Power and Energy aware job scheduling techniques

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02-07-2015



Your business technologists. Powering progress

Top500 HPC supercomputers

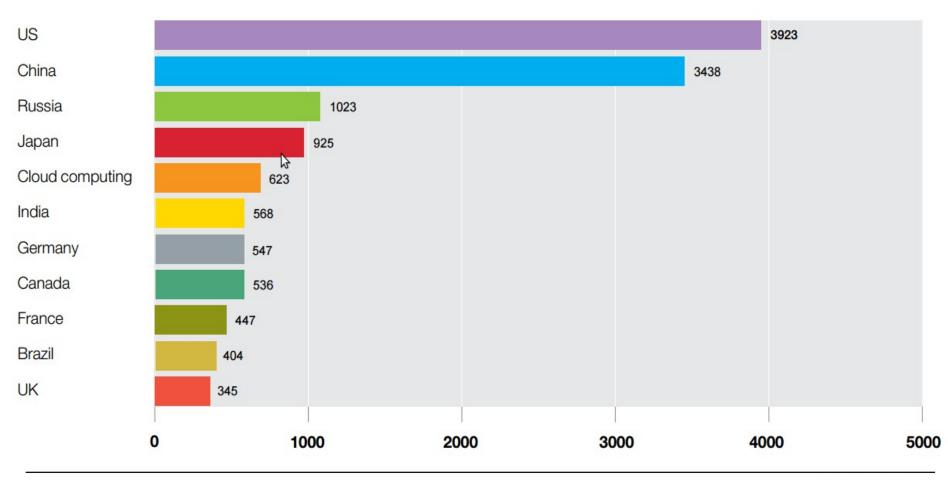
Rank	Site	System	Cores	Rmax (Tflop/s)	Rpeak (Tflop/s)	Power (kW)	
1	NUDT, China	Tianhe-2	3,120,000	33,862.7	54,902.4		17,808
2	ORNL, USA	Titan	560,640	17,590.0	27,112.5		8,209
3	LLNL, USA	Sequoia	1,572,864	₽17,173.2	20,132.7		7,890
4	Riken, Japan	K computer	705,024	10,510.0	11,280.4		12,660
5	ANL, USA	Mira	786,432	8,586.6	10,066.3		3,945
6	CSCS, Switzerland	Piz Daint	115,984	6,271.0	7,788.9		2,325
7	Texas, USA	Stampede	462,462	5,168.1	8,520.1		4,510
8	Juelich, Germany	JUQUEEN	458,752	5,008.9	5,872.0		2,301
9	LLNL, USA	Vulcan	393,216	4,293.3	5,033.2		1,972
10	Government,USA	Cray XC30	72,800	3,577	6,131.8		1,499



IT Energy Consumption

2007 electricity consumption. Billion kwH

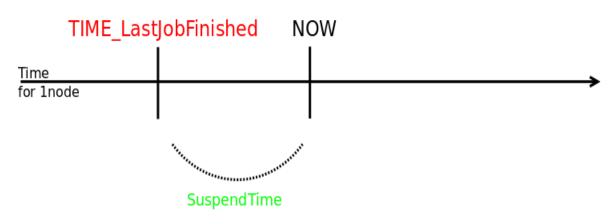
3



http://www.greenpeace.org/international/Global/international/publications/climate/2012/ iCoal/HowCleanisYourCloud.pdf



- Framework for energy reductions through unutilized nodes
 - Administrator configurable actions (hibernate, DVFS, power off, etc)
 - Automatic 'wake up' when jobs arrive



Algorithm for SLURM Energy Reduction Techniques

Nodes Sleep Actions

if SuspendTime > A_PreDefined_Idle_TIME exec SuspendProgram upon SuspendRate nodes per minute

Nodes WakeUp Actions

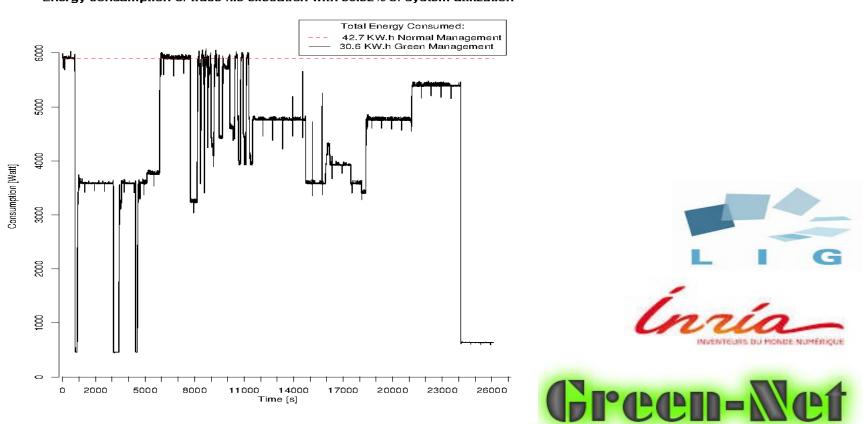
if SleepingNode_isNeeded then exec ResumeProgram upon ResumeRate nodes per minute









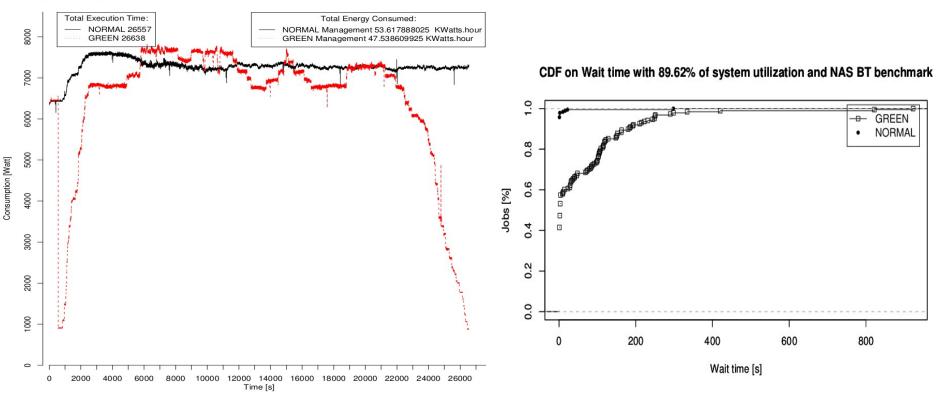


Energy consumption of trace file execution with 50.32% of system utilization

Georges Da Costa, Marcos Dias de Assuncao, Jean-Patrick Gelas, Yiannis Georgiou, Laurent Lefevre, Anne-Cecile Orgerie, Jean-Marc Pierson, Olivier Richard and Amal Sayah Multi-facet approach to reduce energy consumption in clouds and grids: The green-net framework. (In proceedings of e-Energy 2010)

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Energy consumption of trace file execution with 89.62% of system utilization and NAS BT benchmark

Bull atos technologies

Yiannis Georgiou

6 Contributions for Resource and Job Management in High Performance Computing (PhD Thesis 2010)

Total Execution Time: Total Energy Consumed: NORMAL 26557 NORMAL Management 53.617888025 KWatts.hour 8000 **GREEN 26638** GREEN Management 47.538609925 KWatts.hour CDE on Wait time with 89.62% of system utilization and NAS BT benchmark 7000 ssues : 6000 Multiple Reboots: Risks for node crashes or other GREEN NORMAL hardware components problems 5000 Consumption [Watt] •Most of production HPC clusters have a nearly 90% or 4000 higher utilization hence the gain can be trivial •TradeOffs: Jobs Waiting times increases significantly 3000 2000 0 o 000 200 400 600 800 Wait time [s] 0 2000 4000 6000 8000 10000 12000 14000 16000 18000 20000 22000 24000 26000 Time [s]

Energy consumption of trace file execution with 89.62% of system utilization and NAS BT benchmark



Yiannis Georgiou

7 Contributions for Resource and Job Management in High Performance Computing (PhD Thesis 2010)

Power and Energy Management

Issues that we wanted to deal with:

- Attribute power and energy data to HPC components
- Calculate the energy consumption of jobs in the system
- Extract power consumption time series of jobs
- **Control** the Power and Energy usage of jobs and workloads



- Power and Energy monitoring per node
- Energy accounting per step/job
- Power profiling per step/job
- CPU Frequency Selection per step/job

How this takes place :

- In-band collection of energy/power data (IPMI / RAPL plugins)
- Out-of-band collection of energy/power data (RRD plugin)
- Power data job profiling (HDF5 time-series files)
- Parameter for CPU frequency selection on submission commands



- Power and Energy monitoring per node
- Energy accounting per step/job
- Power profiling per step/ich

Overhead: In-band Collection
Precision: measurements and internal calculations
Scalability: Out-of band Collection

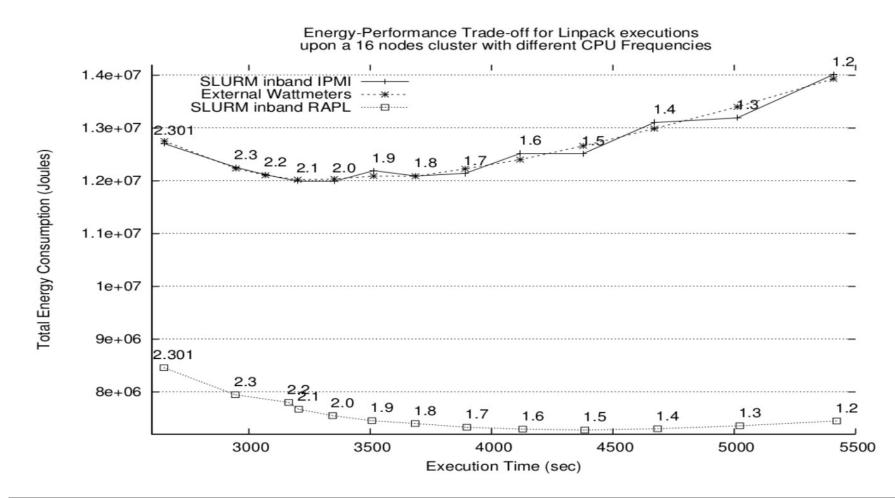
How t

- In-band collection of energy/power data (IPMI / RAPL plugins)
- Out-of-band collection of energy/power data (RRD plugin)
- Power data job profiling (HDF5 time-series files)
- SLURM Internal power-to-energy and energy-to-power calculations



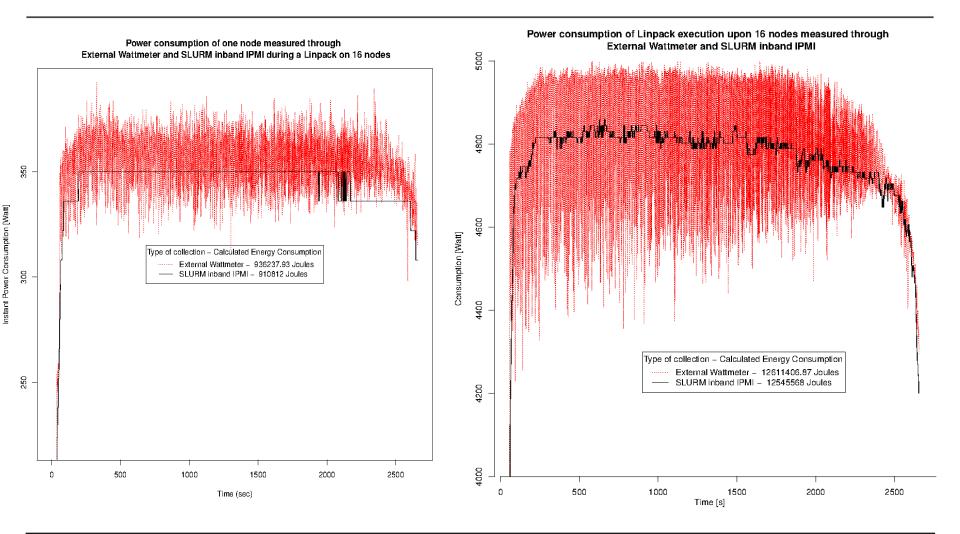
[root@cuzco108 bin]# sacct %3,NodeList%22,State,Start, JobID JobName AllocCPUS Start	End,Elapsed,Consumed					
127 cg.D.32 32 09-12T23:12:51 2013-09-12T2		.11-113] COMPLETED 2013- 490.60KJ				
<pre>[root@cuzco108 bin]# cat extract_127.csv Job, Step, Node, Series, Date_Time, Elapsed_Time, Power 13,0, orion-1, Energy, 2013-07-25 03:39:03,0, 126 13,0, orion-1, Energy, 2013-07-25 03:39:04, 1, 126 13,0, orion-1, Energy, 2013-07-25 03:39:05, 2, 126 13,0, orion-1, Energy, 2013-07-25 03:39:06, 3, 140 13,0, orion-1, Energy, 2013-07-25 03:39:07, 4, 140 13,0, orion-1, Energy, 2013-07-25 03:39:08, 5, 140</pre>						





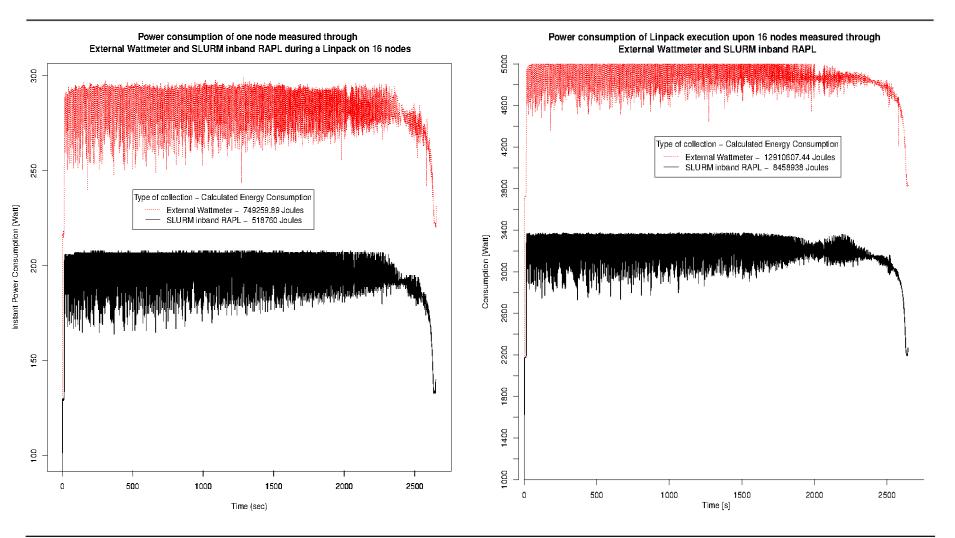
Yiannis Georgiou, Thomas Cadeau, David Glesser, Danny Auble, Morris Jette and Matthieu Hautreux Energy Accounting and Control with SLURM Resource and Job Management System (In proceedings of ICDCN 2014)





Yiannis Georgiou, Thomas Cadeau, David Glesser, Danny Auble, Morris Jette and Matthieu Hautreux Energy Accounting and Control with SLURM Resource and Job Management System (In proceedings of ICDCN 2014)





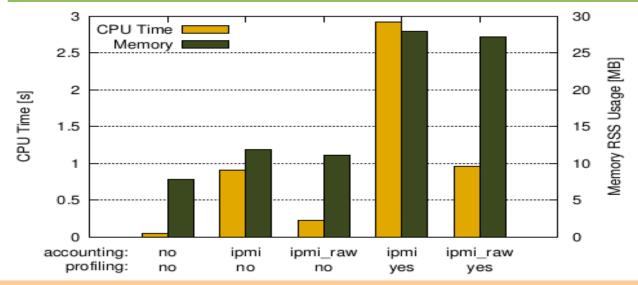
Yiannis Georgiou, Thomas Cadeau, David Glesser, Danny Auble, Morris Jette and Matthieu Hautreux Energy Accounting and Control with SLURM Resource and Job Management System (In proceedings of ICDCN 2014)



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Optimizations of Power and Energy Measurement System

- Based on TUD/BULL BMC firmware optimizations
 - sampling to 4Hz
 - No overhead for accounting



High Definition energy efficiency monitoring based on new FPGA architecture

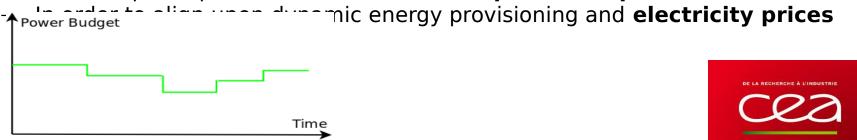
- Sampling to 1000Hz
- Accuracy target to 2 % for energy and power

Daniel Hackenberg, Thomas Ilsche, Joseph Schuchart, Robert Sch["]ne, Wolfgang E. Nagel, Marc Simon, Yiannis Georgiou HDEEM: High Definition Energy Efficiency Monitoring In proceedings E2SC-2014



Power adaptive scheduling

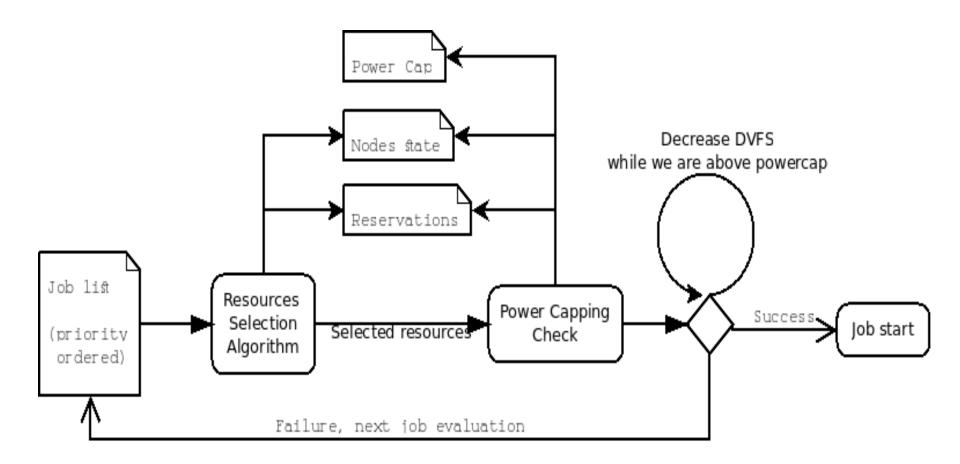
- Provide centralized mechanism to dynamically adapt the instantaneous power consumption of the whole platform
 - Reducing the number of usable resources or running them with lower power
- Provide technique to plan in advance for future power adaptations



- Reductions take place through following techniques coordinated by the scheduler:
 - Letting Idle nodes
 - Powering-off unused nodes
 - Running nodes in lower CPU Frequencies



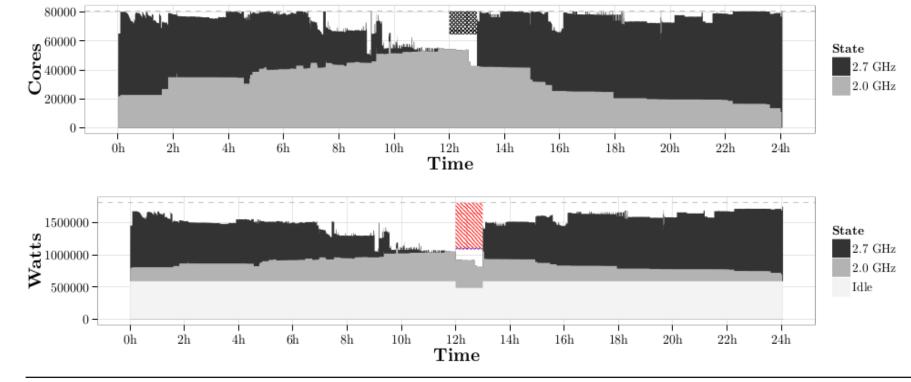
Power adaptive scheduling - algorithm





Power adaptive scheduling

System utilization in terms of cores (top) and power (bottom) for MIX policy during a 24 hours workload of Curie system with a powercap reservation (hatched area) of 1 hour of 40% of total power. Cores switched-off represented by a dark-grey hatched area.

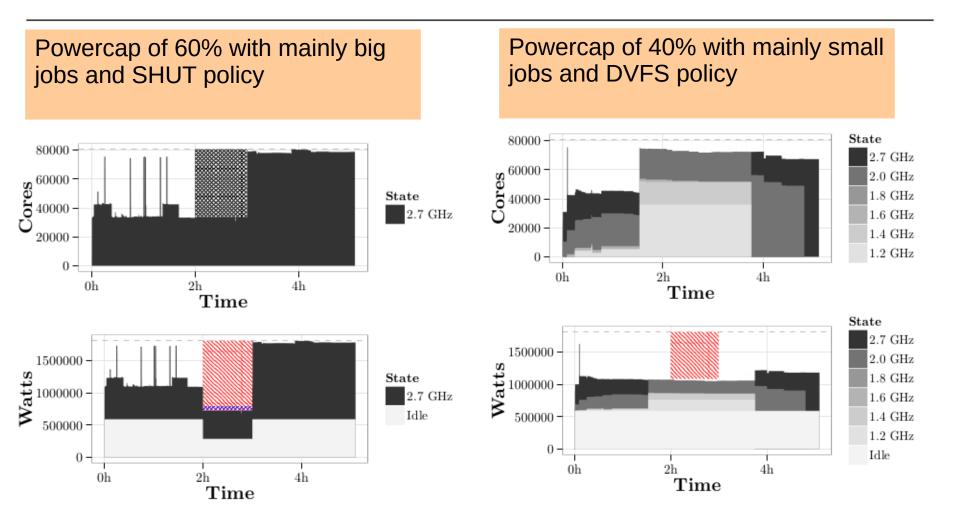


Yiannis Georgiou, David Glesser, Denis Trystram

¹⁸ Adaptive Resource and Job Management for limited power consumption In proceedings of IPDPS-HPPAC 2015



Power adaptive scheduling



 Yiannis Georgiou, David Glesser, Denis Trystram
 Adaptive Resource and Job Management for limited power consumption In proceedings of IPDPS-HPPAC 2015



Energy Fairsharing

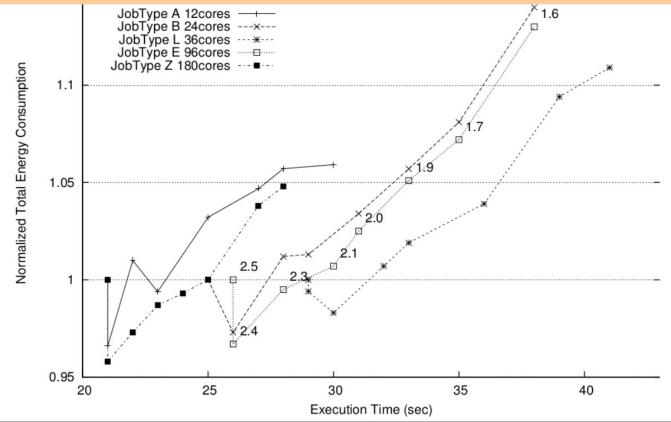
- Fairsharing is a common scheduling prioritization technique
- Exists in most schedulers, based on past CPU-time usage
- Our goal is to do it for past energy usage
- Provide incentives to users to be more energy efficient
 - Based upon the energy accounting mechanisms
 - Accumulate past jobs energy consumption and align that with the shares of each account
 - Implemented as a new multi-factor plugin parameter in SLURM
- Energy efficient users will be favored with lower stretch and waiting times in the scheduling queue





Energy Fairsharing

Performance vs. energy tradeoffs for Linpack applications as calibrated for different sizes and execution times running on an 180-cores cluster at different frequencies.



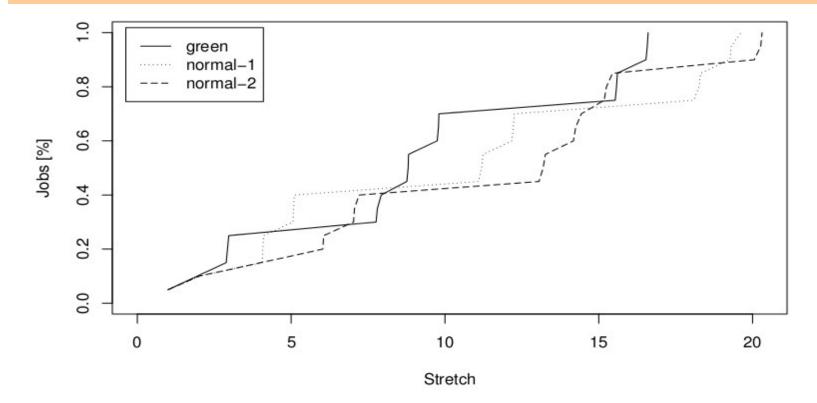
Yiannis Georgiou, David Glesser, Krzysztof Rzadca, Denis Trystram A Scheduler-Level Incentive Mechanism for Energy Eciency in HPC (In proceedings of CCGRID 2015)



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Energy Fairsharing

Cumulated Distribution Function for Stretch with EnergyFairShare policy running a submission burst of 60 similar jobs with Linpack executions by 1 energy-efficient and 2 normal users (ONdemand and 2.3GHz)



 Yiannis Georgiou, David Glesser, Krzysztof Rzadca, Denis Trystram
 A Scheduler-Level Incentive Mechanism for Energy Eciency in HPC (In proceedings of CCGRID 2015)



Ongoing Works - Energy Aware Scheduling

Workload Scheduling

Consider groups of jobs and schedule those that will keep the energy consumption stable

Resources Selection

- Select the **best adapted resources** for lower energy consumption depending on the application profiles (data aware, topology aware, etc)
- Pack jobs in order to leave parts of the cluster unused for powering off
- Select resources based on **temperature** depending of the scope of scheduling



Summary

- Power aware scheduling important for your data center to adapt on the electricity prices and your energy budget
- **Energy fairsharing**: incentive for users to be more energy efficient
- Energy aware scheduling: ongoing works
- Research is published, developments open-source within SLURM
 - CPU Frequency selection parameters since SLURM 2.6 version
 - Energy measurement system plugins since 2.6 version
 - Power aware scheduling to appear in 15.08 version
 - Energy aware scheduling to appear in 16.03 version



Thanks

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12-05-2015

