

## From Shared Virtual Memory to Parameter Servers

Phil Gibbons

15-712 F15

Lecture 16

## Today's Reminders

- No Class Friday or Monday
- My office hours: Today after class

2

## Memory Coherence in Shared Virtual Memory Systems [PODC'86, TOCS 1989]

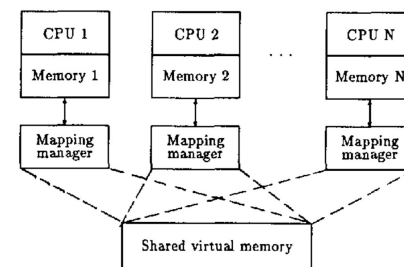
- Kai Li (Princeton)
  - **data**domain co-founder (acquired for \$2.4B !)
  - ACM/IEEE Fellow; NAE
- Paul Hudak (Yale, d. 4/15)
  - ACM Fellow; Co-designed Haskell



*"The paper shows how to simulate coherent shared memory on a cluster, and also introduces directory-based distributed cache-coherence. It spawned an entire research area, and introduced cache coherence mechanisms that are widely used in industry." – SigOps HoF citation*

3

## Shared Virtual Memory



- Page data between the physical memories of the processors (as well as between physical memory & disk)
  - Common mechanism for both
  - Once page in, data access is familiar read/write

4

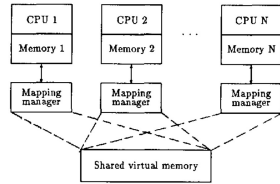
# Memory Coherence Problem

The diagram illustrates a multiprocessor system with shared virtual memory. At the top, there are N processing units, each consisting of a CPU (CPU 1, CPU 2, ..., CPU N) and a local Memory (Memory 1, Memory 2, ..., Memory N). Below each local memory is a Mapping manager. Solid lines connect each CPU to its local memory, and each local memory to its Mapping manager. Dashed lines connect each Mapping manager to a central Shared virtual memory block at the bottom. Ellipses between CPU 2 and CPU N indicate that there are more than two processors in the system.

- **Memory coherence:** read returns value of most recent write to same address
- **Differs from multiprocessor cache coherence**
  - Small caches, fast bus, done in HW => write conflicts incur small delay

“Both theoretical & practical results show MC problem can be solved efficiently on a loosely coupled multiprocessor”

5



- **Memory coherence:** read returns value of most recent write to same address
- **Differs from multiprocessor cache coherence**
  - Small caches, fast bus, done in HW => write conflicts incur small delay

“Both theoretical & practical results show MC problem can be solved efficiently on a loosely coupled multiprocessor”

# Why Shared Virtual Memory Should Have Good Performance

- Unshared data is fine
- Read-only shared data is fine
- Updates to shared data?
  - Each individual thread has good locality in its writes
  - Common goal in designing parallel algorithms is to minimize write contention among threads

- **Unshared data is fine**
- **Read-only shared data is fine**
- **Updates to shared data?**
  - Each individual thread has good locality in its writes
  - Common goal in designing parallel algorithms is to minimize write contention among threads

# Granularity

- **Why larger pages?**
  - Amortize communication overheads
  - 1000s of bytes roughly same cost as 10s of bytes
- **Why smaller pages?**
  - Minimizes chance for contention (false sharing)
- **Choose size to match existing VM page size**
  - Can use existing page protection mechanisms:  
single instructions will trigger page faults & trap to handlers,  
e.g. to enforce memory coherence mechanisms

- **Why larger pages?**
  - Amortize communication overheads
  - 1000s of bytes roughly same cost as 10s of bytes
- **Why smaller pages?**
  - Minimizes chance for contention (false sharing)
- **Choose size to match existing VM page size**
  - Can use existing page protection mechanisms: single instructions will trigger page faults & trap to handlers, e.g. to enforce memory coherence mechanisms

# Maintaining Coherence

- **Each page can be**
  - Read-shared by 1 or more processors, or
  - Exclusively owned by a processor who can write
- **Directory-based coherence (“Fixed Distributed Manager”)**
  - Management of pages is partitioned across processors
  - On fault, consult manager for the page
  - Manager tracks set of read-sharers (“copyset”) or exclusive owner (“owner”) & serves as point of serialization

[Run through example on board]

- **Works well for Cache-coherence.**  
**What’s the issue for Shared Virtual Memory?**
  - Want to avoid extra hop to directory/manager

- **Each page can be**
  - Read-shared by 1 or more processors, or
  - Exclusively owned by a processor who can write
- **Directory-based coherence (“Fixed Distributed Manager”)**
  - Management of pages is partitioned across processors
  - On fault, consult manager for the page
  - Manager tracks set of read-sharers (“copyset”) or exclusive owner (“owner”) & serves as point of serialization

[Run through example on board]
- **Works well for Cache-coherence.**  
**What’s the issue for Shared Virtual Memory?**
  - Want to avoid extra hop to directory/manager

## Dynamic Distributed Manager: Metadata on pages

- $Ptable[p].access = \{read, write, nil\}$
- $Ptable[p].copyset = \text{processors with read copies}$
- $Ptable[p].lock$
- $Ptable[p].probOwner = \text{likely owner}$

9

## Dynamic Distributed Manager

*Read-fault handler:*  
 $Lock(Ptable[p].lock);$   
 ask  $Ptable[p].probOwner$  for read access to  $p$ ;  
 $Ptable[p].probOwner := ReplyNode;$   
 $Ptable[p].access := read;$   
 $Unlock(Ptable[p].lock);$

*Read server:*  
 $Lock(Ptable[p].lock);$   
 IF I am owner THEN BEGIN  
    $Ptable[p].copyset$   
      $:= Ptable[p].copyset \cup \{RequestNode\};$   
    $Ptable[p].access := read;$   
   send  $p$  to  $RequestNode$ ;  
 END  
 ELSE BEGIN  
   forward request to  $Ptable[p].probOwner$ ;  
    $Ptable[p].probOwner := RequestNode;$   
 END;  
 $Unlock(Ptable[p].lock);$

10

## Dynamic Distributed Manager

*Write fault handler:*  
 $Lock(Ptable[p].lock);$   
 ask  $Ptable[p].probOwner$  for write access to page  $p$ ;  
 $Invalidate(p, Ptable[p].copyset);$   
 $Ptable[p].probOwner := self;$   
 $Ptable[p].access := write;$   
 $Ptable[p].copyset := \{\};$   
 $Unlock(Ptable[p].lock);$

*Write server:*  
 $Lock(Ptable[p].lock);$   
 IF I am owner THEN BEGIN  
    $Ptable[p].access := nil;$   
   send  $p$  and  $Ptable[p].copyset$ ;  
    $Ptable[p].probOwner := RequestNode;$   
 END  
 ELSE BEGIN  
   forward request to  $Ptable[p].probOwner$ ;  
    $Ptable[p].probOwner := RequestNode;$   
 END;  
 $Unlock(Ptable[p].lock);$

11

## Speedups for 3D PDE

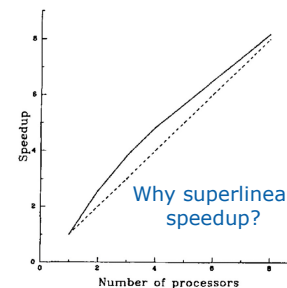


Fig. 5. Speedups of a 3-D PDE where  $n = 50^3$ .

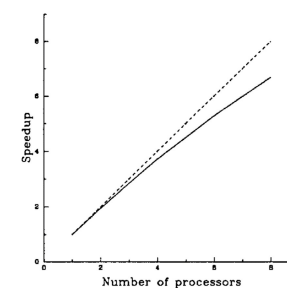


Fig. 7. Speedups of a 3-D PDE where  $n = 40^3$ .

**Note: Static mapping (CM\*) only comes close to SVM  
with heroic programming effort**

12

## Speedups for Sort & Dot-Product

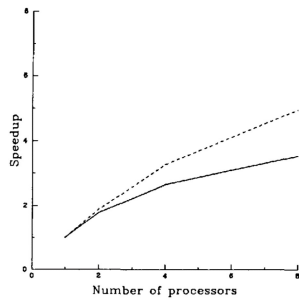


Fig. 8. Speedup of the merge-split sort.

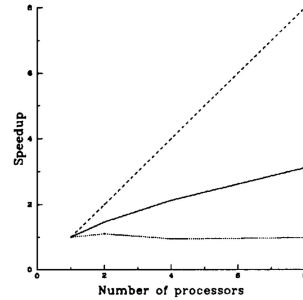


Fig. 9. Speedup of the dot-product program.

13

## Coherence Algorithms

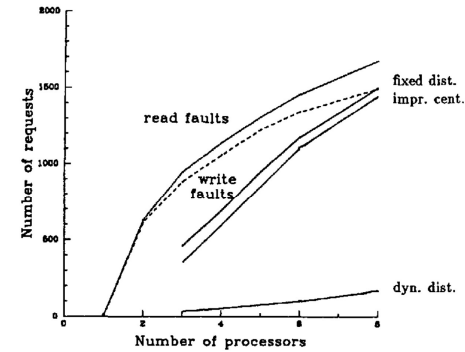


Fig. 11. Forwarding requests.

**3D PDE**  
probOwner field usually correct; page sharing is small

14

## Limitations

- Main classes of programs that would perform poorly:
  - Frequent updates to shared data
  - Excessively large data sets that are only read once
- Only ran on up to 8 processors

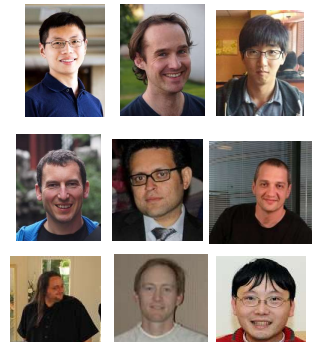
What should this paper get credit for?

15

## Scaling Distributed Machine Learning with the Parameter Server

[OSDI'14]

- Mu Li (CMU)
- David Andersen (CMU)
- Jun Woo Park (CMU)
- Alexander Smola (CMU)
- Amr Ahmed (Google)
- Vanja Josifovski (Pinterest)
- James Long (Google)
- Eugene Shekita (Google)
- Bor-Yiing Su (Google)



16

## Some Big Learning Frameworks

- **GraphLab (Dato)**

- Carlos Guestrin (CMU->Washington)



- **Spark (Databricks)**

- Ion Stoica (UC Berkeley)



- **Petuum**

- Eric Xing, Greg Ganger, Phil Gibbons, Garth Gibson (CMU)



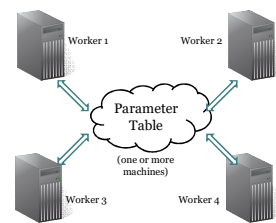
- **Parameter Server (Marianas Labs)**

- Alex Smola, Dave Andersen (CMU)



## Parameter Servers for Distributed ML

- Provides all machines with convenient access to global model parameters
- Enables easy conversion of single-machine parallel ML algorithms
  - “Distributed shared memory” programming style
  - Replace local memory access with PS access



**Single Machine Parallel**

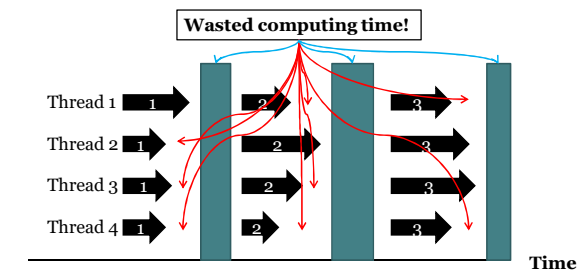
```
UpdateVar(i) {
  old = y[i]
  delta = f(old)
  y[i] += delta
}
```

**Distributed with PS**

```
UpdateVar(i) {
  old = PS.read(y,i)
  delta = f(old)
  PS.inc(y,i,delta)
}
```

† Ahmed et al. (WSDM 2012), Power and Li (OSDI 2010)

## The Cost of Bulk Synchrony

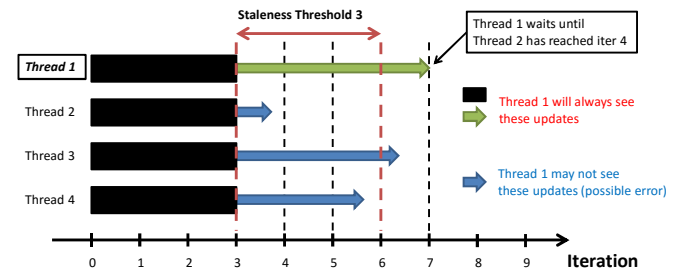


Threads must wait for each other  
End-of-iteration sync gets longer with larger clusters

**Precious computing time wasted**

But: Fully asynchronous => No algorithm convergence guarantees

## Stale Synchronous Parallel (SSP)



Allow threads to usually run at own pace

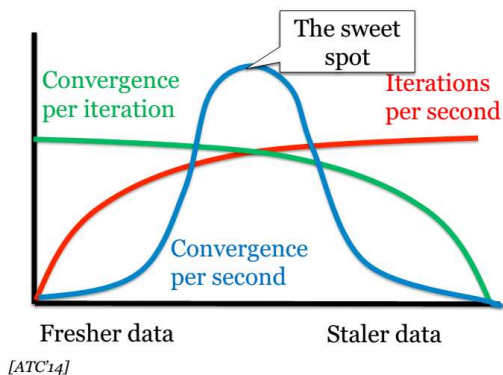
Fastest/slowest threads not allowed to drift >S iterations apart

Protocol: check cache first; if too old, get latest version from network

Consequence: fast threads must check network every iteration

**Slow threads check only every S iterations – fewer network accesses, so catch up!**

## Staleness Sweet Spot

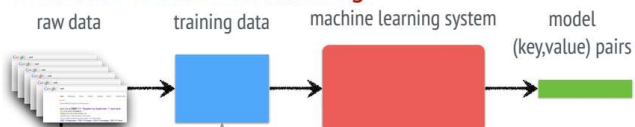


## Enhancements to SSP

- **Early transmission of larger parameter changes, up to bandwidth limit**
- **Find sets of parameters with weak dependency to compute on in parallel**
  - Reduces errors from parallelization
- **Low-overhead work migration to eliminate transient straggler effects**
- **Exploit repeated access patterns of iterative algorithms** (IterStore)
  - Optimizations: prefetching, parameter data placement, static cache policies, static data structures, NUMA memory management

## Parameter Server

### Overview of machine learning



### Scale of Industry problems

- ✦ 100 billion examples
- ✦ 10 billion features
- ✦ 1T – 1P training data
- ✦ 100–1000 machines



- ✦ scale to industry problems
- ✦ efficient communication
- ✦ fault tolerance
- ✦ easy to use

## Distributed Data Analysis Systems

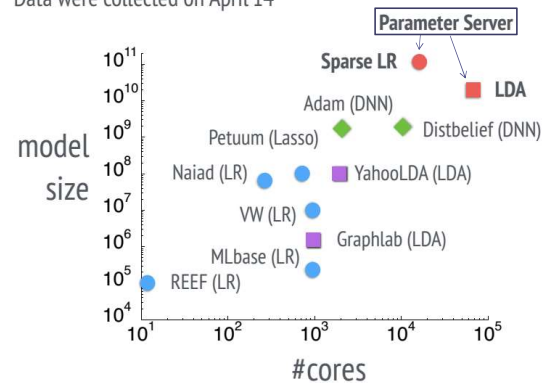
	Shared Data	Consistency	Fault Tolerance
Graphlab [34]	graph	eventual	checkpoint
Petuum [12]	hash table	delay bound	none
REEF [10]	array	BSP	checkpoint
Naiad [37]	(key,value)	multiple	checkpoint
MLbase [29]	table	BSP	RDD
Parameter Server	(sparse) vector/matrix	various	continuous

**Fair characterizations?**

## Parameter Server

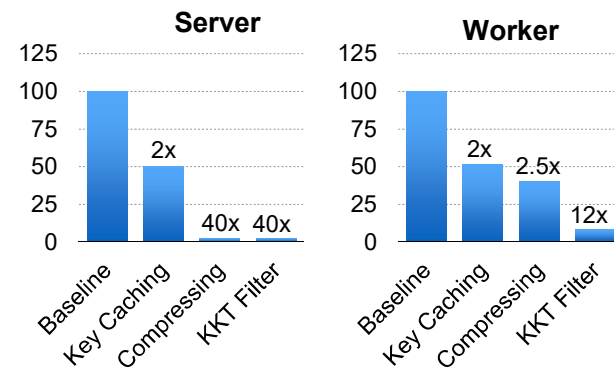
### Largest experiments of related systems

Data were collected on April'14

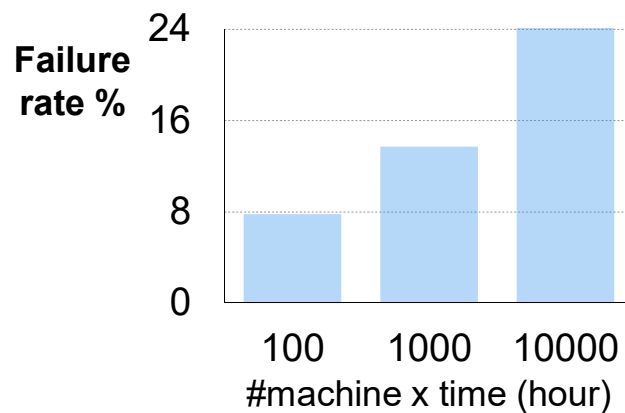


## Traffic Reduction by Filters

### Ad click prediction 636TB data, 1TB model, and 1000 machines

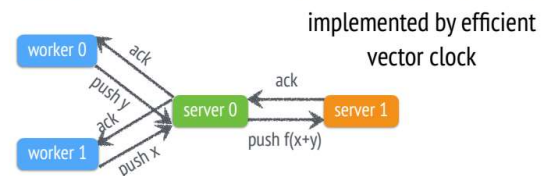


## Machine learning job logs in a 3-month period



## Fault Tolerance

- Model is partitioned by consistent hashing
- Default replication: Chain replication (consistent, safe)
- Option: Aggregation reduces backup traffic (algo specific)



## **Next Wednesday's Papers**

### **Application Performance and Flexibility on Exokernel Systems**

**Frans Kaashoek, Dawson Engler, Greg Ganger, Hector Briceno,  
Russell Hunt, David Mazziere, Thomas Pinckney,  
Robert Grimm, John Jannotti, Kenneth Mackenzie**

**SOSP'97**

### **Safe Kernel Extensions without Run-Time Checking**

**George Necula and Peter Lee**

**SigOps HoF paper**