## Scheduling: today and tomorrow

Larry Carter



# Apologies for

- Misrepresenting your work
- Not knowing very much
- Taking extreme positions

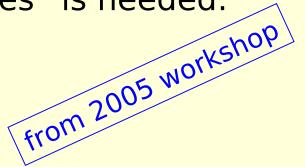
• ...

## Market-based systems are inevitable

A "convergence of technologies" is needed:

- Electronic money
- Allocatable resources
- Trust

- ...



Once this happens, compute power will either be:

- Abundant
- Scarce

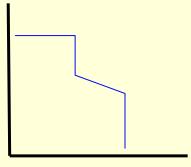
Moore's law will eventually fail; people's imagination won't

÷

## Issues for market-based scheduling

### User's quality measure:

Makespan, stretch, or steady-state throughput \$
User-specific "utility"



#### Completion time

### Server's goal:

Maximize throughput

Maximize profit ("structural unemployment" may be beneficial)

### System wide goal:

Fairness

Equal access to market, but "money talks"

#### User's motivation:

Altruism Profit

### New opportunities:

Brokers, publicists, insurers, ...

from 2005 workshop

# Towards more generality

### Computational platform

- Homogeneous → heterogeneous platform
- Processor time → communication, memory, ... times
- Centralized → distributed decision making
- Reliable → unreliable or collusive processors
- One → multiple administrative domains

### red = new since Aussois 2004

### Application model

- Independent tasks → Job = DAG of tasks
- Constant → changing resource needs
- Uniform tasks → multiple bags of tasks

### Objective of optimization

single goal → individual utility functions

# Combating NP-completeness

NP complete result → adaptive approximation algorithms → (simgrid) simulations

Makespan → Steady state → Trim analysis

## Non-traditional features

	Platform model	Application model	Objective function	Other features
Rosenberg		DAG	maximize   -ism	
Weinberg	multi-level			symbiosis
Lee	memory supercomputer	(real)	user-specified	
Jeannot			reduce errors	colluding users
Dongarra	supercomputer	(real)		
Trystram	multiple domains		multiple	
Agrawal	varying allocation			adversarial allocator
Beaumont	heterogeneous	multiple, divisible		
Detti			reliability	crashes possible
Marchal	heterogeneous	multiple bags	stretch	on-line versus off-
		ET CETERA		···

# What we're accomplishing

### Breadth-first search of new models

- Driven by technology changes
  - Multicore, unreliable processors, ...
- Improving constant from 9/7 to 5/4

Introducing (potentially important) new paradigms, e.g.

- IC-optimality
- Symbiosis
- Collusion-resistance
- Nash equilibrium

# What we're not accomplishing

# New algorithms implemented in "real" system

(Perhaps if we were that successful, we wouldn't be attending this workshop)

# What should we be accomplishing??

Computer science is not a natural science

We get to invent our own models

Discovering properties of random models isn't nearly as interesting as discovering "nature"

We should work towards having an influence

(Well, that's my opinion)

## How to have an influence

"Throw great idea over the wall"

i.e. publish paper

If it's good enough, people will pick it up example: randomized routine

But what's on the other side?

"Not invented here"

Usually, we must do (much) more

# A not-yet influential idea

## Bandwidth-centric scheduling

"A parent node responding to requests from multiple children should give first priority to child with highest bandwidth."

# A not-yet influential idea

## Bandwidth-centric scheduling

"A parent node responding to requests from multiple children should give first priority to child with highest bandwidth."

Why hasn't this been adapted by BOINC??

Bandwidths don't follow one-port model

BOINC doesn't even know the bandwidths

## The other side of the wall

### "Not invented here"

Learning our language and sifting many papers is very difficult

People have their own ideas they think are good

## To overcome these barriers, we need to

Learn about their world

Demonstrate effectiveness on their data

My experience: this effort benefits <u>me</u>
new problems
new ideas

Schoduling in Au

# Other ways to have influence

Often, our ideas are discovered independently by others

At best, our theory can help assure others that the ideas are valid (a constructive interaction)

At worst, we can get into big fights over who deserves the credit or patents

Our work can suggest what general directions are more or less promising (if we can get ourselves in an advisory position).

Perhaps we can demonstrate value of:

Collecting extra information

Providing new capability

## Influence-aware research

Suppose we want to do research on desktop grids for DAG applications

What is a potential application?

What information would be readily available to scheduler in such an application.

Can we argue that our technique is so good it's worth the effort to collect needed parameters?

Can we envision a path towards implementation

Possible target: Chess or Go on BOINC

# Multi-core: a new opportunity

Jack Dongarra's problem (LAPACK)

Even easily parallelized applications will need to tolerate variable execution times and failures

Scheduling for more general progams (e.g. threaded programs) will be needed

Adaptive, self-scheduling techniques are easiest to get adapted

Locality will be very important in future

The cost of moving data is MUCH more than the cost of computation

We must learn to live with unreliable cores

## Conclusion

We're doing excellent work

I'm not suggesting you totally change your research

Perhaps we could do more to get work used

Discussion

# Backup slides

# Symbiotic Scheduling

Symbiosis: Two applications run concurrently take less time than running one then the other.

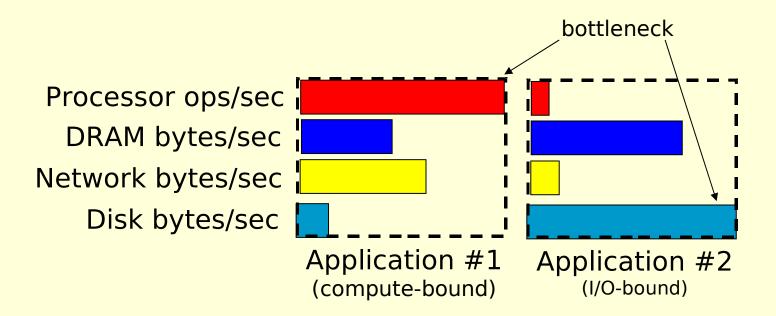
- "Timesharing" on uniprocessors usually isn't symbiotic.
- Symbiosis has been demonstrated for multithreaded processors (Snavely)

### Typical node:

- Multiple processors
- Shared memory (but separate caches)
- Communication network to other nodes
- I/O channel to disks

Opportunity for symbiosis when different applications have different bottlenecks.

## **Symbiosis**



## **Symbiosis**

