

# Fundamentals of Cloud Computing

## Lecture 2

# Economic Fundamentals

# Clouconomics: Quantitative analysis of clouds

- **Common Infrastructure**
  - Benefits of pooling resources and performing statistical multiplexing of workloads
- **Location independent**
  - Relationship between #nodes and latency
- **Online connectivity**
  - Capital and latency tradeoff between P2P vs hub networks
  - Improvement that network optimizations can bring
- **Utility Pricing (pay per use)**
  - When is the cloud cheap compared to owning?
  - Is it beneficial to use cloud if it is more expensive than owning?
- **on-Demand provisioning**
  - If your service is growing/declining exponentially, what is the overhead of fixed capacity?

\* Joe Weinman. Clouconomics: The Business Value of Cloud Computing, Wiley, 2012

# Utility Pricing

- Q: Should you go for the public cloud if the unit price is higher than a home-grown solution?
- Consider a car
  - Buy (lease) for EUR 10 per day vs. Rent a car for EUR 30 a day
  - If you need a car for 2 days in a month, buying would be much more costly than renting
  - It depends on the load/demand

# Utility Pricing: Calculation

- $L(t)$ : load (demand for resources)  $0 < t < T$
- $P = \max( L(t) )$  : Peak Load
- $A = \text{Avg}( L(t) )$  : Average Load
- $B = \text{Baseline (owned) unit cost}$  ;  $B_T = \text{Total Baseline Cost}$
- $C = \text{Cloud unit cost}$ ;  $C_T = \text{Total Cloud Cost}$
- $U = C / B$  : Utility Premium *For the rental car example,  $U=3$*
  
- $B_T = P \times B \times T$  (since Baseline should handle peak load)
- $C_T = \int C \times L(t) dt = A \times U \times B \times T$
- **When is cloud cheaper than owning?**
  - $C_T < B_T$
  - $A \times U \times B \times T < P \times B \times T \rightarrow U < P/A$
- **When Utility premium is less than Peak to Average load ratio**

# Utility Pricing: Summary

- Utility Pricing is good when demand varies over time
  - Usually the case of a start-up or a seasonal business
- When Utility Premium is less than ratio of Peak Demand to Average Demand, Cloud computing is beneficial
- Next, we look at the possible savings that Cloud providers can create using statistical multiplexing

# Common Infrastructure: Multiplexing

- **Resource pooling**

- Allows economies of scale
- Reduces overhead cost
- Allows cloud provider more negotiating power when buying infrastructure (volume purchasing)

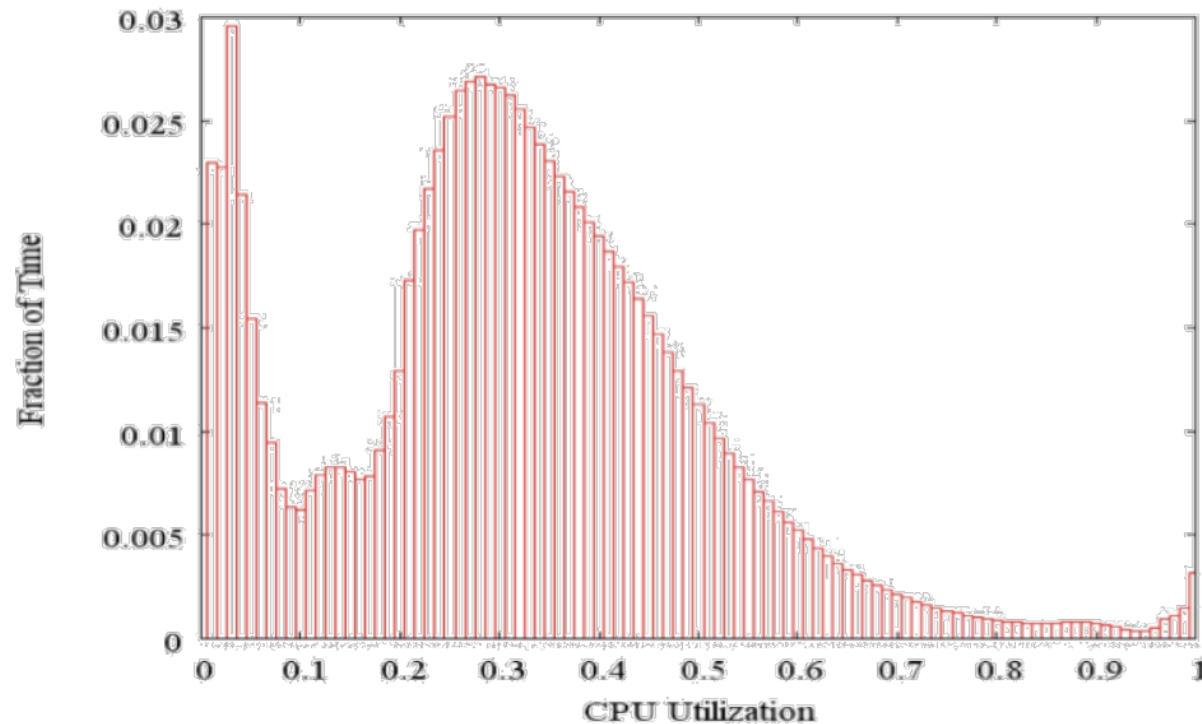
- **Multiplexing (multi-tenancy)**

- Allows statistics of scale

# Common Infrastructure: Multiplexing

- Assume you combine 2 infrastructures into a bigger one
  - One is built to peak requirements
  - The other is built to less than peak

Activity profile of a sample of 5,000 Google Servers over a period of 6 months





# Benefits of multiplexing

- Benefits for the infrastructure built to peak
  - Load multiplexing yields higher utilization and lower cost per delivered resource w.r.t. unconsolidated workloads
- Benefits for part of the system built to less than peak
  - Load multiplexing can reduce the unserved requests
  - Reduces a penalty function associated with such requests (e.g., a loss of revenue or a Service-Level agreement SLA violation payout).

# A Measure of Smoothness

- **Lets define coefficient of (load) variation  $C_v$** 
  - $C_v = \sigma / |\mu|$
  - non-negative ratio of the standard deviation  $\sigma$  to the absolute value of the mean  $|\mu|$ .
  - The larger the mean for a given standard deviation, or the smaller the standard deviation for a given mean, the “smoother” the load curve is
- **Importance of *smoothness***
  - An infrastructure with fixed assets servicing highly variable load will achieve lower utilization than a similar one servicing relatively smooth demand.
  - So for workloads with high  $C_v$ , or low smoothness, fixed asset servicing is bad
- Let's see what happens to  $C_v$  when we multiplex workloads

# Case study: Independent jobs

- Let  $X_1, X_2 \dots X_n$  be  $n$  independent jobs (random variables) with identical standard deviation  $\sigma$  and positive mean  $\mu$ 
  - Hence,  $Cv(X_1)=Cv(X_2)=\sigma/\mu$
- Consider the random variable  $X=X_1+X_2+\dots+ X_n$  (multiplexing)
- Statistics 101
  - $mean(X)=mean(X_1)+mean(X_2)+\dots+mean(X_n)=n\mu$
  - $var(X)=var(X_1)+var(X_2)+\dots+var(X_n)=n\sigma^2$

# Smoothing under multiplexing

- Hence standard deviation of  $X$  is
  - $\text{stdev}(X) = \sqrt{\text{Var}(X)} = \sqrt{n} \cdot \sigma$
- Finally  $\text{Cv}(X) = \sqrt{n} \cdot \sigma / n\mu = \sigma / \sqrt{n}\mu$ 
  - *i.e.*,  $\text{Cv}(X) = \text{Cv}(X_i) / \sqrt{n}$
  - We obtain “smoother” aggregate load with multiplexing!
  - Hence, we have benefits from statistics of scale in addition to those from economies of scale
- Best Case: Negative correlation
  - Consider jobs  $X_i$  and  $1-X_i$
  - Multiplexing:  $X = X_i + 1 - X_i = 1$
  - For random variable  $X = 1$ ,  $\sigma = 0$ ,  $\text{Cv} = 0$
  - Optimally smooth, best CPU utilization!

# Common Infrastructure: Summary

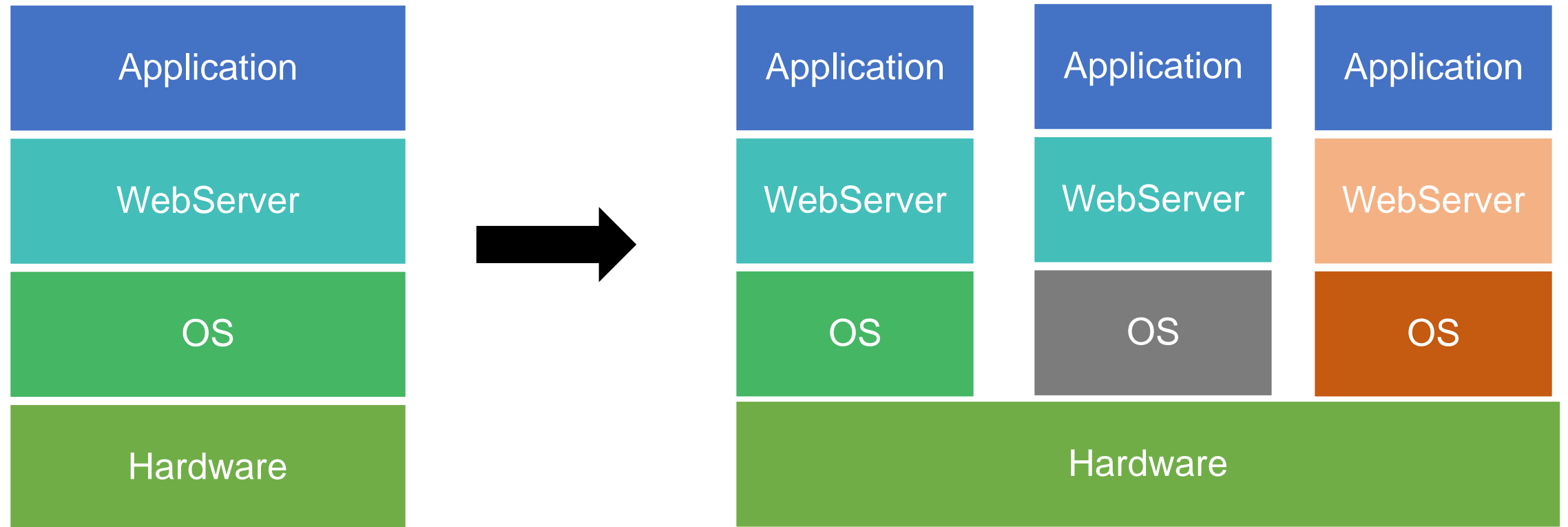
- Negative-correlated jobs
  - Private, mid-size, and large-size providers can experience similar statistics of scale
- Independent jobs
  - Mid-size providers can achieve similar statistical economies to an infinitely large provider
- Takeaway lesson: cloud provider of any size can be profitable!
  - At least according to “Value of Common Infrastructure”

# Infrastructure Fundamentals

## Virtualization

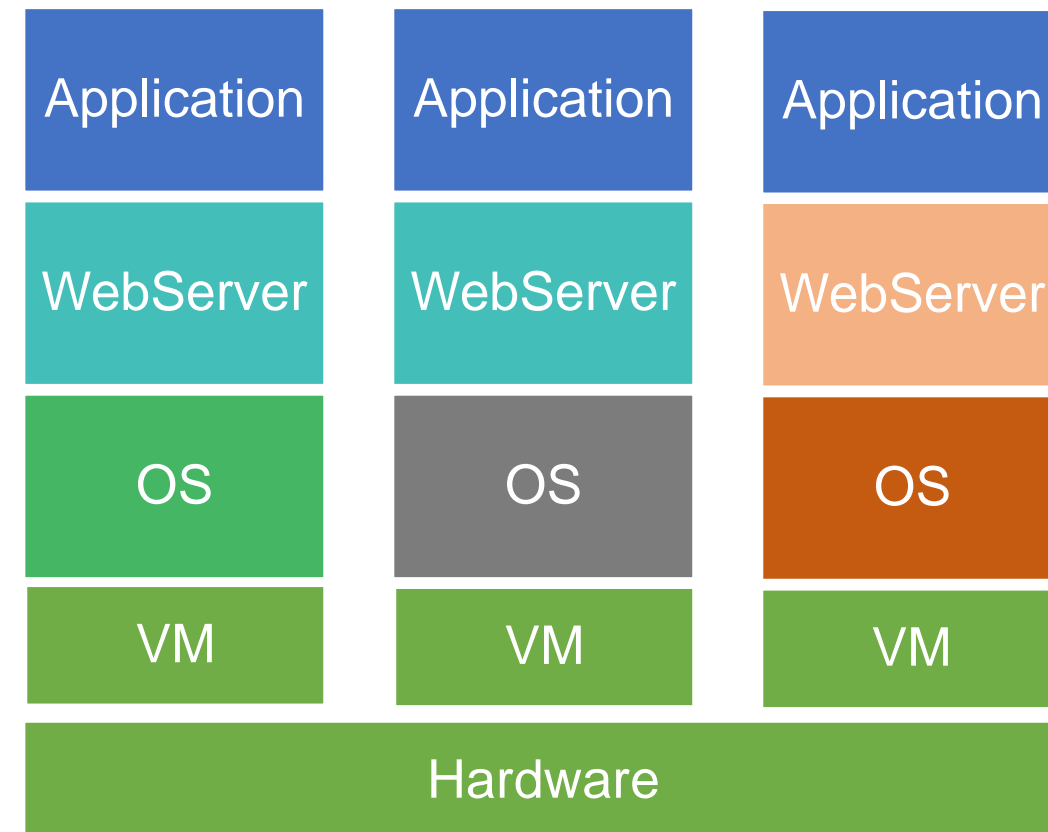
# Sharing Resources

- Economics of cloud computing requires resource sharing
- How do we share a physical computer among multiple applications?



# Virtualization: An Abstraction

- Introduce an abstract model of what a generic computing resource should look like – a “virtual machine” (VM)
  - CPU -> virtual CPU
  - Disk -> virtual Disk
  - NIC -> virtual NIC
- Provide one such “virtual machine” per tenant & host multiple virtual machines on the same physical machine



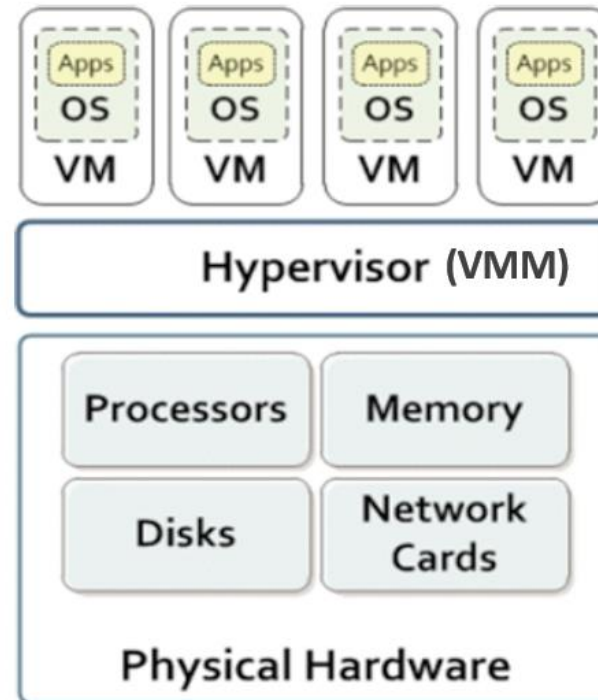


# Virtual machine and Hypervisor

- Virtual Machine:
  - “A fully protected and isolated copy of the underlying physical machine’s hardware.” -- definition by IBM
- Virtual Machine Monitor (aka VMM, aka Hypervisor):
  - “A thin layer of software that's between the hardware and the Operating system, virtualizing and managing all hardware resources”
- Two types of hypervisors
  - Type 1: VMM runs directly on physical hardware
  - Type 2: VMM built on top of a host OS

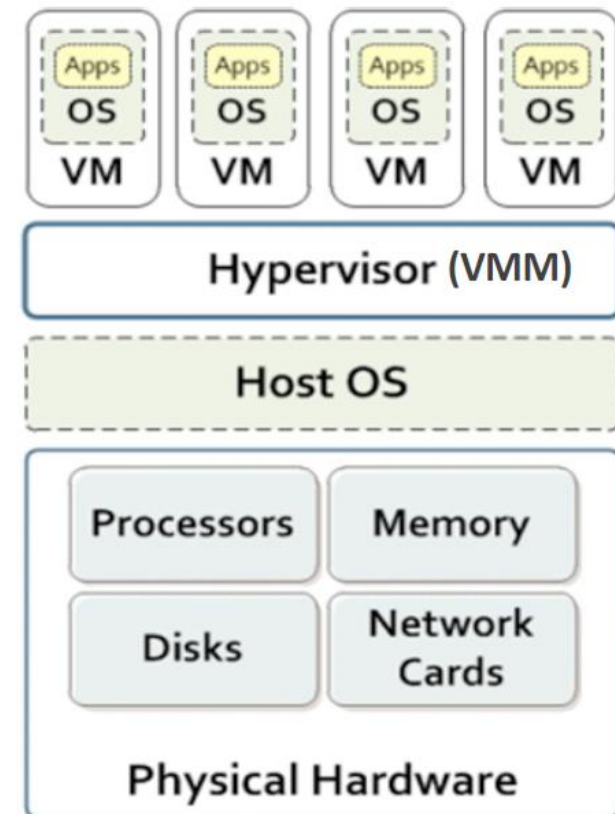
# Type 1 Hypervisor

- VMM directly implemented on physical hardware
- VMM performs scheduling and allocation of resources
- Eg: VMWare ESX Server

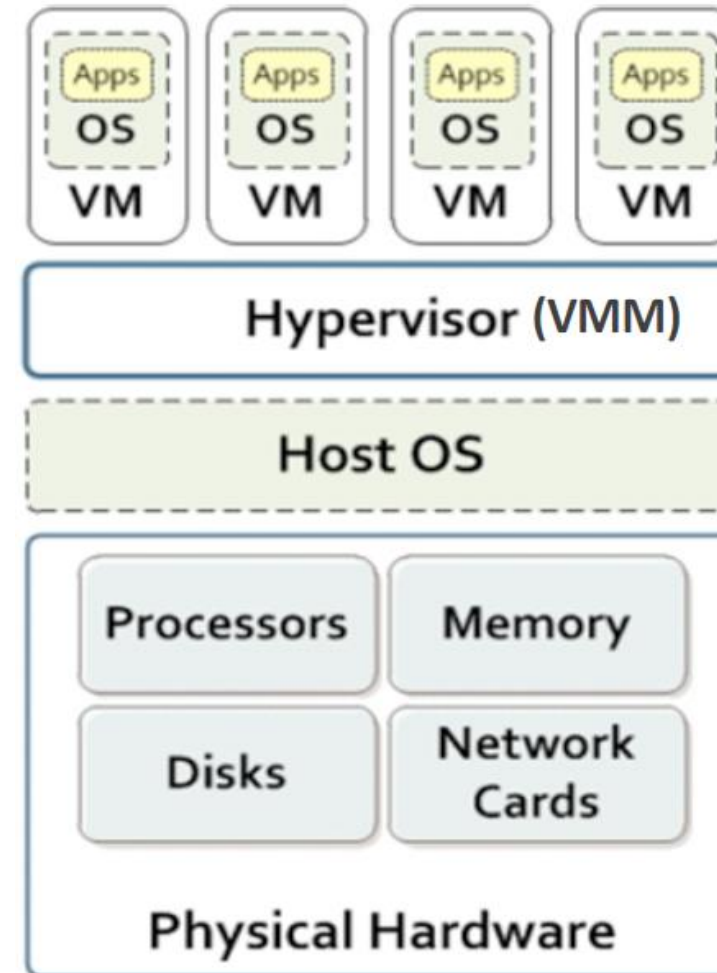
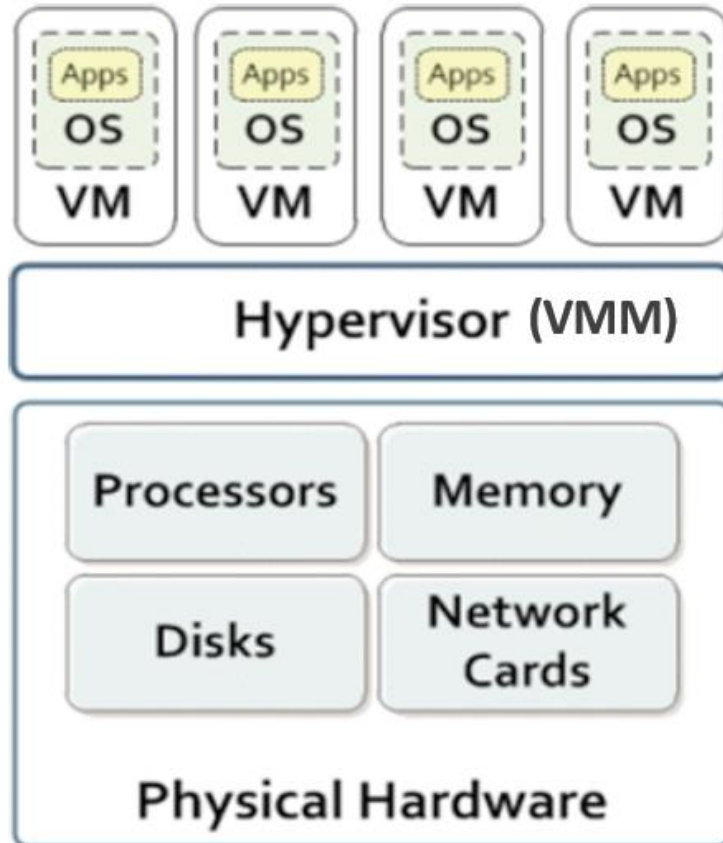


# Type 2 Hypervisor

- VMMs built completely on top of a host OS
- Host OS provides resource allocation and standard execution environment to each “guest OS”
- Example: User-mode Linux (UML), QEMU



# Type 1 vs type 2



# Virtualization: A bit of history (1960s, 70s)

- IBM sold big mainframe computers
  - Companies could afford one
  - Wanted to run apps designed for different OSes.
- Idea: add a level of indirection!
- CP/CMS (Control Program/Cambridge Monitor System), 1968
  - Time sharing providing each user with a single-user OS
  - Allow users to concurrently share a computer
- Hot research topic in 60s-70s; entire conferences devoted to VMs



IBM System/370

# Virtualization fades away (1980s)

- Interest died out in 80s
  - More powerful, cheaper machines
  - Ex: sun workstation
- Could deploy new OS on different machine
  - More powerful OSES (UNIX, BSD, MINIX, Linux)
- No need to use VM to provide multi-user support



Ken Thompson, Dennis Ritchie



Andy Tanenbaum

# New Beginnings (1990s)

- Multiprocessor in the market
  - Innovative Hardware
- Hardware development faster than system software
  - Customized OS are late, incompatible, and possibly buggy
- Commodity OSes not suited for multiprocessors
  - Do not scale due to lock contention, memory architecture
  - Do not isolate/contain faults; more processors, more failures

# It's Disco time

[Disco: Running Commodity Operating Systems on Scalable Multiprocessors](#), Edouard Bugnion, Scott Devine, and Mendel Rosenblum, SOSP'97

- Idea: Insert a software layer -- Virtual Machine Monitor -- between hardware and OS running commercial OS
- Virtualization:
  - Used to be: make a single resource appear as multiple resources
  - Disco: make multiple resources appear like a single resource
- Trading off between performance and development costs

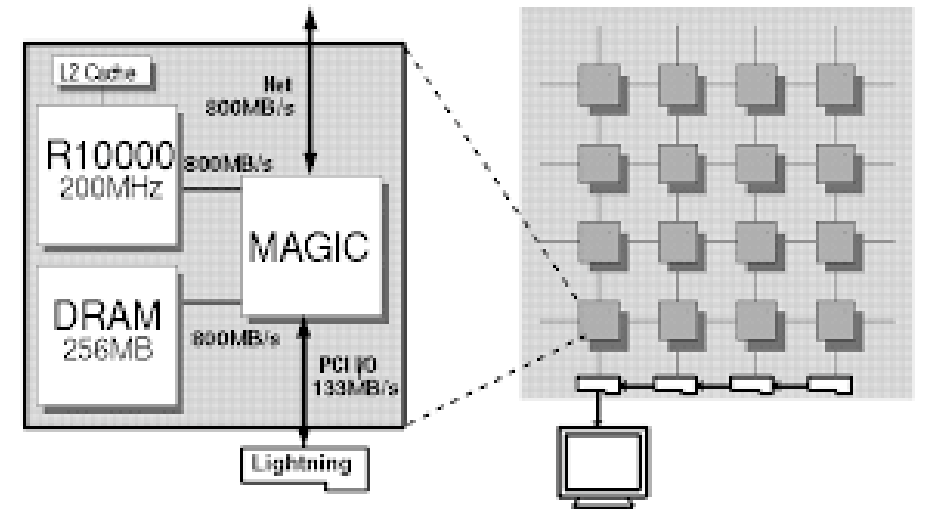


# Disco

- Extend modern OS to run efficiently on shared memory multiprocessors with minimal OS changes
- A VMM built to run multiple copies of Silicon Graphics IRIX operating system on Stanford Flash multiprocessor

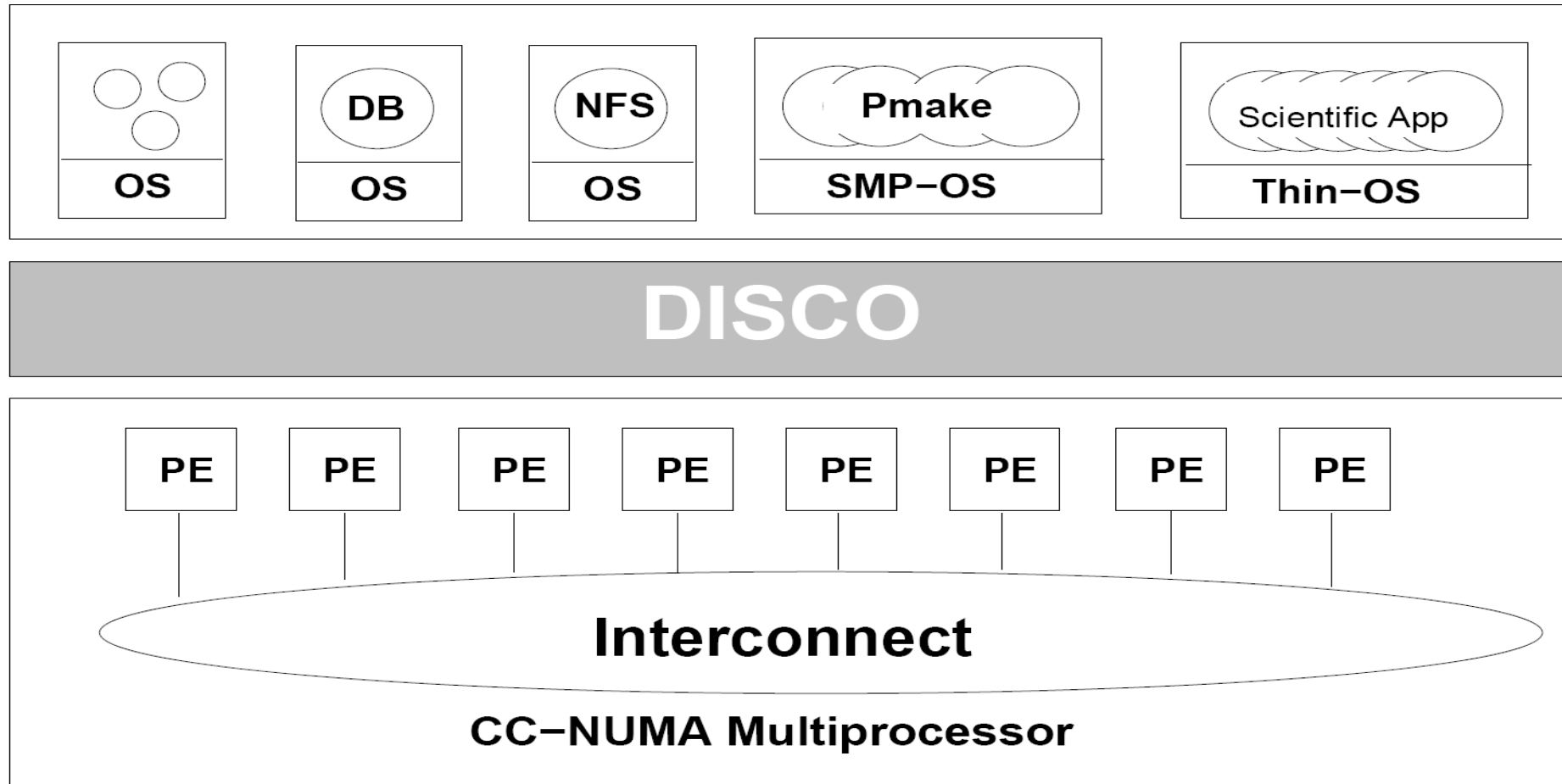


IRIX Unix based OS



Stanford FLASH: cache coherent NUMA

# Disco Architecture



# Disco to VMWare



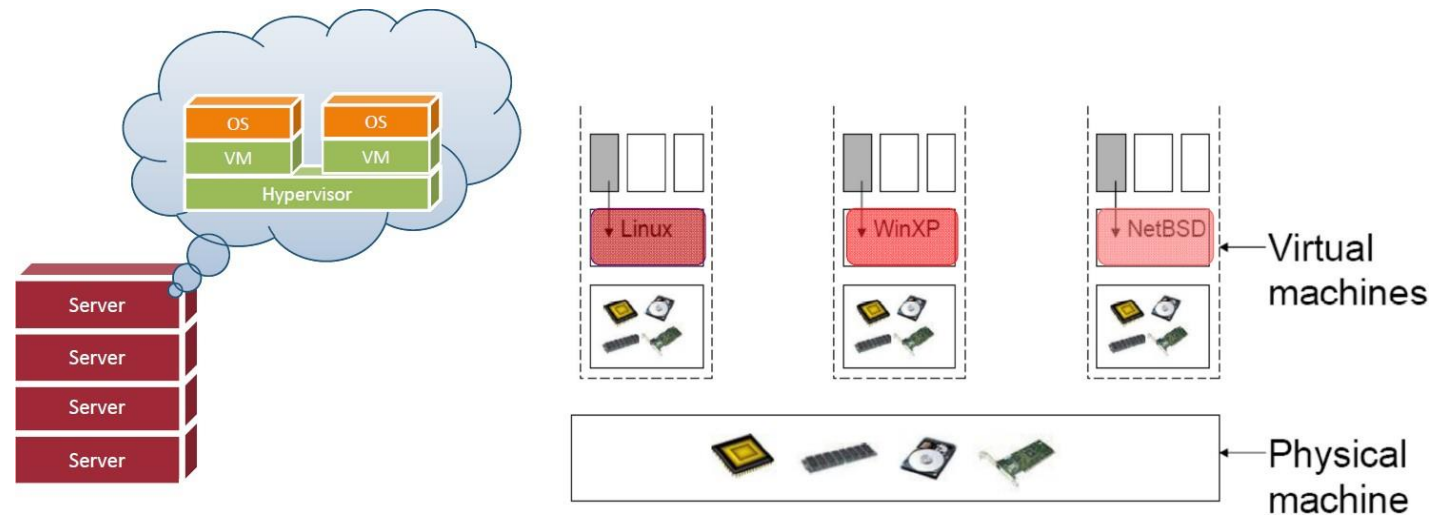
Mendel  
Roseblum  
(Stanford University)

- Started by creators of Disco
- Initial product: provide VMs for developers to aid with development and testing
  - Can develop & test for multiple OSes on the same box
- Actual, killer product: **server consolidation**
  - Enable enterprises to consolidate many lightly used services/systems
  - Cost reduction, easier to manage
  - Eventually over 90% of VMWare's revenue

# Virtualization and the cloud: IaaS recap

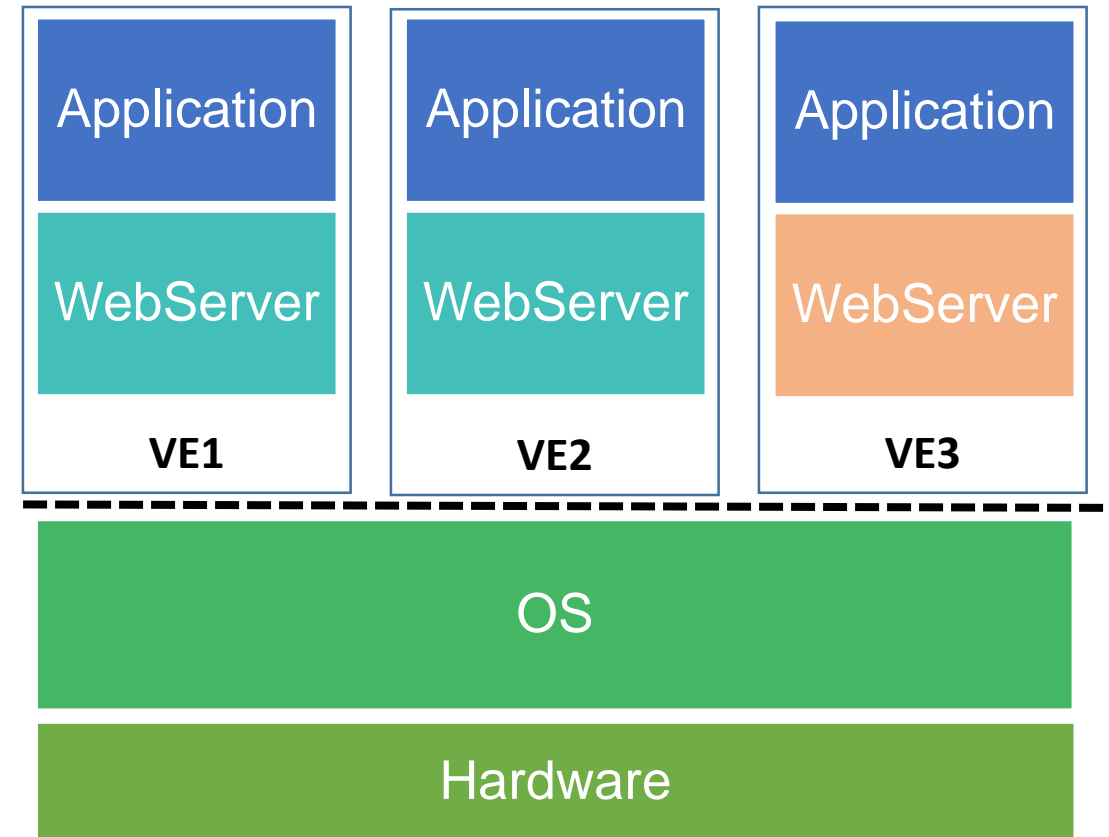
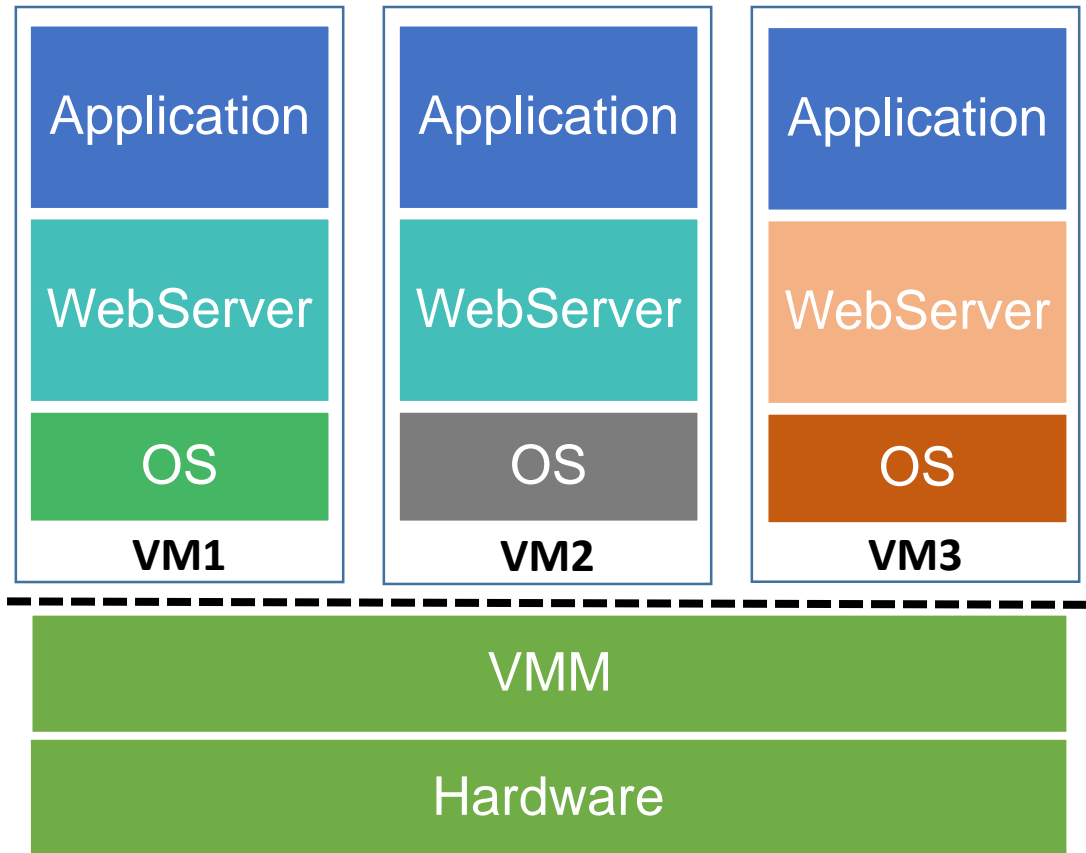
## Virtualization is the enabler of IaaS

- The cloud provider leases to users Virtual Machine Instances (i.e., computer infrastructure) using the virtualization technology
- The user has access to a standard Operating System environment and can install and configure all the layers above it



# From VMs to Containers

# Is full hardware emulation the only option for virtualization?



Can we raise the abstraction one level?  
Can we support OS-level virtualization to create “Virtual Environments”?

# Use case for OS-level virtualization

- Hosting providers & PaaS providers
  - Need to host multiple applications/tenants on a single server
- Full hardware virtualization still expensive
  - Full OS + libraries + application per tenant => reduced multitenancy
  - Use of VMMs means license fee
- OSes have always supported multitenancy
  - Multiple processes share same OS
  - Virtual memory & file systems for sharing memory and storage
- Can we extend OS to securely isolate multiple applications?
  - Observe and control resource allocation across groups of processes
  - Limit visibility and communication across groups of processes

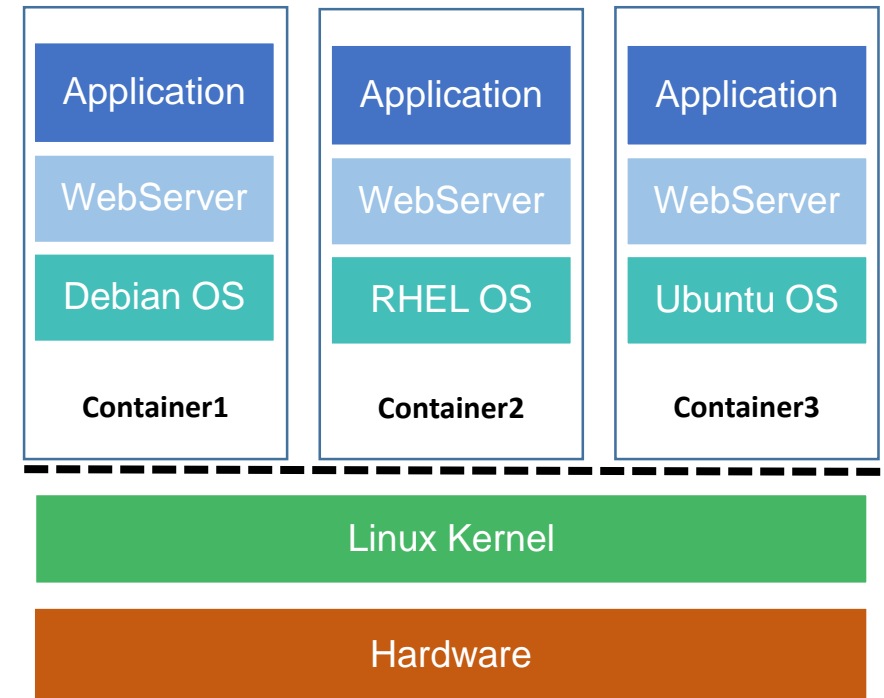
# Linux kernel functionality to support virtualization

- Cgroups (control groups)
  - Metering and limiting resources used by a group of processes
  - Ex: Partitioning resources in a server so that
    - Memory: Researchers: 40%, Professors: 40%, Students: 20%
    - CPU: Researchers: 50%, Professors: 30%, Students: 20%
    - Network: WWW browsing (20%), NFS (60%), Others (10%)
- Namespaces: Limiting what processes can view
  - Abstract a global system resource and make it appear as a separated instance to processes within a namespace.
    - Just like what a process within a VM can “see”
  - Example namespaces
    - Net: Each process group gets its own net interfaces, routing tables,
    - Pid: Processes can only see other processes in same PID namespace
    - Mount, IPC, ...



# From virtual machines to Linux containers

- Cgroups + namespaces provide (for most part) everything needed to create “containerized” processes.
  - SELinux and a few other pieces for security
- LXC (Linux Containers)
  - User-land tools that allows creation and running of multiple isolated Linux virtual environments (VE) on a single host
  - Apart from linux kernel, everything can be isolated--userland, libraries, runtimes, and applications
  - Ex: can be used to run multiple LINUX distros as containers

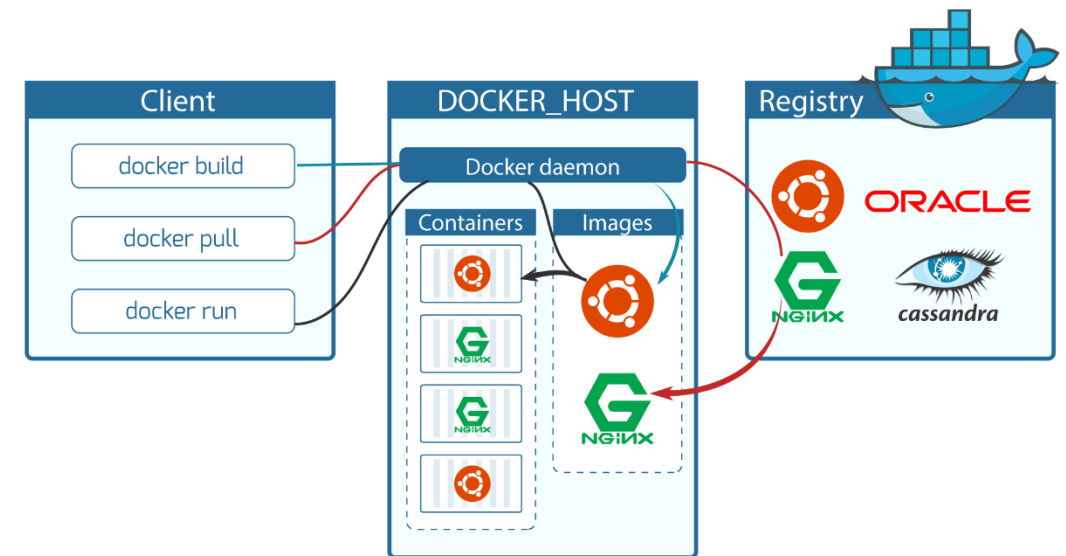


# From LXC to Docker

- Early LXC was built for sysadmins
  - No support for moving images, copy-on-write, sharing previously created images
- But developers have different needs
  - Application developed on dev servers, tested in build servers, and deployed in across production servers.
  - How do we make sure configuration is the same? Dependencies are met?
- Docker was developed by Solomon Hykes and others at dotCloud in 2013 to solve this problem with containers
  - **Package system** that can pack an application and all dependencies as a container image after development
  - **Transport system** that ensures that the application image runs exactly similar on test and production systems

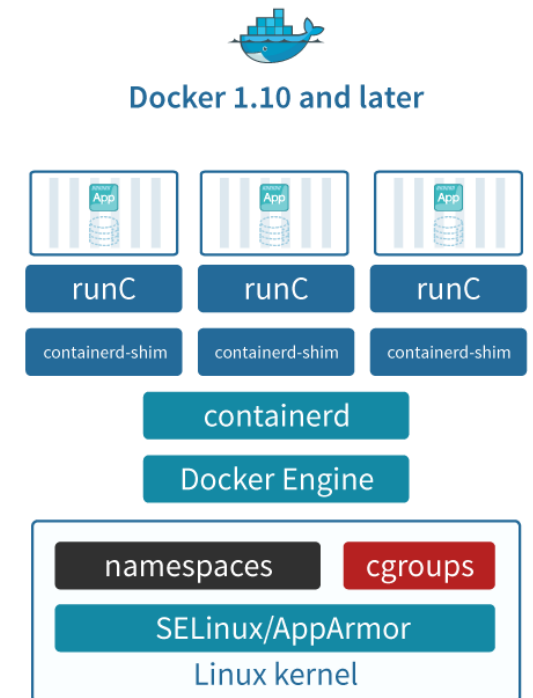
# More on Docker

- Docker was originally built on LXC and provided
  - A container image format
  - A method for building container images (Dockerfile/docker build)
  - A way to manage container images (docker images, docker rm , etc.)
  - A way to manage instances of containers (docker ps, docker rm , etc.)
  - A way to share container images (docker push/pull)
  - A way to run containers (docker run)



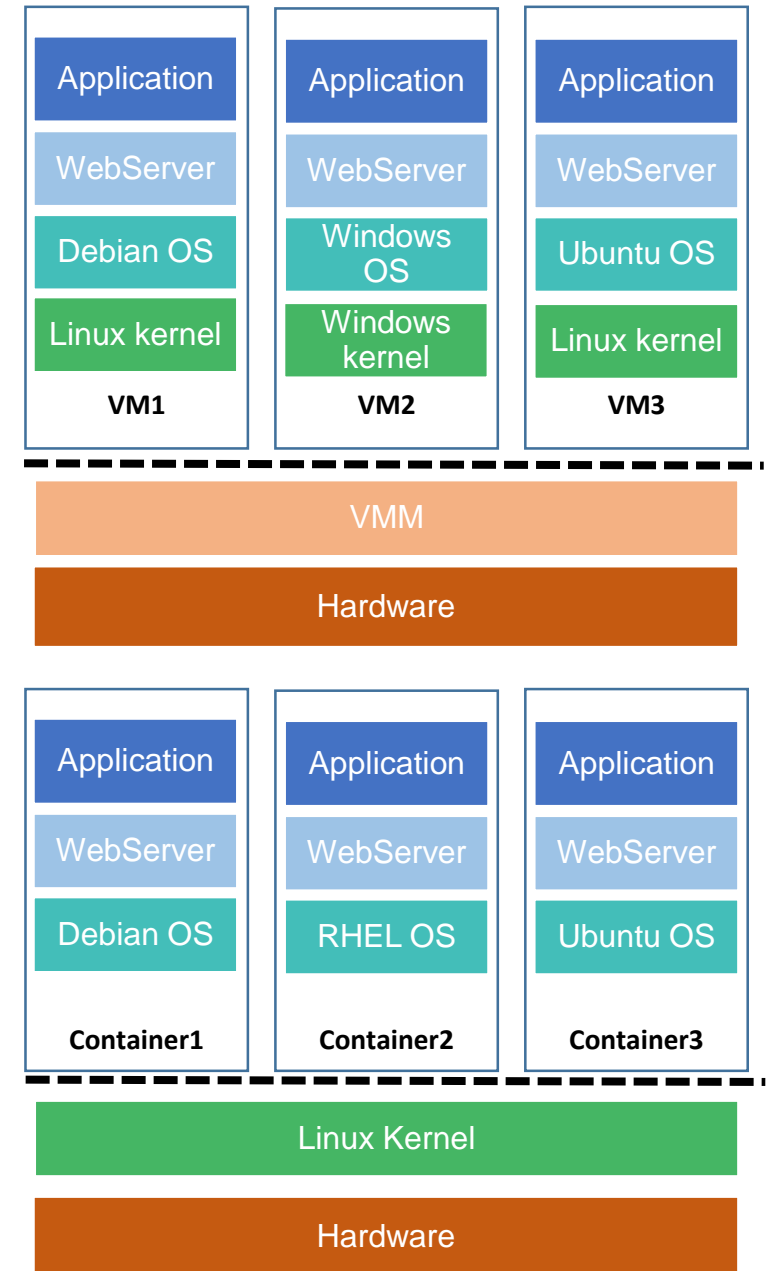
# Container ecosystem today

- Low-level Container runtimes
  - Set up and manage namespaces, cgroups and container execution
  - Ex: runC (open container initiative), lxc, rkt
- High-level container runtimes
  - Image format, image management, sharing
  - Ex: cri-o, containerd from Docker (builds on runc)
- Docker today is a collection of components
  - Docker engine: user-facing daemon, REST API, CLI
  - “Runtime”: Containerd, container-shim, runC



# Advantages of containers

- Abstraction levels
  - Hypervisors work at hardware abstraction level
  - Containers work at OS abstraction level
- Containers offer higher density
  - VMs need O(GB) vs containers that need O(MB)
  - Can pack many more containers per server
- Containers improve elasticity
  - Easy to “scale up” container than a VM
  - Reason for container adoption in hosting and PaaS environments
  - Example: Everything in Google from gmail to search is containerized
- Native CPU performance
  - No virtualization overhead
- Dramatically improves software development lifecycle
  - Easy to build, test, deploy software without worrying about portability

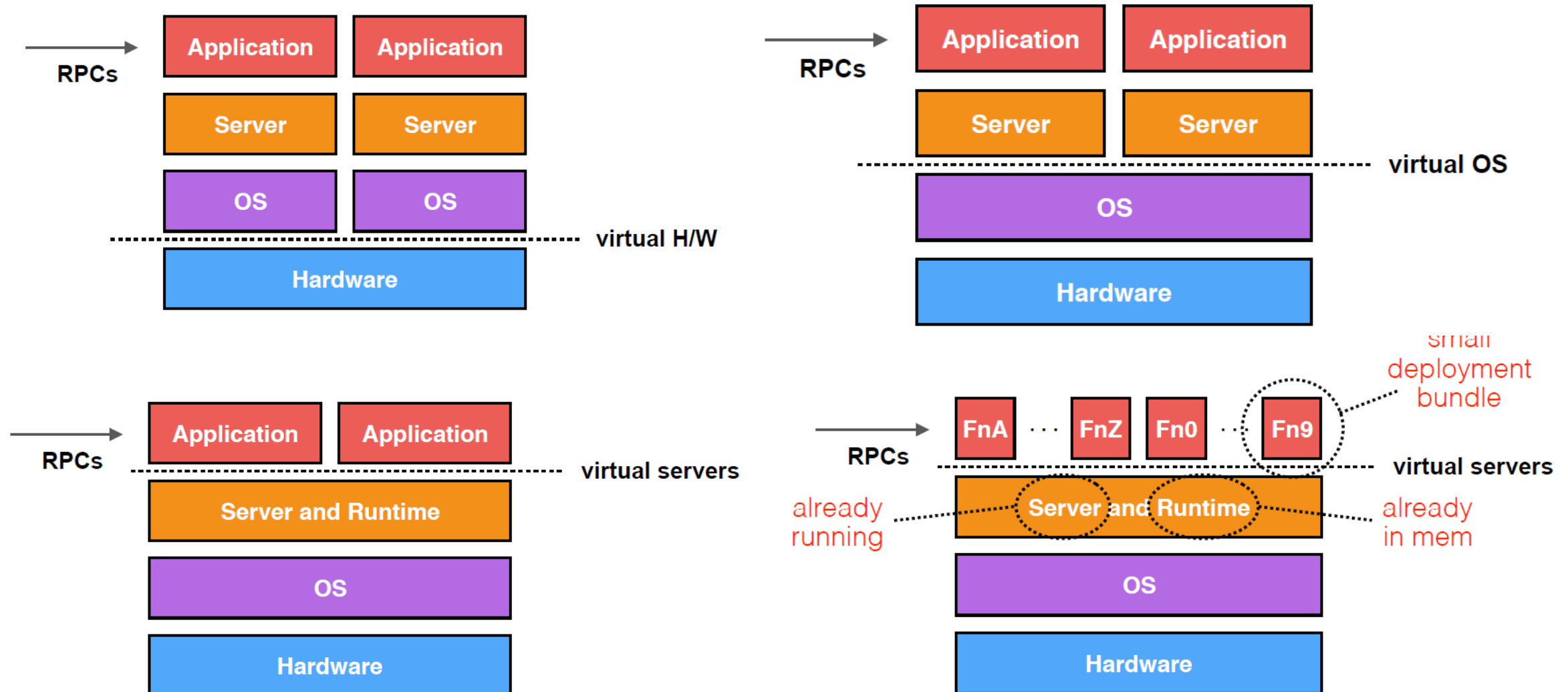


# From Containers to Serverless Functions

# Container Case Study

- Google Borg
  - Internal container platform at Google (<http://static.googleusercontent.com/media/research.google.com/en//pubs/archive/43438.pdf>)
  - 25 second median startup!
    - 80% of time spent on package installation
- What if we have an light-weight HTTP app?
  - Say < 1,000 LoC that runs for 200 msecs on each HTTP request?
  - In a container, the app would take too long to start
- Containers cannot deal with flash crowds, load balancing, interactive development, etc

# Three generations of virtualization

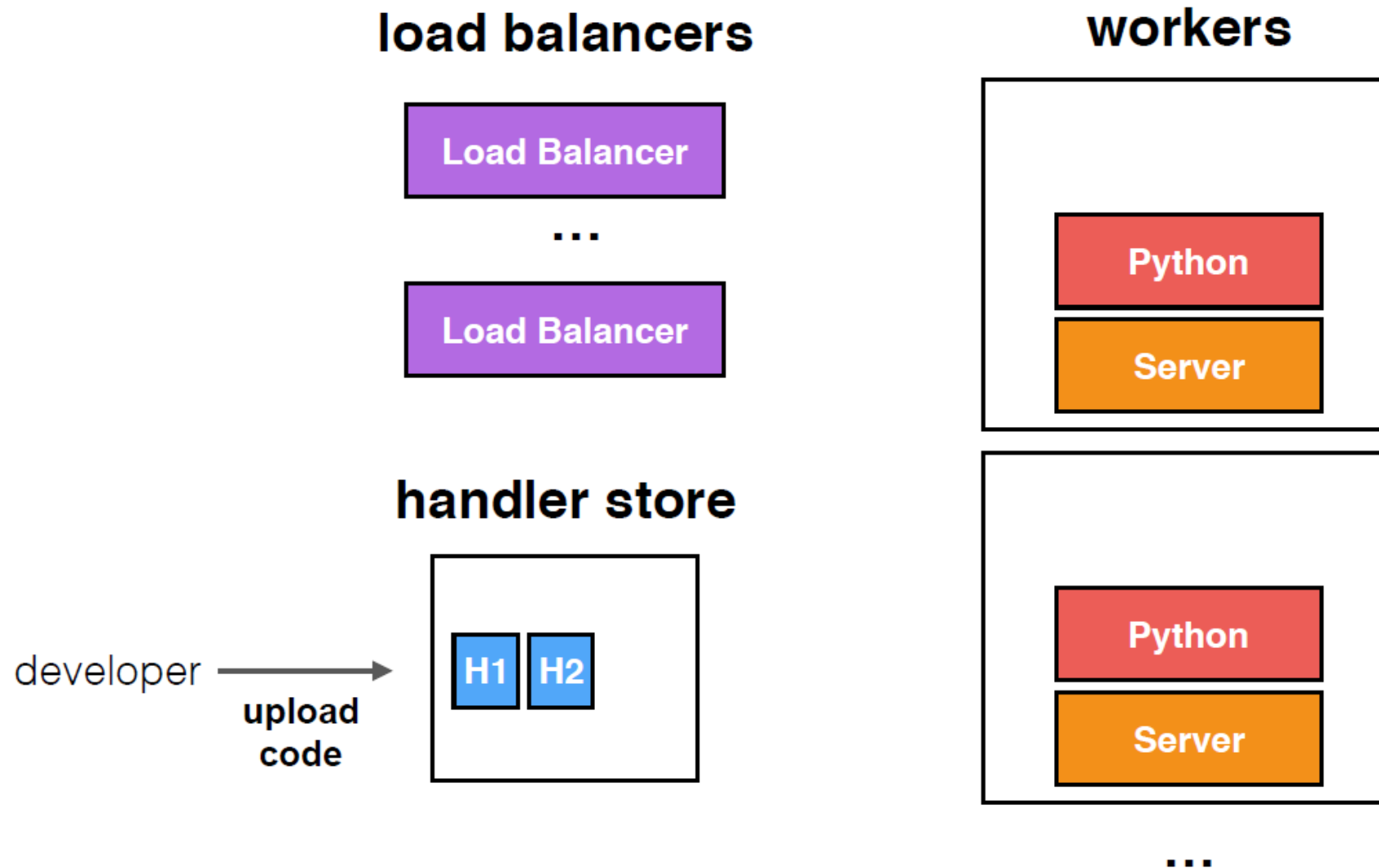




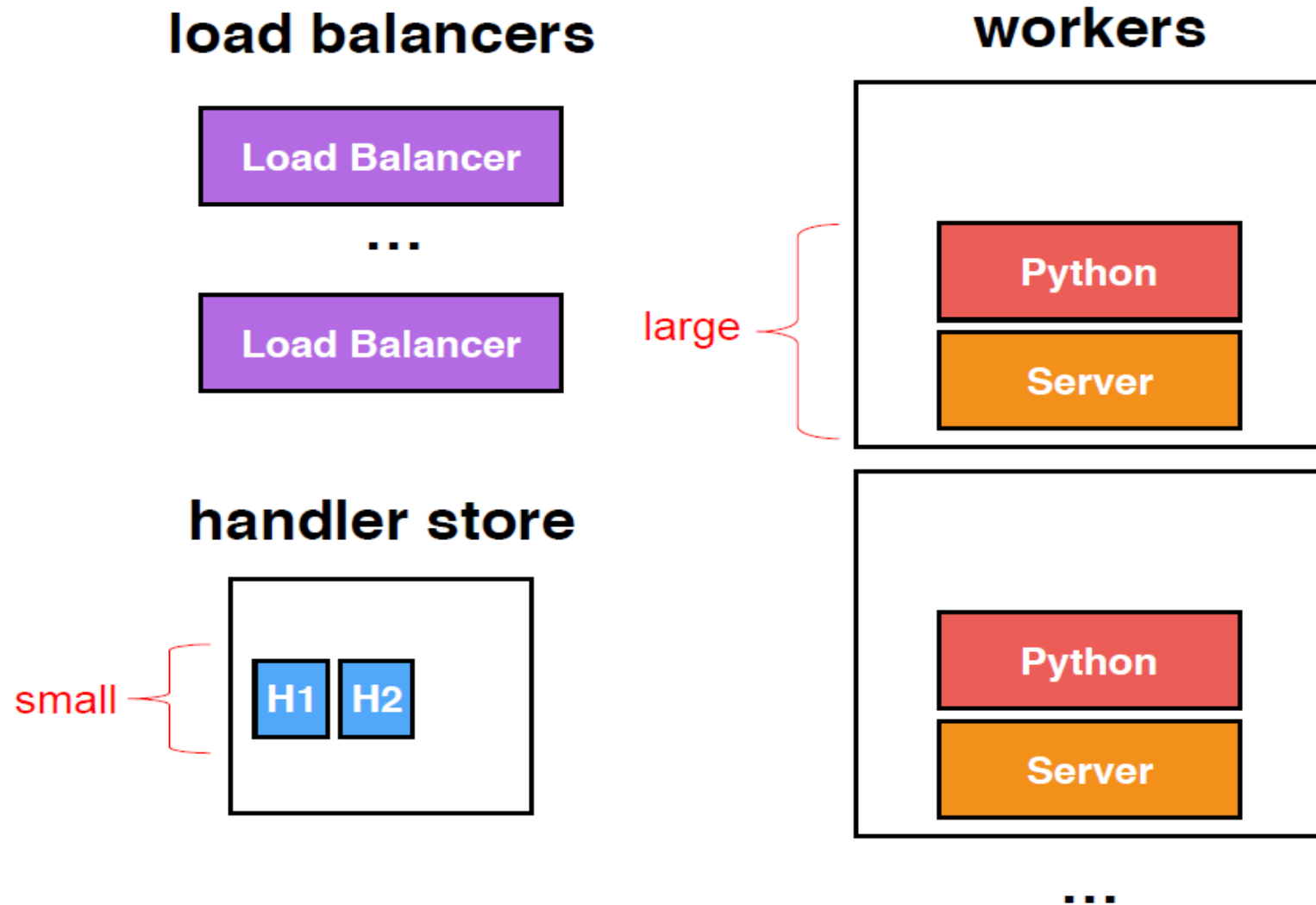
# Serverless functions: Model

- Run user handlers in response to events
  - web requests (RPC handlers)
  - database updates (triggers)
  - scheduled events (cron jobs)
- Pay per function invocation
  - actually pay-as-you-go
  - no charge for idle time between calls
  - e.g., charge  $\text{actual\_time} * \text{memory\_cap}$
- Share server pool between customers
  - Any worker can execute any handler
  - No spinup time
  - Less switching
- Encourage specific runtime (C#, Node.JS, Python)
  - Minimize network copying
  - Code will be in resident in memory

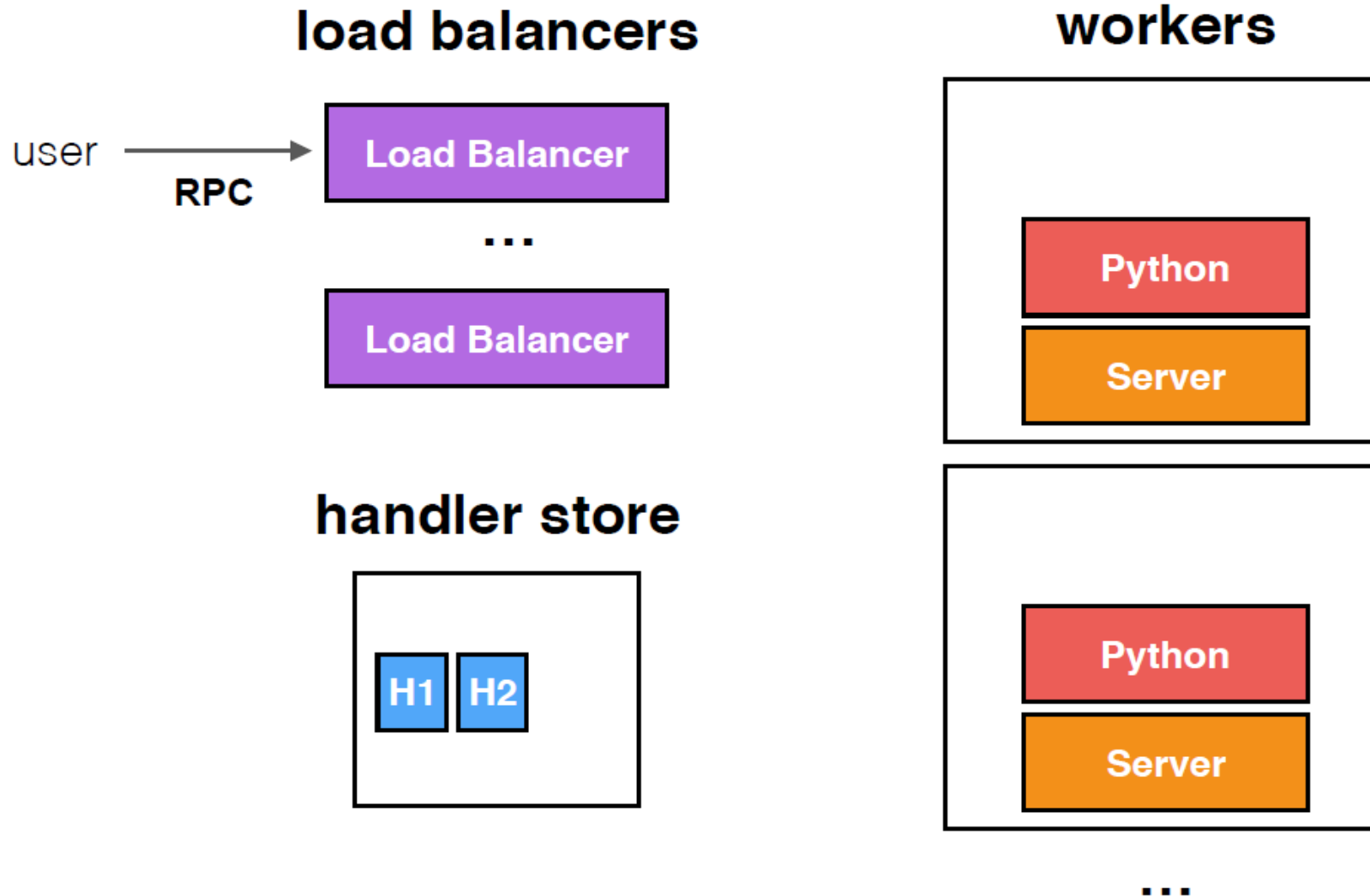
# Architecture



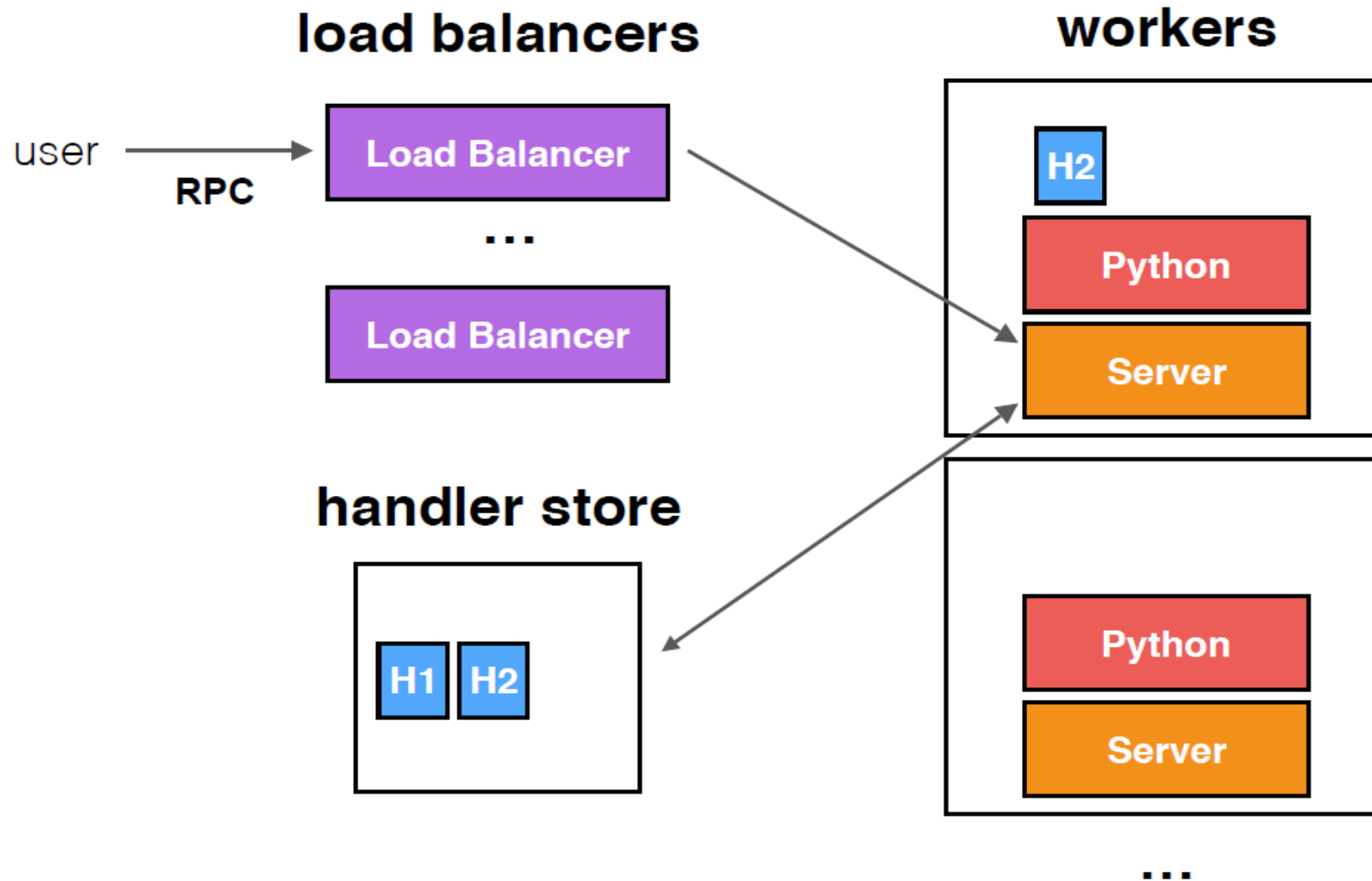
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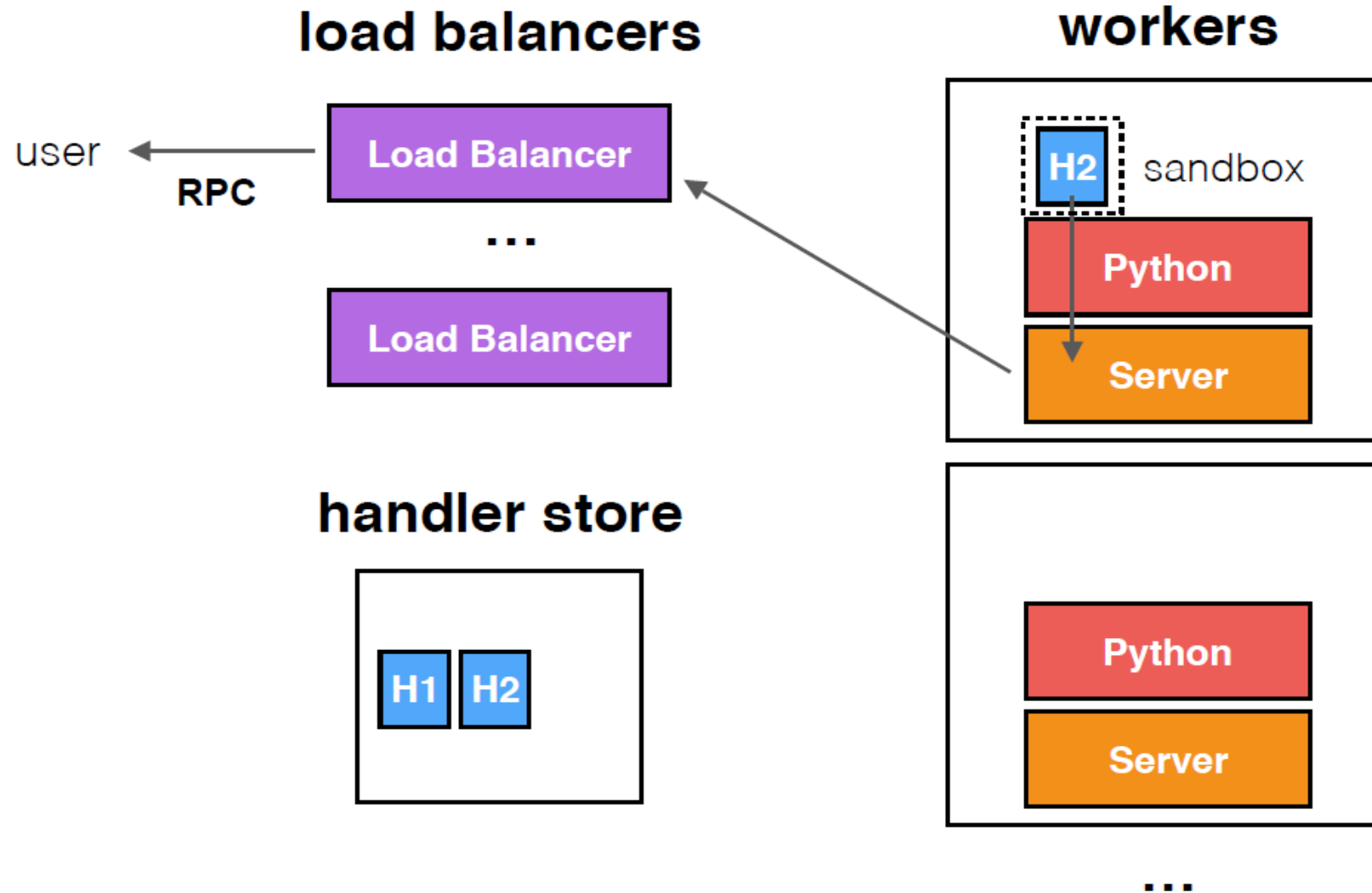
# Architecture



# Architecture



# Architecture



# Functions vs containers

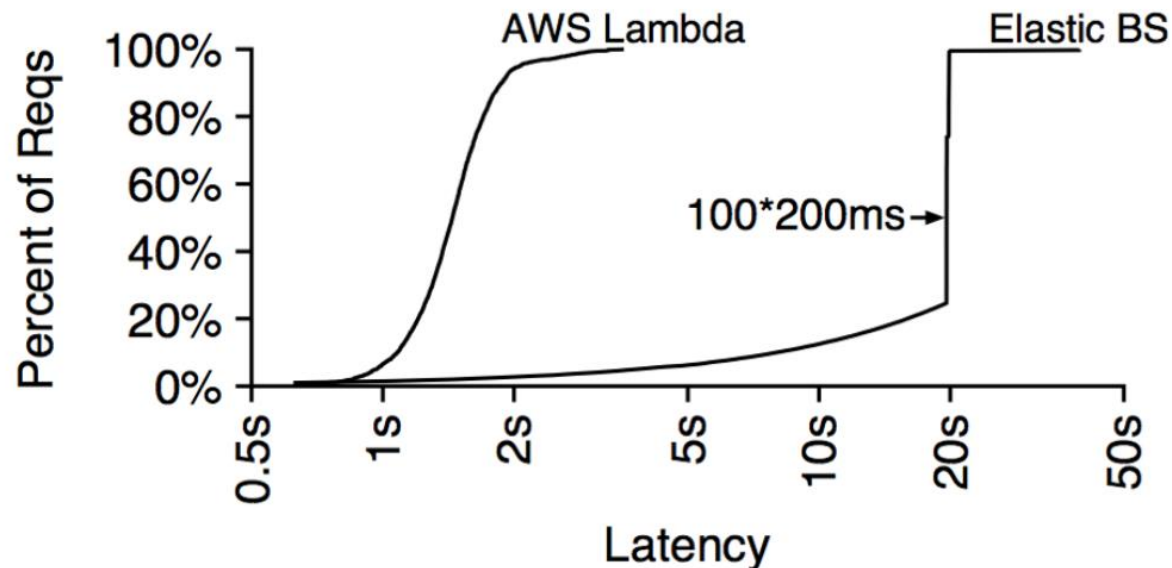
## Serverless Computation with OpenLambda,

Scott Hendrickson, Stephen Sturdevant, Tyler Harter, Venkateshwaran Venkataramani†, Andrea C. Arpaci-Dusseau, Remzi H. Arpaci-Dusseau

- Experimental setup:
  - Amazon Elastic Beanstalk
    - Autoscaling cloud service
    - Build applications as containerized servers, service RPCs
    - Rules dictate when to start/stop (various factors)
  - AWS Lambda serverless functions
- Workload
  - Simulate a small short burst
  - Maintain **100 concurrent requests**
  - Use **200 ms** of compute per request
  - Run for **1 minute**

# Scalability result

- AWS Lambda RPC has a median response time of only 1.6s
  - Lambda was able to start 100 unique worker instances within 1.6s
- An RPC in Elastic BS often takes 20s.
  - All Elastic BS requests were served by the same instance; as a result, each request had to wait behind 99 other 200ms requests.





# Functions vs explicit provisioning

- With VMs or containers, we need to decide
  - What type of instances?
  - How many to spin up?
  - What base image?
  - What price spot?
  - And then wait to start.....
- Functions truly delivery the promise of the cloud
  - finally pay-as-you-go
  - finally elastic
  - will fundamentally change how people build scalable applications