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## Coordinating multiple managers to achieve specified power-performance tradeoffs

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IBM Austin: Charles Lefurgy, Freeman Rawson



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# **Team Members**



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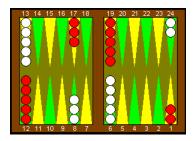


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### Watson Lab

Jeff Kephart

## **Austin Lab**



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## Background

- The costs and constraints associated with electrical power are becoming an increasing concern for data center operation
  - 50% of data centers will have insufficient power and cooling capabilities by 2008 (Gartner)
  - Power will be the second-highest operating cost (after labor) in 70% of data centers
  - According to Berkeley RAD lab, 30% reduction in power would save \$15B and 100M metric tons of CO<sub>2</sub> emissions per year in US (1.7% of total emissions)
- Power efficiency is becoming the subject of energy and environmental regulations by governments around the world
  - EPA report to Congress due in mid-2007; Energy Star standards for systems, data centers
  - Other agencies such as DoE and other governments involved as well

- Non-governmental organizations & industry groups actively studying the problem
  - Green Grid: consortium of IT companies developing best practices for reducing power consumption in data centers
    - AMD, Dell, HP, IBM, Sun Microsystems, Microsoft, VMWare, ...
  - SPEC working on a power/performance benchmark
- Industry is aggressively developing and marketing power-conserving hardware and software solutions



## Agenda

Background

### Power-Performance Research

- Algorithms
- Results
- Commercialization

### Universal algorithms for managing power and performance

Universally Optimal Power Management Algorithm

Algorithm MaxPowerSavings()
For each object 0 in DataCenter
TurnOff(0)

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### Universal Performance Management Algorithm

### Algorithm MaxPerformance()

For each object O in DataCenter

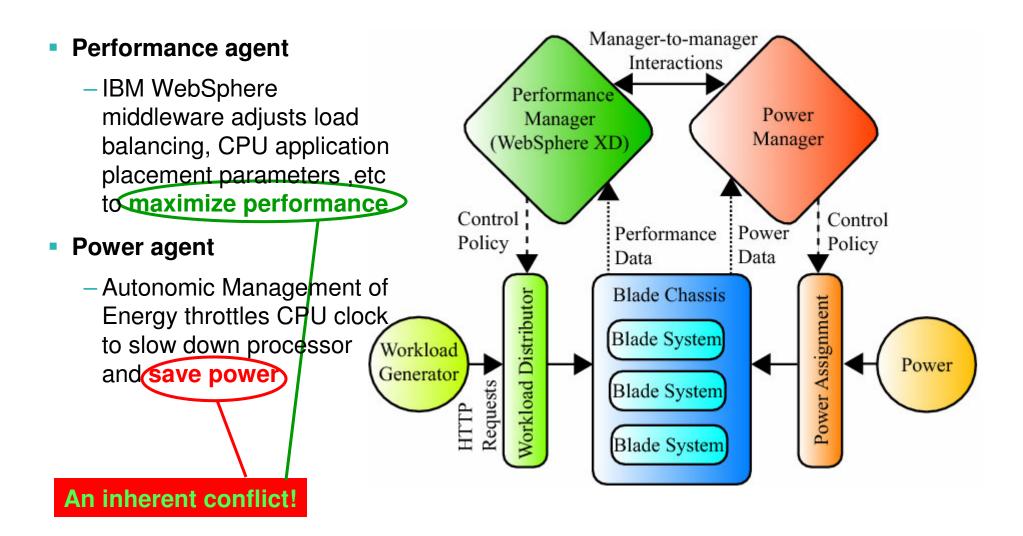
TurnOnCompletely(0)

PerformanceManage(0)

Now all we need to do is combine these algorithms somehow...?!?

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## **Power and Performance Management**

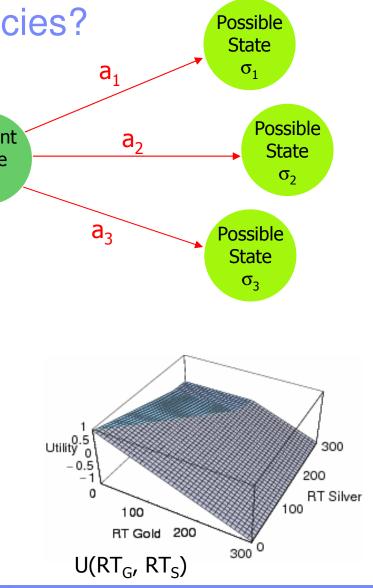




#### Kephart and Walsh, Policy04

#### How to represent high-level policies? $a_1$ Utility functions map any possible state of a system to a scalar value Current $a_2$ State They can be derived from S - a service level agreement **a**<sub>3</sub> preference elicitation techniques - simple templates, e.g. specify response time thresholds and "importance" levels They are a generally useful representation for high-level objectives, e.g. - Minimize power while meeting SLA Utility 0.5 0

- Maximize performance while meeting power constraint
- Range in between



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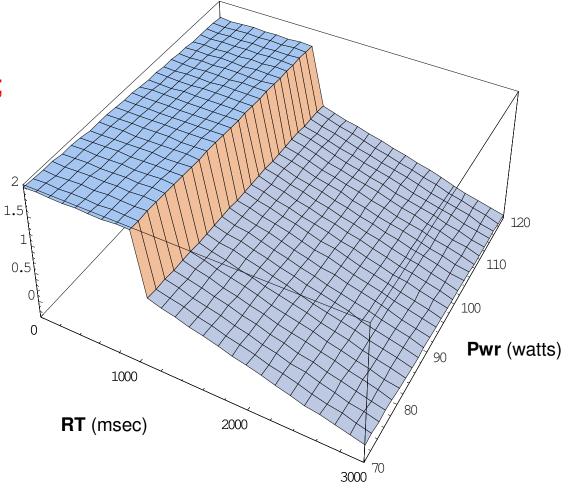
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## **Power-performance utility functions**

- U(perf, pwr) = U(perf) ε Pwr; Pwr < Pwr<sub>Max</sub>
- U(perf, pwr) = U(perf)/Pwr; Pwr < Pwr<sub>Max</sub>

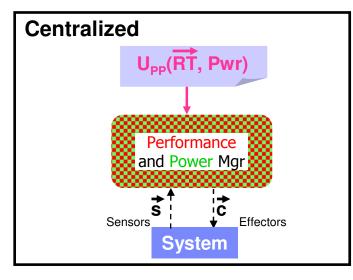
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### Multiagent approach to Power-Performance optimization

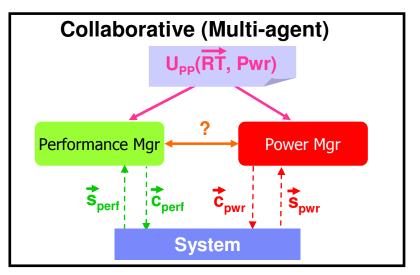
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Set effectors c to maximize  $U_{PP}$ 

Conceptually easiest

Not very practical!



#### What info should be exchanged, and how?

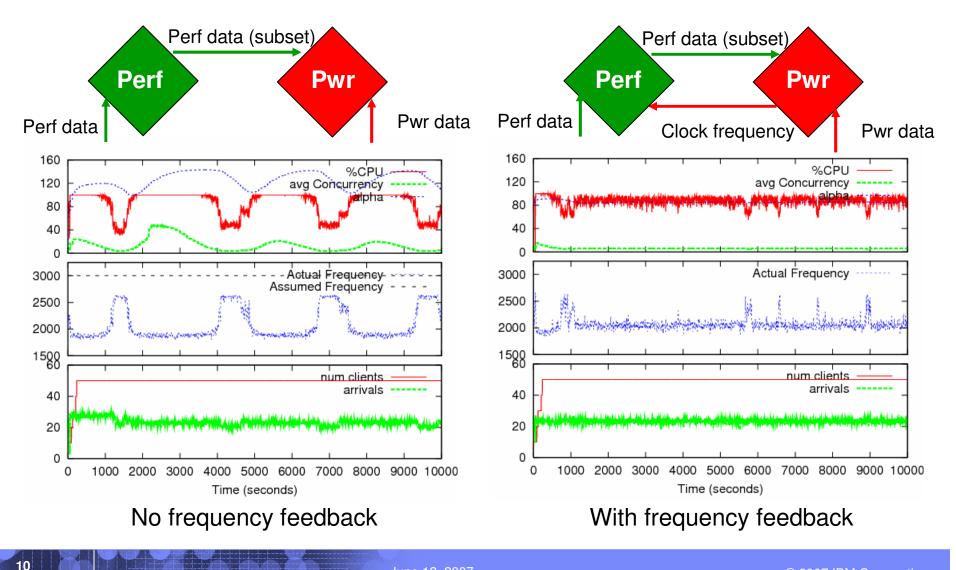
- Do we need negotiation?
- Do we need mediation?
- What are the right power control knobs?

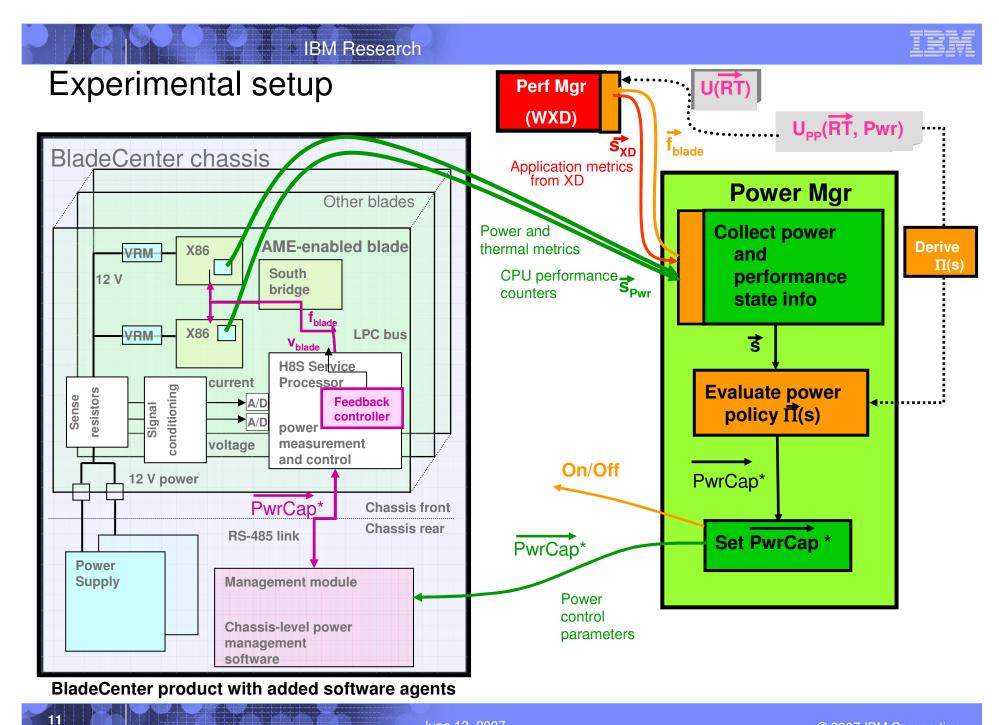
#### Strategy: Try the simplest method first

- No negotiation or mediation
- Power control knobs = power cap settings
- Minimize changes to WXD
- · Add complexity only if/when really needed

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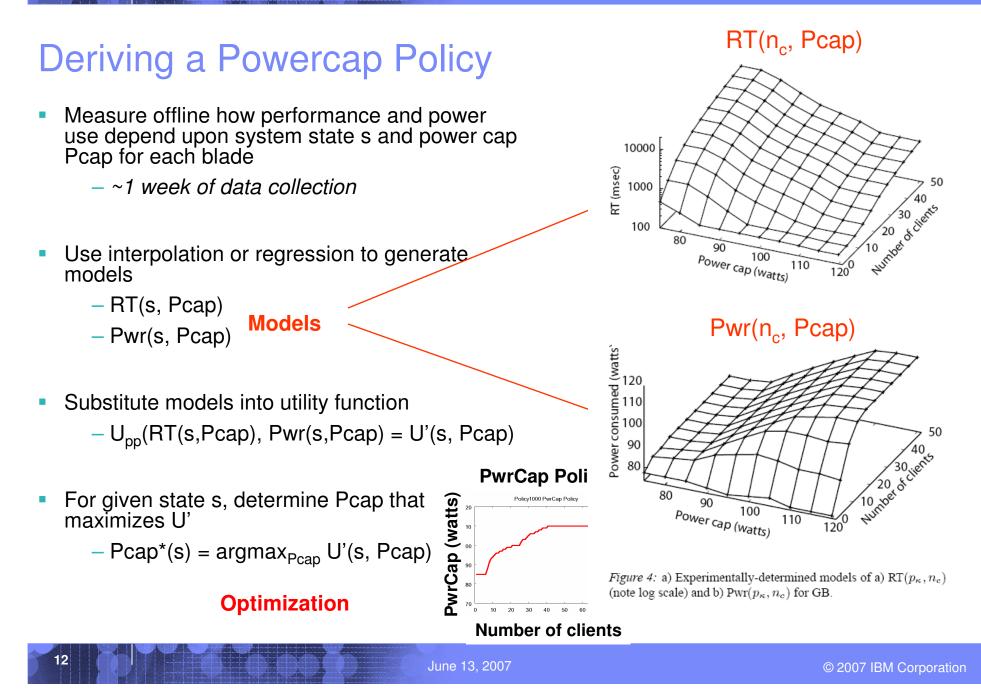
### What needs to be communicated?





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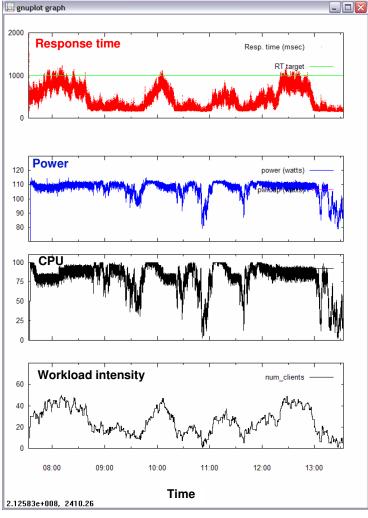


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## Experiment: Hand-crafted Power Policy

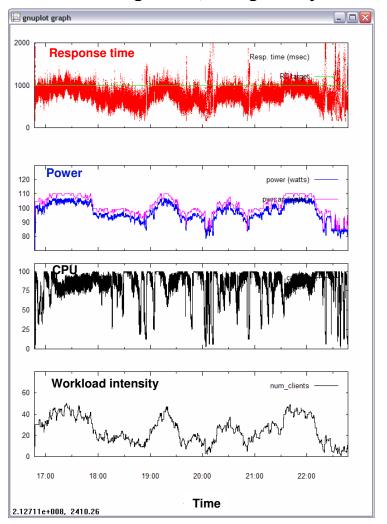
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#### No power management



#### Avg power = 107.9 watts

#### Power management, using Policy HC01



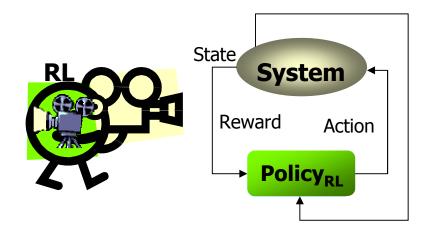
Avg power = 96.6 watts (savings: 11.3 watts = 10.5%)



## Hybrid Reinforcement Learning

#### Tesauro et al., ICAC 2006

- Reinforcement learning methods learn to make good decisions by
  - observing <state, action, reward> tuples
  - learning long-range value functions V(state, action)
  - Abiding by optimal policy: when in state s, take action a that maximizes V(s, a)
- Typically, they learn by updating V(state, action) starting from random assumptions
- This can take a long time, and performance can be very poor during the learning phase
- We invented a new RL technique, Hybrid RL, that starts from an existing policy, and improves upon it
- Very general method that automatically improves *any* existing systems management policy
  - No knowledge engineering needed



State = {power, performance metrics}

Action = {powercaps}

Reward = U(perf, pwr)

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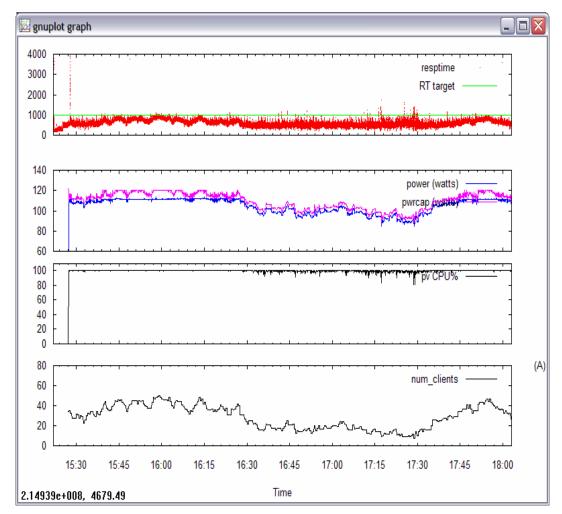
### Hybrid RL improves our initial power policy!

- Good power savings
  - 8.9% less power than for unmanaged case
  - Was 10.5% for handtuned policy

### Reduced SLA violations

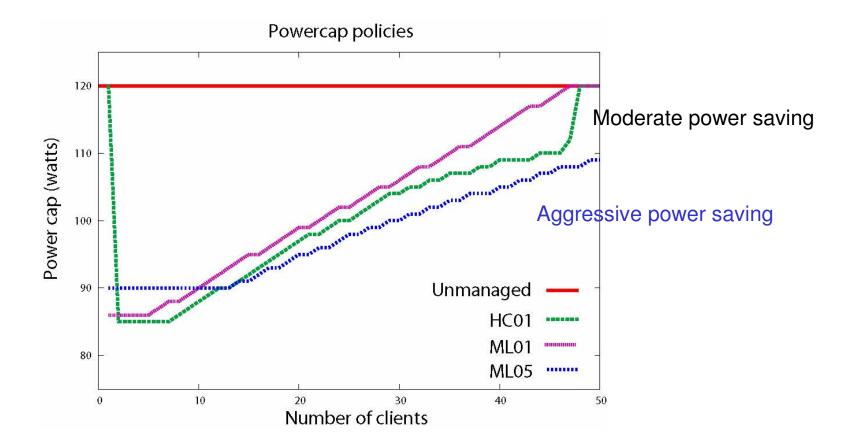
- 1.5% of response times exceed threshold
- Was 21% for hand-tuned policy

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## **Powercap policies**

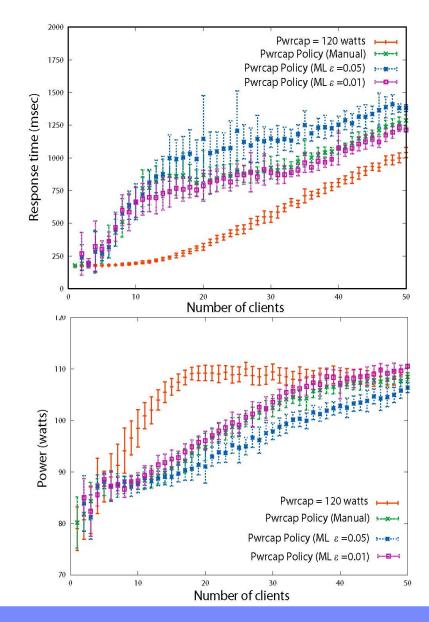


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### Where is the power being saved?

- Power is saved primarily when the number of clients is moderate
  - For low workloads, power consumption is not constrained by powercaps
  - At high workloads, utility is maximized by setting high powercaps

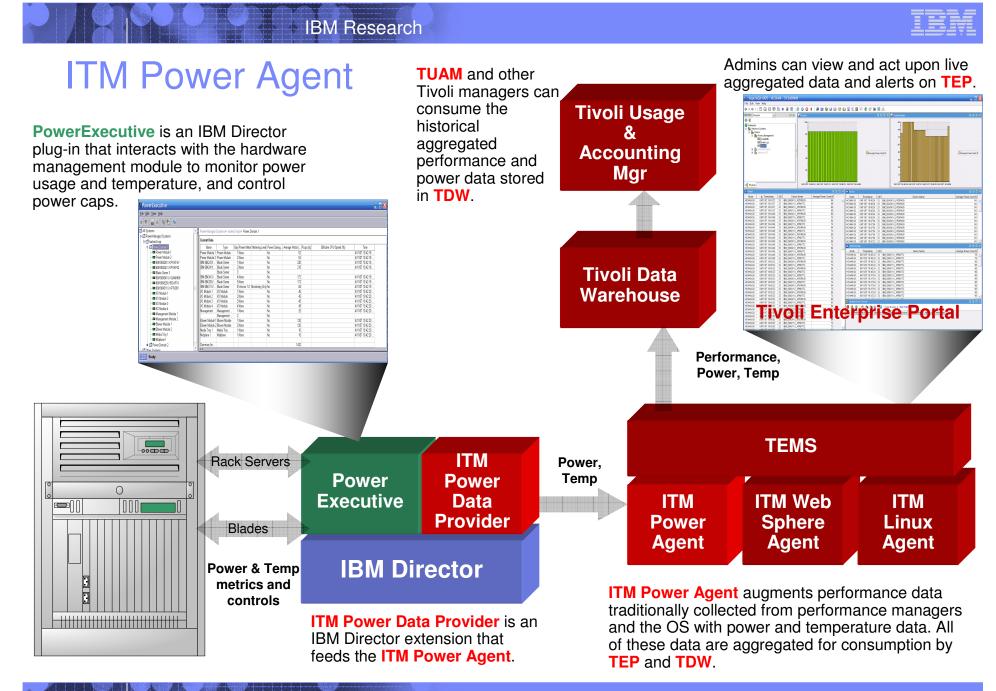
Policy	Avg watts	Savings
Unmanaged	104.9	0%
HC01	95.3	9.2%
ML01	96.1	8.4%
ML05	92.7	11.6%





## Agenda

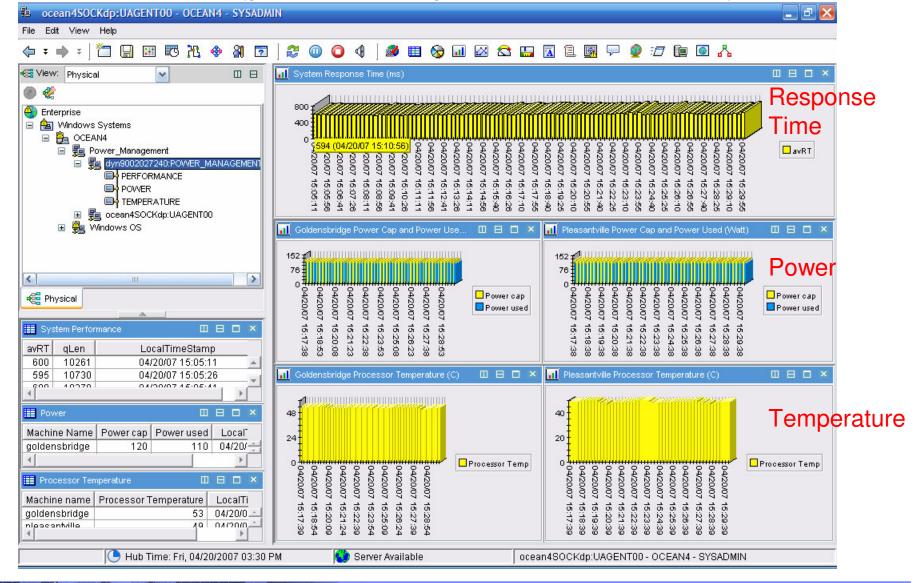
- Introduction
- Power-Performance Research
  - Algorithms
  - Results
- Commercialization



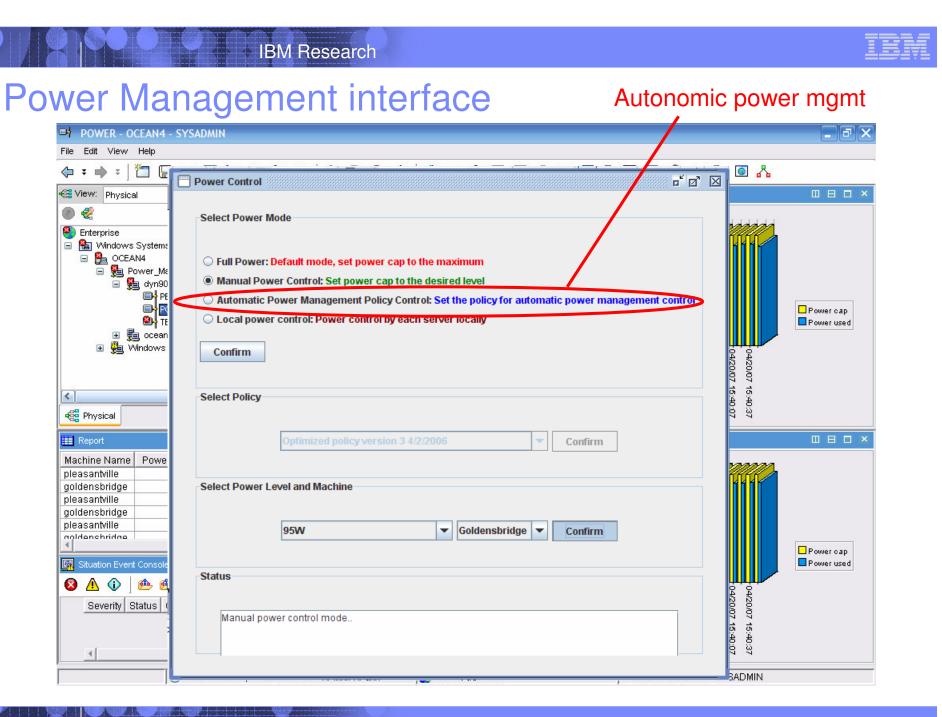
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## Admin console (power and performance data)

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# Summary and Future Directions

### Thus far, we have achieved coordination across

-Multiple levels (from chip to application)

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- -Two management disciplines (performance, power)
- -Approach
  - Express joint objectives in terms of utility functions
  - Combine modeling, optimization and state-of-art ML technologies
    - ML can save time and yield better policy

### We are currently exploring

- -Dynamic voltage and frequency scaling
- Dynamic power-off of servers, exploiting virtualization and load balancing

### > 30% power savings in initial tests

-An additional management discipline: availability

### Opportunities to extend to data center level