Effects of Active Cooling on Workload Management in High Performance Processors

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Abstract: This paper presents an energy-efficient workload scheduling methodology for multi-core multi-processor systems under actively cooled environment that improves overall system power performance with minimal response time degradation. Using a highly efficient miniature-scale refrigeration system, we show that active-cooling by refrigeration on a per-server basis not only leads to substantial power-performance improvement, but also improves the overall system performance without increasing the overall system power including the cost of cooling. Based on the measured results, we present a model that captures different relations and parameters of multi-core processor and the refrigeration system. This model is extended to illustrate the potential of power optimization of multi-core multi-processor systems and to investigate different methodologies of workload scheduling under the actively cooled environment to maximize power efficiency while minimizing response time. We propose an energy-efficient workload scheduling methodology that results in total consumption comparable to the spatial subsetting scheme but with faster response time under the actively cooled environment. The actively cooled system results in ≥29% of power reduction over the non-refrigerated design across the entire range of utilization levels. The proposed methodology is further combined with the G/G/m-model to investigate the trade-off between the total power and target SLA requirements.

1 INTRODUCTION

Increase in energy consumption due to the tremendous growth in the number and size of data centers presents a whole new set of challenges in maintaining energy-efficient infrastructure. While data centers’ energy consumption had accounted for 2% of the total energy budget of the USA in 2007, it is expected to reach 4% by the year 2011. This number is equivalent to $7.4 billion per year on electric power where this number has changed by 60% since 2006 (U.S. EPA, 2007).

Worldwide trend of energy consumption in data centers tracks the US trend (Rajamani, 2008). Fig. 1 shows the number of data center installations, worldwide new server spending, and electric power and cooling costs. Despite the steady increase of installed base of data centers over the last decade, new server spending has stayed relatively constant due to the decrease in electronic costs. As the data center infrastructure becomes denser, power density has been increasing by approximately 15% annually (Humphreys, 2006), hence increasing electricity consumption for operating servers and cooling. It is likely that IT operating cost will soon outweigh the initial capital investment. Detailed energy breakdown of different types of data centers can be found in (Rajamani, 2008), (Tschudi, 2003), (Lawrence Berkeley National Labs, 2007), (Patterson, 2008). A data center can consist of hundreds or even thousands of server racks where each rack can draw more than 20kW of power. Relative percentage of various contributors to energy usage varies considerably among data centers, but up to 90% of the total energy is attributed to the energy dissipated by the computer load and the energy required by the Computer Room Air Conditioning unit (CRAC) Additionally, there is a strong relation between the energy consumed by the computer load and the CRAC units since any reduction in electronic heat can be compounded in the cooling system. For example, CRAC energy efficiency of data centers can increase by 1% per degree Celsius.
Reducing the overall energy usage is an area of interest across multiple disciplines. The focus of most efforts on energy/power saving in server systems is on processor elements (Jing, 2011), (Chaparro, 2007), (Ma, 2003), (Tschanz, 2003), (Brooks, 2000), (Sato, 2007), (Ghosh, 2011), (Rabaey, 2003). Approaches for power saving of processors often adopt both the software-based energy-aware workload scheduling (Jing, 2011), (Lin, 2011), (Luo, 2013) and hardware-based circuit and architectural power management techniques to effectively optimize energy usage. A typical software-based workload scheduling algorithm controls energy by distributing workloads to processors in a way to reduce both the electric and cooling costs. The basis of these approaches relies on powering-off servers that are not utilized by concentrating the workload on a subset of the servers. This method is known as spatial subsetting, and has been shown to successfully tackle the issue of idle server power consumption (Jing, 2011), (Pinheiro, 2001), (Chase, 2001).

Moreover, energy savings from the off-power servers is compounded in the cooling systems that consume power to remove the heat dissipated in the servers. While this approach significantly reduces idle power, it raises a concern of degraded response time in computing systems, due to the power-latency trade-off.

To address the problem of degraded response time in spatial subsetting, one solution is to employ an over-provisioning scheme (Chen, 2005), (Ahmad, 2010). The over-provisioning algorithm can be considered as a power and response time optimization problem. By predicting how many servers are required to service the requested workload, the workload management software assigns a subset of processors to remain at idle state to absorb sudden increases in the load. Determining the number of server to be held at idle state often relies on a good model that successfully plans capacity depending on the upcoming workloads. For instance, G/G/m-models from queueing theory have been used to obtain useful measures like average execution velocity and average wait time to support capacity and workload planning of multi-processor systems in order to satisfy target SLA requirements (Chen, 2005), (Ahmad, 2010), (Müller-Colsternann, 2007).

Often the software utilizes special hardware supports (Jing, 2011), such as dynamic voltage frequency scaling (McGowen, 2006), (Burd, 2000),
(Nowka, 2002), or thread migration (Zhang, 2005), stopping a processor through power gating (Tschanz, 2003), (Zhang, 2005), (Henzler, 2005), or body biasing (Tschanz, 2003). However, these techniques not only induces area penalty but also require some transition time in and out of the low-power state and imply performance degradation.

Along with these techniques, actively cooling the processors using refrigeration has attracted recent interest as a practical option to ease the power problems in high performance computing units (Copeland 2005), (Mahajan, 2006), (Nnanna, 2006), (Chu, 2004), (Trutassanawin, 2006). Operating CMOS circuitry at sub-ambient temperatures for higher performance has been shown over the past few decades (Carson, 1989), (Aller, 2000). While the speed improvement can be traded for lower power dissipation of the electronics, the cost of cooling can limit the overall system power performance. Recent work (Park, 2010), (Park, 2010), has shown that active cooling not only can lead to overall power improvement that includes the cost of cooling power without performance degradation; the results show that the amount of power savings is roughly proportional to the ratio of leakage power to total power due to the exponential sensitivity of leakage power to temperature, irrespective of type of workload. For instance, cooling a processor that dissipates 175.4W of power with 30% electronic leakage power resulted in a total system power consumption of 133W. This performance is 25% better than the non-cooled reference design (Park, 2010). Focus of this paper is to explore the effectiveness of workload scheduling to improve power efficiency of multi-core multi-

processor systems in an actively cooled environment using a highly efficient refrigeration system. Results presented in this paper suggest that there exists a methodology under actively cooled environment that optimizes power efficiency while minimizing response time in and out of the low-power state. Furthermore, we combine our proposed methodology with the G/G/m-models to reduce both total power and response time degradation while meeting target SLA requirements.

2 MULTI-CORE PROCESSOR UNDER THE ACTIVELY-COOLED ENVIRONMENT

A miniature-scale refrigeration system for electronic cooling that is capable of operating at a reduced temperature with high efficiency has been developed and experimentally tested in (Park, 2010). The compressor used in our miniature refrigeration system has cooling capacity in the several hundred-watt ranges, indicating that this refrigeration system can potentially be configured to simultaneously cool multi-processor servers. We envision a possible configuration of the HPC server unit as illustrated in Fig. 2.

A layout and photograph of the refrigeration system for electronic cooling is shown in Fig. 3. A configuration of the refrigeration system charged with R-134a refrigerant consists of a compressor, condenser, an expansion valve, a cold plate, evaporator, and a cooling fan. A 12V power supply provided the required power. Additionally, a motor drive board is installed to control the compressor speed and modulate the refrigeration capacity at different loads. K-type bead probes are taped to the evaporator and the condenser for temperature measurements. Power meters are used to measure power consumptions of the cooler and the heat source. By controlling the speed of the compressor, we cool the microprocessor at different heat loads and temperatures in order to obtain minimal total system power. Specific chip junction temperature would be the temperature that resulted in the lowest system power. The detailed description of the experimental setup and performance of our miniature refrigeration system for electronic cooling can be found in (Park, 2010). We characterize the power performance of a 4-core processor at different operating conditions using this refrigeration system. It is important to mention that while our analysis
uses a system that can be enclosed in a server chassis, and vapor compression refrigeration systems can achieve considerably higher efficiency with larger cooling capacity at the expense of larger volume. Such systems can potentially cool entire racks of servers with the coolant distributed with parallel flow through the server blades as shown in Fig. 4. The results discussed in this paper can be directly applied.

The mechanisms for power dissipation of digital CMOS ICs are well understood. The total power dissipation can be estimated by the sum of the active power and leakage power (Rabaey, 2003), (Chandrakasan, 1992).

\[ P_{\text{electric}} = P_{\text{active}} + P_{\text{leakage}} \]  

\[ P_{\text{electric}} = \alpha \cdot C_{\text{switched}} \cdot f_{\text{clk}} \cdot V_{dd}^2 \]  

\[ P_{\text{leakage}} = V_{dd} \cdot I_0 \cdot \exp\left(\frac{-V_{th}}{kT_{\text{junction}}/q}\right) \]

The active power, \( P_{\text{active}} \), depends on the activity factor, \( \alpha \), and the amount of power that dissipates charge/discharge capacitive nodes between the supply voltage (\( V_{dd} \)) and ground when executing the logic. \( C_{\text{switched}} \cdot f_{\text{clk}} \cdot V_{dd}^2 \). At nano-meter scale technology, the switches that implement the logic results in a leakage current to flow through each logic gate even when the logic is not active. This leakage becomes a significant component of total chip power in modern era processors. The leakage power, \( P_{\text{leakage}} \), has an exponential relation with the degree that a transistor’s ON/OFF threshold, \( V_{th} \), exceeds the thermal voltage, \( kT_{\text{junction}}/q \). The \( P_{\text{leakage}} \) equation simplifies the dependence of leakage power by lumping (1) the number and size of logical switching paths in a computational unit, (2) the carrier properties in the transistor, and (3) dependence of leakage on the logical structure of each logic gate of a digital processor into a single constant \( I_0 \).

For a digital processor, power dissipation and computing performance are closely related. The Equation (4) shows this relationship for the delay of a logic gate. The current is a function of temperature and primarily depends on the carrier mobility. A designer can typically trade-off any improved speed performance by reducing the supply voltage, \( V_{dd} \).

\[ \text{Delay} \alpha RC \frac{V_{dd}}{I_{\log \text{ic}}} \]

Lower temperatures lead to improved performance of electronic devices. Lower power and higher speed results from (1) an increase in carrier mobility and saturation velocity, (2) an exponential reduction in sub-threshold currents from a steeper sub-threshold slope (\( KT/q \)), (3) an improved metal conductivity for lower delay, and (4) better threshold voltage control enabling to lower \( V_{th} \).

For coolers and refrigerators, the efficiency is represented in terms of COP defined by

\[ \text{COP} = \frac{P_{\text{electric}}}{P_{\text{cooling}}} \]

where \( P_{\text{cooling}} \) represents the cooling power of the refrigeration system required to lift the total amount of heat (\( P_{\text{electric}} \)) generated by the processor. Furthermore, the cooling power can be expressed in terms of COP and the COP of the Carnot cycle with Eq. (6) and (7) where \( T_{\text{evap}} \) is the cold-end temperature of the evaporator, \( T_{\text{cond}} \) is the temperature at the condenser, and \( \eta \) is the second law of efficiency.
Figure 6: Power consumption across different utilization level before and after cooling using PG and CG as the core stopping techniques for (a) 2-core and (b) 1-core processor.

\[
\frac{P_{\text{electric}}}{P_{\text{cooling}}} = \frac{\eta \cdot \text{COP}_{\text{carnot}}}{\eta \cdot \left( \frac{T_{\text{evap}}}{T_{\text{cond}}} - 1 \right)} \quad (6)
\]

\[
P_{\text{cooling}} = \frac{P_{\text{electric}}}{\eta} \cdot \left( \frac{T_{\text{cond}}}{T_{\text{evap}}} - 1 \right) \quad (7)
\]

Equating Eq. (1) and (7) results in total system power of

\[
P_{\text{total}} = P_{\text{electric}} + P_{\text{cooling}} \quad (8)
\]

that includes the cooling power consumption in order to quantify whether the system offers an overall power reduction at different operating temperature. The model serves as a useful tool to evaluate overall system performance including optimal operating temperatures and the amount of total power reduction.

Using this approach, we explore the optimal operating conditions of the system across different processor utilization. In order to experimentally quantify the power consumption of compute-intensive processors, the workload used in all our experiments is Intel’s LINPACK, workload, which is CPU bound. Note that our model tracks well with measured data (Park, 2010), (Park, 2010). It is also important to emphasize that our experiment not only uses voltage scaling to trade-off the improved speed performance into a power reduction but also controls refrigeration system to modulate the cooling capacity in order to obtain minimal system power. The speed performance of the processor is kept constant across utilization level. The results are used to build a model of the 4-core processor operating at reduced temperatures and applied to multi-core multi-processors in later sections.

The 4-core processor can be configured such that 1, 2 or 4 cores are active while unused cores are completely turned off to address the problem of idle power consumption. The refrigeration system is used to cool the microprocessor at different configurations. The amount of total power before and after cooling and the associated power saving across different process utilization levels for different number of cores is shown in Fig. 5. Here, total power before cooling represents forced air cooling that includes the fan power. As can be seen, the result shows that the total power savings of at least 3, 7 and 13 percent can be obtained across the entire range of processor utilization for 1, 2, and 4 cores respectively. The detailed temperature and voltage operating points and the breakdown of the total system power in terms of active, leakage, and cooling components at different utilization with and without active cooling components are shown in (Park, 2010). Effectiveness of cooling is proportional to utilization level. This result suggests that the energy-conscious provisioning would need to concentrate the workload on a minimal active set of cores that run near a maximum utilization level, while other excess cores transition to low-power states to reduce the energy cost. However, using power gating (PG) technique to power on/off cores comes at a price of response time degradation since powering up a core that is completely shut down requires up to 1000 cycles (Kumar, 2003).

Figure 7: Generic workload scheduling management for multi-core multi-processor computing system.
On the other hand, a simpler way to stop a core with minimal response time degradation is to clock gate (CG) the core (Tschanz, 2003), (Kurd, 2001). Main advantage of this power saving technique is the state of the processor can be preserved since supply voltage is not cut. This provides a response time which is orders of magnitude faster than waking from power gating or powering up a processor. However, in terms of power consumption, this technique stops dynamic power dissipation, but since power is not entirely cut-off, the core continues dissipating leakage power. Operating CMOS circuitry at reduced temperatures substantially reduces the power since leakage power depends exponentially on temperature. The result that captures the impact of CG at reduced temperatures is shown in Fig. 6.

Before cooling, the CG processor consumes considerably higher power as compared to the PG processor, due to the increase in leakage power. As expected, lowering the temperature of the CG processor exponentially reduces leakage power and results in total power that is comparable to PG processors. At 100% utilization level, power savings from cooling with CG and PG are 36% and 20%, respectively, for a 2-core processor. Results are more significant for a 1-core processor where power savings from cooling with CG and PG are 40% and 12%, respectively. For both cases, CG appears to be a better core stopping technique under the actively cooled environment. In this way, response time significantly improves at the expense of negligible (~2.5W) power penalty.

The model that captures different relations and parameters of our 4-core processor and the refrigeration system is extended to illustrate the potential of power optimization of multi-core multi-processor systems and investigate different methodologies of workload scheduling under the actively cooled environment.

### 3 WORKLOAD SCHEDULING METHODOLOGY

With our model derived in Section 2, energy-aware workload scheduling algorithms assign incoming workload to available processors such that power consumption is minimized as constrained by response time requirements.

The server platform we analyze consists of a 4-processor server system with 4-cores per processor under the actively cooled environment. In particular, we are interested in aspects where the effects of electronic cooling change the conventional way of assigning workloads. Detailed results and discussions are presented in this section.

Fig. 7 provides generic management architecture for multi-core multi-processor computing systems where 5 out of 16 cores are utilized. This particular HPC server unit has total of 100% utilization level where each core is responsible for 6.25%. The methodologies we evaluate are the following:

- **Spatial Subsetting (S.S.)**: We assume that unused cores power off by PG. The next core can turn up upon arrival of the workload when the current core is fully occupied.
- **1 Core Over-provision with PG (1 O.P. w/ PG)**: Similar to spatial subsetting but one core remains at idle state to absorb sudden peaks in loading.
- **1 Core Over-provision with CG (1 O.P. w/ CG)**: Similar to 1 Core over-provision with PG but uses CG for the core stopping mechanism.
- **Processor based Over-provision (Processor based O.P.)**: Neither PG nor CG is employed and unused cores remain at idle state.

For all cases, the next processor powers up after all four cores within the active processor are fully utilized to prevent idle power consumption.

For comparison purposes, we show the amount of total power consumption before and after cooling for different types of methodologies across varying
utilization levels in Fig. 8 (a) and (b). Note that the total power after cooling includes the cost of cooling. The impact of cooling on different schemes can be seen through the associated power savings as illustrated in Fig. 9. Several observations can be made based on the results. First, spatial subsetting clearly consumes the least amount of power, but the advantage diminishes under the cooled environment. Second, the processor based over-provision scheme dissipates the largest amount of power but has no response time degradation. Third, the 1 core over-provision with CG scheme achieves an excellent compromise that provides the largest amount of power reduction from cooling. Finally, since the next processor powers up after all four cores within the active processor are fully utilized, three power-up transition delays are unavoidable for all cases. They occur from 25% to 31.25%, 50% to 56.25%, and 75% to 81.25%.

Figure 9: Associated power savings at different utilization level from electronic cooling for different workload assignment methodologies.

Next, we show a new way of assigning workloads under refrigerated cooling and the approach is described in Fig. 10. We demonstrate that the proposed way reduces both the power consumption and the response time requirements at reduced temperature, resulting in power comparable to spatial subsetting but provides a similar response time as 1 core over-provision with CG. Example of the approach is shown in Fig. 10.a; given a workload that requires 4 cores at 100% utilization, the workload scheduling is such that 4 cores are assigned equally to 2 processors. Total power consumption of 196W and 127W is measured, before and after cooling, resulting in a 35% power reduction. On the other hand, the system employing (b) the spatial subsetting scheme and (c) 1 core over-provision with CG scheme consumes 179W and 127W and 199W and 140W before and after cooling, respectively. The amount of total power saving of the proposed approach is considerably higher compared to (b) and (c), which has 29% and 30% of power savings.

To be complete, we show the proposed workload scheduling methodology for different utilization levels in Fig. 11. Since power-up events are necessary when a new processor is brought online, three power-up transition delays are unpreventable. These events occur when (B) transitions to (C), (D) transitions to (E), and (F) transitions to (G). In between these transitions and at higher utilizations beyond (G), performance does not degrade with increasing utilization besides the response delay of a few cycles due to CG. Fig. 12 plots the power dissipation and the percentage savings before and after cooling for each of the conditions shown in Fig. 11.

Although conclusions in this section are drawn from a given platform, the intent is not to restrict to a particular platform. The absolute amount of power saving number would be different as different type of systems would have different electronic profile. However, we suggest applying the idea to larger systems where the proposed workload scheduling methodology is applied after cooling. Leveraging the benefits of clock gating at reduced temperatures, our methodology reduces both the power consumption and the response time requirements at reduced temperatures, resulting in power comparable to the spatial subsetting scheme but provides a faster response time since our scheme does not power-off processor cores.
4 ASSIGNMENT OF WORKLOAD BASED ON SLA

As an extension to the proposed methodology, we combine it with the G/G/m-model to reduce both the total power consumption and the response time degradation while meeting specific SLA requirements. Results from the queuing theory have been used to obtain measures like average execution velocity and average wait time to support capacity and workload planning of multi-processor systems for different workload variability (z). Using the approximation formulas for a G/G/m-model, we can reach an optimal agreement between high utilization of the processors (energy-conscious provisioning) and the target SLA requirements. For simplicity, consider a scenario where a specific workload requires 8 cores at 98% of utilization level, and assume that this workload can be linearly mapped to 9 and 10 cores at 87% and 78%, respectively as shown in Fig. 13. Fig. 14 shows the execution velocity for each of these 3 workload scenarios.

Execution velocity is the average ratio for the total amount of workload units that are served without any delay. The value ranges from 0 to 100 where the value 100 means that the workload does not encounter any wait delays for the system resources while the value 0 means that all work is delayed. Fig. 14 is derived using the formula given in (Müller-Colstermann, 2007). When setting the SLA for execution velocity of >60%, using 10 processors to a utilization of 78% satisfies the requirement. On the other hand, using 8 processors result in an unacceptable execution velocity of 6.5%. Moreover, it is important to note that by increasing the number of processors, there is no transition delay due to powering up a processor, and the only performance degradation results from the response delay of CG.
Next, we evaluate the normalized average wait time, $E[W]$, for different values of workload variability, $z$, where the normalization is performed with respect to the service time to the length of one unit. Here, workload variation represents the variation of request inter-arrival times and request sizes. We consider $0 \leq E[W] \leq 0.5$ for the good quality of service level. Similarly, notice how the system requires 10 processors at 78% of utilization to meet the average wait time requirements for $z \leq 5$ (see Fig. 15).

Finally, we summarize the results of our proposed methodology by comparing with spatial subsetting. The total amount of power consumption before and after cooling for the two schemes is shown in Fig. 16. As expected, the actively cooled system with the proposed methodology dissipates power that is comparable to the spatial subsetting scheme but enables superior response time for different levels of SLA. Analysis also shows that the overall system power savings of 35, 30, and 29% are obtained when using 8, 9, and 10 cores, respectively. It is worth noting that the amounts of saving decreases as we increase the number of cores as the cores now operate at lower utilization levels. Using a larger number of cores at lower utilization levels inevitably increases the total power consumption, but the system operates with much improved SLA. For instance, using 10 processors instead of 8 increase the total power consumption by 25%, but the system now operates at execution velocity of $>60\%$ and normalized wait time of $\leq 0.5$. 

![Figure 13](image1.png)  
**Figure 13:** Required utilization level across different number of processors for the proposed methodology. 

![Figure 14](image2.png)  
**Figure 14:** Execution velocity vs. number of processors. Number in the figure represents required utilization level.

![Figure 15](image3.png)  
**Figure 15:** Normalized average wait time vs. number of processors as function of workload variability $z=0.5, 1.0, 2.0, 5.0$. 

Figure 16: Normalized average wait time vs. number of processors as function of workload variability $z=0.5, 1.0, 2.0, 5.0$. 

Finally, we summarize the results of our proposed methodology by comparing with spatial subsetting. The total amount of power consumption before and after cooling for the two schemes is shown in Fig. 16. As expected, the actively cooled system with the proposed methodology dissipates power that is comparable to the spatial subsetting scheme but enables superior response time for different levels of SLA. Analysis also shows that the overall system power savings of 35, 30, and 29% are obtained when using 8, 9, and 10 cores, respectively. It is worth noting that the amounts of saving decreases as we increase the number of cores as the cores now operate at lower utilization levels. Using a larger number of cores at lower utilization levels inevitably increases the total power consumption, but the system operates with much improved SLA. For instance, using 10 processors instead of 8 increase the total power consumption by 25%, but the system now operates at execution velocity of $>60\%$ and normalized wait time of $\leq 0.5$. 

![Figure 16](image4.png)
5 CONCLUSIONS

An energy-efficient workload scheduling methodology for HPC servers is presented using a highly efficient miniature scale refrigeration system for electronic cooling. By leveraging the benefits of clock gating at reduced temperatures, our proposed methodology results in total power consumption that is comparable to the spatial subsetting scheme. Moreover, it provides a response time of disabling clock gating which is orders of magnitude faster than waking from power gating or powering up a processor. Our actively cooled system results in $\geq 29\%$ power reduction over the non-refrigerated design across the entire range of utilization levels. Furthermore, combining our proposed methodology with the G/G/m-model, we show the trade-off between power and SLA requirements. Setting the target SLA requirement to execution velocity of $>60\%$ and normalized wait time of $\leq 0.5$, the number of required processors to execute a particular workload inevitably increased, leading to the $25\%$ increase in total power consumption. Nevertheless, this still maintains $29\%$ of power reduction, compared to non-cooled design.

While the results discussed in this paper can be directly applied to large-scale multi-server systems, overall system realization is still a big challenge and some important design issues of building such systems are overall power consumption, reliability, and cost. Furthermore, thorough understanding of the strong coupling between refrigerated server racks and CRAC units ( cascaded cooling system) is needed for future research. Nevertheless, current data centers can consume up to $90\%$ of the total energy from computer load and the energy required by the CRAC units. Decreasing the power dissipated or by the computer load is imperative as any reduction in electronic heat can be compounded in the cooling system.

Finally, it would be interesting to explore different feedback-driven control solutions that provide capability to adapt to diverse environment, workload, and user constraints. This is relegated to future work. A model-based software framework that predicts and senses upcoming workloads and provides real-time information to refrigeration and electronic systems to tune compressor speed, temperature, and supply voltage are worthy of being studied in order to achieve optimal power performance.

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