

# Virtualization

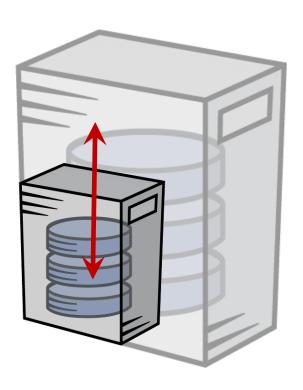


"It was much nicer before people started storing all their data in the Cloud."

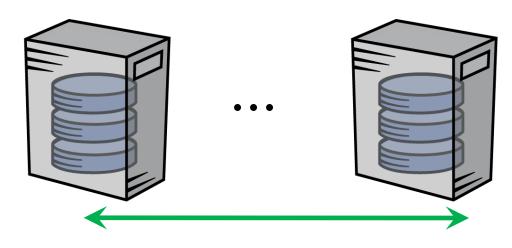


# **Hardware Scalability**

Vertical scaling: expand machine



Horizontal scaling: more (smaller) machines





# **Vertical Scaling: Supercomputer**







## **Horizontal Scaling: Cluster**

- Goal: more compute power, fault tolerance cheap
  - Commodity hardware
- Approach: horizontal scalability
- cluster = (loosely or tightly) connected computers working together,
   appearing as single system
  - each node same task
  - clustering middleware = software controlling & scheduling
- Related
  - Amdahl's Law: predict theoretical speedup when using multiple processors
  - more recently: Playstation clusters, Xbox clusters



# **Horizontal Scaling: Beowulf Cluster**



[Hoffman & Hargrove, ORNL]



# Horizontal Scaling: Supercomputers Today

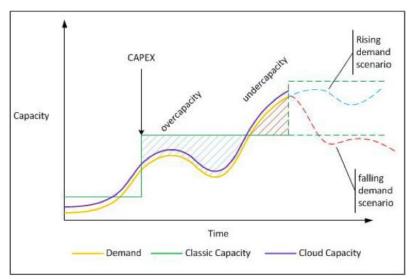
TaihuLight: 10,649,600 cores in 40,960 nodes; 1,3 TB RAM; 93 PFlop/s





### **Virtualization**

- Problem: just-in-time resource provisioning
- Approach:
  - Outsourcing to service provider
  - Virtual Machine (VM) to share computer resources on demand
- Many commercial providers
  - including Amazon AWS, Microsoft Azure, T-Systems, Hetzner, ...
- laaS, PaaS, SaaS, ...



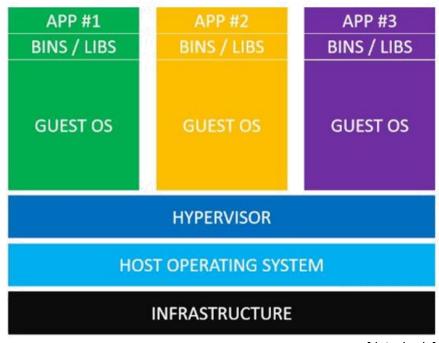
Capacity vs Utilization curves 8

[rackspace.com]



## **Virtual Machines**

- Virtual Machine (VM) = computer application
  - resembling a complete "computer"
    - Host system running 1..\* guest systems
- Technically:
  - application invokes guest OS services
  - Guest OS calls intercepted, forwarded to host OS
  - Host OS fulfills request
- Hypervisor = virtual machine monitor
  - resource orchestration: VM start, operation, stop, ...
  - Running on host
- Data can be local or mounted from remote (ex: SAN)

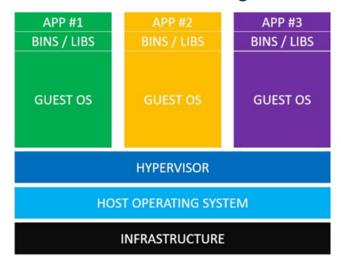


[datavizz.in]



### **Virtual Machines vs Containers**

- Problem: Large VM overhead of Virtual Machine
  - Oversized: most parts not needed → launch time ~1min, costly updates
- Approach: Containerization = link only parts required
- PS: Microservice idea lifting off!



APP #1

BINS / LIBS

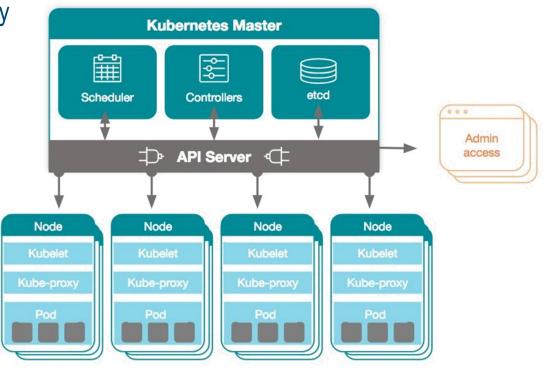
Virtual Machines

**Docker Containers** 



#### **Kubernetes**

- automating deployment, scaling, management of containerized applications
- group containers that make up an application into logical units
  - easy management & discovery
- Open source by Google: kubernetes.io





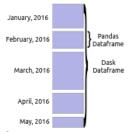
### Dask

- parallelism for python analytics, enabling performance at scale
  - Dynamic task scheduling
  - "Big Data" collections larger-than-memory / distributed environments
- Open source: dask.org

# Numpy Numpy Array Dask Array

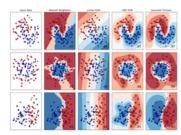
Dask arrays scale Numpy workflows, enabling multi-dimensional data analysis in earth science, satellite imagery, genomics, biomedical applications, and machine learning algorithms.

#### **Pandas**



Dask dataframes scale Pandas workflows, enabling applications in time series, business intelligence, and general data munging on big data.

#### Scikit-Learn



Dask-ML scales machine learning APIs like Scikit-Learn and XGBoost to enable scalable training and prediction on large models and large datasets.