Full-System Power Analysis and Modeling for Server Environments

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Motivation

- Costs of power and cooling
 - Electricity now ~50% of data center costs (ComputerWorld, 4/06)
 - Data center cooling consumes ~1W per W consumed by system
- Power density and compaction
- Thermal failures
 - − 10C temperature increase →
 50% reliability decrease
- Environmental issues
 - EnergyStar Enterprise Server and Data Center Efficiency Initiative, 2006



Goals: Prerequisites to Optimizing Power

- Understand server power
 - Across different types of systems
 - Component breakdowns
 - Temporal variation
 - Within and between workloads
- Develop model for server power
 - Fast, online model deployable in a data center scheduler
 - Zero hardware cost to the end user
 - Input: accessible OS metrics; Output: "good enough" (within 5-10%) estimate of power

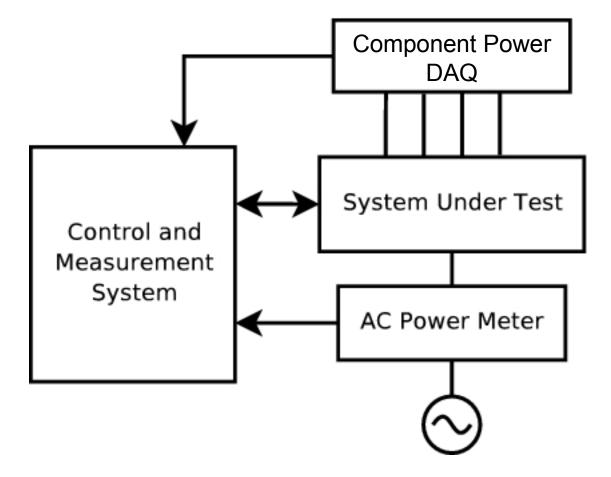
Outline

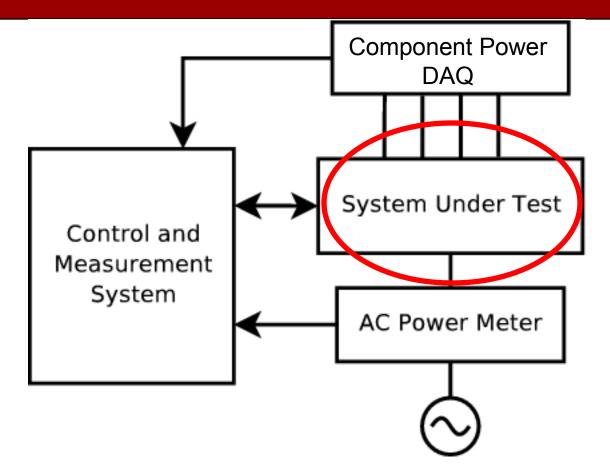
- Motivation
- Experimental setup
- Power characterization
- Power modeling
- Future work
- Conclusions

Test Machines

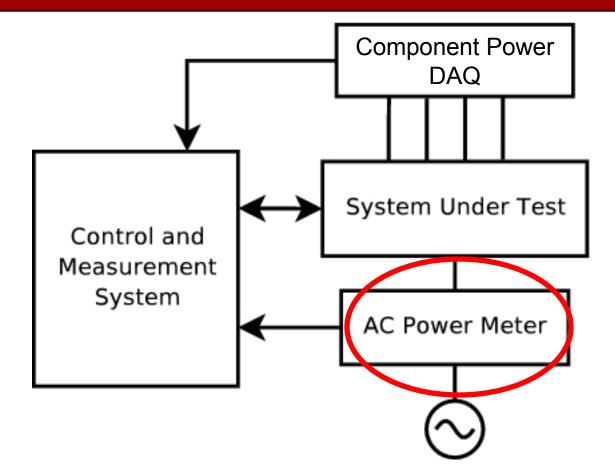
- **Power-optimized** blade server
 - Low-power processor states
- Compute-optimized Itanium server
 - Zero power-saving technology in processors
 - Resources imbalanced in favor of processors

| | Blade Server | Itanium Server | |
|---------|-------------------------|------------------------|--|
| CPU | 1 * AMD Turion, 2.2 GHz | 4 * Itanium 2, 1.5 GHz | |
| Memory | 512 MB SDRAM | 1 GB DDR | |
| Storage | 1 HDD, 40 GB, 2.5" | 1 HDD, 36 GB, 3.5" | |
| Network | 10/100 Ethernet | 10/100 Ethernet | |





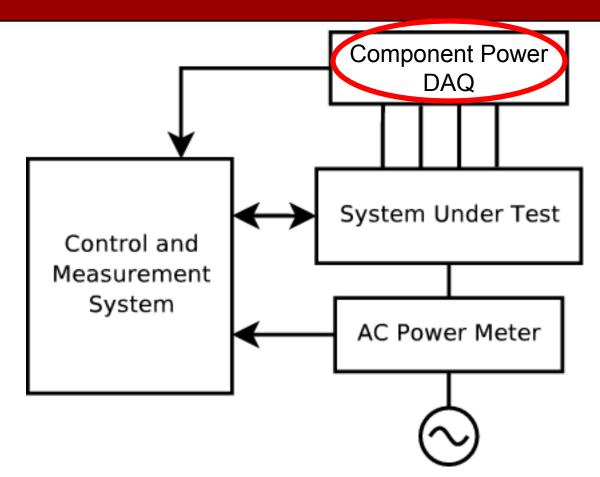
- System Under Test: Blade or Itanium server
- Runs **benchmark** + low-overhead **performance monitors** (e.g. sar, caliper) at 1 sample/sec



Insert measurement between machine and wall to measure overall power

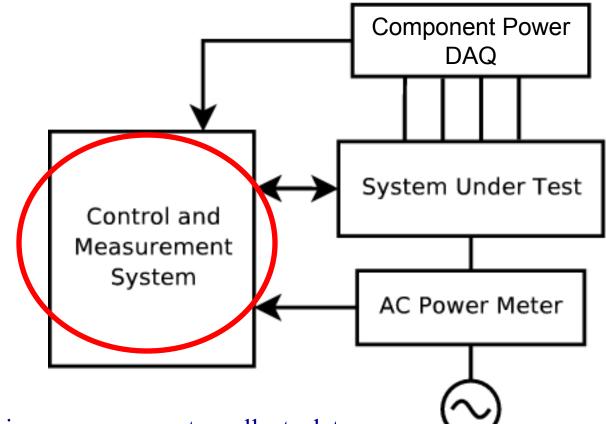
•Blade server: 1 sample/sec

•Itanium server: Currently 20 sample/sec



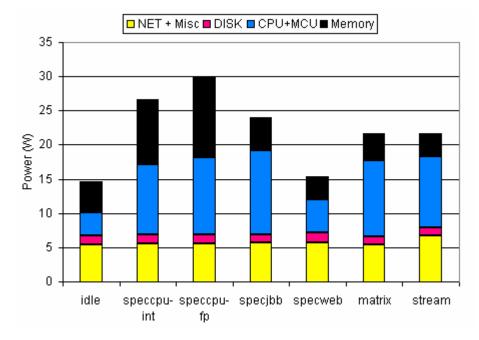
• We cut into and instrumented the individual *power planes* of the servers, to capture component-level DC power (~20 samples/sec)

• This is NOT required for our model

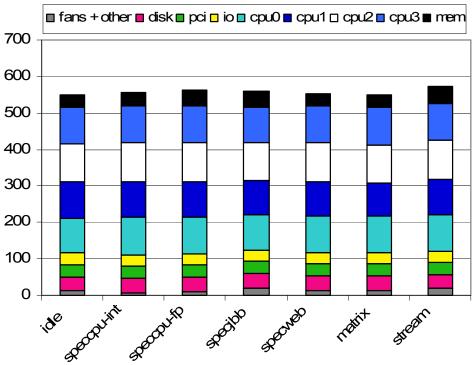


PC: synchronizes measurements, collects data

- Performance metrics from system under test
- Overall power from AC power meter
- Component power from ADC

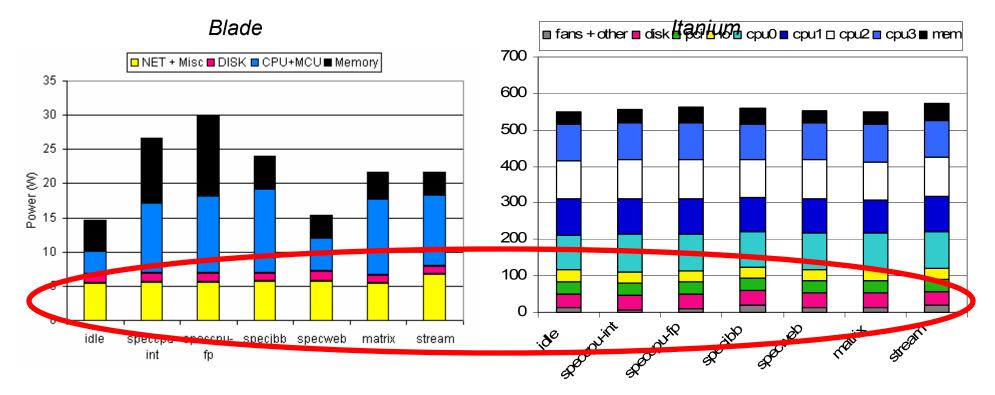


Blade



Itanium

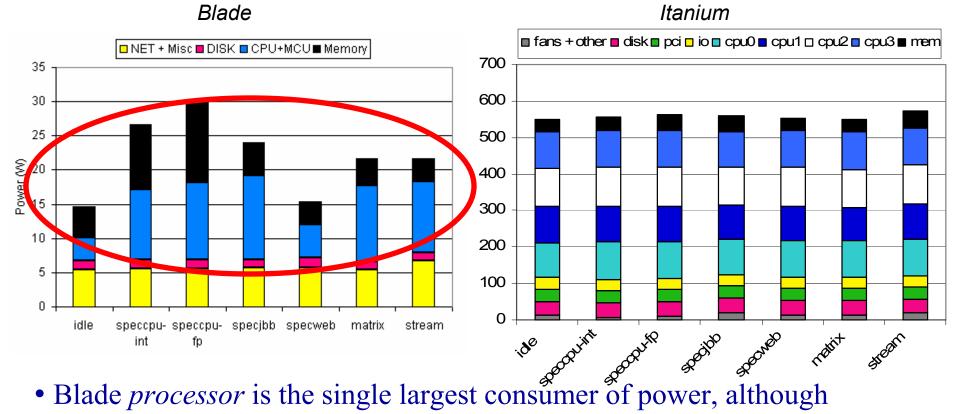
- Average DC power of components
- Benchmarks: *idle, SPECint, SPECfp, SPECjbb, SPECweb, matrix multiply, streams*



• Disk, net, fan, and misc components

•Non-negligible contributors to power

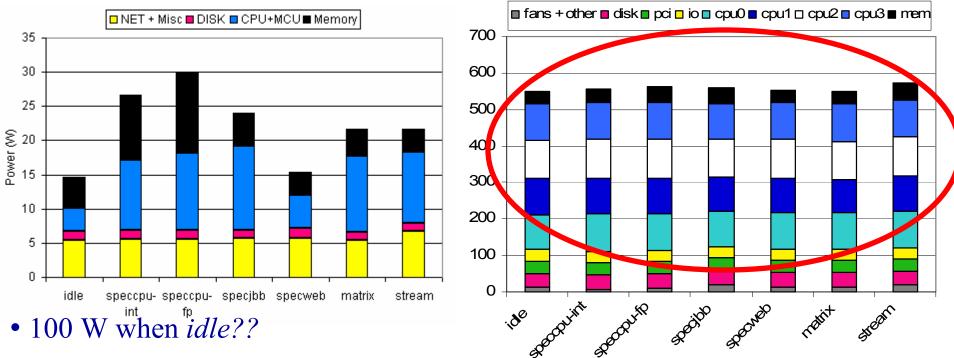
•Small variation in average power consumption (occasional spikes)



memory is close behind

• High variation in processor power consumption shows that blade is optimized for power

Blade



Itanium

•Not much variation (30%) between idle and max power in Itanium

- •So the 4 processors dominate
- High variation in memory, percentage-wise

Power Characterization Conclusions

• Conventional wisdom

- After CPU, memory is the next bottleneck
- Lots of variation in CPU power if chip is optimized for power; otherwise runs near 100% at all times

• More surprising

- The assorted "misc" components the arcane circuits on different power planes – really matter (~20% of blade power). Optimizing these may be worthwhile
- Disk contribution is relatively small
- Enormous idle power on the Itanium system

Power Modeling

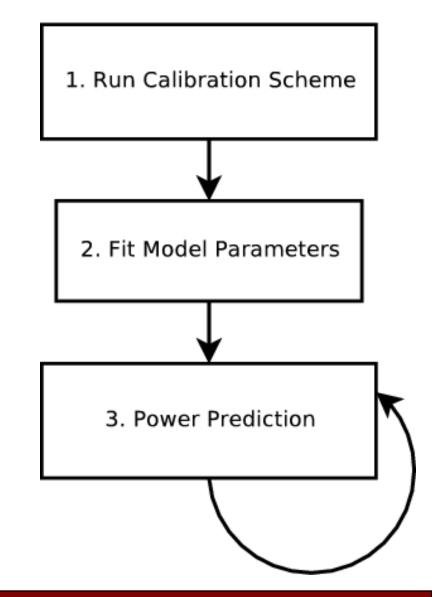
- Goal: Develop an online model for use in data center schedulers
- Model requirements
 - Full-system
 - Non-intrusive; easy for end user
 - Fast enough for online use
 - Reasonably accurate (within 5-10%)
 - Inexpensive
 - Generic (applicable to different types of systems)

Power Modeling: Past Approaches

- Simulation-based detailed models
 - Inexpensive, arbitrarily accurate
 - Not full-system
 - Tailored specifically to particular systems & components
- Direct hardware measurements
 - Accurate, fast, easy
 - Expensive (especially over many machines)
- The Mantis Question
 - Can high-level combined metrics give a good approximation?

Power Modeling

- Run <u>one-time</u> calibration scheme (possibly at vendor)
 - *Inputs*: performance metrics, AC power measurements
 - Workloads that stress individual components: CPU, memory, disk, network
- Fit model parameters to calibration data
 - Linear model for simplicity
- Use model to predict power
 - Inputs: performance metrics (as from sar or caliper) at each point in time
 - Output: estimation of AC power at each point in time



Calibration

- Stress each system component in isolation to develop a model
- Used *gamut* program (J. Moore, 2005) to stress CPU, memory, disk, network at varying degrees of utilization
 - Could use any program that can selectively stress components
 - *Gamut* can't always stress each component to the absolute maximum
 - Runs as a user program on top of the OS, so incomplete control of the hardware
 - *Getting CPU power to the absolute max. may require architectural knowledge*
 - Overheads (program and OS) prevent it from maxing out subsystems

Model Creation

- GOAL: Predict instantaneous power within 10% using a simple, fast model
 - Inputs: OS-level utilization metrics + AC power for calibration suite
 - Output: An equation which relates power to these metrics
- INPUT: Utilization metrics
 - u_{cpu} = CPU utilization (%)
 - u_{mem} = Off-chip memory access count
 - u_{disk} = Hard disk I/O rate
 - u_{net} = Network I/O rate
- OUTPUT: For linear model, an equation of form

$$p_{pred,i} = A + B * u_{cpu,i} + C * u_{mem,i} + D * u_{disk,i} + E * u_{net,i}$$

Model Inputs

• Input is a matrix *M*, e.g.:

• And a vector p_{meas} , e.g.:

. . .

$$p_{meas}, t=0$$

 $p_{meas}, t=1$
 $p_{meas}, t=2$

. . .

Model Creation

• *LP solution*: a vector of weights for each utilization metric

$$\vec{p}_{pred} = M\vec{s}$$

• Errors

$$\varepsilon_{i} = \frac{p_{pred,i} - p_{meas,i}}{p_{meas,i}}$$

• *Objective*: minimize absolute error of models over all calibration programs

$$\min \sum_{n=1}^{N} (t_n^+ - t_n^-)$$

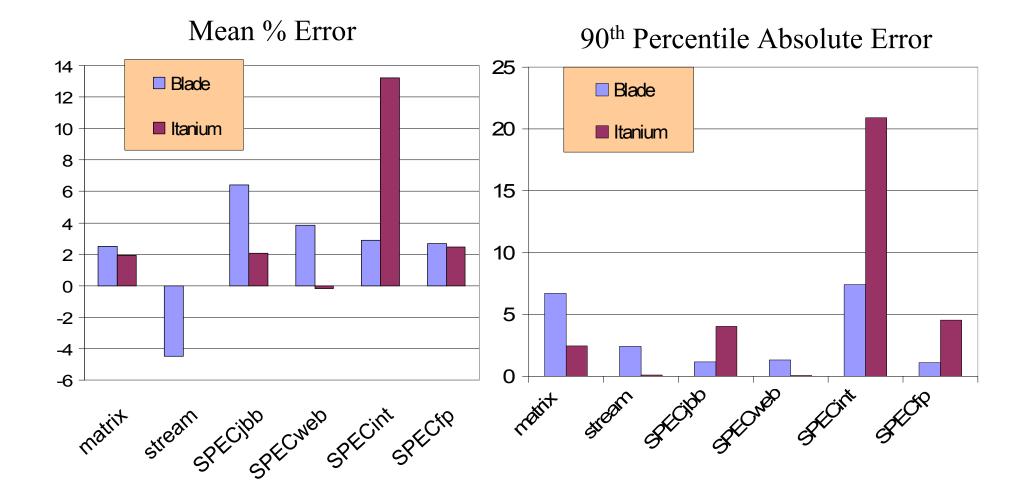
Models Developed

Power prediction equation:

$$p_{pred,i} = A + B * u_{cpu,i} + C * u_{mem,i} + D * u_{disk,i} + E * u_{net,i}$$

| | A (const) | B (cpu) | C (mem) | D (disk) | E (net) |
|---------|-----------|---------|-----------|----------|----------------------|
| Blade | 14.45 | 0.236 | 4.47*10-8 | 0.00281 | 3.1*10 ⁻⁸ |
| Itanium | 635.62 | 0.1108 | 4.05*10-7 | 0.00405 | 0.0 |

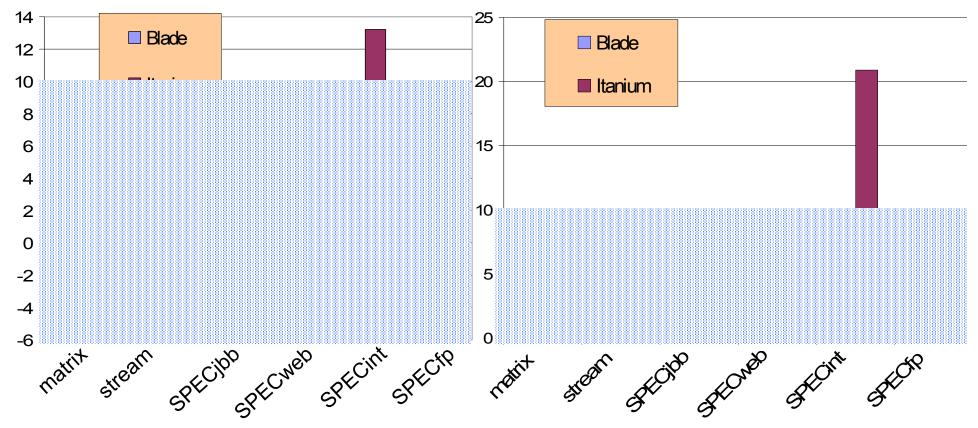
Evaluation



Evaluation

Mean % Error

90th Percentile Absolute Error



Generic model works (within 10%) on 2 very different systems over a varied set of benchmarks

Applications and Future Work

- Improving models
 - Component-level modeling and validation
 - Exploring nonlinear models
 - Adding/replacing CPU utilization % with a generic measurement of ILP
- Data center resource provisioning
 - Estimate power costs at different granularities (server, enclosure, rack...)
 - Power-aware scheduling and mapping
- Data center thermal optimizations
 - Replace expensive external thermal sensors with Mantis estimates
 - Generate data center thermal map
- Fan control
 - Dynamically set fan speed in response to estimated power
 - With component-level models, turn on fans aimed at high-power components

Conclusions

- Goals:
 - Understand server power consumption
 - Develop power model that can be used online in data centers
- Understanding server power
 - Quantitative component/temporal power breakdown
 - Confirming conventional wisdom: CPU is biggest consumer, memory is next
 - Need cooperation of software for low power
 - "Misc" component is worth paying attention to
- Developing a power model
 - High-level metrics give a reasonable approximation of power
- Future work
 - Improve model (ILP metrics, non-linear models...)
 - Use model in a data center scheduler