Using git and mapping reads

Analysis of High Throughput Sequencing Data 2-7-2017

Agenda

- Git
 - Version control
 - Cloud storage
- Read mapping
 - Aligning short reads to a reference genome
 - Naive string searchign
 - Hashing
- Big O notation
 - You will hate this

Local storage and lazy version control are bad

- Keeping things on just your computer is dangerous!
 - What if your dog eats your computer?
- Losing track of versions is dangerous!
 - Sometimes fixing a bug completely breaks everything
 - Sometimes you want to make a big change

Cloud storage and version control

- Dropbox is pretty common
- The "cloud"
 - A bunch of computers that are redundantly backed up
- Version control
 - Systematically keeping track of updates, and keeping backups of all updates

git and github make this easy!

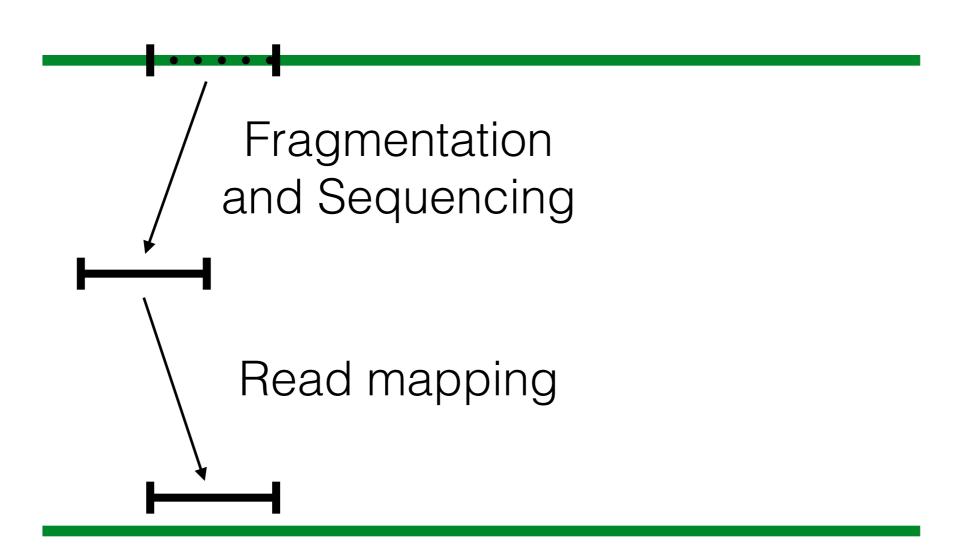
- git is a version control software
 - Repositories
 - Branches
 - Commits
- github is a cloud storage service that uses git

Learning git

- Tutorial in the problems
- Typical workflow
 - git pull
 - do stuff that modifies a file
 - git add file_that_you_changed
 - git commit -m 'what you changed in the file'
 - git push
- apt-get install on Linux, brew install on MacOS

Read mapping

Can't see it



Can see it

Casting as a string searching problem

- Genome is a string of A's, C's, G's and T's
- Read is a substring of the genome
 - Possibly noisy!
- Want to find part of the string where the substring matches

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How long does this take?

- We have a large genome
 - Human genome is 3 billion base pairs
- We have a lot of reads
 - Hundreds of thousands to tens of millions depending on the application
- Want to estimate how long something will take before we start
 - How it will scale as we change the input data

Big O notation

- How the SLOWEST part of the algorithm scales with INPUT size
- Common examples
 - O(n): linear
 - O(n log n): log-linear
 - O(n²): quadratic

Complexity of the naive substring searching algorithm

- String of length n
- Substring of length m
- Need to test if the substring matches for all n positions in the string
- For each position in the string need to check if all m positions match
- O(n*m)
- If we do this k times (i.e. we have k reads) then it's O(k*n*m)!

Tips for computing complexity

- Every nested loop that runs for n_i iterations adds a factor of n_i
- For i in range(n1):
 - For j in range(n2):
 - For k in range(n3):
 - •
- O(n1n2n3)

Hashing provides a far more efficient algorithm

- The slow part is searching the whole genome every single time
- Dictionaries (aka hash tables) provide a fast way to do string matching
 - Looking up a string in a hash table takes O(1) time
- If we has the genome, we can search way faster!

Two step algorithm to search with hashes

- read has length m
- genome has length n
- for i in range(len(n)):
 - add genome[i:(i+m)] as key to dictionary
- for every read:
 - check if the read is in the dictionary

Next time

- The way that this is done in practice
- Actually using bioinformatics software and not writing your own code!