



# **DOE Joint Genome Institute**

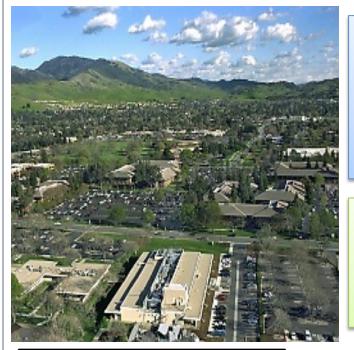
## **Computing Requirements**

## Shane Canon, Rob Egan, David Goodstein Victor M Markowitz

NERSC BER Requirements for 2017 September 11-12, 2012

## JGI DOE JOINT GENOME INSTITUTE DOE Joint Genome Institute (JGI)





Established in 1999, located in Walnut Creek

Supported by the DOE Office of Science.

Budget of ~70 Million/ yr ~ 300 Employees

#### **Mission**

User facility for large scale genomics to enable bioenergy & environmental research

#### **Mission Areas**

**Bioenergy** 

**Carbon Cycling** 







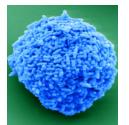
#### JGI Programs



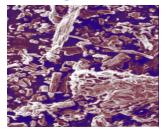




Microbes

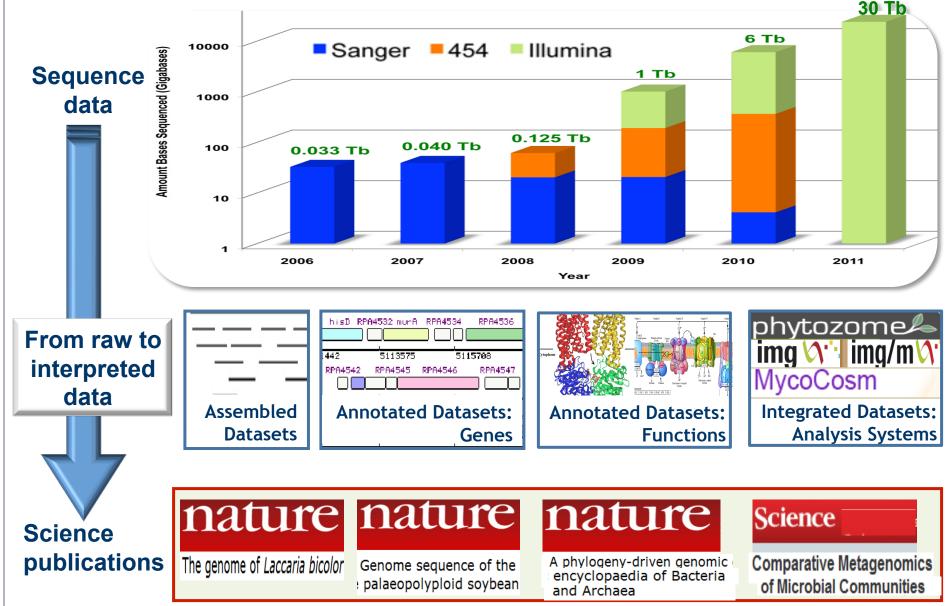


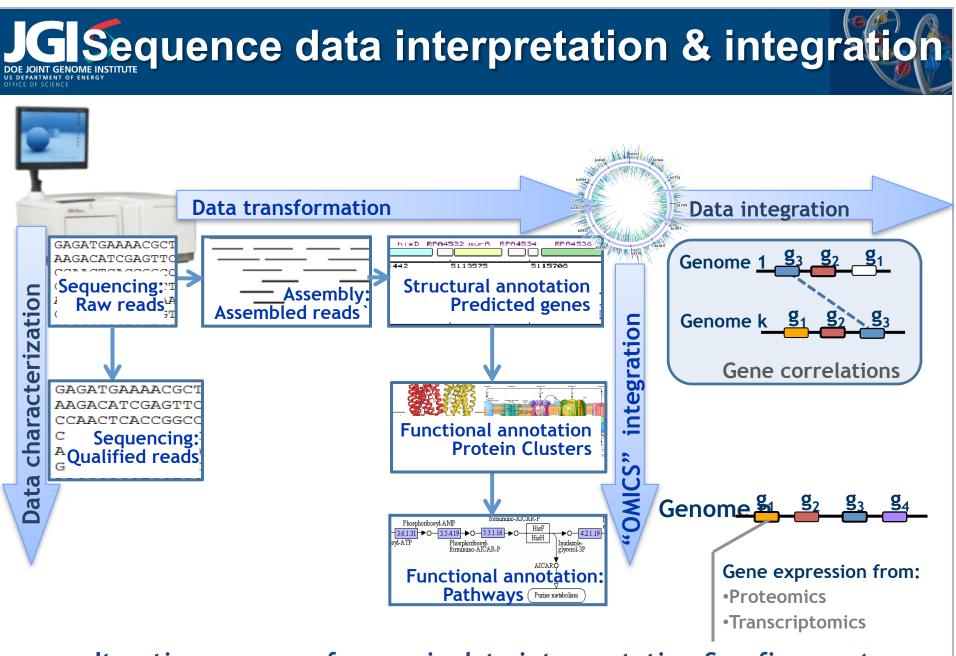
#### **Metagenomes**



#### **G S J G J G I is a production sequencing facility** E JOINT GENOME INSTITUTE **Energy and Environmental Science** User **Experimental** Data **Data Generation** Interpretation Interactions Assembly of **Data Distribution** Sequencing **Sequence Data**

#### JGI JOINT GENOME INSTITUTE DE JOINT GENOME INSTITUTE US DE JOINT GENOME INSTITUTE OFFICE OF SCIENCE

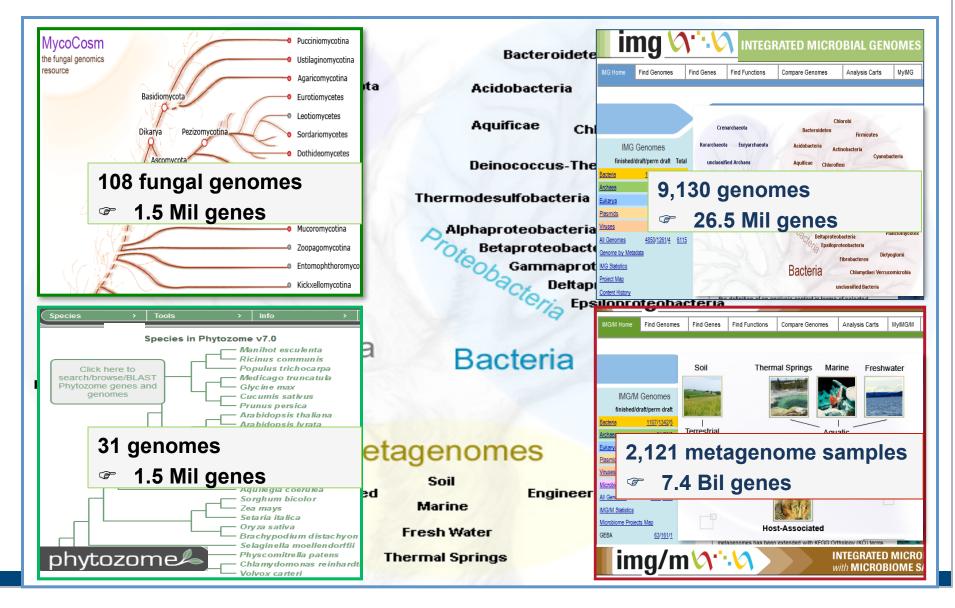


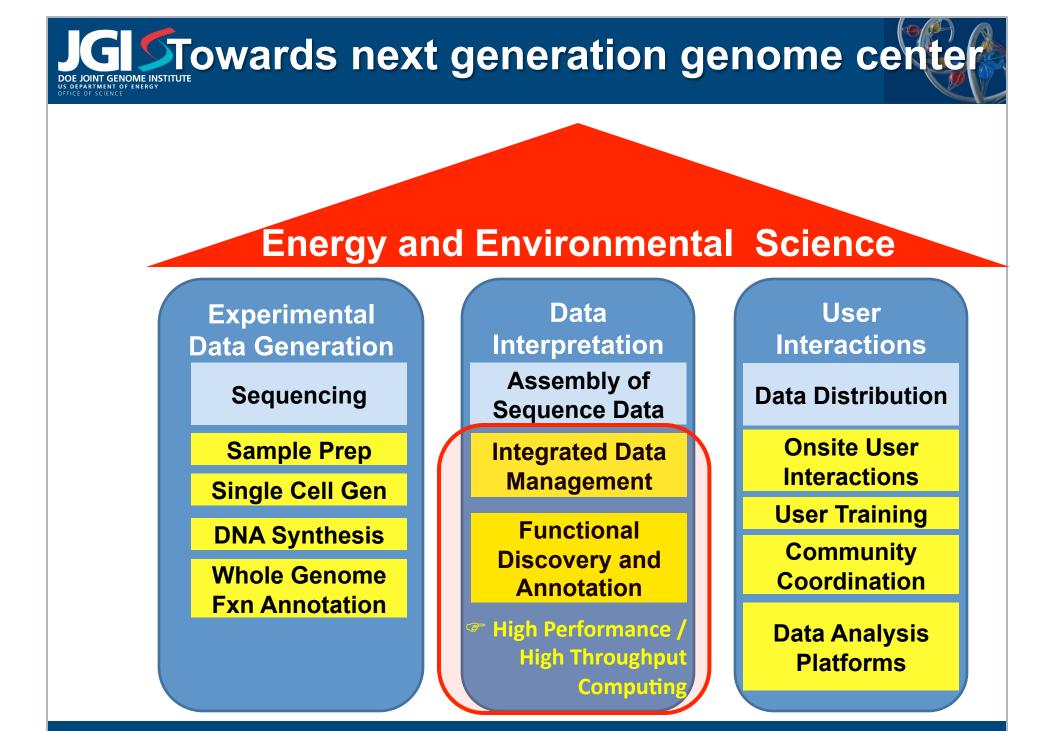


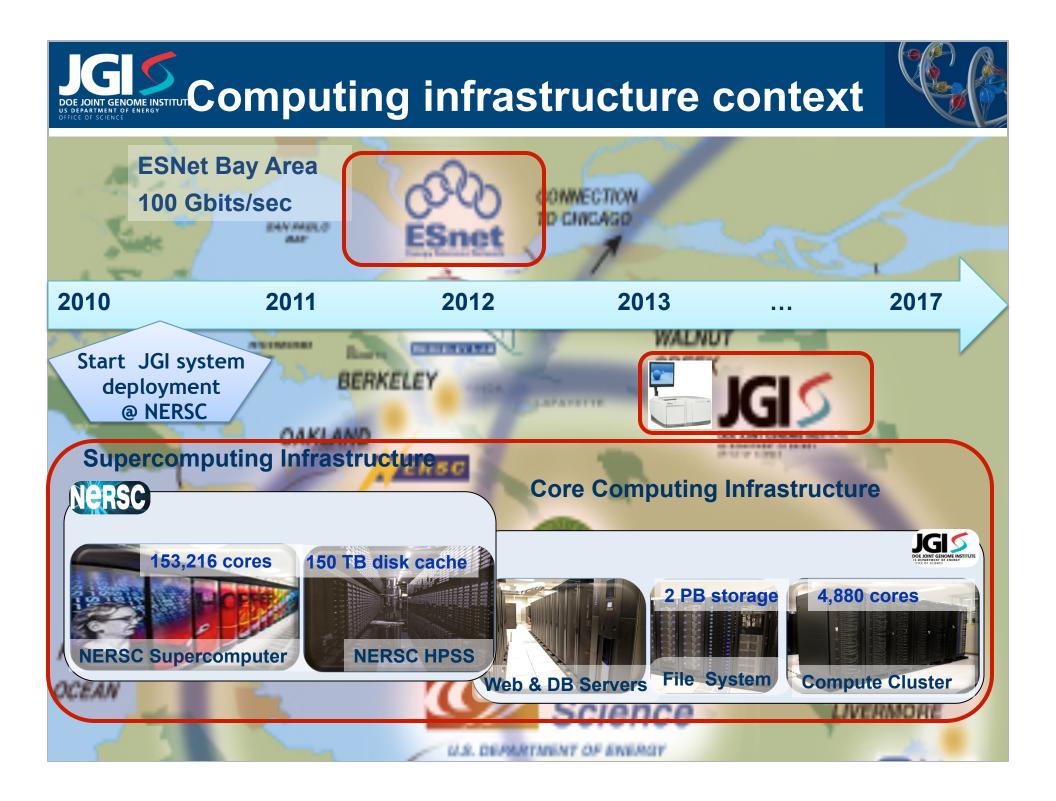
Iterative process of genomic data interpretation & refinement

## JG Context for conducting scientific studies

#### Data integrated framework for (meta) genome sequence data analysis









## Scaling challenge

### Computing needs

**Existing** sequence data processing methods, pipelines, systems **cannot keep up** with rapid increase in size & number of sequence datasets

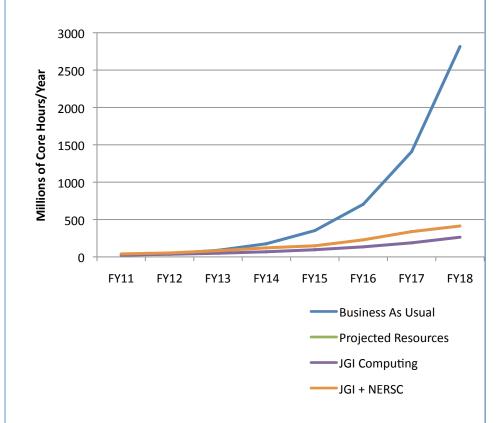
#### Alternatives

#### □ Increase supply

- Buy more
- Use other resources
  - Use the cloud

#### Reduce consumption

- Improve efficiency
- Reduce data processing





## Addressing scaling challenge



### Optimize

**Re-engineer key methods/tools/workflows** 

### Innovate

Design new analysis methods

### Prioritize

**Selective computations** 

## Limit

Sequence what you can analyze

## Adapt

Ask different questions

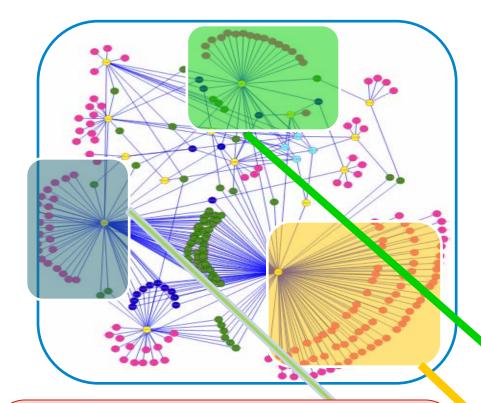
**Re-engineered workflows** 

addressed scaling challenge for microbial genome & metagenome data processing

Advanced data management (Fastbit) techniques applied to gene cassette conservation (operon) analysis results in substantial performance boost

## Metagenome analysis challenge





#### **Protein clustering**

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Allow comparing metagenome samples in terms of known & potentially new protein families

- requires 206 M CPU core hours annually for maintaining clusters using linear gene pledging
- alternative strategies that can provide 10X acceleration are being explored

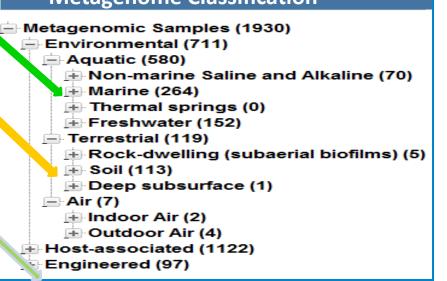
#### IMG/M: 7 Billions Genes

Metagenome analysis limited to known protein families from sequenced organisms

☞ ignores about 70% of genes of

potentially new organisms that are not similar to these protein families

img/monities



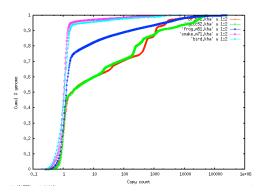
# JGI Case study: Sorghum bicolor var. Rio

(~700Mb plant genome, with Sanger reference)

Meraculous (Chapman): scales up to 1000+ cores (full cluster); peak (bottleneck) memory ~10-20Gb. 2-3 days elapsed time. 486Mbp scaff. total; 74.4kb scaffN50; 7.9kb contigN50

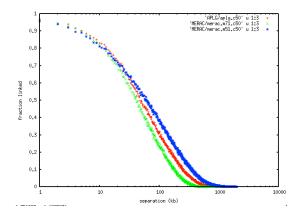
APLG (Labutti): 15 days elapsed time; peak memory 365Gb; not "clusterable". 438Mbp scaff. total; 69kb scaffN50; 19.4kb contigN50

SGA (Sunkara): 11 days elapsed; 30Gb peak memory; (in principle) parts can be run on cluster. 425Mbp scaff total; 7.9kbp scaffN50; 6.4kbp contigN50



Plants are harder than (typical) vertebrates to assemble due to repeat content and polymorphism

Correctness matters! (N50 is often our only metric, but it is a pretty poor one)



"C50" (only measurable with some known reference sequence) APLG 48kbp; meraculous 62kbp.

2-4X more global misjoins in APLG result



Machines currently using: Carver and Hopper

#### Hours used in 2012: ~12M core hours

- □ Primarily on Hopper using the taskfarmer
- □ Key applications included BLAST, hmmsearch, and usearch
- □ Scales 1k-12k (some larger runs too)
- Native MPI applications included the JGI developed K-mernator and the ANL developed Kiki
- Reading ~16 GB and ~TB of output
- Try to use around 16 GB per node (partition references to fit in memory)
- Heavy Use of TaskFarmer
- Typically stick to scratch for performance

# JGIS HPC genome assembly applications

Software	HPC Scope	State	Speed on 1TBase, Reduced Metagenome
ABySS	Contigs Only	Mature & Active	3h / 1800 cores (Reduced 787Gbase)
Ray	Scaffolds	Mature & Active	Unsuccessful
Kiki	Contigs Only	In Development	Unsuccessful
Pasha	"Pre-Graph" Only	Mature & Active	Not Attempted
Forge	Scaffolds	Not Mature, Inactive	Not Able to Compile
Contrail (Hadoop)	Scaffolds	In Development	Unsuccessful
Kmernator	Data Reduction & Contig Extension	Mature & Active	3h / 7200 cores (Full 1.115 TB raw data)

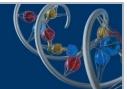
# JGI Expanded use of HPC for JGI projects

- Solution State State
- HPC allocation will be used to address other needs
  - Problems beyond Genepool's scale
  - Development and testing new methods
  - □ Accelerating Time Sensitive Operations
  - □ Handling Demand during Peak Times
  - MPI-enabled applications benefiting from interconnect
  - □ Non-throughput workloads (parallel assemblers)

# JGI Future HPC requirements for JGI projects

- Core hours should keep pace with Sequence Growth -> ~75M core hours in 2017
- Changes in sequence technology will likely change workload characteristics
  - □NGS created an assembly challenge
  - □Single molecule have longer reads with more errors
- Increased support for data intensive workloads





Software	Targeted Application	Speedup*	Comments
GPU HMMER (2.0)	Homology Search	20x	HMMER 3.0 is superior
GPU-BLAST	Homology Search	3-5x	Protein only
CUDASW++	Sequence Alignment	10x	
MUMerGPU	Sequence Alignment	3-10x	
SOAP3	Sequence Alignment	10x	
CuSHAW	Sequence Alignment	8-10x	
GSNP	Variant Detection	7x	
DecGPU	Read Correction	N/A	MPI based

\* Speedup is not the appropriate metric; most calculations are against the (sometimes poorly optimized) single threaded code or a single core. Power consumption and capital costs should be rigorously evaluated against a large data set.





- Support for non-relational database services (NoSQL, Key-Value stores)
- Continue and enhanced support for Hadoop
- Enhanced support for High-Throughput Workloads (enable more workloads to move from Genepool to the big systems)



## Summary



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## **Thank You**