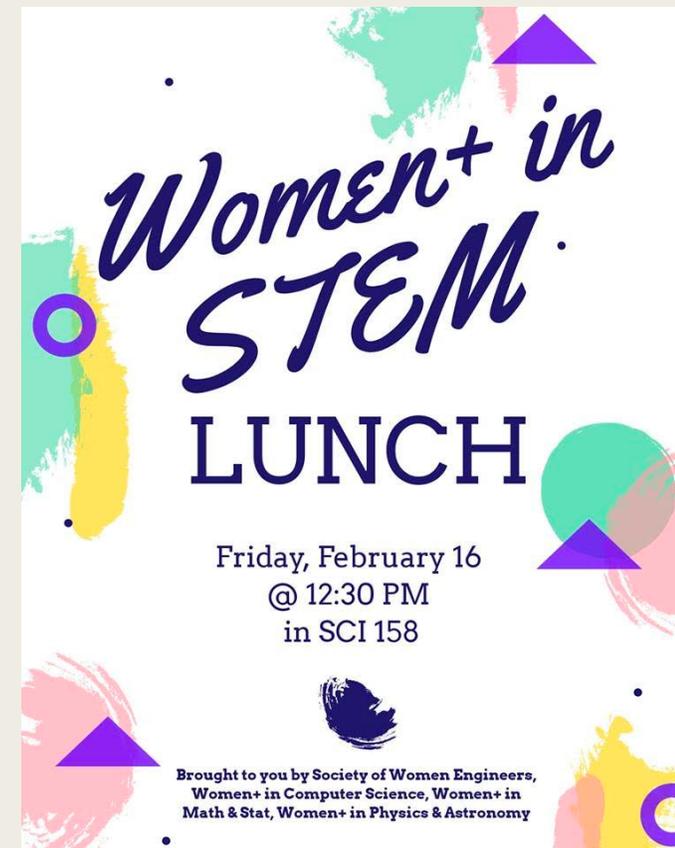


# Outline: Feb 16

- Finish: FM-index and practical read mapping
- Today: link between “getting data” and “using data”
- Introduce phylogenetic trees and evolutionary process of mutations

## Notes:

- Remote edit new instructions on Piazza (and website)
- Welcome to use pandas (make sure it works on lab machines)



# Lab 4 Notes

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- For the reverse BWT part (i.e. getting back the original string from the BWT), you should not be creating any additional data structures beyond what you needed for mapping a read
- Optional: implement sorting the rotations (cyclic permutations) from the original string only

..... *i* abcg..... *j* abcd..... \$

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`g > d therefore Rotation(i) > Rotation(j)`

..... *i* abc**g** ..... *j* abcd ..... \$

	<i>A</i>
abcd..... \$..... abcd.....	<i>j</i>
⋮	
abcg..... abcg..... \$.....	<i>i</i>

Quick jump back to local alignment

Bioinformatics is not very different from mathematics; the literature is populated with many Amerindian eggs. My favorite example is the [Smith-Waterman algorithm](#), an algorithm for local alignment [published by Temple Smith and Michael Waterman in 1981](#). The Smith-Waterman algorithm is a simple modification of the [Needleman-Wunsch algorithm](#):

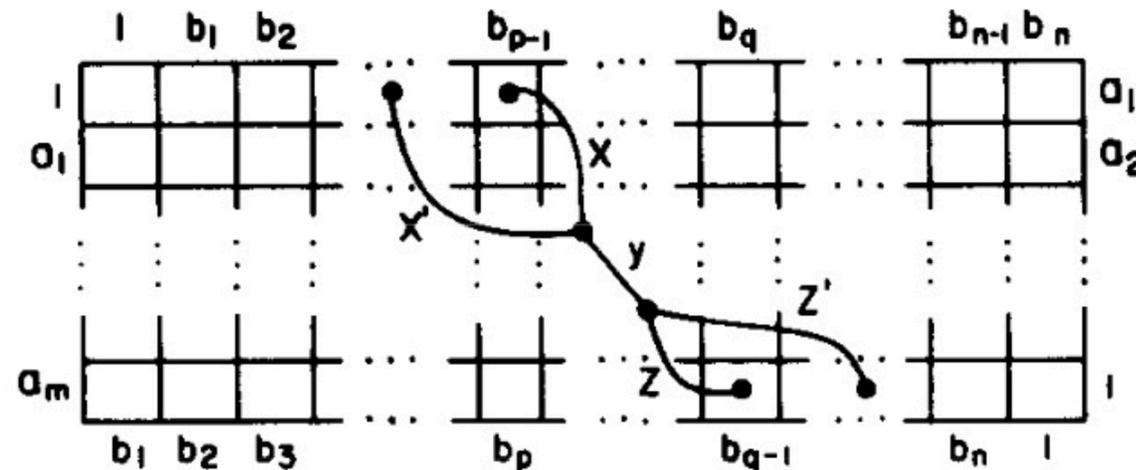
1970

	Smith-Waterman algorithm	Needleman-Wunsch algorithm
Initialization	First row and first column are set to 0	First row and first column are subject to gap penalty
Scoring	Negative score is set to 0	Score can be negative
Traceback	Begin with the highest score, end when 0 is encountered	Begin with the cell at the lower right of the matrix, end at top left cell

From Lior Pachter's  
(professor at  
Caltech) blog

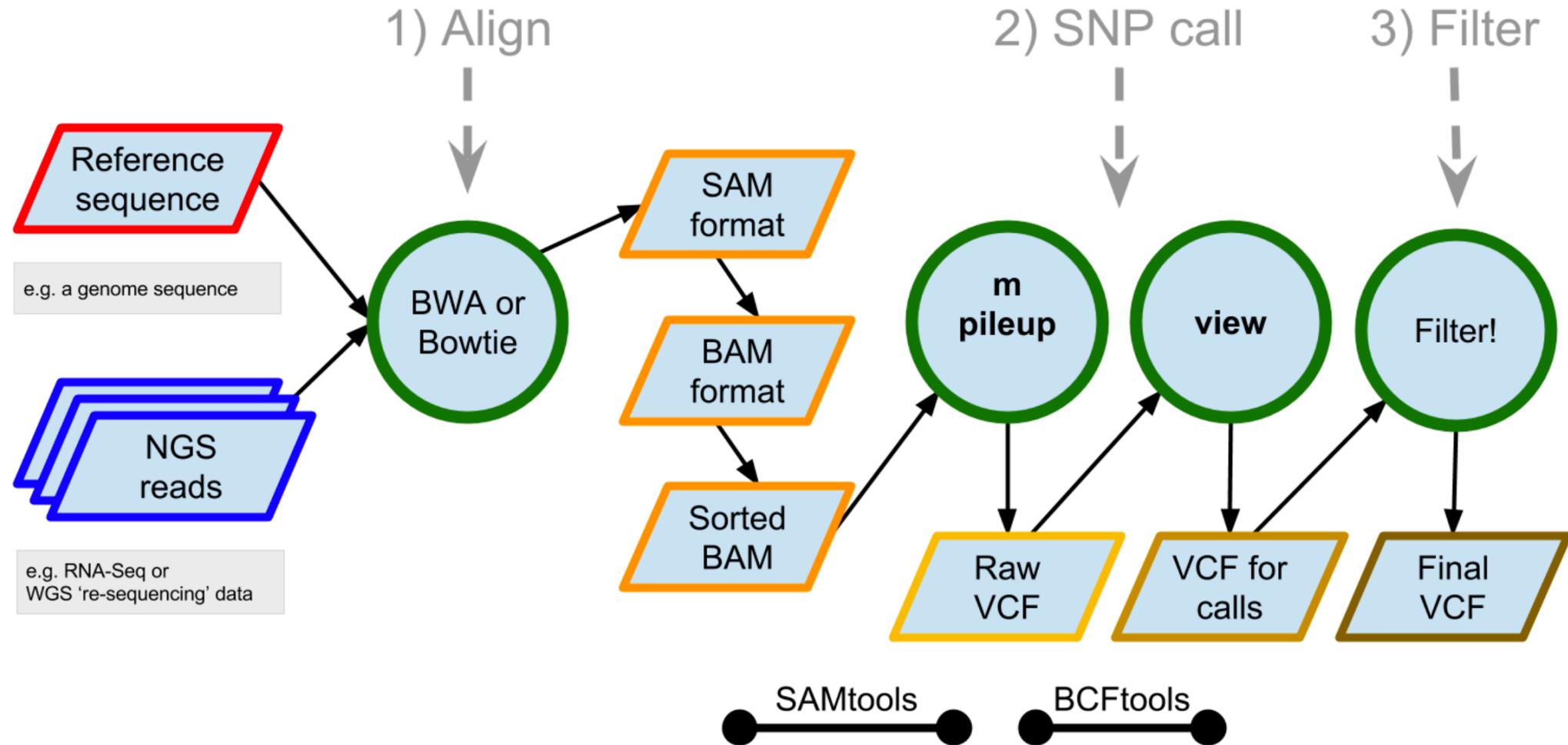
Feb. 15, 2018

The table above shows the differences. **That's it!** This table made for a (highly cited) paper. Just initialize the Needleman-Wunsch algorithm with zeroes instead of a gap penalty, set negative scores to 0, trace back from the highest score. In fact, it's such a minor modification that when I first learned the details of the algorithm I thought "This is obvious! After all, it's *just* the Needleman-Wunsch algorithm. Why does it even have a name?! Smith and Waterman got a highly cited paper?! For *this?!*" My skepticism lasted only as long as it took me to discover and read [Peter Sellers' 1980 paper](#) attempting to solve the same problem. It's a lot more complicated, relying on the idea of "inductive steps", and requires untangling mysterious diagrams such as:



# Practical Read Mapping

# Pipeline overview



# Bowtie and BWA (posted reading)

- First practical read aligners to use the Burrows-Wheeler Transform

## Fast and accurate short read alignment with Burrows–Wheeler transform

Heng Li, Richard Durbin  [Author Notes](#)

*Bioinformatics*, Volume 25, Issue 14, 15 July 2009 Pages 1754–1760,

<https://doi.org/10.1093/bioinformatics/btp324>

**Published:** 18 May 2009 **Article history** ▼

BWA

Bowtie

## Ultrafast and memory-efficient alignment of short DNA sequences to the human genome

[Ben Langmead](#) , [Cole Trapnell](#), [Mihai Pop](#) and [Steven L Salzberg](#)

*Genome Biology* 2009 10:R25

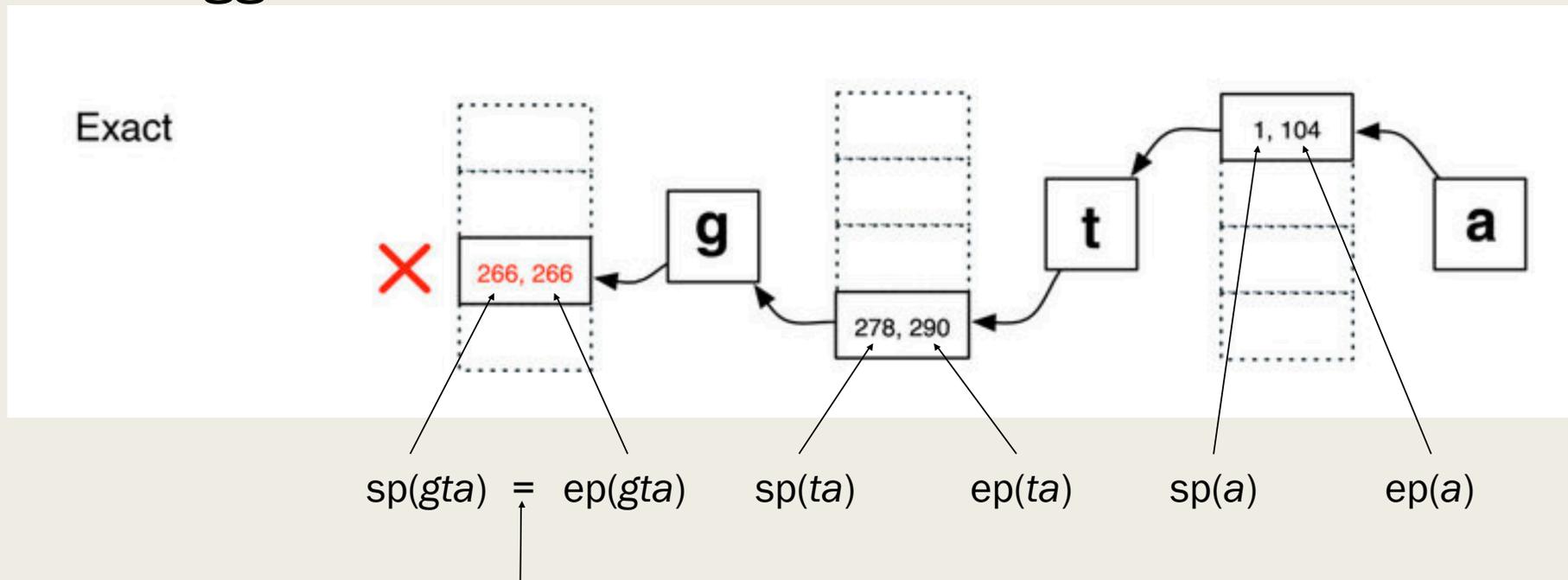
<https://doi.org/10.1186/gb-2009-10-3-r25> | © Langmead et al.; licensee BioMed Central Ltd. 2009

Received: 21 October 2008 | Accepted: 4 March 2009 | Published: 4 March 2009

# Bowtie: exact matching

- Figure 2 from the Bowtie paper: exactly what we have done in class, except exclusive of end-point

Read: 'ggta'

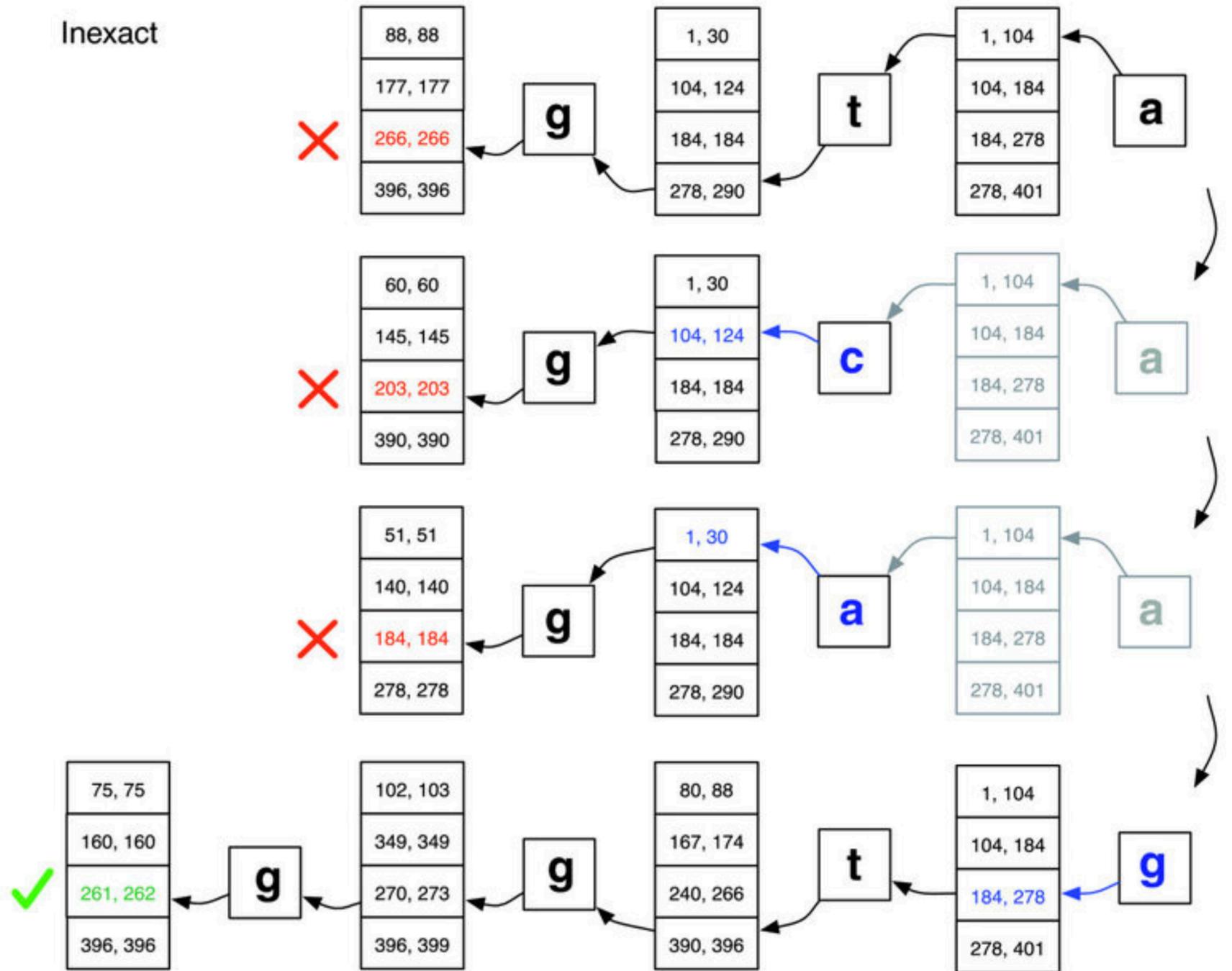


Equal (in this case), means no occurrences!  
(for us, start > end means no occurrences)

# Bowtie: inexact matching

Read: 'ggta'

Did not find the read, but we did find: 'ggtg'



# Comparison of Bowtie and BWA

**Table 2.**

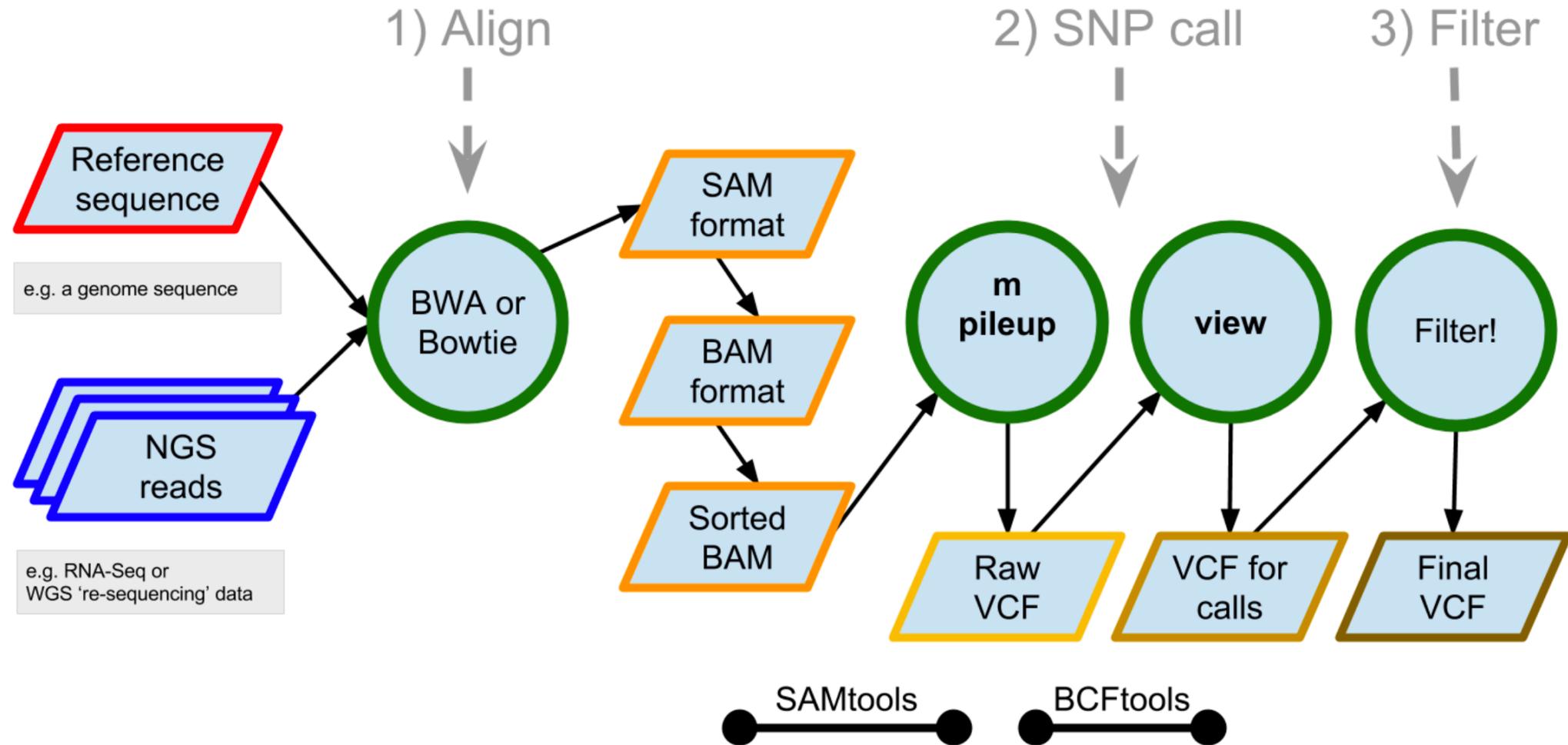
Evaluation on real data

Program	Time (h)	Conf (%)	Paired (%)
Bowtie	5.2	84.4	96.3
BWA	4.0	88.9	98.8
MAQ	94.9	86.1	98.7
SOAP2	3.4	88.3	97.5

The 12.2 million read pairs were mapped to the human genome. CPU time in hours on a single core of a 2.5 GHz Xeon E5420 processor (Time), percent confidently mapped reads (Conf) and percent confident mappings with the mates mapped in the correct orientation and within 300 bp (Paired), are shown in the table.

What comes after read-mapping?

# Pipeline overview







# SAM -> BAM -> sorted BAM -> VCF

- SAM: “Sequence Alignment/Map” file format
- Shows the locations of every read alignment, as well as the quality scores of each base of the alignment
- BAM: “Binary Alignment/Map” file format is a compressed version of SAM
- Next step: we need to sort the SAM or BAM file by the start position of each alignment
- Then we need to “call” the variation, finally giving us a VCF file
- VCF file: “Variant Call Format” shows the position of each SNP in order (usually we have one file for each chromosome)

# VCF file format

**VCF header**

```
##fileformat=VCFv4.0
##fileDate=20100707
##source=VCFtools
##reference=NCBI36
##INFO=<ID=AA,Number=1,Type=String,Description="Ancestral Allele">
##INFO=<ID=H2,Number=0,Type=Flag,Description="HapMap2 membership">
##FORMAT=<ID=GT,Number=1,Type=String,Description="Genotype">
##FORMAT=<ID=GQ,Number=1,Type=Integer,Description="Genotype Quality (phred score)">
##FORMAT=<ID=GL,Number=3,Type=Float,Description="Likelihoods for RR,RA,AA genotypes (R=ref,A=alt)">
##FORMAT=<ID=DP,Number=1,Type=Integer,Description="Read Depth">
##ALT=<ID=DEL,Description="Deletion">
##INFO=<ID=SVTYPE,Number=1,Type=String,Description="Type of structural variant">
##INFO=<ID=END,Number=1,Type=Integer,Description="End position of the variant">
```

**Mandatory header lines**

**Optional header lines** (meta-data about the annotations in the VCF body)

**Body**

#CHROM	POS	ID	REF	ALT	QUAL	FILTER	INFO	FORMAT	SAMPLE1	SAMPLE2
1	1	.	ACG	A,AT	.	PASS	.	GT:DP	1/2:13	0/0:29
1	2	rs1	C	T,CT	.	PASS	H2;AA=T	GT:GQ	0 1:100	2/2:70
1	5	.	A	G	.	PASS	.	GT:GQ	1 0:77	1/1:95
1	100	.	T	<DEL>	.	PASS	SVTYPE=DEL;END=300	GT:GQ:DP	1/1:12:3	0/0:20

**Deletion**

**SNP**

**Large SV**

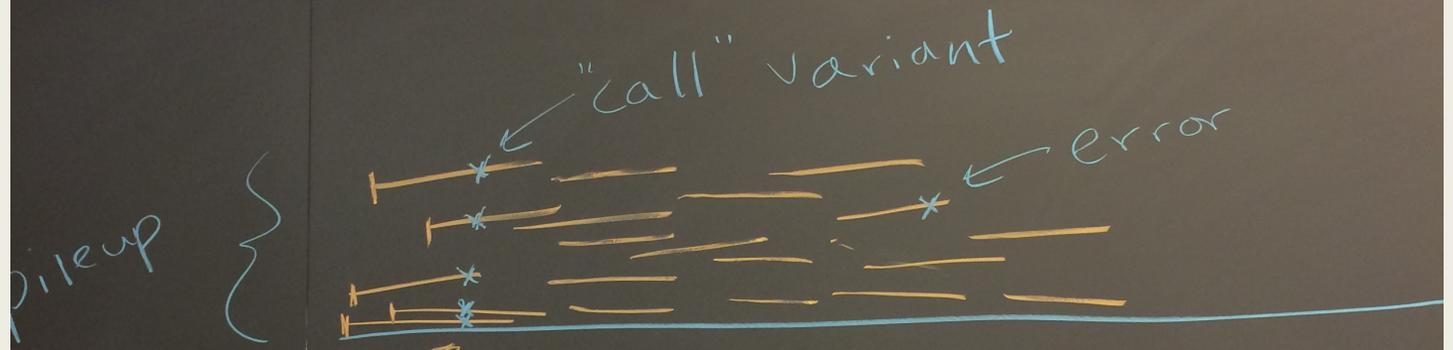
**Insertion**

**Other event**

**Reference alleles (GT=0)**

**Alternate alleles (GT>0 is an index to the ALT column)**

**Phased data** (G and C above are on the same chromosome)



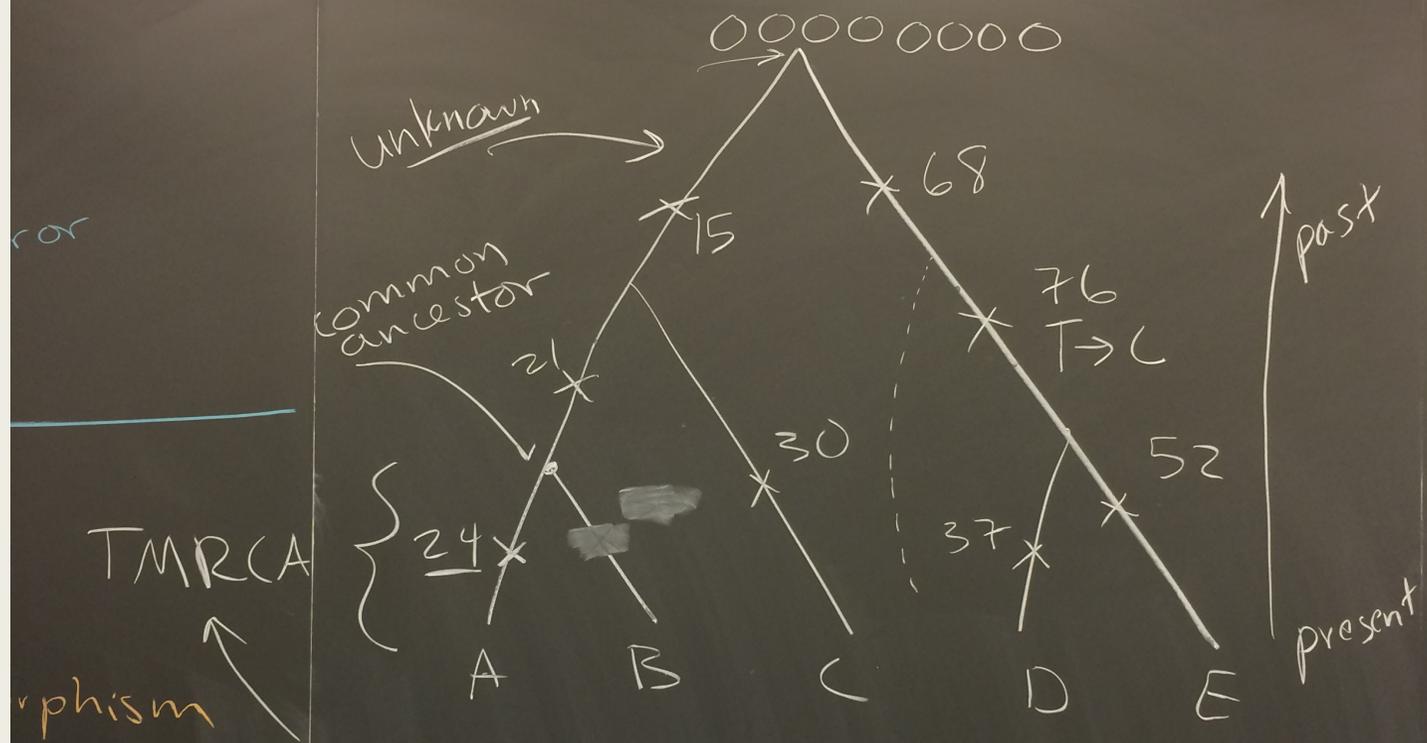
reference

SNP: single nucleotide polymorphism

TMRCA

reference

Begin: using variation to reconstruct  
evolutionary events



TMRCA

phism

time to most recent  
common ancestor

ast

	15	21	24	30	37	52	68	76
A	1	1	1	0	0	0	0	0
B	1	1	0	0	0	0	0	0
C	1	0	0	1	0	0	0	0
D	0	0	0	0	1	0	1	1
E	0	0	0	0	0	1	1	1

resent

$\frac{5}{5}$   
 $\frac{2}{5}$

↑  
 private mutation  
 (singleton)

matrix:  
 (V(F))  
 (transposed)

haplotype: a sequence from a single chromosome (humans are diploid)

SNP: single nucleotide polymorphism (caused by a single mutation)

Allele: a type of observed variation

Ancestral: allele of the ancestor of all seqs in sample ("G")

Derived: allele of the variant observed in sample ("T")

Ref: whoever we sequence first

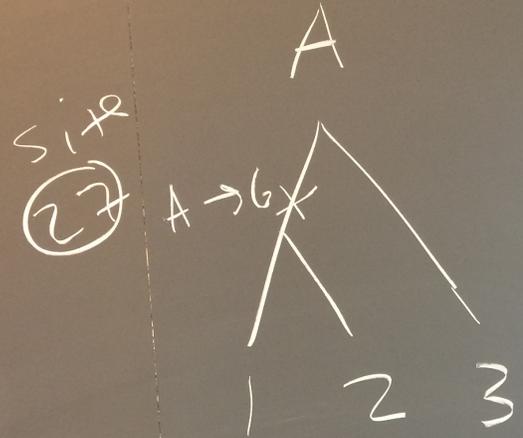
Alt: allele of observed non-ref individuals

most variation  
is biallelic

rare: tri-allelic

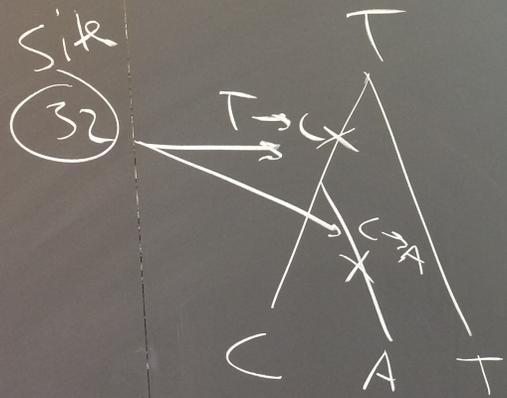
"T"

"G"



1	G → 1
2	G → 1
3	A → 0

↑  
bi-allelic



1	C
2	A
3	T

tri-allelic

8

	A	B	C	D	E
A	0	1	3		
B		0			
C			0		
D				0	
E					0

pileup

SNP

# of pairwise differences