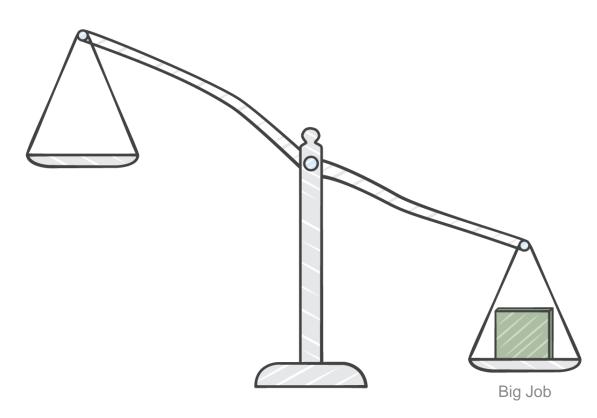
High Performance Computing on AWS

Wen DAI (代闻) Solutions Architect Amazon Web Services wendai@amazon.com

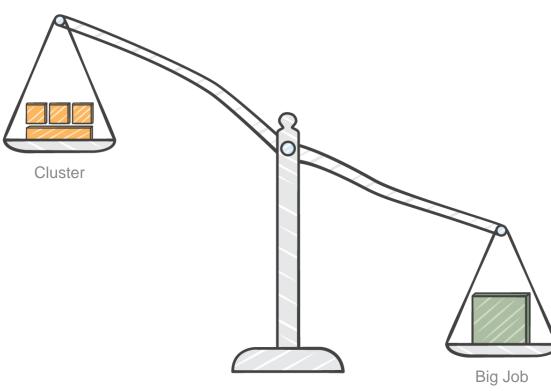


Take a typical big computation task...



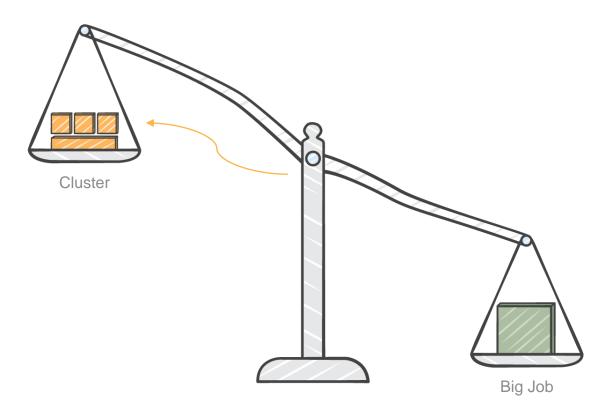


...that an average cluster is too small (or simply takes too long to complete)...



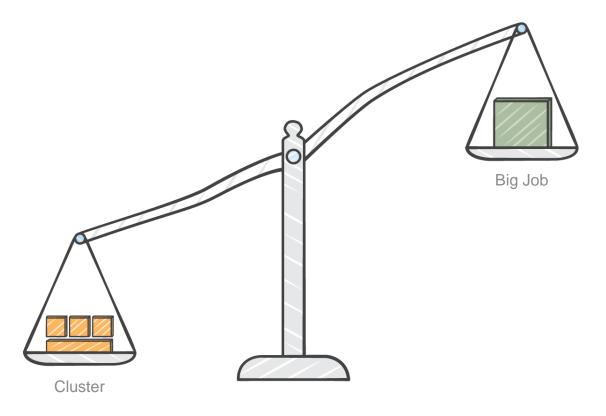


...optimization of algorithms can give some leverage...



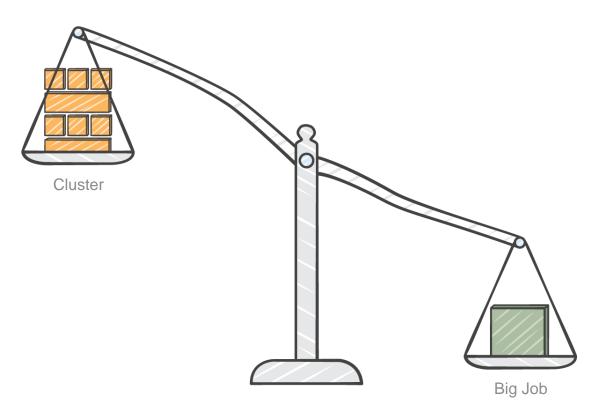


...and complete the task in hand...



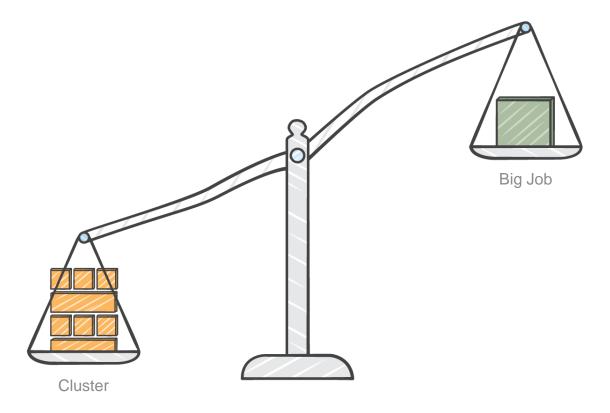


Applying a large cluster...



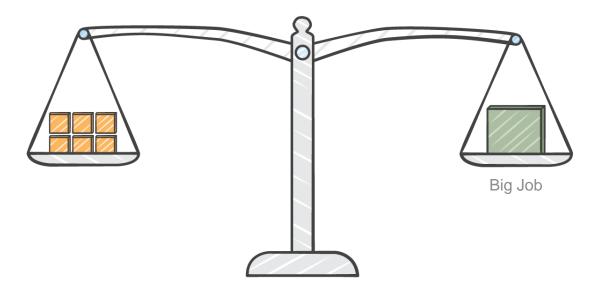


...can sometimes be overkill and too expensive



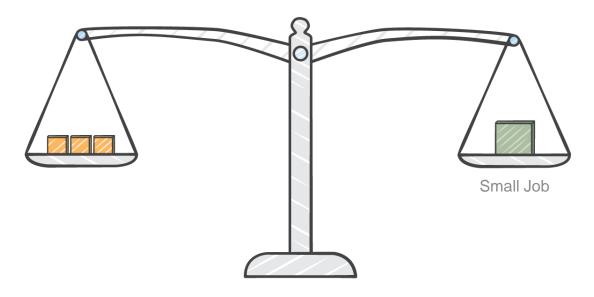


AWS instance clusters can be balanced to the job in hand...



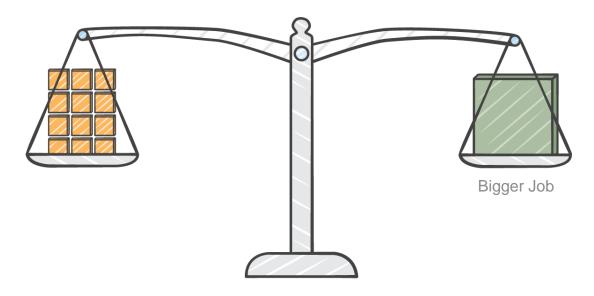


...nor too large...



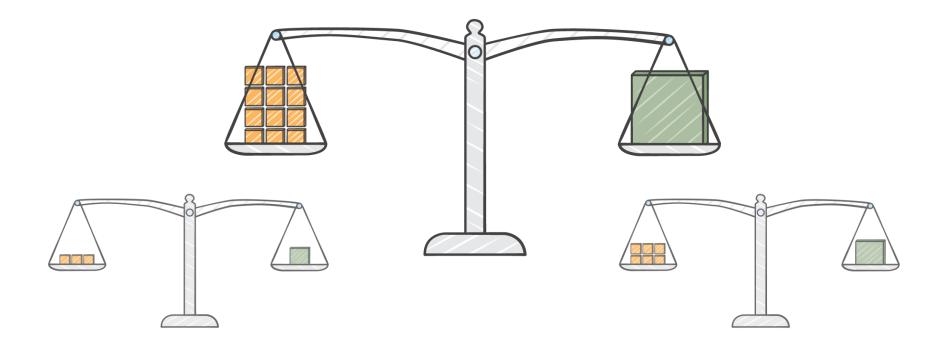


...nor too small...

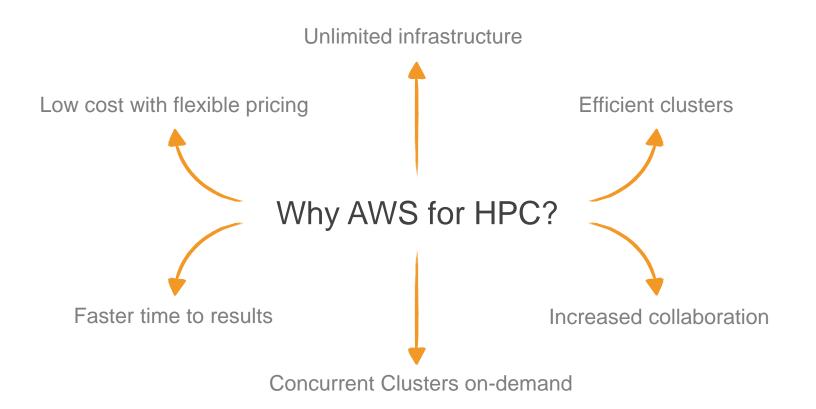




...with multiple clusters running at the same time









Customers running HPC workloads on AWS



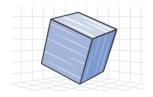


Popular HPC workloads on AWS

Genome processing



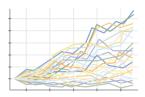
Modeling and Simulation



Government and Educational Research



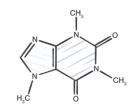
Monte Carlo Simulations



Transcoding and Encoding



Computational Chemistry





Across several key industry verticals

Utilities

Biopharma

Schneider Electric



Materials Design



Manufacturing

Autodesk

Academic research

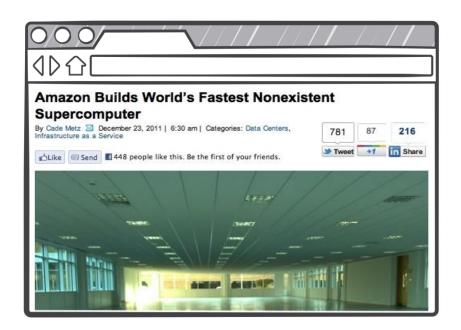


Auto & Aerospace





TOP500: 76th fastest supercomputer on-demand



Jun 2014 Top 500 list

484.2 TFlop/s

26,496 cores in a cluster of EC2 C3 instances

Intel Xeon E5-2680v2 10C 2.800GHzprocessors

LinPack Benchmark



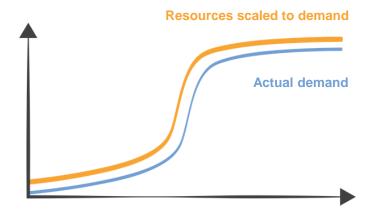


Benefits of Agility



Rigid On-Premises Resources





Elastic Cloud-Based Resources





Unilever: augmenting existing HPC capacity

The key advantage that AWS has over running this workflow on Unilever's existing cluster is the ability to scale up to a much larger number of parallel compute nodes on demand.

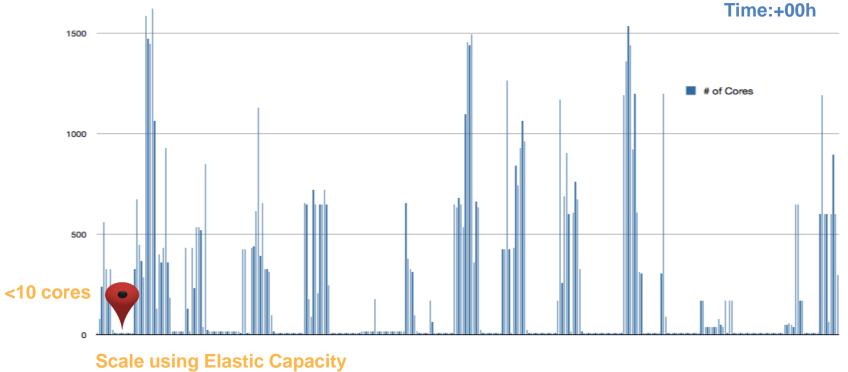
> Pete Keeley Unilever Researchs eScience IT Lead for Cloud Solutions



Unilever's digital data program now processes
genetic sequences twenty times faster

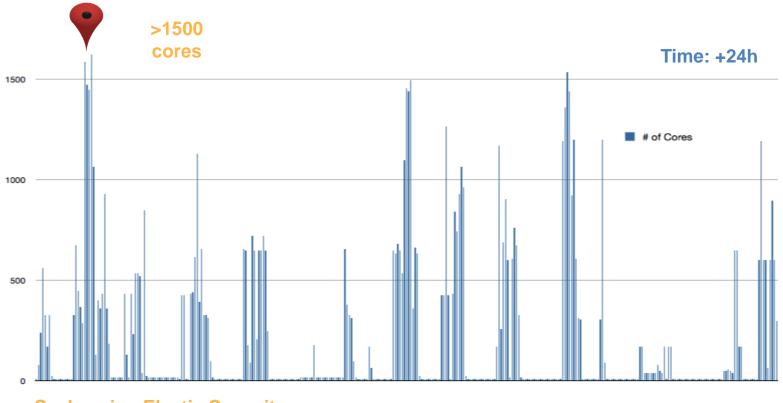


Scalability on AWS





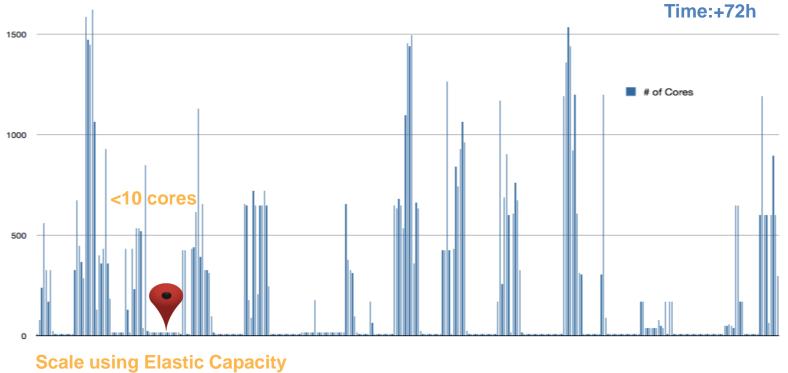
Scalability on AWS





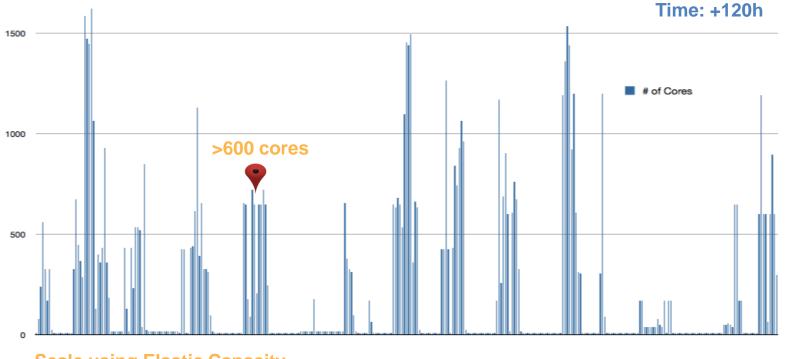
Scale using Elastic Capacity

Scalability on AWS



web services

Scalability on AWS



Scale using Elastic Capacity



Schrodinger & CycleComputing: computational chemistry

Simulation by Mark Thompson of the University of Southern California to see which of 205,000 organic compounds could be used for photovoltaic cells for solar panel material.

Estimated computation time 264 years completed in 18 hours.

SCHRÖDINGER.

CYCLECOMPUTING

- 156,314 core cluster across 8 regions
- 1.21 petaFLOPS (Rpeak)
- \$33,000 or 16¢ per molecule



Cost Benefits of HPC in the Cloud

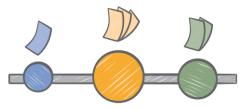
On-Premises



Capital Expense Model

High upfront capital cost High cost of ongoing support





Pay As You Go Model

Use only what you need Multiple pricing models



Many pricing models to support different workloads

Free Tier

Get Started on AWS with free usage & no commitment

For POCs and getting started



On-Demand

Pay for compute capacity by the hour with no long-term commitments

For spiky workloads, or to define needs

Reserved

Make a low, one-time payment and receive a significant discount on the hourly charge

For committed utilization

Spot

Bid for unused capacity, charged at a Spot Price which fluctuates based on supply and demand

For time-insensitive or transient workloads



Dedicated

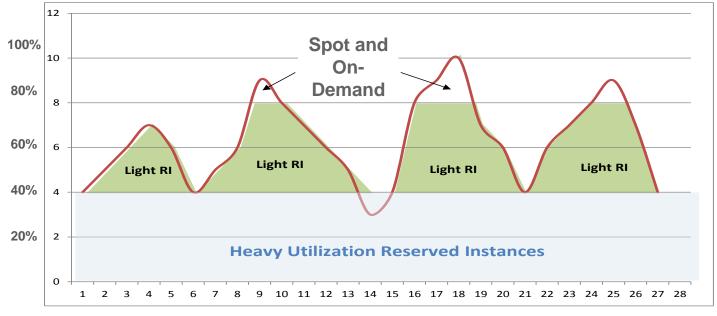
Launch instances within Amazon VPC that run on hardware dedicated to a single customer

For highly sensitive or compliance related workloads





Optimize Cost by using various EC2 instance pricing models



Utilization Over Time



Harvard Medical School: simulation development

The combination of our approach to biomedical computing and AWS allowed us to focus our time and energy on simulation development, rather than technology, to get results quickly. Without the benefits of AWS, we certainly would not be as far along

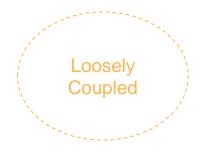
as we are.

Dr. Peter Tonellato, LPM, Center for Biomedical Informatics, Harvard Medical School



- Leveraged EC2 spot instances in workflows
- 1 day worth of effort resulted in 50% in cost savings





Embarrassingly parallel Elastic Batch workloads







Embarrassingly parallel Elastic Batch workloads

Data management Task distribution Workflow management





Embarrassingly parallel Elastic Batch workloads

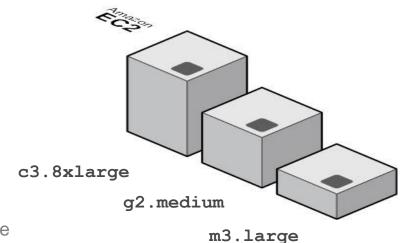
Data management Task distribution Workflow management



Compute Services

Elastic Compute Cloud (EC2)

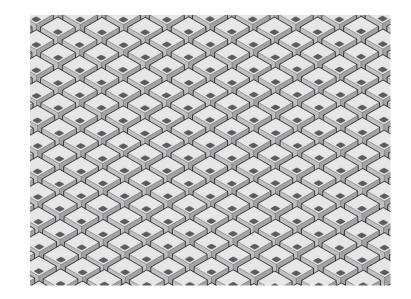
Basic unit of compute capacity, virtual machines Range of CPU, memory & local disk options Choice of instance types, from micro to cluster compute





Automation & Control

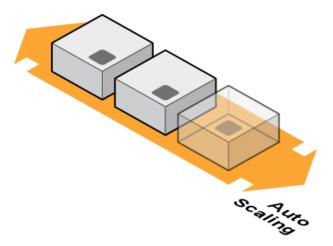
CLI, API and Console Scripted configurations





Auto Scaling

Automatic re-sizing of compute clusters based upon demand and policies







Embarrassingly parallel Elastic Batch workloads

Data management Task distribution Workflow management



What if you need to:

Implement MPI? Code for GPUs?



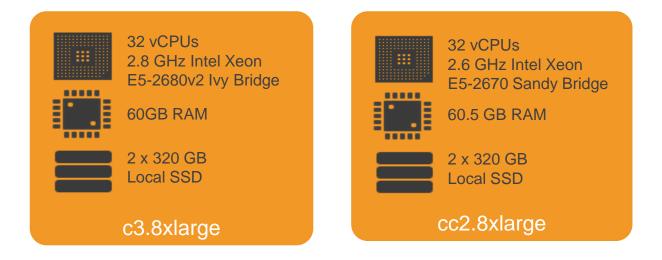
Tightly coupled

Cluster compute instances

Implement HVM process execution

Intel® Xeon® processors

10 Gigabit Ethernet –c3 has Enhanced networking, SR-IOV







Tightly coupled

Network placement groups

Cluster instances deployed in a Placement Group enjoy low latency, full bisection 10 Gbps bandwidth



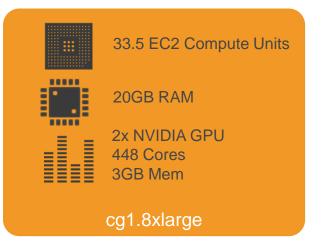
10Gbps

Tightly coupled

GPU compute instances

CG1 instances

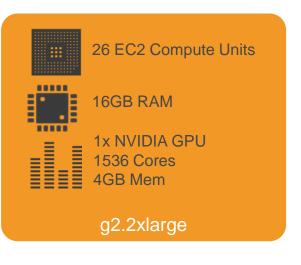
Intel® Xeon® X5570 processors 2 x NVIDIA Tesla "Fermi" M2050 GPUs I/O Performance: Very High (10 Gigabit Ethernet)





G2 instances

Intel® Intel Xeon E5-2670 1 NVIDIA Kepler GK104 GPU I/O Performance: Very High (10 Gigabit Ethernet)





National Taiwan University: shortest vector problem

Our purpose is to break the record of solving the shortest vector problem (SVP) in Euclidean lattices...the vectors we found are considered the hardest SVP anyone has solved so far.

Prof. Chen-Mou Cheng Principle Investigator of Fast Crypto Lab



• \$2,300 for using 100x Tesla M2050 for ten hours



| Desibles | Dimension | Euclidean norm | Seed | Contestant | Solution | Algorithm | Subm. Date |
|----------|-----------|-------------------|------|--------------------------------|----------|-----------|---------------|
| Position | Dimension | | | | | | |
| 1 | 120 | 2851 | 0 | Po-Chun Kuo, Michael Schneider | vec | ENUM, BKZ | 2011-04-6 |
| 2 | 116 | 2825 | 0 | Po-Chun Kuo, Michael Schneider | vec | ENUM, BKZ | 2011-04-1 |
| 3 | 114 | 2778 | 0 | Po-Chun Kuo, Michael Schneider | vec | ENUM, BKZ | 2011-03- |



Tightly coupled

CUDA & OpenCL

Massive parallel clusters running in GPUs NVIDIA Tesla cards in specialized instance types



Characterizing HPC



Embarrassingly parallel Elastic Batch workloads

Data management Task distribution Workflow management Interconnected jobs Network sensitivity Job specific algorithms



Supporting Services

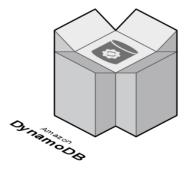
Data management

Fully-managed SQL, NoSQL, and object storage

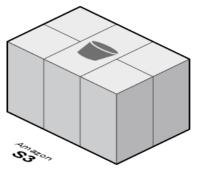


Relational Database Service

Fully-managed database (MySQL, Oracle, MSSQL, PostgreSQL)



DynamoDB NoSQL, Schemaless, Provisioned throughput database



Object datastore up to 5TB per object Internet accessibility

S3



Moving compute closer to the data

"Big Data" changes dynamic of computation and data sharing





TRADERWORX: Market Information Data Analytics System

For the growing team of quant types now employed at the SEC, MIDAS is becoming the world's greatest data sandbox. And the staff is planning to use it to make the SEC a leader in its use of market data

> Elisse B. Walter, Chairman of the SEC Tradeworx



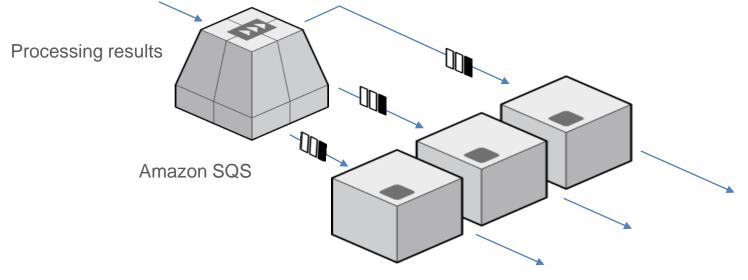
- Powerful AWS-based system for market analytics
- 2M transaction messages/sec; 20B records and 1TB/day



Supporting Services

Feeding workloads

Using highly available Simple Queue Service to feed EC2 nodes



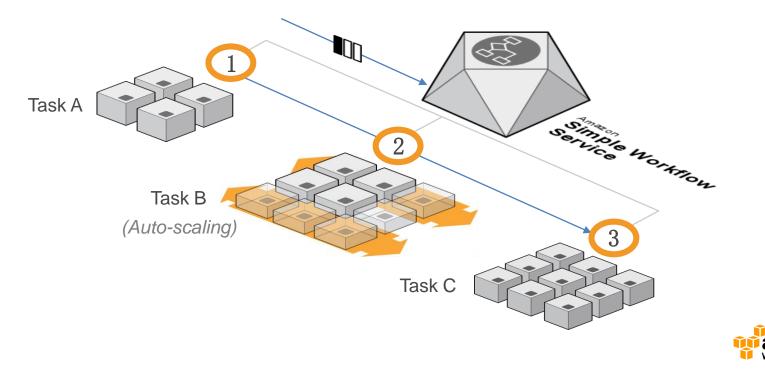
Processing task/processing trigger



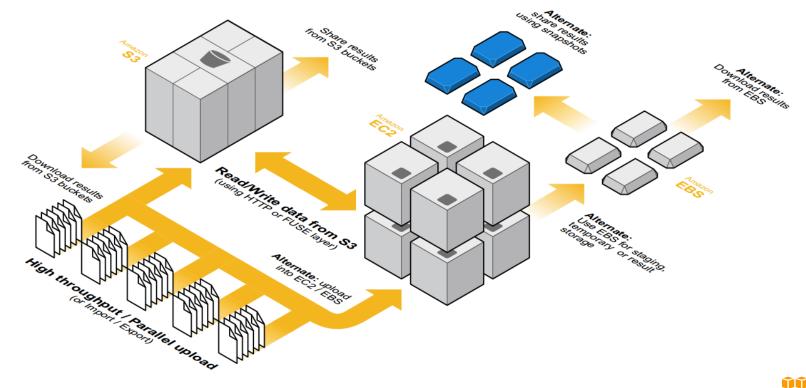
Supporting Services

Coordinating workloads & task clusters

Handle long running processes across many nodes and task steps with Simple Workflow



Architecture of large scale computing and huge data sets





NYU School of Medicine: Transferring large data sets

Transferring data is a large bottleneck; our datasets are extremely large, and it often takes more time to move the data than to generate it. Since our collaborators are all over the world, if we can't move it they can't use it.

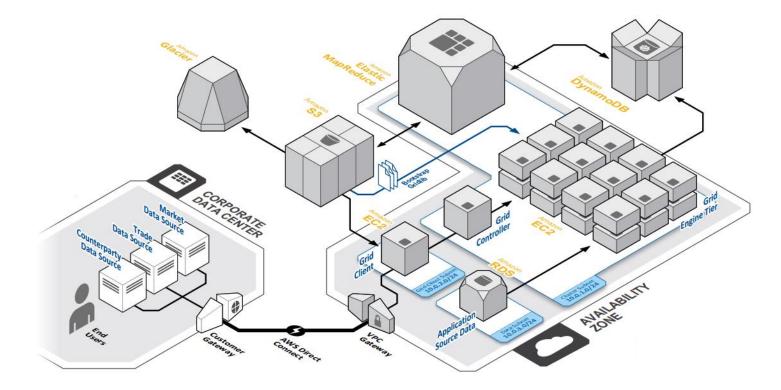
> **Dr. Stratos Efstathiadis** Technical Director of the HPC facility, NYU



- Uses Globus Online
- Data transfer speeds of up to 50MB/s



Architecture of a financial services grid computing





Bankinter: credit-risk simulation

With AWS, we now have the power to decide how fast we want to obtain simulation results. More important, we have the ability to run simulations that were not possible before due to the large amount of infrastructure required.

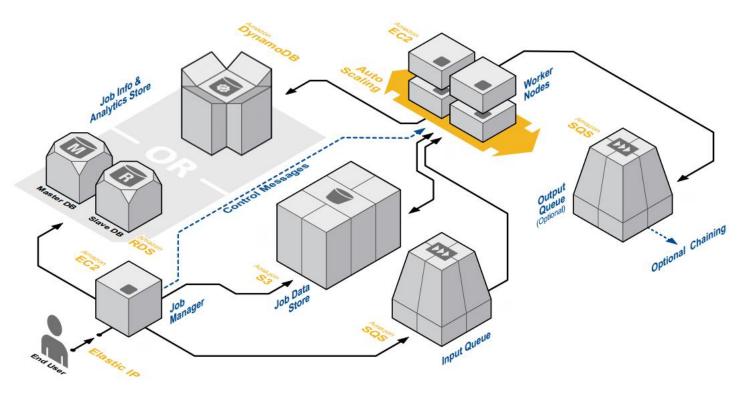
> Javier Roldán Director of Technological Innovation, Bankinter

bankinter.

• Reduced processing time of 5,000,000 simulations from 23 hours to 20 minutes



Architecture of queue-based batch processing





When to consider running HPC workloads on AWS

Improvement



Remove the queue Hardware refresh cycle Reduce costs Collaboration of results Increase innovation speed Reduce time to results

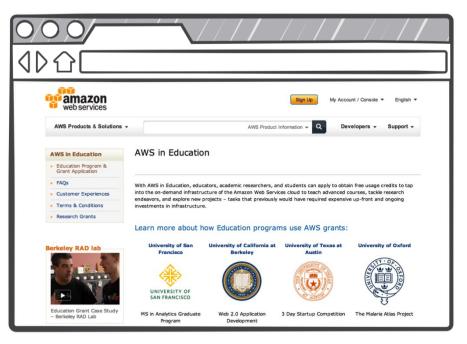
New ideas



New HPC project Proof of concept New application features Training Benchmarking algorithms



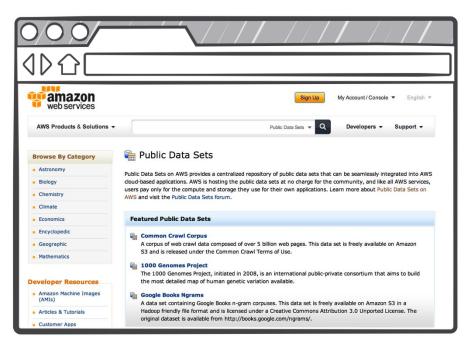
AWS Grants Program



aws.amazon.com/grants



AWS Public Data Sets

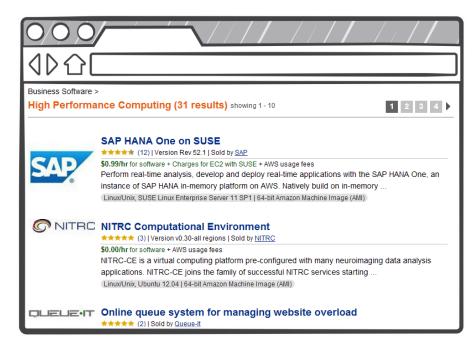


aws.amazon.com/datasets

free for everyone



AWS Marketplace – HPC category

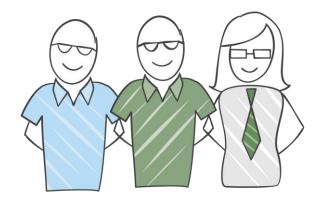


aws.amazon.com/marketplace





Getting Started with HPC on AWS



- Sales and Solutions Architects
 - Enterprise Support
 - **Trusted Advisor**
 - **Professional Services**

aws.amazon.com/hpc

contact us, we are here to help



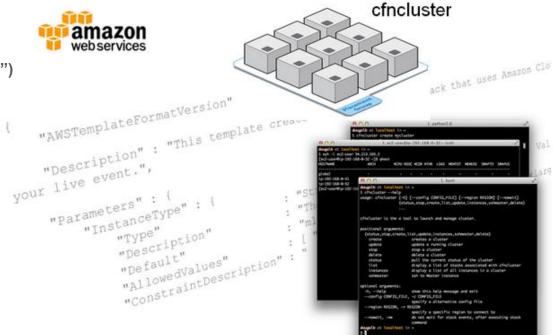
Try out our HPC CloudFormation-based demo

cfncluster ("CloudFormation cluster")

Command Line Interface Tool

Deploy and demo an HPC cluster

For more info: aws.amazon.com/hpc/resources



HPC Partners and Apps









GreenButton™





















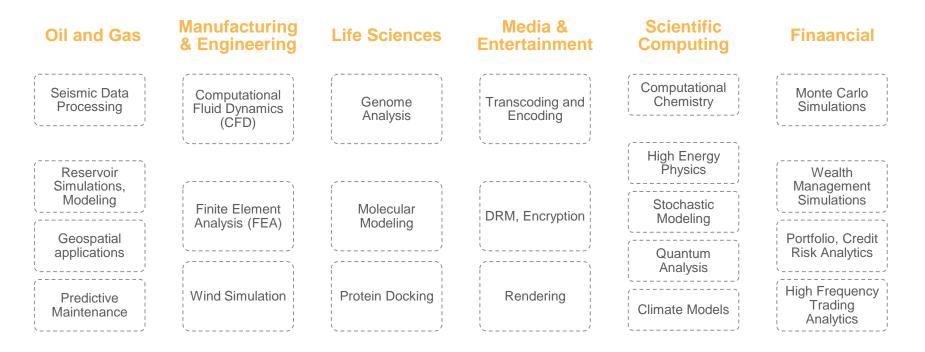




LEADER IN CONDOR GRID COMPUTING SOLUTIONS



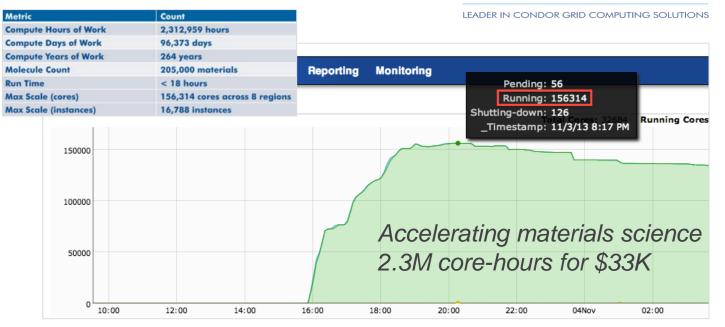
Costomers are using AWS for more and more HPC workloads





So What Does Scale Mean on AWS?







Cyclopic energy: computational fluid dynamics

AWS makes it possible for us to deliver state-of-the-art technologies to clients within timeframes that allow us to be dynamic, without having to make large investments in physical hardware.

> **Rick Morgans** Technical Director (CTO), cyclopic energy



Two months worth of simulations finished in two days



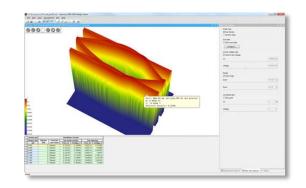
Mentor Graphics: virtual lab for design and simulation

Thanks to AWS, the Mentor Graphics customer experience is now fast, fluid, and simple.

> Ron Fuller Senior Director of Engineering, Mentor



• Developed a virtual lab for ASIC design and simulation for product evaluation and training





AeroDynamic Solutions: turbine engine simulation

We're delighted to be working closely with the U.S. Air Force and AWS to make time accurate simulation a reality for designers large and small.

> George Fan CEO, AeroDynamic Solutions



• Time accurate simulation was turned around in 72 hours with infrastructure costs well below \$1,000



HGST: molecular dynamics simulation

HGST is using AWS for a higher performance, lower cost, faster deployed solution vs buying a huge on-site cluster.

> Steve Philpott CIO, HGST



• Uses HPC on AWS for CAD, CFD, and CDA





Pfizer: large-scale data analytics and modeling

AWS enables Pfizer's Worldwide Research and Development to explore specific difficult or deep scientific questions in a timely, scalable manner and helps Pfizer make better decisions more quickly.

> Dr. Michael Miller Head of HPC for R&D, Pfizer



• Pfizer avoiding having to procure new HPC hardware by being able to use AWS for peak work loads.



Thank you!

Please visit aws.amazon.com/hpc for more info

