# AtSNP Infrastructure

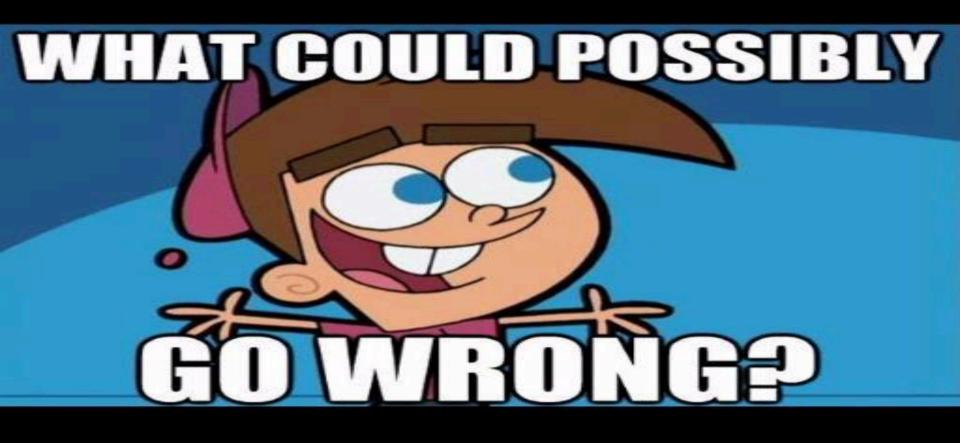
a case study for searching billions of records while providing significant cost savings over cloud providers

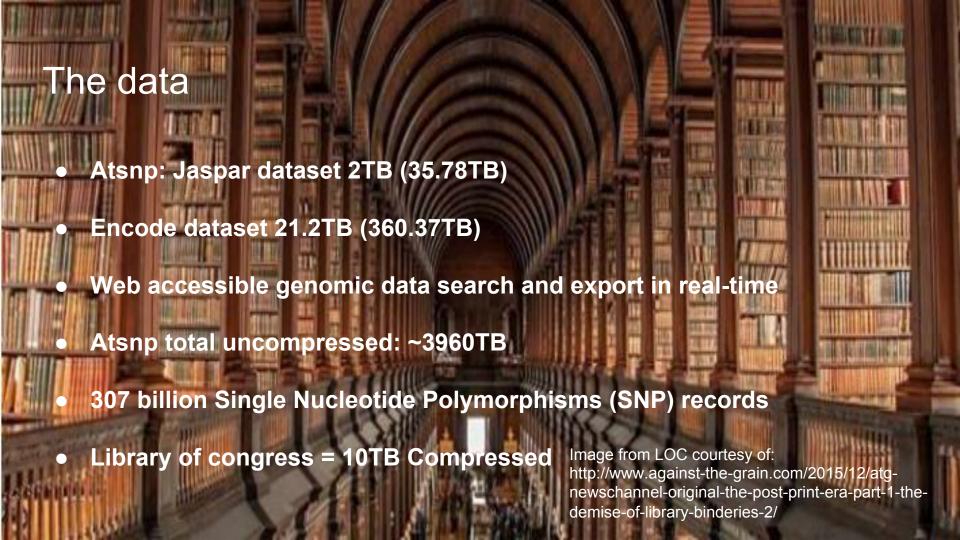
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# The atSNP story

- Hallway conversation
- Want to put 2TB of data on the web
- Have an another dataset to put online in the future
- Post-Doc will work with you
- Let me know what you need





### What is at SNP

- Software developed to evaluate SNP-Transcription factors-DNA interactions
- 115,500 CPU hours to compute SNP to Position Weight Matrix (Big Data)
  - Computed using HTCondor UW-CHTC and OSG
  - Wanted to make this compute power available to researchers without this amount of compute at hand
- Calculate p-values
- Determine SNP-PWM motif's
- Motif images for each of the 307 bill (+)
  - Originally a PNG for each SNP-PWM
  - Would have consumed 3.7Petabytes

TATTCTCCTCCTCAAA

Best Match with the SNP Allele

TATTCTCTTTTCTTCTTCAAA

Best Match with the SNP Allele

TATTCTCTTTTCTTTTCTTAAA

1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17

GCCACGCCCCCTCAAA

1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17

Best Match with the Reference Allele

SP4 2 Motif Scan for rs115414042

SNP

(+)

Ref

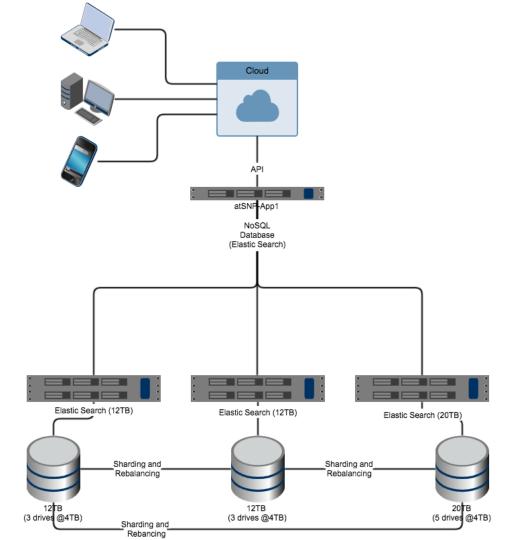
### Constraints

- Cost
- Supportability (personal time, monitoring, domain knowledge)
- Speed to implementation
- Data center rackspace
- Query result times

# Feasibility Candidates

- Objective: use a DB with a large usage and support base
- Cassandra
  - NoSQL known for quick access and search
- MySQL (or MariaDB)
  - Oldie and goodie
- Elasticsearch
  - Indexes log data
- Others
  - We needed quick turn around and widely supported platforms

Infrastructure for our initial feasibility testing



### Cassandra

#### Pro's

- Fast searches
- Fast imports (ETL) (14,664records/sec)
- Auto rebalancing on node failure

#### Con's

- No range query support\*
- No team domain expertise

# MySQL (MariaDB)

#### Pro's

- Team domain expertise
- Range query support

#### Con's

- Slow ETL (ETL 1023records/sec)
- Partitioning of data across systems manually
- Auto rebalancing on node failure

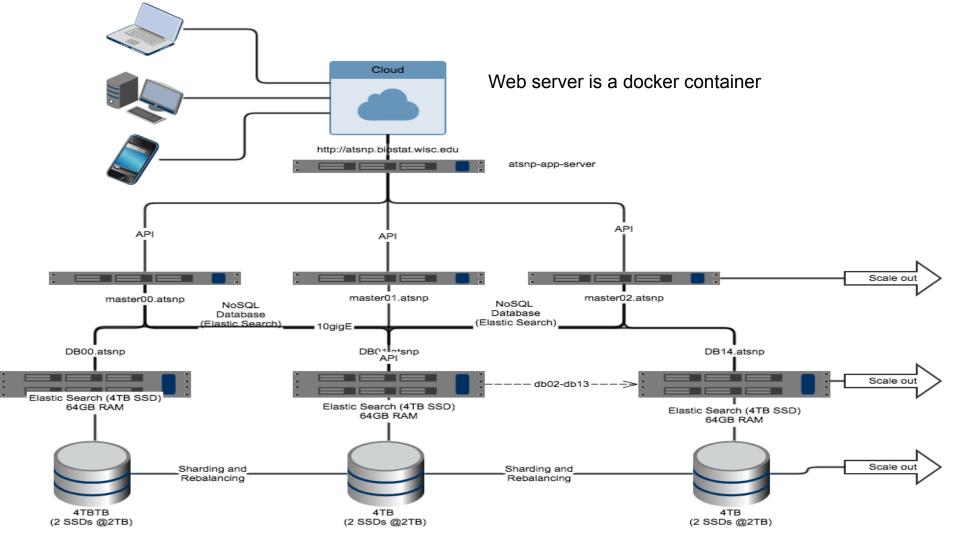
### Elasticsearch

#### Pro's

- Range queries
- Reasonable Load times (ETL- 11,944records/sec)
- Auto rebalancing on node failure

#### Con's

- No domain expertise
- Data loading took longer than Cassandra



### Results of final infrastructure

- Final results proved elasticsearch was a viable option for
  - loading
  - searching
  - and retrieving of data
- Scale-out infrastructure
  - Can add more nodes as data needs change/grow
  - Response time is critical for genomics data searches
  - Future improvements can be easily integrated
- Cost
  - Amazon, \$0.135/GB/Month
  - Our final cost \$0.039/GB/Month
  - 3.4x Cost Savings over Amazon

# **Key Contributions**

- Feasibility testing is important for application infrastructure deployments
- Cloud providers are not always the lowest cost provider
- NoSQL databases are great for scalability and work for genomic data stores
- atSNP website:
  - http://atsnp.biostat.wisc.edu
- System engineers are rockstars

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- My Family

