

Simultaneous Multithreading Applied to Real Time

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Abstract

Existing models used in real-time scheduling are inadequate to take advantage of simultaneous multithreading (SMT), which has been shown to improve performance in many areas of computing, but has seen little application to real-time systems. The SMART task model, which allows for combining SMT and real time by accounting for the variable task execution costs caused by SMT, is introduced, along with methods and conditions for scheduling SMT tasks under global earliest-deadline-first scheduling. The benefits of using SMT are demonstrated through a large-scale schedulability study in which we show that task systems with utilizations 30% larger than what would be schedulable without SMT can be correctly scheduled.

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1 Introduction

Simultaneous multithreading (SMT) is a technology developed in the 1980s and 90s that allows multiple processes to issue instructions to different processor contexts, or threads, on a single physical computing core, creating the illusion of multiple cores for every one core that is actually present. It was designed to increase system utilization, particularly in the presence of memory latency [6, 26]. SMT became widely available in 2002, when it was made available on Intel processors [18]. Early experiments on the Pentium 4 showed that SMT could increase throughput by a factor of more than 1.5 in the best case [1, 2, 25]. The first attempt to utilize SMT in a real-time context was made in 2002 by Jain et al. [15], who showed that, by enabling SMT and making every thread available for real-time work, it is possible to schedule workloads with total utilizations up to 50 percent greater than what would be possible on the same platform without SMT. While Jain et al. gave ample



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experimental evidence that SMT can enable systems with higher utilization to be supported, neither they nor anyone else, to our knowledge, has provided a schedulability test that takes SMT into account.

Unfortunately, SMT's increase in throughput comes at the cost of longer and less predictable execution times, caused by contention for limited hardware resources. Apparently, the real-time systems community decided that this uncertainty makes SMT inappropriate for real-time work. We question the validity of this assessment for soft real-time (SRT) systems that may tolerate some tardiness. Evidence suggest that even others begin to question this assessment in the context of safety-critical domains. In particular, the U.S. Federal Aviation Administration has received requests to certify safety-critical applications that use SMT, though they currently lack adequate techniques for doing so [21]. (We defer considerations of safety-critical applications to future work.)

As evidence of the potential benefits of SMT, we present a sample of our results in Fig. 1; a platform with 16 cores is capable of scheduling task systems with total utilizations exceeding 20. We discuss this graph and others in Sec. 5.

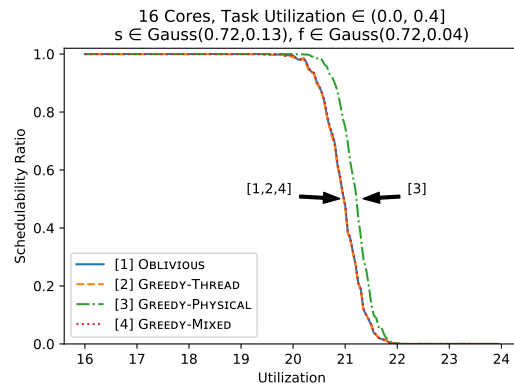
Considered problem. We consider the problem of defining a scheduler for SRT systems that reaps the benefits of SMT without sacrificing execution-cost predictability. Existing models for analyzing real-time workloads do not allow us to specify how enabling SMT affects a task, so to quantify the per-task effects of SMT, we introduce a new task model, SMART (**S**imultaneous **M**ultithreading **A**ppplied to **R**eal **T**ime). Using the SMART model, we attack our problem by dividing it into three sub-problems:

- **Sub-Problem 1:** Determine execution costs for tasks with SMT enabled. "Costs" is plural for each task; one worst-case execution cost is not enough to define a task.
- **Sub-Problem 2:** Decide which tasks should use SMT. How using SMT will affect any given task is a function of what other tasks are using SMT.
- **Sub-Problem 3:** Schedule so tasks using SMT do not interfere with tasks not using SMT.

The second sub-problem is particularly interesting. In general, allowing a task to execute with SMT will decrease the demand the task places on the hardware platform but increase the time needed for the task to execute. To address our problem, we need to balance the advantages of decreasing platform demand with the disadvantages of increasing task execution time. It is not enough to evaluate a task in isolation; every task that uses SMT may influence every other task that uses SMT.

Motivation. Processors are expensive. For any workload, real time or not, it is desirable to minimize the hardware cost needed to obtain a given level of performance. SMT is a

■ **Figure 1** Schedulability on 16 cores with SMT. Note that the horizontal axis begins at utilization 16 and that schedulability does not begin to drop until utilization 20. Effectively, more than 20 cores worth of capacity can be had on a 16-core platform. We discuss this graph and others like it in detail later.



means to get the most work out of a given processor. Presently, SMT is widely implemented, meaning there is a high chance that users are paying for SMT even if they are not using it. A better understanding of SMT would allow for better use of existing hardware resources.

Related works. Snively and Tullsen demonstrated that SMT performance is dependent on which tasks share a core and introduced the term "symbiosis" to describe this concept [24]. We have already mentioned Jain et al.'s work on SMT and real-time scheduling from 2002 [15]. Since then, Cazorla et al. [3], Gomes et al. [12, 13], and Zimmer et al. [28] have proposed ways to eliminate the timing uncertainties associated with SMT by means of detailed control over program execution and, in the case of Zimmer et al., a purpose-built processor, FlexPRET. Cazorla et al. [3] and Lo et al. [17] gave methods to limit real-time work to a small number of threads, leaving the remaining threads to execute only when doing so will not interfere with real-time work. Mische et al. [20] proposed to use SMT to hide context-switch times by using threads to switch task state in and out in the background. Early work on the performance of tasks executed by hardware threads was done by Bulpin [1], Bulpin and Pratt [2], Huang et al. [14], and Tuck and Tulsen [25]. Detailed analysis of Intel's microarchitecture, including the resource constraints that are relevant with SMT, have been performed by Fog [11]. A preliminary version of our paper was presented as a work in progress at RTSS 2018 [22].

Contribution and organization. We introduce the SMART task model, a method for scheduling SMART tasks, and a related schedulability test. While other works focus on modifying hardware to make SMT more predictable, our work allows for SMT-supported real-time work to run on existing hardware and operating systems. We give results of benchmark tests measuring the performance impacts of SMT with regard to execution times. We show, using a schedulability study based on our benchmark results, that it is possible to correctly schedule task systems with utilizations more than 30% greater than what would be schedulable on the same platform without SMT enabled.¹

The rest of this paper is organized as follows. In Sec. 2, we give a brief overview of SMT technology, discuss the shortcomings of the sporadic task model with regard to SMT, and introduce the SMART model. In Sec. 3, we address Sub-Problems 2 and 3, showing how SMT can be used to schedule otherwise unschedulable task systems. In Sec. 4, we address Sub-Problem 1, how to determine appropriate costs. (Note that we address our sub-problems in reverse order.) In Sec. 5, we present our schedulability experiments and results. In Sec. 6, we conclude and suggest future directions for our research.

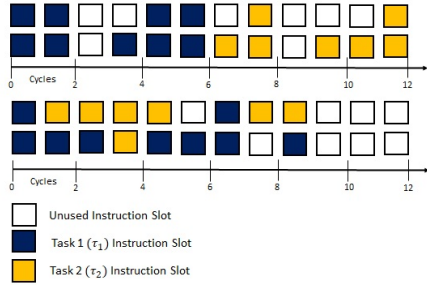
2 What is a SMART Task?

Here we give a brief overview of SMT technology alongside the sporadic task model and its limitations. We introduce SMART as an alternative model to address SMT.

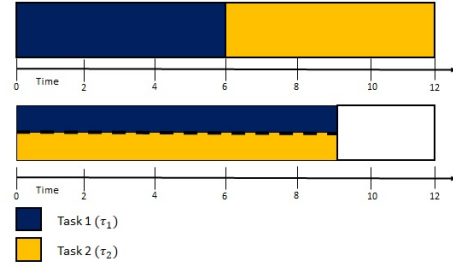
2.1 SMT Basics

Cores with SMT enabled accept multiple instructions per cycle from multiple tasks, reducing wasted instructions per cycle. A detailed explanation is available in Eggers et al.[6], but we illustrate the essentials in Example 1.

¹ While Jain et al. [15] were able to schedule systems with up to 50% greater utilization, they define a "correctly scheduled system" as one having a low number of observed deadline misses, whereas we define correctness as all tasks having analytically guaranteed bounded tardiness.



■ **Figure 2** Top: task execution without SMT. Bottom: task execution with SMT.



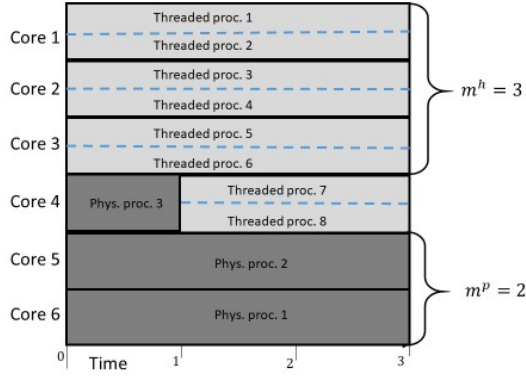
■ **Figure 3** Two tasks executing without SMT (top) and with SMT (bottom). With SMT, each task requires more time to complete individually, but time for both tasks to complete is reduced.

► **Example 1.** Fig. 2 shows the effect of enabling SMT. At the top of the figure, tasks τ_1 and τ_2 execute sequentially without SMT on a processor that can accept two instructions per cycle. When less than two instructions are available for execution, as at times 2, 3, and elsewhere, processor cycles are lost. τ_1 finishes at time 6 and τ_2 at time 12. At the bottom of the figure, the same tasks execute in parallel with SMT enabled, reducing the number of lost processor cycles. Both tasks finish at time 9. In this case, SMT has the effect of delaying the completion of τ_1 , but speeding up the completion of τ_2 , since it does not have to wait for τ_1 to complete before beginning its own execution. ◀

Fig. 3 gives a more task-centric view of the two tasks seen in Fig. 2. For the remainder of this paper, we will conceptualize tasks as seen in Fig. 3; we are interested in how long a task takes to execute and how much of a core it uses, not an exact cycle-by-cycle accounting. As shown in Fig. 3, SMT can cause individual tasks to take longer to complete, but total throughput is potentially increased, since the number of wasted instruction slots can be decreased. The challenge for real-time scheduling is to take advantage of this increased throughput without allowing increased execution costs to render the system unschedulable. The effect of SMT on task execution times is not constant across tasks; how much a task's execution time is increased by SMT depends on both the task itself and on other tasks that might be executing on the same core.

To discuss SMT more easily, we make a distinction between a core and a processor. A *core* is the hardware unit responsible for executing instructions. A *processor* is a single instruction context on a core. Every computer core, by definition, supports at least one processor, but computer cores capable of SMT may support multiple processors. We define a *physical processor* as a processor that occupies an entire core, while a *threaded processor* corresponds to a single hardware thread. Different threaded processors on the same core are *sibling processors*. Tasks scheduled on sibling processors are said to be *co-scheduled*.

We focus on a platform π that has m cores where every core supports one physical processor or two threaded processors at a time. For example, Fig. 4 shows a system of six cores. Cores 1-3 have SMT enabled and support two threaded processors each. Cores 5 and 6 have SMT disabled and support one physical processor each. Core 4 initially has SMT disabled and supports one physical processor, but at time 1, SMT is enabled on core 4, causing the single physical processor to be replaced by two threaded processors. We only consider two threads per core because this is what Intel currently supports.



■ **Figure 4** Example of cores supporting threaded processors only, physical processors only, or both.

2.2 Task Model

In the traditional implicit-deadline sporadic task model, a task $\tau_i = (T_i, C_i)$ is defined by its *period*, T_i , and its worst-case execution *cost*, C_i . The *utilization* of τ_i is given by $u_i = \frac{C_i}{T_i}$. Every task releases an unlimited number of *jobs*, with the k^{th} job released by τ_i denoted by $\tau_{i,k}$. Jobs of τ_i are released at least T_i units of time apart and have an implicit deadline of T_i . If the jobs of each task τ_i are released exactly T_i units apart, then the task system is *periodic*. We consider only SRT systems here, in which some deadline misses are acceptable. In our model, a job's *tardiness* is the difference between its completion time and deadline, if the job completes after its deadline, and zero otherwise. A task's tardiness is the maximum tardiness of any of its jobs. We define an SRT system as being *correctly scheduled* if all tasks have guaranteed bounded tardiness. A task system is *SRT-schedulable* under scheduling algorithm A if it can be correctly scheduled by the specific algorithm A , and *SRT-feasible* if it is SRT-schedulable by some algorithm A . An algorithm is *SRT-optimal* if it can schedule all SRT-feasible task systems.

Given a platform π consisting of m identical cores and no SMT, a task system τ is SRT-feasible if and only if

$$\forall \tau_i \in \tau \quad u_i \leq 1 \quad \text{and} \quad \sum_{i=1}^n u_i \leq m \quad (1)$$

both hold [4].

The SMART model. The shortcoming of the sporadic model in regard to SMT is that it only allows one worst-case execution cost per task, and therefore cannot adequately characterize a task system's behavior in the presence of SMT. For example, it is not possible to specify the task behavior seen in Fig. 3 using the sporadic model. To address this shortcoming, we introduce the SMART model. In this model, every task is modeled as $\tau_i = (T_i, (C_{i:j}))$. All parameters must be rational. As in the sporadic model, T_i is the period of τ_i . The parameter $(C_{i:j})$ is a list of costs that indicate the worst-case execution cost of a job of τ_i given that the entire job is co-scheduled with one or more jobs of τ_j . We define $C_{i:i}$ to be τ_i 's cost when it executes on a normal physical processor. For all $i \neq j$, $C_{i:j} \geq C_{i:i}$.² We define $u_{i:j} = \frac{C_{i:j}}{T_i}$.

² In the rare event that $C_{i:j} < C_{i:i}$ holds, the two are likely close in value, and we can simply redefine $C_{i:j}$ to equal $C_{i:i}$.

Notice that we are implicitly making four simplifying assumptions here: **(i)** τ_i 's worst-case execution time can be determined by examining how it is interfered with when co-scheduled with each other task *individually*; **(ii)** when τ_i is co-scheduled with τ_j , every portion of τ_i receives the same amount of interference from every portion of τ_j ; **(iii)** the two threads of a given core are identical; and **(iv)** the hardware-level priority of τ_i and τ_j , when co-scheduled, is fixed. In practice, (i) and (ii) will not necessarily hold, but we maintain that our model is sufficient for non-safety critical SRT workloads. Currently, (iii) and (iv) hold on Intel architectures [11]. We discuss (i) and (ii) further in Sec. 4 when we delve into the issue of how to actually determine execution costs.

► **Definition 2.** *The execution rate of τ_i given that it is co-scheduled with τ_j is given by $r_{i:j} = \frac{C_i}{C_{i:j}}$, where both C_i and $C_{i:j}$ are maximum observed execution times.*

We assume no relationship between $r_{i:j}$ and $r_{j:i}$; in fact, as we show in our benchmark experiments, the two can differ significantly. Our definition assumes two hardware threads per core, but could be expanded to allow for additional threads. In general, $r_{i:j} > 0.5$ indicates that τ_i could benefit from being co-scheduled with τ_j assuming that $C_{i:j} \leq T_i$ and $C_{j:i} \leq T_j$ hold.

► **Example 3.** Suppose Fig. 3 depicts one job each of SMART tasks τ_1 and τ_2 . $C_1 = 6$ and $C_2 = 6$, but $C_{1:2} = 9$ and $C_{2:1} = 9$, giving $r_{1:2} = r_{2:1} = \frac{2}{3}$. Task τ_2 benefits from SMT; the job completes at time 9 with SMT as opposed to time 12 without. If both jobs are released at time 0 and have a deadline at time 10, then SMT allows for both jobs to complete on time, whereas without SMT, τ_2 's job misses its deadline. ◀

Scheduling SMART tasks. We need to schedule n tasks that have n costs each; this problem poses obvious difficulties. In the next section, we show how we can schedule SMART tasks similarly to traditional sporadic tasks without sacrificing the advantages of SMT.

3 Scheduling Physical and Threaded Tasks

Not all tasks will benefit from SMT. We label tasks that should and should not use SMT as *threaded tasks* and *physical tasks*, respectively. Physical tasks can execute only on physical processors and threaded tasks only on threaded processors. To keep the task types separate, we divide them into two task subsystems, τ^p and τ^h , that we schedule separately.

► **Definition 4.** *Subsystem τ^p is the set of all physical tasks in τ . $n^p = |\tau^p|$. Subsystem τ^h is the set of all threaded³ tasks in τ . $n^h = |\tau^h|$.*

After partitioning τ into τ^p and τ^h , physical tasks have cost C_i^p and utilization $u_i^p = \frac{C_i^p}{T_i}$; threaded tasks have cost C_i^h and utilization $u_i^h = \frac{C_i^h}{T_i}$. Costs for physical tasks are no different than costs in a sporadic task system, but costs for threaded tasks are a function of how the task system is divided. These cost parameters are a simplification of the full SMART parameters; we will show how to obtain them in Sec. 3.2.

► **Definition 5.** *The total utilizations of τ^p and τ^h are given by $U^p = \sum_{i=1}^{n^p} u_i^p$ and $U^h = \sum_{i=1}^{n^h} u_i^h$ respectively. To measure the total demand placed on the platform, we define effective utilization, $U^E = U^p + \frac{U^h}{2}$. U^h is halved in the sum to reflect the fact that each threaded task requires only half a core at a time to execute. ◀*

³ We use h rather than t for threaded so as to avoid confusion with t for time.

3.1 Sub-Problem 3: Scheduling Task Subsystems

In this section, we give conditions for scheduling τ^p and τ^h on π . We assume the decision of which tasks should be physical and which should be threaded has already been made. Our current problem is how to schedule those tasks, but the best way to do so is not clear.

► **Example 6.** Suppose we attempt to schedule a task system τ using global earliest-deadline-first scheduling (GEDF). Let τ_1 be a threaded task and τ_2 a physical task such that at time t , a job of τ_1 with a deadline of $t + 1$ is contending for a single core with a job of τ_2 with a deadline of $t + 2$. Following GEDF, the job of τ_1 should be given priority over that of τ_2 . However, if no other threaded task has an active job at time t , then doing so will cause the second threaded processor of a core in π to be unused, negating any advantage gained by having τ_1 be threaded. If we avoid this problem by giving priority to τ_2 , then we are not wasting processor capacity, but we are violating EDF priority rules. If we co-schedule the tasks on threaded processors despite τ_2 being a physical task, then unanticipated task interference may ensue, potentially invalidating assigned per-task worst-case execution costs. None of these approaches is particularly satisfactory. ◀

To address the problems raised in Example 6, we divide π into *sub-platforms* π^p and π^h .

► **Definition 7.** π^p is the sub-platform of π that schedules only tasks in τ^p . It includes $m^p = \lfloor U^p \rfloor$ fully available cores and one partially available core. Given a length- W interval, denoted a window, the partially available core belongs to π^p for $a^p W$ time units per window, where $a^p = U^p - \lfloor U^p \rfloor$. π^p can exist only if $U^p \leq m$.

► **Definition 8.** π^h is the sub-platform of π that schedules only tasks in τ^h . It has $m^h = m - \lfloor U^p \rfloor$ fully available cores and one core available for $a^h W$ time units per window, where $a^h = \lceil U^p \rceil - U^p$. Consequently, $m^h + a^h = m - U^p$. If $a^p > 0$, then $a^h = 1 - a^p$.

We refer to the core shared by both platforms as the *shared core*. If there is no shared core, then $a^p = a^h = 0$. Note that $m^p + a^p + m^h + a^h = m$ must hold.

► **Example 9.** In Fig. 4, π^p is shown in dark gray and π^h in light gray. The sub-platforms are defined by $m^p = 2$, $m^h = 3$, $W = 3$, $a^p = \frac{1}{3}$, and $a^h = \frac{2}{3}$. ◀

We now give schedulability results for τ^p and τ^h individually and then combine those conditions to get an overall schedulability result. For the most part, we will focus on the case where a shared core exists. Our results are based on Devi and Anderson's EDF-high-low (EDF-hl) algorithm [5]. EDF-hl gives schedulability conditions and tardiness bounds for "low" SRT tasks that are scheduled according to GEDF but are subject to interruption from periodic "high" hard real-time tasks, with at most one such task fixed on each processor. For our purposes, we can view τ^p as a set of low tasks scheduled on $m^p + \lceil a^p \rceil$ processors and subject to preemption by a single high task with period W and cost $a^h W$. This reflects the fact that, from the perspective of τ^p , work on the shared core is periodically preempted. Likewise, we can view τ^h as a set of low tasks scheduled on $2(m^h + 1)$ processors that are periodically preempted by two high tasks, both with period W and cost $a^p W$. The following definitions apply to the EDF-hl results.

► **Definition 10.** Devi and Anderson define τ_H as the set of all high tasks, τ_L as the set of all low tasks, $u_{\max}(\tau_L)$ as the highest-utilization task within τ_L , U_{sum} as the total utilization of both τ_H and τ_L , U_H is the sum of all the utilizations of all tasks in τ_H , and U_L is the sum of the $\min(\lceil U_{\text{sum}} \rceil - 2, n)$ largest utilization of tasks in τ_L .

We state an abridged version of Theorem 1 in [5] here. The full version defines the tardiness bound B as a function of the task system and platform. We omit that portion of the theorem due to space constraints.

► **Theorem 11.** *EDF-hl ensures a tardiness bound of at most B to every task τ_i of τ_L if $|\tau_H| \leq m$ and $U_{sum} \leq m$ and at least one of (2) or (3) holds.*

$$m - |\tau_H| - U_L > 0 \quad (2)$$

$$m - \max(|\tau_H| - 1, 0)u_{max}(\tau_L) - U_L - U_H > 0 \quad (3)$$

Returning to our problem, our schedulability conditions rely on the following assumptions. These assumptions allow us to schedule τ^p and τ^h as if they both consisted of standard sporadic tasks. We will show how to support Assumptions 1 and 2 in Sec. 3.2.

▷ **Assumption 1.** Tasks have been divided into threaded and physical tasks such that $\forall \tau_i^p \in \tau^p, u_i^p \leq 1$ and $\forall \tau_i^h \in \tau^h, u_i^h \leq 1$ both hold. Without loss of generality, we assume that the tasks in each of the sets τ^p and τ^h are indexed in decreasing-utilization order, e.g., u_1^p (resp., u_1^h) is the largest utilization in τ^p (resp., τ^h).

▷ **Assumption 2.** Worst-case costs for physical and threaded tasks have been determined.

▷ **Assumption 3.** Physical tasks are not permitted to execute on threaded processors.⁴

► **Lemma 12.** *τ^p is schedulable on π^p under GEDF such that all tasks have guaranteed bounded tardiness if (4) holds.*

$$U^p \leq m^p + a^p. \quad (4)$$

Proof. If $a^p = 0$, then the result restates the SRT feasibility condition for m identical, fully available processors from (1). GEDF is known to be SRT-optimal [4], so the result follows.

If $m^p = 0$, then it can easily be shown that the system is schedulable only if $U^p \leq a^p$.

In the rest of the proof, we consider the remaining possibility, i.e., that $a^p > 0$ and $m^p > 0$ both hold. For this case, we show that Theorem 11 can be applied.

From the perspective of τ^p , there exists a set of low tasks τ^p with total utilization U^p , one high task with utilization a^h , and $m^p + 1$ processors. Thus, we want to apply Theorem 11 with the substitutions $m \leftarrow m^p + 1$, $\tau_L \leftarrow \tau^p$, $U_{sum} \leftarrow U^p + a^h$, and $|\tau_H| = 1$. With these substitutions, (4), and Def. 8, it is straightforward to see that both $|\tau_H| \leq m$ and $U_{sum} \leq m$ hold, as required by Theorem 11. We now show that (2) holds, from which bounded tardiness for the tasks in τ_L , i.e., those in τ^p , follows. To see this, note that from Def. 8 and $U_{sum} = U^p + a^h$, we have

$$\begin{aligned} U_L &= \sum_{i=1}^{\min(\lceil U^p + a^h \rceil - 2, m^p)} u_i^p \\ &= \{\text{by Defs. 7 and 8, } U^p + a^h = m^p + 1\} \\ U_L &= \sum_{i=1}^{\min(m^p - 1, m^p)} u_i^p \\ &\Rightarrow \{\text{because } u_i^p \leq 1 \text{ holds, by Assumption 1}\} \\ U_L &< m^p. \end{aligned}$$

From this inequality, we have $m - |\tau_H| - U_L = m^p + 1 - 1 - U_L > 0$, as required by (2). ◀

⁴ When the shared core belongs to π^p , it supports a physical processor, not a threaded processor.

The schedulability condition for τ^h is slightly more complicated, due to it potentially having two partially available processors.

► **Lemma 13.** τ^h is schedulable on π^h under GEDF such that all tasks have guaranteed bounded tardiness if (5) and at least one of (6) or (7) hold, where $u_{\max}(\tau^h)$ denotes the maximum task utilization in τ^h .

$$U^h \leq 2(m^h + a^h) \quad (5)$$

$$2m^h > \sum_{i=1}^{\min(2m^h, n^h)} u_i^h \quad (6)$$

$$2(m^h + a^h) - u_{\max}(\tau^h) > \sum_{i=1}^{\min(2m^h, n^h)} u_i^h \quad (7)$$

Proof. As in the prior proof, the proof is straightforward if either $a^h = 0$ holds or $m^h = 0$ holds, so we focus on the remaining possibility, i.e., $m^h > 0$ and $a^h > 0$ both hold; note that the latter implies that $a^p > 0$ holds as well. As before, we will use Theorem 11. In this case, we are attempting to schedule a set of low tasks τ^h with total utilization U^h on $2(m^h + 1)$ processors given two high tasks, each with utilization a^p . Thus, we want to apply Theorem 11 with the substitutions $m \leftarrow 2(m^h + 1)$, $\tau_L \leftarrow \tau^h$, $U_{\text{sum}} \leftarrow U^h + 2a^p$, and $|\tau_H| = 2$. With these substitutions, (5), and Def. 8, it is straightforward to see that both $|\tau_H| \leq m$ and $U_{\text{sum}} \leq m$ hold, as required by Theorem 11. In the rest of the proof, we show that, with these substitutions, (6) implies (2) and (7) implies (3), from which bounded tardiness for the tasks in τ_L , i.e., those in τ^h , follows.

To see that (6) implies (2), first note that, because m^h is an integer, we have $\lceil U_{\text{sum}} \rceil - 2 \leq \lceil m \rceil - 2 = \lceil 2(m^h + 1) \rceil - 2 = \lceil 2m^h \rceil = 2m^h$. Therefore,

$$\begin{aligned} 2m^h &> \sum_{i=1}^{\min(2m^h, n^h)} u_i^h \\ \Rightarrow \{ \text{because } \lceil U_{\text{sum}} \rceil - 2 &\leq 2m^h \} \\ 2m^h &> \sum_{i=1}^{\min(\lceil U_{\text{sum}} \rceil - 2, n^h)} u_i^h \\ &= \{ \text{by the definition of } U_L \text{ in Def. 10} \} \\ 2m^h &> U_L, \end{aligned}$$

i.e., $2m^h - U_L > 0$ holds, which is equivalent to (2), since $m = 2(m^h + 1)$ and $|\tau_H| = 2$.

To see that (7) implies (3), observe that

$$\begin{aligned} 2(m^h + a^h) - u_{\max}(\tau^h) &> \sum_{i=1}^{\min(2m^h, n^h)} u_i^h \\ \Rightarrow \{ \text{reasoning as above} \} \\ 2(m^h + a^h) - u_{\max}(\tau^h) &> U_L \\ = \{ \text{because } a^h &= 1 - a^p \} \\ 2(m^h + 1 - a^p) - u_{\max}(\tau^h) &> U_L, \\ = \{ \text{in our context } u_{\max}(\tau^h) &= u_{\max}(\tau_L), |\tau_H| - 1 = 2, U_H = 2a^p, \text{ and } m = 2(m^h + 1) \} \\ m - \max(|\tau_H| - 1, 0)u_{\max}(\tau_L) - U_H &> U_L, \end{aligned}$$

which is equivalent to (3). ◀

A special case applies when there is no shared core.

► **Lemma 14.** *If $a^h = 0$, then τ^h is schedulable on π^h under GEDF if and only if $U^h \leq 2m^h$ holds.*

Proof. With no shared core, the platform consists of $2m^h$ identical cores. The standard SRT feasibility test given by (1) applies. ◀

Our next step is to give a schedulability condition for τ^p and τ^h combined on π . This condition is a straightforward extension of the preceding lemmas, but it has the benefit of letting us focus on τ rather than on how π is partitioned.

► **Theorem 15.** *Platform π can be partitioned such that τ^p is schedulable on π^p and τ^h is schedulable on π^h , both under GEDF, if (8) and at least one of (9) or (10) hold.*

$$U^E \leq m \tag{8}$$

$$2(m - \lceil U^p \rceil) > \sum_{i=1}^{\min(2(m - \lceil U^p \rceil), n^p)} u_i^h \tag{9}$$

$$2(m - U^p) - u_1^h > \sum_{i=1}^{\min(2(m - \lceil U^p \rceil), n^p)} u_i^h \tag{10}$$

Proof. In order to define m^p and a^p so that $m^p + a^p = U^p$ holds, as in Def. 7, we merely require $U^p \leq m$ to hold, and by Def. 5, this is implied by (8). Note that $m^p + a^p = U^p$ satisfies Condition (4) in Lemma 12.

Schedulability of τ^p on π^p is implied by (8):

$$\begin{aligned} U^E &\leq m \\ &= \{\text{by Def. 5, } U^E = U^p + \frac{U^h}{2}\} \\ U^p &\leq m \\ &= \{\text{by Def. 7, } m^p + a^p = U^p\} \\ U^p &= m^p + a^p, \end{aligned}$$

which is the condition for τ^p per Lemma 12.

We next show that (8) implies Condition (5) of Lemma 13. To see this, observe that, by Def. 5, $U^E \leq m \Rightarrow \frac{U^h}{2} \leq m - U^p$. Also, by Def. 8, $m^h + a^h = m - U^p$. Putting these facts together, we have $U^h \leq 2(m^h + a^h)$, which is (5).

We conclude the proof by showing that (9) is equivalent to Condition (6) of Lemma 13, and that (10) is equivalent to Condition (7) of Lemma 13. To see the former, note the following.

$$\begin{aligned} 2(m - \lceil U^p \rceil) &> \sum_{i=1}^{\min(2(m - \lceil U^p \rceil), n^h)} u_i^h \\ &= \{\text{by Def. 8, } m - \lceil U^p \rceil = m^h\} \\ 2m^h &> \sum_{i=1}^{\min(2m^h, n^h)} u_i^h \end{aligned}$$

Similarly, to see that (10) holds, note the following.

$$\begin{aligned}
 2(m - U^p) - u_1^h &> \sum_{i=1}^{\min(2(m - \lceil U^p \rceil), n^h)} u_i^h \\
 &= \{\text{by Def. 8, } m - \lceil U^p \rceil = m^h.\} \\
 2(m^h + a^h) - u_1^h &> \sum_{i=1}^{\min(2m^h, n^h)} u_i^h.
 \end{aligned}$$

Having verified all conditions of Lemmas 12 and 13, we conclude that τ^p is schedulable on π^p and τ^h is schedulable on π^h . \blacktriangleleft

Again, a special case applies if U^p is integral.

► **Corollary 16.** *If U^p is integral, then both τ^p and τ^h are schedulable on their respective sub-platforms under GEDF so long as $U^E \leq m$ holds.*

Proof. Similar to the proof of Lemma 14. \blacktriangleleft

It is not strictly necessary that π^p be defined as we do here. If we allow other design considerations, such as maximizing cache affinity or minimizing tardiness, different platform definitions may be preferable, but we defer those possibilities to future work.

By themselves, the results of this section are not very useful, since there are an exponential number of possible ways to partition π . In the next section, we show how to efficiently find τ^p and τ^h that will be schedulable under Theorem 15.

3.2 Sub-Problem 2: Dividing the Tasks

We have addressed how to schedule a task system τ for a given pair of subsystems τ^p and τ^h . Here, we show how we arrive at Assumption 1— τ has already been divided—and weaken Assumption 2, which states that all execution costs have been determined, to the following:

▷ **Assumption 4.** If τ_i is a threaded task, then $C_i^h = \max_{\forall \tau_j \in \tau} C_{i:j}$.

Oblivious scheduling. We first work through a simple example of dividing a task system and then formalize that approach into what we term *symbiosis-oblivious partitioning*.⁵ We then show how our approach can be improved by modifying Assumption 4.

► **Example 17.** Let τ consist of four SMART tasks,

$$\begin{aligned}
 \tau_1 &= (8, (7, 10, 10, 9.\bar{3})), & \tau_2 &= (4, (4, 1, 2, 1.\bar{3})), \\
 \tau_3 &= (4, (3, 2.\bar{6}, 2, 2.5)), & \tau_4 &= (8, (6, 6, 5.\bar{3}, 4)).
 \end{aligned}$$

Under traditional sporadic scheduling, where we consider only physical costs, τ has total utilization $\frac{7}{8} + \frac{1}{4} + \frac{2}{4} + \frac{4}{8} = 2.125$ and will require three cores to be feasibly scheduled (recall that $C_{i:i}$ gives τ_i 's cost with nothing co-scheduled, i.e., without SMT). Based on Assumption 4, we see that $C_1^h = 10$ if τ_1 is threaded. Because $T_1 = 8$, making τ_1 threaded would give $u_1^h = \frac{10}{8}$, making the system unschedulable. For τ_2 , C_2^h would be at most τ^2 's period, but $C_2^h = 4$ would be more than twice $C_2^p = 1$. Part of the schedulability condition given in

⁵ The terms symbiosis-oblivious and symbiosis-aware scheduling were previously used by Jain et al.[15].

```

1: for all  $\tau_i \in \tau$  do
2:    $C_i^h \leftarrow \max_{\forall j \leq n} C_{i:j}$ 
3:   if  $C_i^h \leq T_i$  and  $\frac{C_i}{C_i^h} \geq 2$  then
4:      $\tau^h \leftarrow \tau^h \cup \tau_i$ 
5:   else
6:      $C_i^p \leftarrow C_{i:i}$ 
7:      $\tau^p \leftarrow \tau^p \cup \tau_i$ 
8:   end if
9: end for
10: if  $|\tau^h| < 2$  then
11:    $\tau^p \leftarrow \tau^p \cup \tau^h$ 
12:    $\tau^h \leftarrow \emptyset$ 
13: end if
14: return  $\tau^p, \tau^h$ 

```

Algorithm 1: Oblivious Partitioning

Theorem 15 is that $U^E \leq m$. Because U^E is defined as $U^E = U^p + \frac{U^h}{2}$ (Def. 5), placing τ_2 in τ^h would increase U^E more than placing τ_2 in τ^p , so we do not wish for τ_2 to be threaded. For both τ_3 and τ_4 , $\max C_{i:j} \leq T_i$ and $\min(\frac{C_i}{C_{i:j}}) \geq .5$ both hold, so letting those tasks be threaded would decrease U^E compared to placing them in τ^p without violating $u_i^h \leq 1$, so we allow those tasks to be threaded, giving $u_3^h = \frac{3}{4}$ and $u_4^h = \frac{6}{8}$. The resulting partition has $U^p = \frac{7}{8} + \frac{1}{4}$, $U^h = \frac{3}{4} + \frac{6}{8}$, and $U^E = 1.875$. It can, per Theorem 15, be scheduled on only two cores. ◀

We formally state the steps we just took in Algorithm 1, which partitions τ into τ^p and τ^h so as to minimize U^E subject to $u_i^h \leq 1$ for all threaded tasks and $|\tau^h| \geq 2$. The resulting partition is then schedulable if Theorem 15 holds. We require that $|\tau^h| \geq 2$ holds since allowing a single threaded task will give no schedulability advantage compared to letting all tasks be physical. We refer to partitions of τ that obey both these constraints as *legal*. We will examine the effectiveness of Algorithm 1 in our schedulability study.

► **Definition 18.** A partition of τ into τ^p and τ^h is legal if and only if $\forall \tau_i^h \in \tau^h, u_i^h \leq 1$ and $|\tau^h| \neq 1$ hold.

A more complex cost model. Under Assumption 4, the only variable that influences the cost of τ_i is whether τ_i is physical or threaded. However, Assumption 4, and consequently Algorithm 1, is highly pessimistic with regard to assigning C_i^h values. Returning to Example 17, we declared $C_3^h = 3$ on the grounds that $\forall j, \max C_{3:j} = 3$ holds. However, there is a limitation to that logic; $C_3^h = 3$ is based on the assumption that τ_1 can interfere with τ_3 , but in our example, we decided that τ_1 should not be threaded. We can remove this limitation, thereby improving our model, by replacing Assumption 4 with Assumption 5. The difference is that under Assumption 5, C_i^h is only based on other threaded tasks, not on all tasks in τ .

▷ **Assumption 5.** If τ_i is threaded, then $C_i^h = \max_{\forall \tau_j \in \tau^h} C_{i:j}$.

The difference is that while Assumption 4 considers interference from all tasks in τ , Assumption 5 considers interference only from other tasks in τ^h . While this approach removes some of the pessimism present in symbiosis-oblivious scheduling, it has the disadvantage that every time a task is added to or removed from τ^h , C_i^h may change for all tasks in τ^h .

We refer to task-partitioning algorithms that incorporate Assumption 5 as *symbiosis-aware partitioning*. We give a brief demonstration of symbiosis-aware partitioning in Example 19, using the same task set as in Example 17.

► **Example 19.** We first decide that τ_1 must be physical, since $\forall j \neq 1, C_{1:j} > T_1$. Knowing that no task will be co-scheduled with τ_1 , we have $C_2^h = 2$ and $C_3^h = 2.\bar{6}$, giving $u_2^h = \frac{2}{4}$ and $u_3^h = \frac{2.\bar{6}}{4}$, but leaving u_4^h unchanged. (In Example 17, we made τ_2 a physical task and τ_3 a threaded task with $u_3^h = \frac{3}{4}$.) Now we make all of τ_2 , τ_3 , and τ_4 threaded, with τ_3 having a lower utilization than before. We now get $U^p = \frac{7}{8}$ and $U^h = \frac{2}{4} + \frac{2.\bar{6}}{4} + \frac{6}{8}$, so that $U^E = 1.8\bar{3}$, a reduction from $U^E = 1.875$ in Example 17. Again, τ^p and τ^h are schedulable on two cores per Theorem 15.

A greedy approach to schedulability. We propose to use Algorithm 2 to partition τ . The algorithm seeks to minimize U^E by repeatedly moving a task from τ^p to τ^h , or vice versa, to give the greatest decrease in U^E . It does so until either a specified maximum number of attempts has been made or it reaches a partition that cannot be improved by the movement of any single task. The algorithm is not optimal, even given an unlimited number of attempts, as there may exist partitions of τ that cannot be improved by moving any one task but can be improved by moving two or more tasks.

The **for** loop of lines 3 through 16 determines, for every τ_i in τ^p , the benefit of moving that task to τ^h . Line 4 tests what C_i^h would be if τ_i were in τ^h . Lines 10 through 13 calculate the change to tasks already in τ^h caused by moving τ_i , and line 15 gives the total change to U^E caused by moving τ_i to τ^h .

Similarly, the **for** loop of lines 19 through 23 determines the benefit of moving τ_j to τ^p , for every τ_j currently in τ^h . Line 20 gives the change to tasks remaining in τ^h caused by moving τ_j , and line 22 gives the total change to U^E caused by moving τ_j to τ^p . The **if** of line 25 guarantees that no task will be moved unless moving that task will decrease U^E , preventing the algorithm from placing τ into any one partition more than once.

The algorithm returns a partition that can be tested for schedulability by Theorem 15.

Algorithm 2 assumes, and maintains as an invariant, that the partition is legal, as defined in Def. 18. To begin Algorithm 2, τ must already be in a legal partition. We propose three ways to achieve this. First, in the *greedy-threaded* approach, we begin with all tasks in τ^h and then place into τ^p all tasks for which any possible C_i^h value will give $u_i^h > 1$. Intuitively, putting tasks in τ^h whenever possible should be beneficial, so we should start with as many tasks in τ^h as possible.

Second, in the *greedy-physical* approach, we start with all tasks in τ^p apart from the single pair of tasks that will give the greatest decrease to U^E . This can be done by defining the decrease to U^E associated with a single pair of tasks (τ_i, τ_j) as

$$\forall (i, j), \Delta(i, j) = u_i^p + u_j^p - \frac{1}{2} \left(\frac{C_{i:j}}{T_i} + \frac{C_{j:i}}{T_j} \right)$$

and adding to τ^h the pair of tasks that maximize $\Delta(i, j)$ subject to $\frac{C_{i:j}}{T_i} \leq 1$ and $\frac{C_{j:i}}{T_j} \leq 1$. When τ_i and τ_j are placed into τ^h , u_i^p and u_j^p are no longer part of U^p and can be subtracted from U^E . However, we must add half of the new U^h value, $\left(\frac{C_{i:j}}{T_i} + \frac{C_{j:i}}{T_j} \right)$, to U^E . We expect this approach will be more efficient than the first one in task systems where u_i^p is typically large or $\frac{C_{i:j}}{C_{i:i}}$ is typically small, since there will be relatively few tasks that can be placed in τ^h , making it more efficient to begin with the majority of tasks in τ^p . If no satisfactory pair of tasks exists, then we conclude that SMT should not be used with this task system.

Require: τ partitioned such that $\forall \tau_i \in \tau^h, u_i^h \leq 1$ and $|\tau^h| \geq 2$

```

1: for  $\ell \leftarrow 1 \dots \text{maxLoops}$  do
2:    $\triangleright$  Identify best move from  $\tau^p$  to  $\tau^h$ 
3:   for all  $\tau_i \in \tau^p$  do
4:      $C_i^h = \max_{\tau_i \in \tau^h} C_{i:j}$ 
5:      $u_i^h = \frac{C_i^h}{T_i}$ 
6:     if  $u_i^h > 1$  then
7:       continue
8:     end if
9:      $\triangleright$  Calculate how adding  $\tau_i$  to  $\tau^h$  will affect tasks already in  $\tau^h$ 
10:    if moving  $\tau_i$  to  $\tau^h$  will cause  $u_j^h \geq 1$  for any  $\tau_j \in \tau^h$  then
11:      continue
12:    end if
13:     $I(\tau_i^h) \leftarrow$  total increase in util. of tasks already in  $\tau^h$  caused by moving  $\tau_i$ 
14:     $\triangleright \Delta(i)$  gives decrease to  $U^E$  caused by moving  $\tau_i$ .
15:     $\Delta(i) \leftarrow u_i^p - \frac{u_i^h + I(\tau_i^h)}{2}$ 
16:  end for
17:   $\triangleright$  Identify best move from  $\tau^h$  to  $\tau^p$ 
18:  if  $|\tau^h| > 2$  then
19:    for all  $\tau_j \in \tau^h$  do
20:       $D(\tau_j^h) \leftarrow$  total decrease in util. of tasks already in  $\tau^h$  caused by moving  $\tau_j$ 
21:       $\triangleright \Delta(j)$  gives decrease to  $U^E$  caused by moving  $\tau_j$ .
22:       $\Delta(j) \leftarrow \frac{u_j^h + D(\tau_j^h)}{2} - u_j^p$ 
23:    end for
24:  end if
25:  if no task has a positive  $\Delta$  value then
26:    break
27:  end if
28:  Move task with maximum  $\Delta$  to other subsystem and update threaded costs
29: end for
30: return( $\tau^p, \tau^h$ )

```

Algorithm 2: Greedy Partitioning

Third, in the *greedy-mixed* approach, we first run Algorithm 1 and use the partition given by doing so as our starting point. Intuitively, Algorithm 1 by itself should give a partition with a lower U^E value than either of the other two approaches, so using it as a starting point should yield better results. As with the greedy-physical approach, if Algorithm 1 places no tasks in τ^h , then we conclude that SMT should not be used. We compare these three approaches in our schedulability experiments, presented in Sec. 5. We found that for all three versions of Algorithm 2, there existed task systems that were schedulable according to that version alone. In fact, the greedy-physical approach seemed to find more schedulable task systems than the other two.

■ **Table 1** Baseline Execution Times (ns)

Benchmark	max	mean	$CV \left(\frac{\text{std. dev.}}{\text{mean}} \right)$
adpcm_dec	167,380	151,914	0.006659
adpcm_enc	158,053	147,394	0.006463
ammunition	47,979,870	47,899,553	0.001589
cjpeg_transupp	844,791	827,661	0.002087
cjpeg_wrbmp	32,420	26,712	0.010552
dijkstra	15,740,782	15,719,309	0.000445
epic	665,837	649,170	0.002284
fmref	154,776	99,280	0.068863
gsm_dec	470,193	463,592	0.002546
gsm_enc	1,337,465	1,320,787	0.001934
h264_dec	93,361	82,045	0.006340
huff_enc	247,232	234,213	0.005431
mpeg2	135,009,849	134,898,300	0.000248
ndes	21,600	15,426	0.015071
petrinet	3,682	62	0.215268
rijndael_dec	965,022	940,081	0.007688
rijndael_enc	872,400	858,645	0.002224
statemate	11,928	6,495	0.026602
susan	10,958,260	10,932,188	0.000379

4 Sub-Problem 1: SMT and Execution Times

Current literature does not address how SMT affects worst-case execution costs. While the early 2000s saw multiple detailed analyses of the performance effects of SMT [1, 2, 25], little work of this type has been done since then. While ongoing research into scheduling with SMT exists outside of real time [7, 8, 10, 23], this current research does not suit our needs for two reasons. First, it tends to be oriented towards total throughput and average execution costs, whereas we need information on worst-case execution costs. Second, the current works we are aware of compare different methods of implementing SMT, but do not compare systems that use SMT to those that do not use it.

4.1 Benchmark Experiments

To analyze the effects of SMT on worst-case execution costs, we ran a series of experiments using the TACLeBench sequential benchmarks [9], which consist of 23 C implementations of functions commonly found in embedded and real-time systems. All of our experiments were conducted in Linux on an Intel Xeon Silver 4110 2.1 GHz CPU with eight cores, each capable of supporting two threaded processors, running Linux.⁶

To get baseline results for execution times without SMT enabled, we looped each benchmark 1,000 to 100,000 times—lower cost benchmarks got more loops—and timed the execution of each loop using a nanosecond resolution timer. Between loops, an array the size of the L3

⁶ The code used for these experiments is available at <https://github.com/JoshuaJB/SMART-ECRTS19> and <http://jamesanderson.web.unc.edu/papers/>

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interfering benchmark \ measured benchmark	adpcm_dec	adpcm_enc	ammunition	cjpeg_transupp	cjpeg_wrbmp	dijkstra	epic	fmref	gsm_dec	gsm_enc	h264_dec	huff_enc	mpeg2	ndes	petrinet	rijndael_dec	rijndael_enc	statemate	susan	MINIMUM	max CV
adpcm_dec	0.97	0.96	0.98	0.98	0.99	0.99	0.94	0.96	0.96	0.98	0.97	0.97	0.97	0.96	1.00	0.92	0.97	1.00	0.94	0.92	0.012617
adpcm_enc	0.91	0.85	0.91	0.94	0.97	0.95	0.94	0.93	0.95	0.95	0.94	0.94	0.93	0.95	0.96	0.94	0.96	0.92	0.92	0.85	0.011473
ammunition	0.67	0.66	0.67	0.65	0.69	0.68	0.69	0.64	0.64	0.66	0.68	0.68	0.66	0.68	0.69	0.68	0.70	0.71	0.68	0.64	0.001080
cjpeg_transupp	0.68	0.67	0.70	0.63	0.65	0.64	0.72	0.63	0.62	0.66	0.63	0.65	0.64	0.67	0.77	0.68	0.68	0.68	0.62	0.62	0.010452
cjpeg_wrbmp	0.69	0.63	0.63	0.62	0.59	0.65	0.68	0.69	0.60	0.65	0.54	0.55	0.60	0.63	0.74	0.66	0.52	0.66	0.62	0.52	0.061954
dijkstra	0.70	0.69	0.74	0.68	0.71	0.70	0.74	0.67	0.66	0.69	0.70	0.71	0.69	0.71	0.79	0.72	0.72	0.72	0.70	0.66	0.002617
epic	0.54	0.53	0.57	0.54	0.54	0.59	0.57	0.57	0.56	0.54	0.57	0.55	0.51	0.55	0.59	0.55	0.55	0.54	0.54	0.51	0.013115
fmref	0.75	0.75	0.76	0.73	0.75	0.66	0.77	0.74	0.73	0.73	0.71	0.71	0.73	0.75	0.79	0.76	0.76	0.76	0.67	0.66	0.060101
gsm_dec	0.64	0.63	0.64	0.61	0.65	0.63	0.68	0.60	0.60	0.61	0.62	0.63	0.61	0.63	0.71	0.64	0.64	0.65	0.61	0.60	0.011700
gsm_enc	0.59	0.58	0.60	0.56	0.61	0.62	0.64	0.57	0.57	0.59	0.60	0.61	0.58	0.61	0.65	0.61	0.62	0.63	0.59	0.56	0.012556
h264_dec	0.91	0.92	0.90	0.86	0.87	0.87	0.96	0.85	0.84	0.87	0.85	0.88	0.88	0.91	1.00	0.88	0.79	0.75	0.87	0.75	0.030283
huff_enc	0.73	0.70	0.74	0.67	0.69	0.66	0.79	0.71	0.67	0.69	0.69	0.71	0.66	0.72	0.79	0.69	0.71	0.67	0.69	0.66	0.023466
mpeg2	0.72	0.71	0.73	0.66	0.72	0.70	0.75	0.69	0.68	0.70	0.69	0.70	0.67	0.71	0.64	0.72	0.72	0.72	0.70	0.64	0.144416
ndes	0.66	0.68	0.69	0.67	0.71	0.69	0.72	0.70	0.67	0.57	0.66	0.59	0.61	0.55	0.74	0.70	0.56	0.72	0.67	0.55	0.029245
petrinet	6.11	1.24	0.91	5.93	1.05	5.68	0.88	0.94	1.20	0.78	1.18	0.69	0.78	0.82	0.71	0.60	1.22	0.82	0.98	0.60	0.875009
rijndael_dec	0.58	0.59	0.58	0.64	0.65	0.65	0.67	0.64	0.61	0.63	0.65	0.65	0.62	0.63	0.64	0.61	0.61	0.64	0.63	0.58	0.016778
rijndael_enc	0.56	0.56	0.57	0.61	0.63	0.64	0.65	0.62	0.59	0.60	0.63	0.63	0.59	0.60	0.60	0.59	0.58	0.61	0.61	0.56	0.013859
statemate	0.66	0.99	0.88	0.61	0.84	0.86	1.00	0.89	0.61	0.68	0.77	0.97	0.97	0.96	0.95	0.55	0.97	0.73	0.93	0.55	0.027924
susan	0.62	0.62	0.61	0.56	0.58	0.55	0.67	0.59	0.56	0.60	0.56	0.58	0.58	0.61	0.69	0.61	0.62	0.64	0.57	0.55	0.004294

■ **Figure 5** Effect of SMT on execution times. Measured benchmarks execute with the listed $r_{i,j}$ values when sharing a thread with a given interfering benchmark. Shading is darkest on smallest values. Right column shows the maximum coefficient of variation experienced by each measured benchmark over all interfering benchmarks.

cache was allocated and set, so that every execution started with a cold cache. Benchmarks were assigned a Linux real-time priority, prioritizing them above all normal tasks, pinned to a single processor, and executed sequentially. We excluded four benchmarks from the set—anagram, audiobeam, g723_enc, and huff_dec—as they would not correctly execute in a loop. Results of our baseline experiments are summarized in Table 1. The last column gives the coefficient of variation, defined as the standard deviation divided by the mean.

For threaded execution times, every task was executed alongside every other task. For each pair, the measured task was executed the same number of times as in the baseline experiments while an interfering task executed continuously at equal priority on the second thread of the same core. Our results are summarized in Fig. 5, which shows $r_{i,j}$ for every pair of tasks, with the measured task as τ_i and the interfering task as τ_j . Observed rates ranged from 0.51 (mpeg2 interfering with epic) to 1.00, with the exception of values involving petrinet. Petrinet has an extremely short execution time, as indicated in Table 1; we suspect its strange behavior is merely random noise.

We cannot guarantee that our experiments captured the maximum interference to τ_i caused by τ_j . However, the low coefficients of variation recorded in Fig. 5 imply that different interleavings of τ_i and τ_j will cause only minor variations in the cost of τ_i . As discussed in Sec. 4.3 below, SRT systems may tolerate some cost overruns.

While we have defined $C_{i,i}$ as the cost of τ_i with no co-schedule, the main diagonal of Fig. 5 shows how much slower a task runs when executed with a second copy of itself. This is irrelevant for real-time systems in which task parallelism is forbidden, but is relevant to systems in which different jobs of the same task may execute in parallel, as discussed by Voronov, Anderson, and Yang [27]. Prior to performing our experiments, we had expected that tasks executed alongside copies of themselves would have very low $r_{i,j}$ values, due to competing for the same resources, but our experiments show this is not necessarily the case.

4.2 Benchmark Characterization

In our results, we observe that tasks are relatively consistent both in how vulnerable they are to interference from other tasks and in how much interference they cause to other tasks. This is similar to other results in the literature [1, 2, 14, 25]. We say that tasks that experience little interference from other tasks—i.e. tasks τ_i for which $r_{i:j}$ tends to be high—are *strong*, and that tasks which cause little interference to other tasks—i.e. τ_i for which $r_{j:i}$ tends to be high—are *friendly*. When we define a *strength score* $s_i = \text{mean}_j(r_{i:j})$ and *friendliness score* $f_i = \text{mean}_j(r_{j:i})$, no task has a Pearson correlation⁷ with absolute value greater than 0.14 between s_i and f_i values. Bulpin’s work on the behavior of threaded tasks discusses this lack of correlation further [1, 2].

For both values, we centered and standardized each row and column before fitting them to several common statistical distributions via a log-likelihood maximization. We found the Gaussian distribution to best approximate the results from our experiments. Applying a maximum likelihood (MLE) estimation, we found that mean 0.72 and standard deviation 0.13 were the best for s_i while mean 0.72 and standard deviation 0.04 were best for f_i .

4.3 Reliability of Measured Worst-Case Costs

We stated in Assumption 4 that C_i^h is no more than $\max_{\tau_j \in \tau^h} C_{i:j}$. While we are confident that violations will be rare, we cannot guarantee there will not be any. In particular, our assumption that all portions of τ_i receive the same amount of interference from all portions of τ_j is a potential source of timing violations. For example, let $\tau^h = \{\tau_1^h, \tau_2^h, \tau_3^h\}$ be such that $C_{1:2} = C_{1:3} = 6$. Under Assumption 5, the worst-case execution time for τ_1 is 6. Suppose τ_1 can be broken into two segments, τ_{1a} and τ_{1b} , such that $C_{1a:2} = 4$, $C_{1b:2} = 2$, $C_{1a:3} = 2$, and $C_{1b:3} = 4$. If τ_{1a} is co-scheduled with τ_2 and τ_{1b} is co-scheduled with τ_3 , τ_1 ’s total execution time would be 8, violating our stated worst-case execution costs. At present, our benchmark tests and model do not discover or account for task inter-leavings as in this scenario. In the future, we would like to resolve this with finer-grained timing analysis and a model that does not assume task interference is independent from location within the task. In particular, breaking tasks into segments, determining execution costs per segment, as in our example, and conducting an analysis similar to this paper, but at a finer granularity, seems like a promising way forward. For now, we reiterate that we are only considering applications that are not safety-critical and where some tardiness is acceptable.

Generally, precise timing analysis on multicore is hard and contains uncertainty regardless of the added SMT challenge. Fortunately, Mills and Anderson have shown SRT systems to have expected tardiness bounds based on average rather than worst-case execution times [19]. Their approach relies on designating per-task execution budgets so that if any one job overruns its budget, it will not receive further execution time until a subsequent job of the same task could have been executed had the first job completed. These budgets come from average execution times. Therefore, so long as our costs are greater than the true average costs, any system τ that can be scheduled as we have described will remain so, though possibly with increased tardiness, even if our stated costs are not true worst-case costs.

Concerning our results here, our true interest is not in these specific times, but rather in developing a sense of how SMT-enabled tasks behave so that we can create synthetic tasks for our schedulability study that are good representations of reality.

⁷ A Pearson correlation of ± 1 indicates total positive or negative linear correlation; 0 indicates no correlation.

5 Schedulability Experiments

Having shown how to schedule SMT-enabled systems and analyzed the behavior of our benchmark tasks when using SMT, it remains to be seen whether we can schedule otherwise unschedulable systems. To answer this question, we ran a series of schedulability experiments.

5.1 Experimental Procedure

To run our experiments, we created synthetic task systems to be scheduled on platforms with m cores, $m \in \{4, 8, 16\}$ such that the total system utilization ranged from m to $2m$. For each task system, we partitioned the system into τ^p and τ^h using Algorithm 1 and all three versions of Algorithm 2. We then tested for schedulability per Theorem 15. We created enough task systems that each data point in our graphs represents the composite schedulability of approximately 1,000 task systems. We created over 300 graphs, with a few thousand to hundreds of thousands of task systems per graph. Creating task sets, partitioning task sets, and testing for schedulability consumed over 30 days of CPU time.

We plotted our results on a series of schedulability graphs with total utilizations on the horizontal axis and the proportion of systems that were schedulable on the vertical axis. Since we started at utilization m , and the standard SRT feasibility condition given by (1) requires that $\sum_{i=1}^n u_i \leq m$ hold, every system we created was infeasible without using SMT. Every system that we could schedule is an argument for adapting SMT in real-time systems.

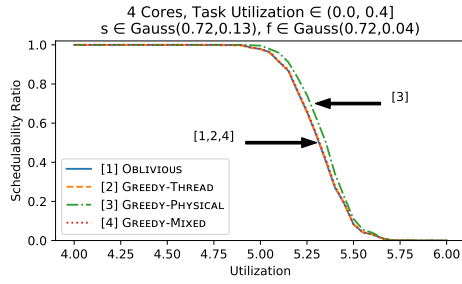
Each graph shows results for tasks created using a common set of utilization and $r_{i:j}$ values. Task utilizations were assigned from one of four ranges: the uniform distributions $(0, .4]$, $[.3, .7]$, $[.6, 1]$, and $(0, 1]$. We used two approaches for determining $r_{i:j}$ values. In the *Gaussian-average* approach, we drew s_i and f_i from the Gaussian distributions with mean 0.72 for both values and standard deviations ranging from 0.13 to 0.39 for s_i and from 0.04 to 0.12 for f_i . These parameters are based on distributions we fitted to our models, as discussed in the previous section. We allowed larger standard deviations than we obtained from our benchmarks to make our results more widely applicable.

In the *uniform-normal* approach, both s_i and f_i come from one of four uniform distributions: $[.65, 1]$, $[.7, 1]$, $[.75, 1]$, or $[.8, 1]$. The two ranges may differ for a given graph. Each $r_{i:j}$ value was then chosen from a normal distribution with mean $s_i f_i$ and standard deviation σ , where σ is .01, .05, or .1. Negative values or those greater than 1 are clamped to 0 or 1 respectively. The intuition behind the uniform-normal approach is to create $r_{i:j}$ values broadly similar to the benchmark values we obtained, but via different methods than Gaussian-average so as to avoid having our results be overly dependent on that model. While high s_i values in this context still indicate tasks that receive little interference from other tasks, and high f_i values indicate tasks that cause little interference to others, they are used differently here than in the Gaussian average approach and should not be directly compared.

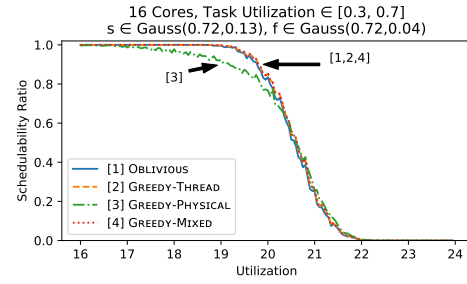
5.2 Results

Due to space constraints, we present only a small portion of our graphs to highlight general trends. A full set of graphs is available in our online appendix.⁸ For all of our graphs, the horizontal axis begins at m ; all of our task systems would be infeasible without SMT.

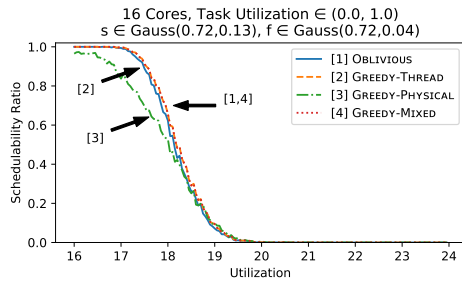
⁸ The appendix and code is available at <http://jamesanderson.web.unc.edu/papers/>. Code is also at <https://github.com/JoshuaJB/SMART-ECRTS19>.



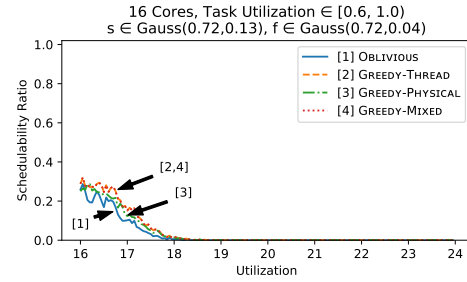
■ **Figure 6** Graph shape is similar to Fig. 1, which has more cores.



■ **Figure 7** Schedulability similar to Figs. 1 and 6, despite higher task utils.



■ **Figure 8** Despite same expected per-task util. as Fig. 7, schedulability is reduced.



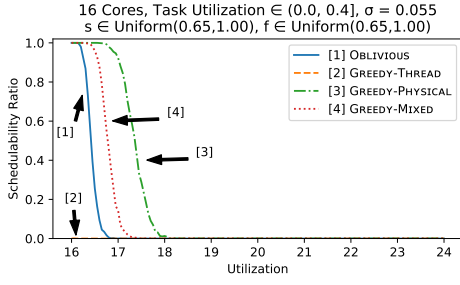
■ **Figure 9** Given high per-task utilizations, only small schedulability gains can be achieved.

▷ **Observation 1.** Given favorable task parameters, virtually all task systems with utilizations as high as $1.25m$, and roughly half of task systems with utilizations of $1.33m$, are schedulable. Favorable task parameters are high means and low standard deviations for friendliness and strength values combined with low per-task utilizations. Examples of these results are seen in Figs. 1, 6, and 7.

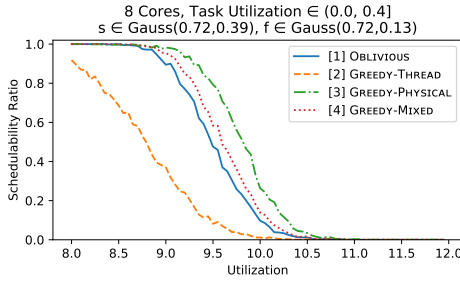
▷ **Observation 2.** Task systems with low per-task utilization received the greatest improvement in schedulability, and task systems with high utilization saw the least. Since threading tasks increases individual execution costs, it will typically not be possible to thread tasks that already have high utilizations. Fig. 6, in our introduction, shows schedulability for task systems with individual utilizations drawn from the uniform distribution $(0, 0.4]$, and shows that the majority of systems considered are schedulable with utilizations as high as 5.34 . Fig. 9 has the same parameters as Figs. 1 and 6, but draws utilizations instead from the range $[.6, 1]$. This graph shows virtually no improvement when run with SMT.

▷ **Observation 3.** Algorithm 1, oblivious partitioning, competes with the more complex algorithms. In our best results, such as Figs. 1, 6, 7, and 13, Algorithm 1 is indistinguishable from the greedy algorithms. When Algorithm 1 does not perform as well as the variants of Algorithm 2, the difference is small enough that the lower algorithm complexity might still make it a better choice.

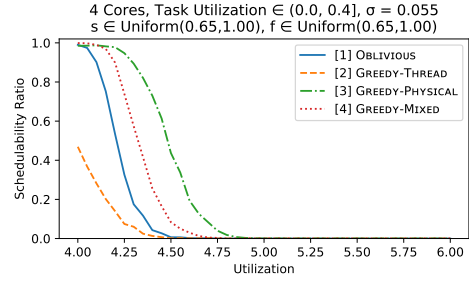
▷ **Observation 4.** Lower $r_{i,j}$ variability yields improved schedulability. In Fig. 12, the task systems sample from the same utilization range as those of Figs. 1 and 6, but here the standard deviation of the distribution from which s_i and f_i are sampled is larger. This increased variance causes fewer task sets to be schedulable Fig. 12 in than in Figs. 1 and 6.



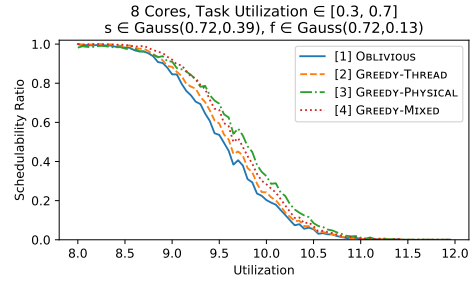
■ **Figure 10** Uniform-normal $r_{i,j}$ values on 16 cores. Note variations in algorithm performance.



■ **Figure 12** Gaussian approach with higher variance. Gains from SMT are reduced compared to Figs. 1, 6, and 7.



■ **Figure 11** Uniform-normal $r_{i,j}$ on 4 cores. Unlike the Gaussian model, core count influences gains from SMT here.



■ **Figure 13** Underperformance of GREEDY-THREAD as in Fig. 12 disappears as utilizations increase.

▷ **Observation 5.** Schedulability benefits of our methods are not limited to task systems generated using a single model. While the Gaussian approach created systems that saw more improvement from SMT, the benefits of SMT are not limited to task systems created under that model, suggesting that SMT can benefit a wide variety of task systems.

6 Conclusion

We have given a task model, SMART, that allows for reasoning about SMT-enabled task systems by defining multiple cost parameters per task. We have shown how to decide which tasks should and should not use SMT and how to take advantage of SMT to schedule otherwise unschedulable task systems. We measured the execution times of benchmark tasks with and without SMT enabled, with the SMT-enabled case covering interference from all other tasks in the set. We conducted an extensive schedulability study using synthetic tasks modeled on our benchmark tasks and showed that for task systems consisting of low utilization tasks, it is possible to schedule virtually all systems with utilization as large as $1.25m$ and to schedule many task systems with utilizations approaching $1.33m$.

In the future, we plan to improve our timing analysis to the point that hard real-time systems, where no tardiness is permitted, becomes an option. In addition, we want expand our soft real-time work by partitioning both task systems and hardware platforms to minimize tardiness, rather than simply maximizing schedulability. Making tasks threaded tends to decrease demand on the platform, potentially reducing tardiness, but will increase execution costs, potentially increasing tardiness [4, 5, 16]. While the potential gains shown in this paper are substantial, we have only begun to expose the potentials of hardware multithreading.

A Appendix

This section contains the full set of graphs from our schedulability studies. They are organized in groups of four with each of the four representing the behavior of the parameter set in the title under four different per-task utilization windows. The individual groups are then ordered by core count on the simulated system, from 4 cores to 16 cores.

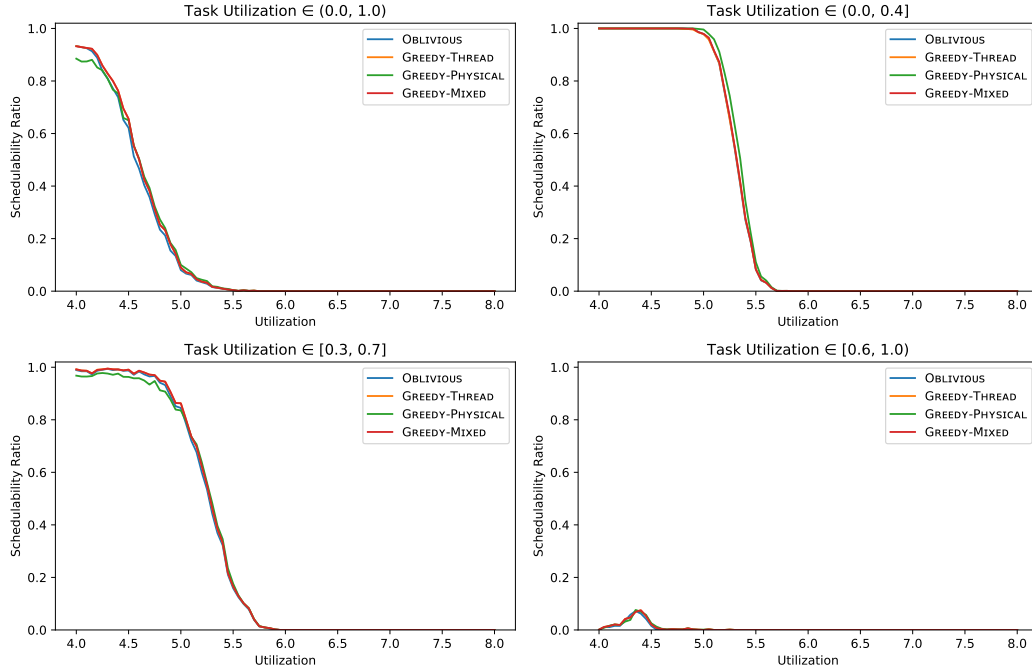
A.1 Schedulability Studies with Gaussian-Average Interference

In this subsection, we show all graphs depicting task systems with $r_{i:j}$ values determined under the Gaussian-average approach, where

$$r_{i:j} = \frac{s_i + f_j}{2}.$$

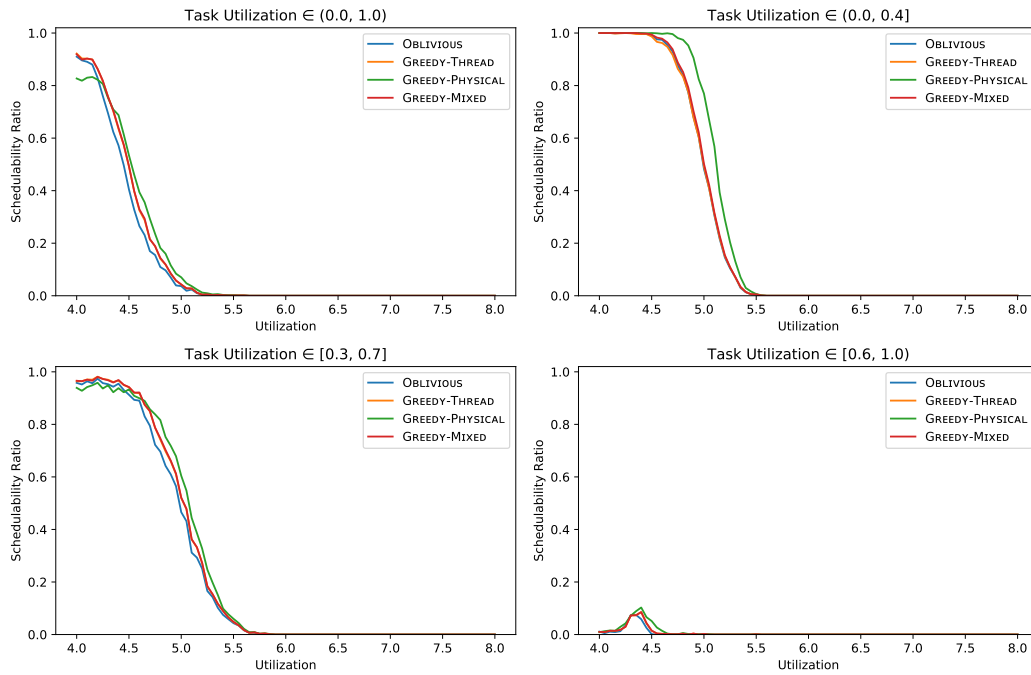
Strength scores are drawn from the Gaussian distributions $(0.72, 0.13)$, $(0.72, 0.26)$, and $(0.72, 0.39)$; friendliness scores from the distributions $(0.72, 0.04)$, $(0.72, 0.08)$, and $(0.72, 0.12)$. The graphs show data for all possible combinations of core count, per-task utilization ranges, and friendliness and strength distributions.

4 Cores, $s \in \text{Gauss}(0.72, 0.13)$, $f \in \text{Gauss}(0.72, 0.04)$

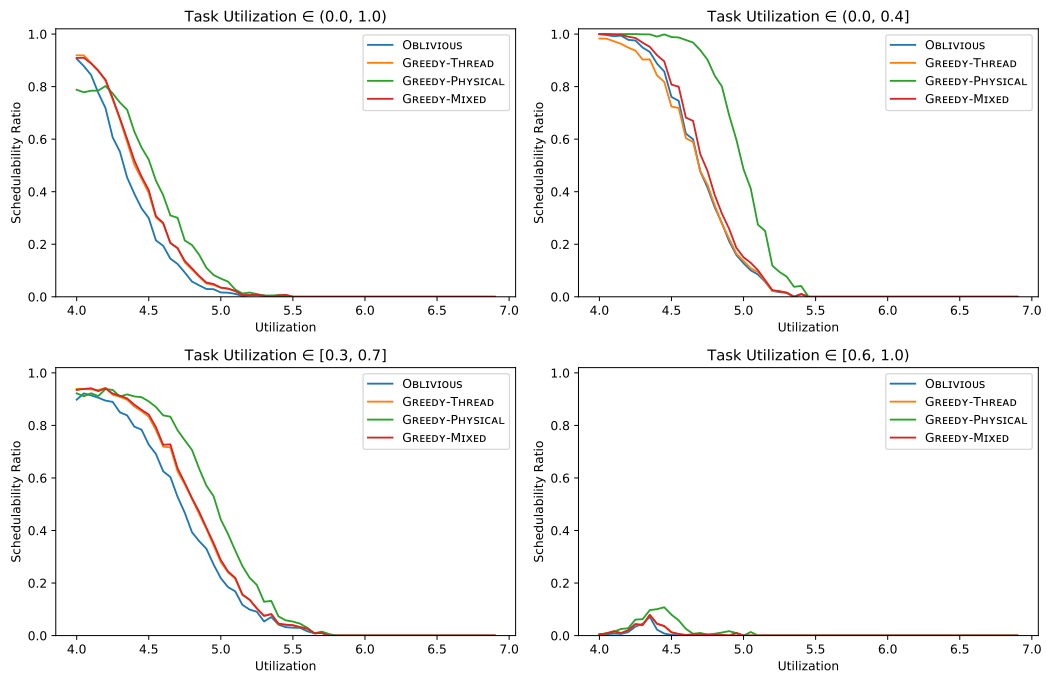


22:22 Simultaneous Multithreading Applied to Real Time

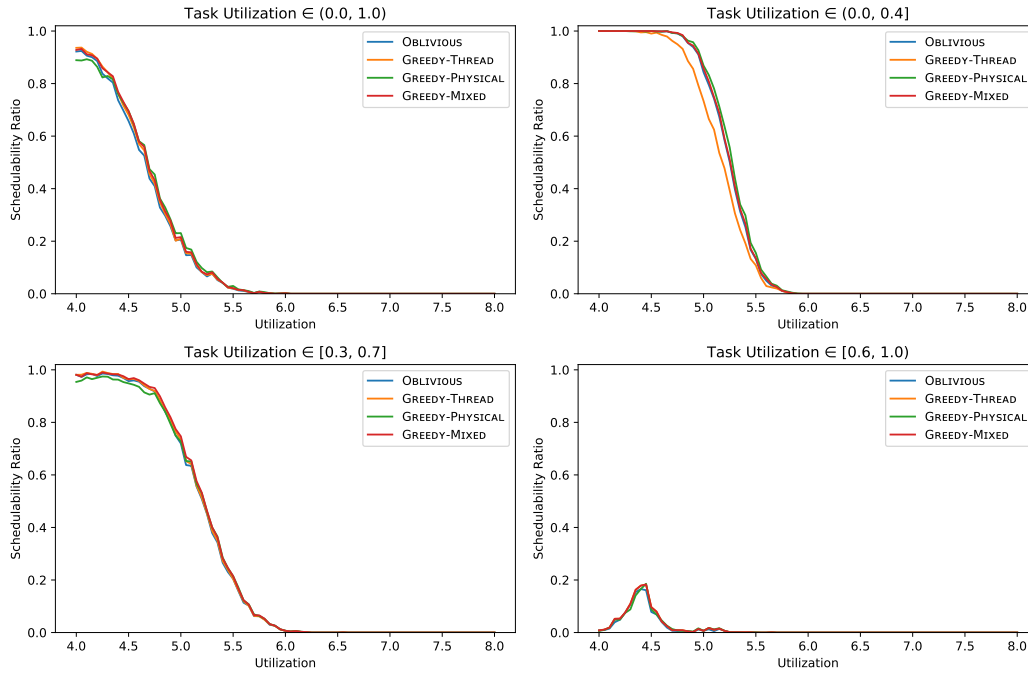
4 Cores, $s \in \text{Gauss}(0.72, 0.13)$, $f \in \text{Gauss}(0.72, 0.09)$



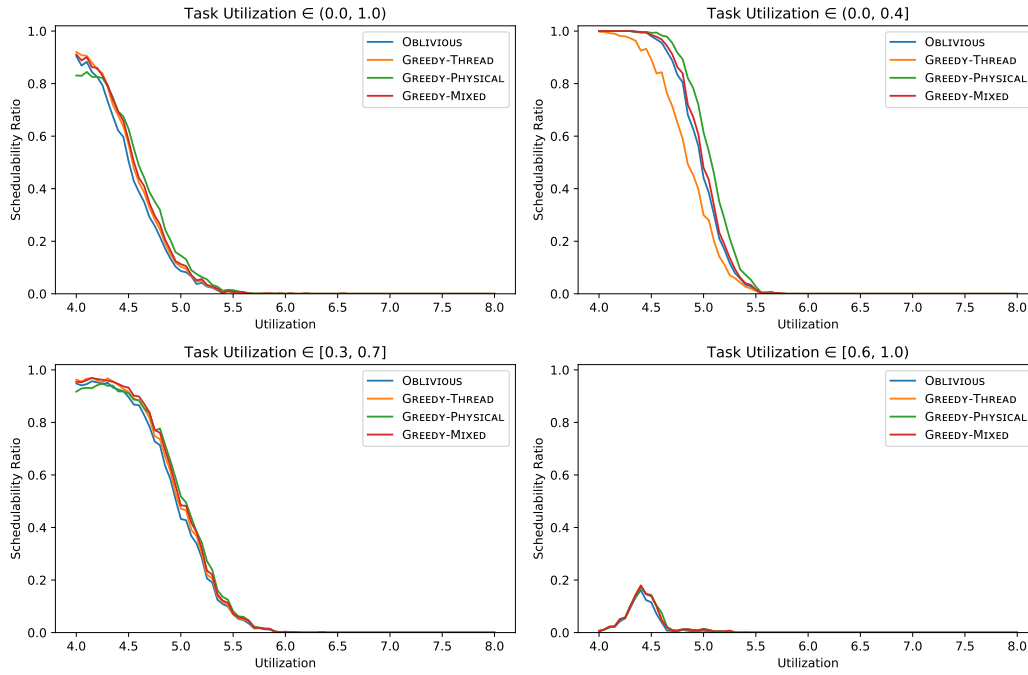
4 Cores, $s \in \text{Gauss}(0.72, 0.13)$, $f \in \text{Gauss}(0.72, 0.13)$



4 Cores, $s \in \text{Gauss}(0.72, 0.26)$, $f \in \text{Gauss}(0.72, 0.04)$

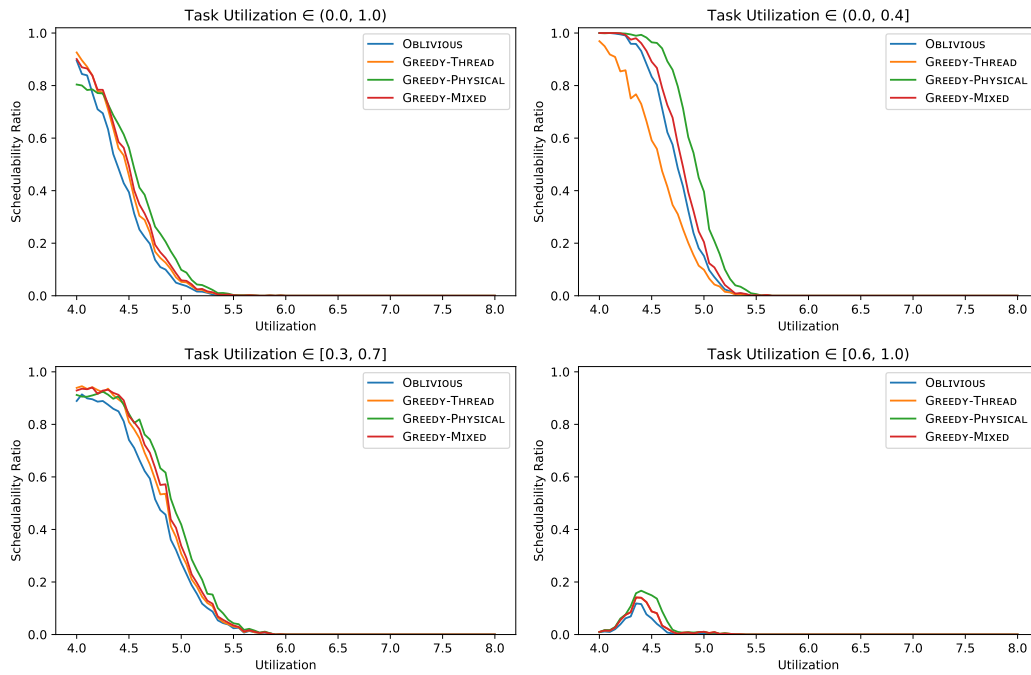


4 Cores, $s \in \text{Gauss}(0.72, 0.26)$, $f \in \text{Gauss}(0.72, 0.09)$

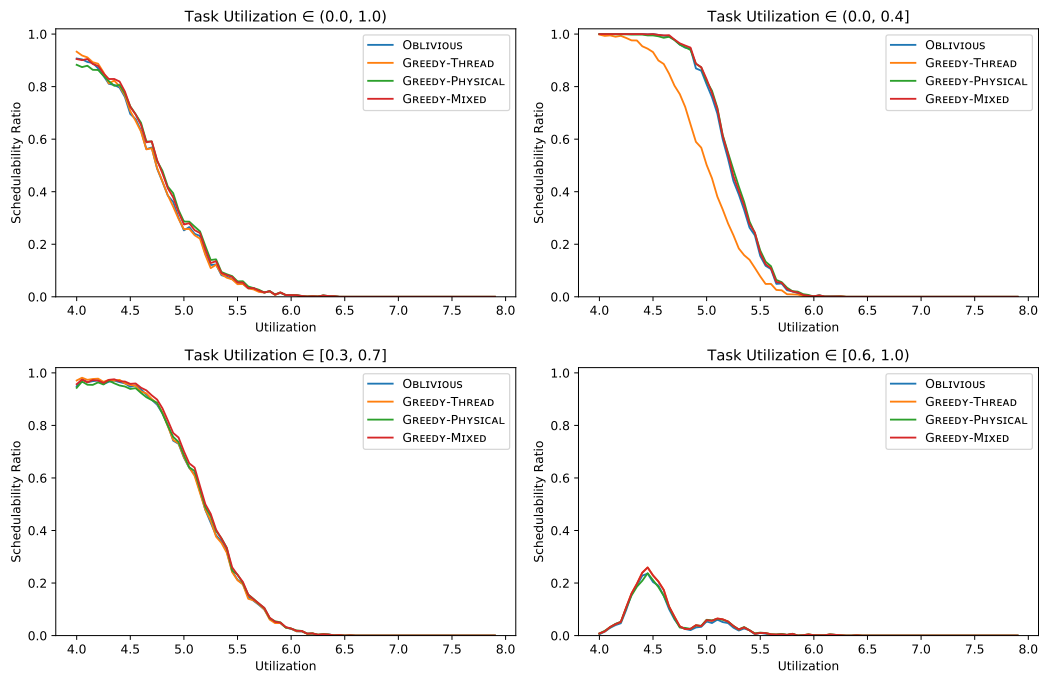


22:24 Simultaneous Multithreading Applied to Real Time

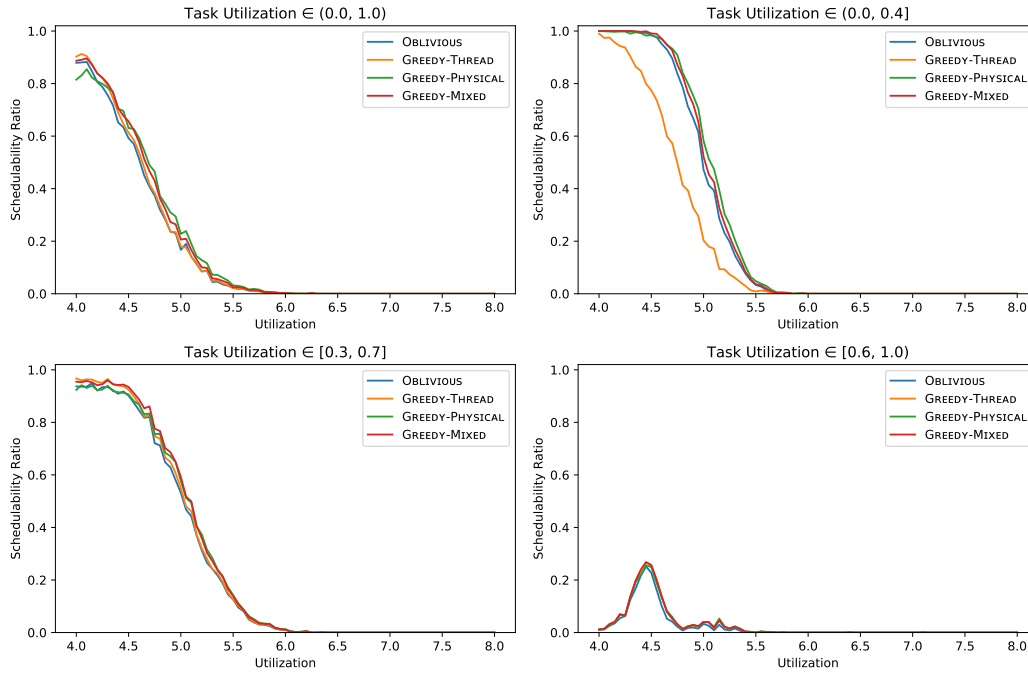
4 Cores, $s \in \text{Gauss}(0.72, 0.26)$, $f \in \text{Gauss}(0.72, 0.13)$



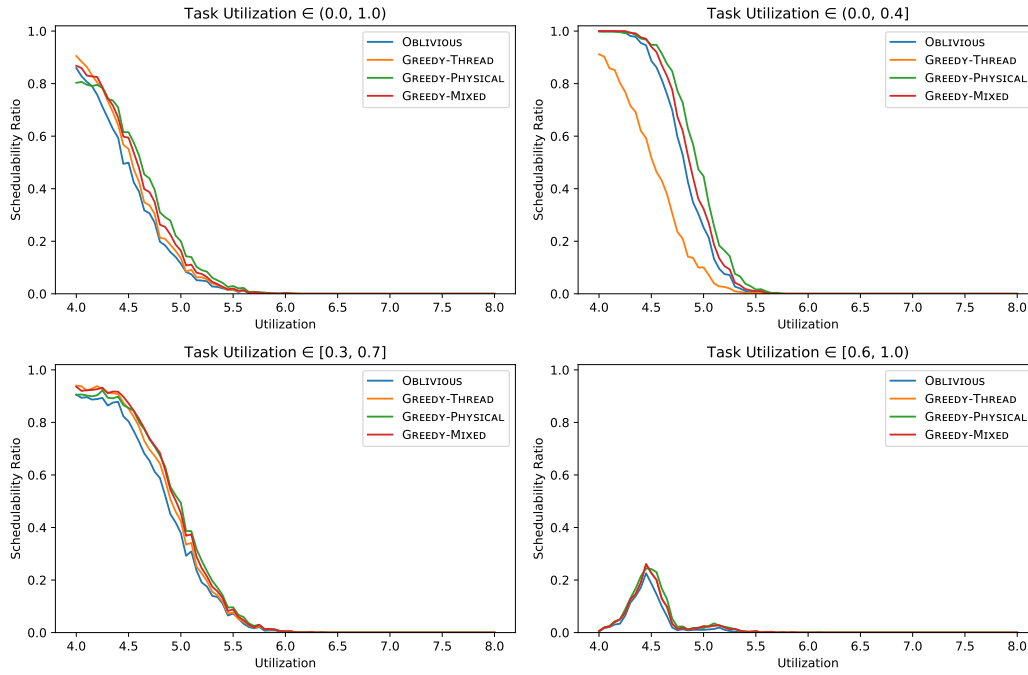
4 Cores, $s \in \text{Gauss}(0.72, 0.39)$, $f \in \text{Gauss}(0.72, 0.04)$



4 Cores, $s \in \text{Gauss}(0.72, 0.39)$, $f \in \text{Gauss}(0.72, 0.09)$

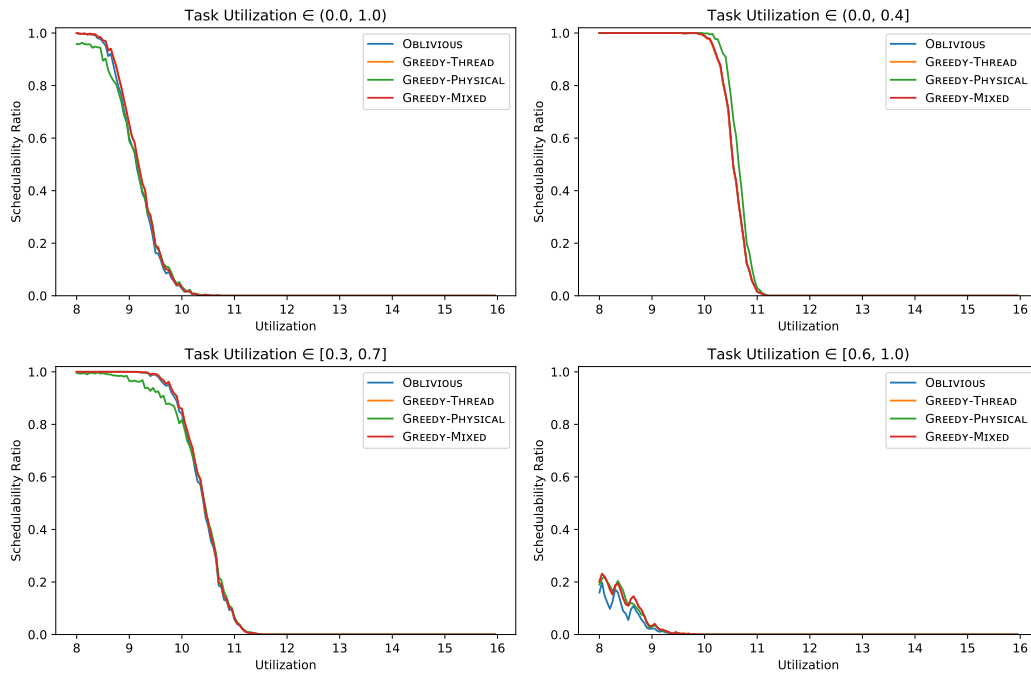


4 Cores, $s \in \text{Gauss}(0.72, 0.39)$, $f \in \text{Gauss}(0.72, 0.13)$

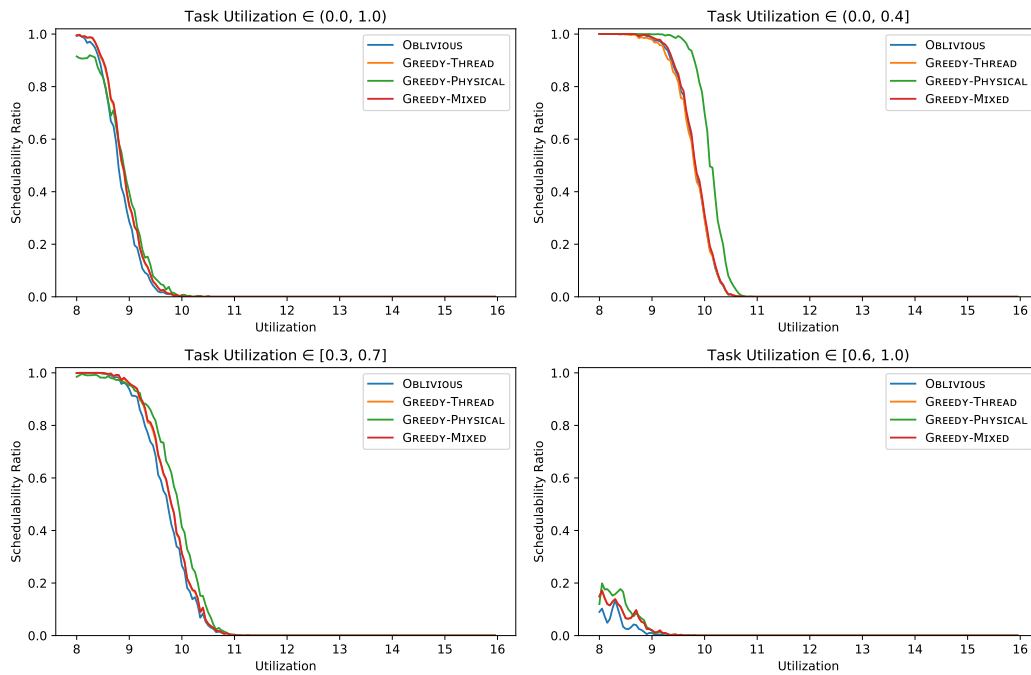


22:26 Simultaneous Multithreading Applied to Real Time

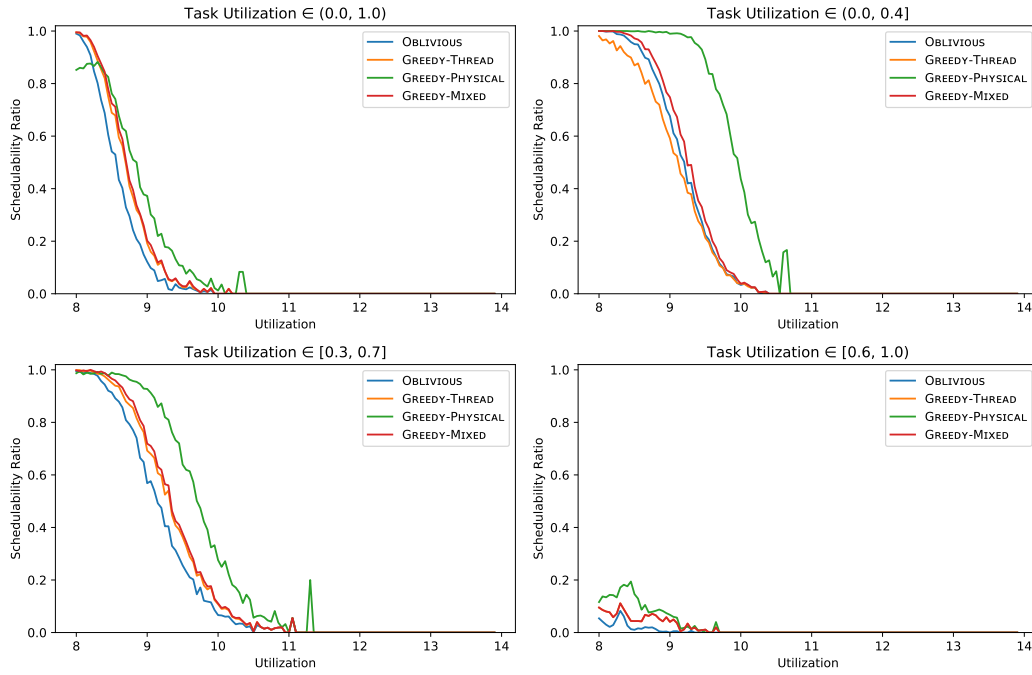
8 Cores, $s \in \text{Gauss}(0.72, 0.13)$, $f \in \text{Gauss}(0.72, 0.04)$



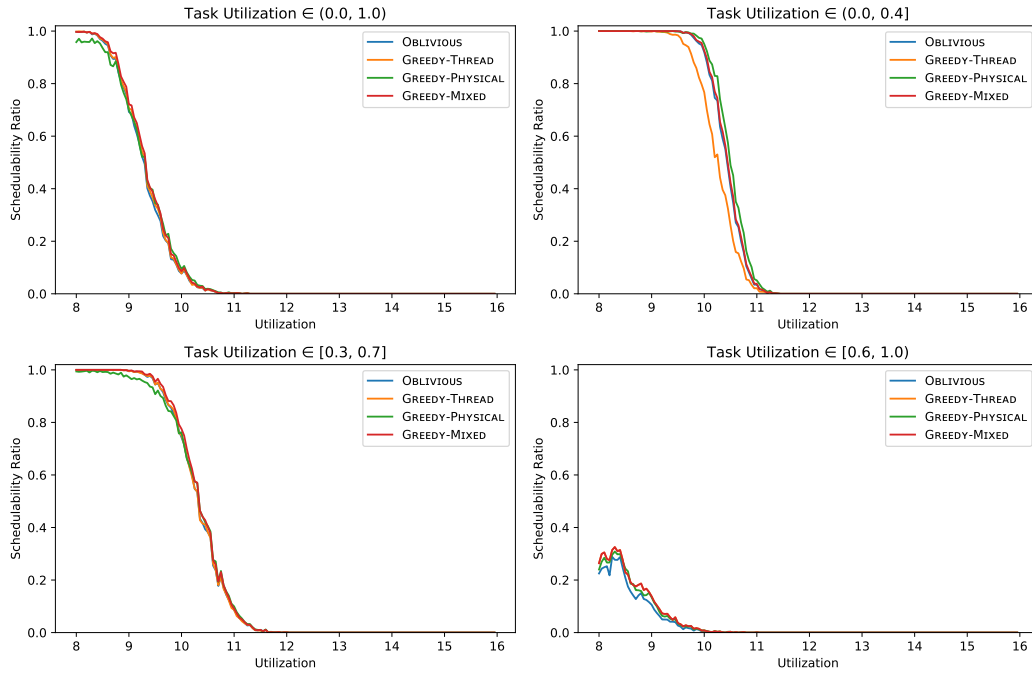
8 Cores, $s \in \text{Gauss}(0.72, 0.13)$, $f \in \text{Gauss}(0.72, 0.09)$



8 Cores, $s \in \text{Gauss}(0.72, 0.13)$, $f \in \text{Gauss}(0.72, 0.13)$

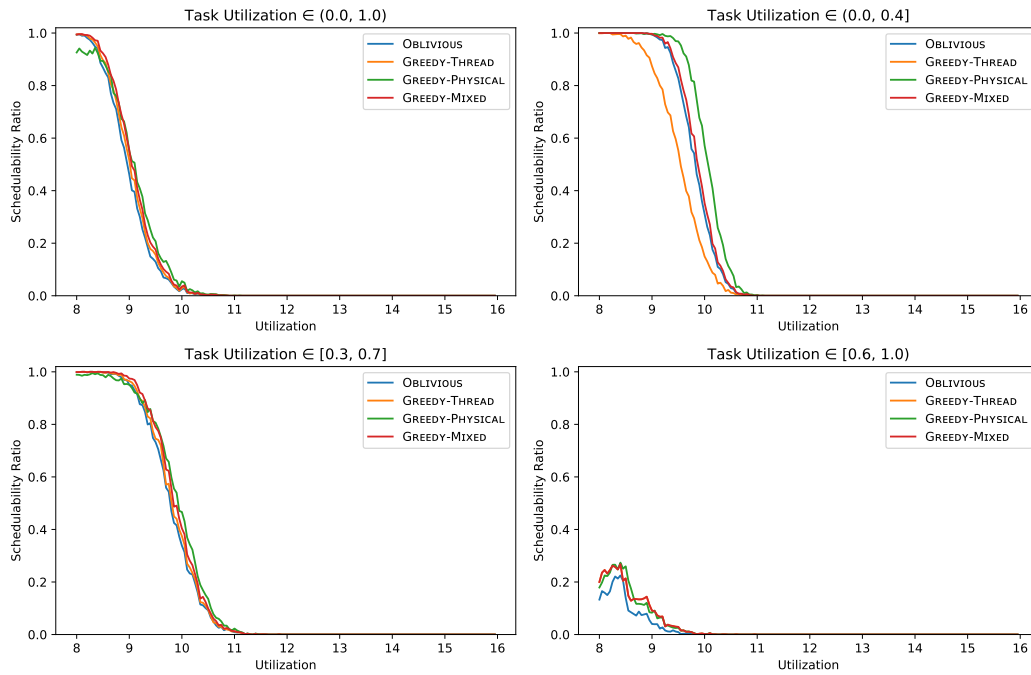


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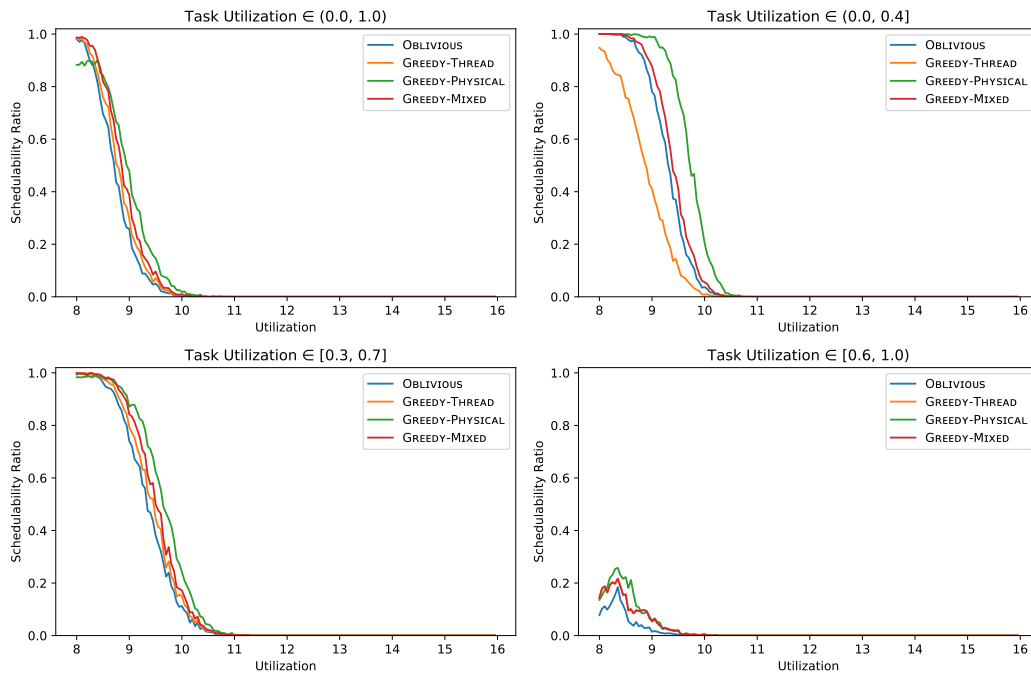


22:28 Simultaneous Multithreading Applied to Real Time

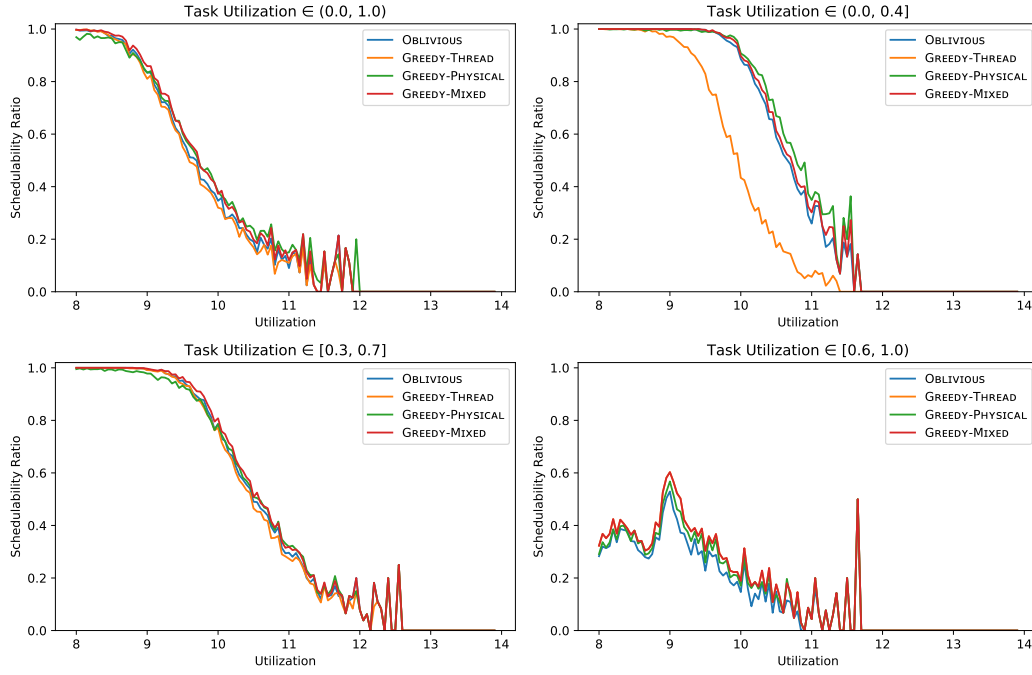
8 Cores, $s \in \text{Gauss}(0.72, 0.26)$, $f \in \text{Gauss}(0.72, 0.09)$



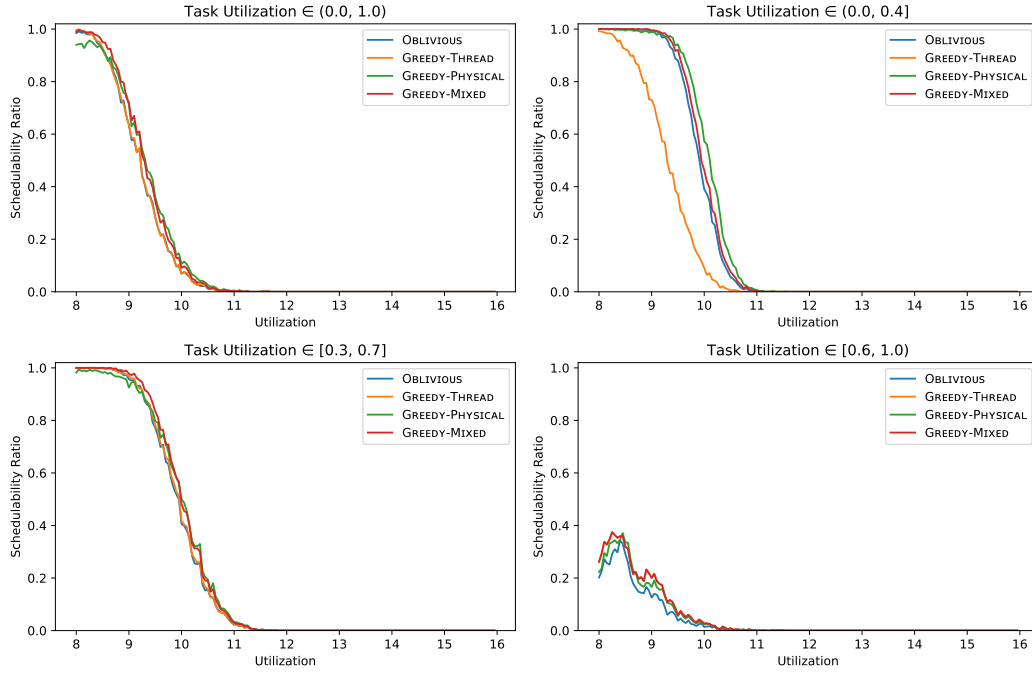
8 Cores, $s \in \text{Gauss}(0.72, 0.26)$, $f \in \text{Gauss}(0.72, 0.13)$



8 Cores, $s \in \text{Gauss}(0.72, 0.39)$, $f \in \text{Gauss}(0.72, 0.04)$

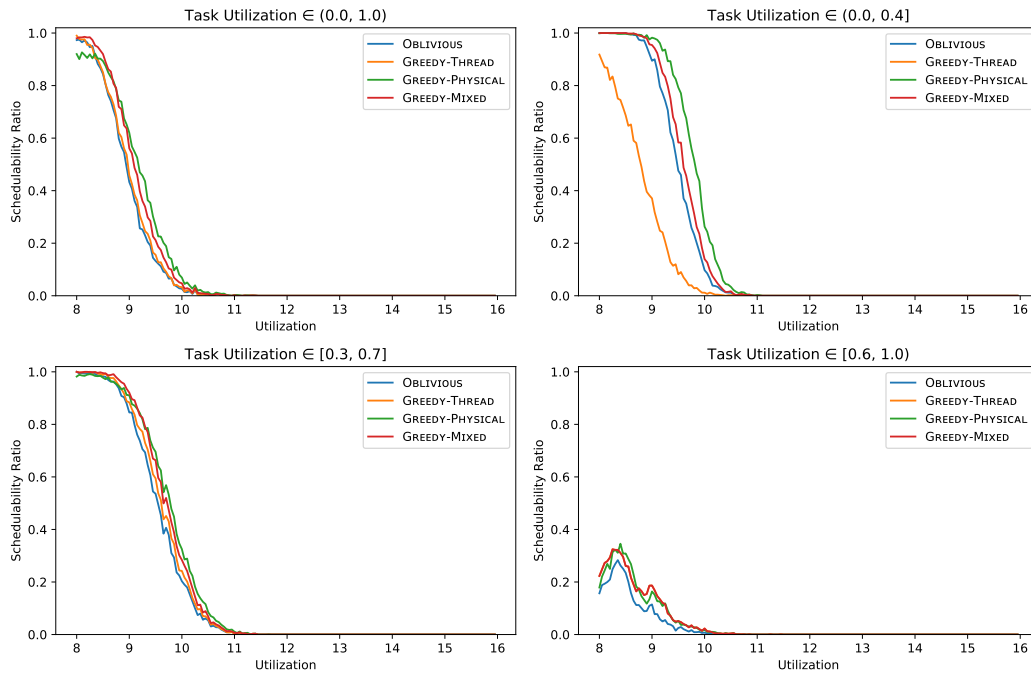


8 Cores, $s \in \text{Gauss}(0.72, 0.39)$, $f \in \text{Gauss}(0.72, 0.09)$

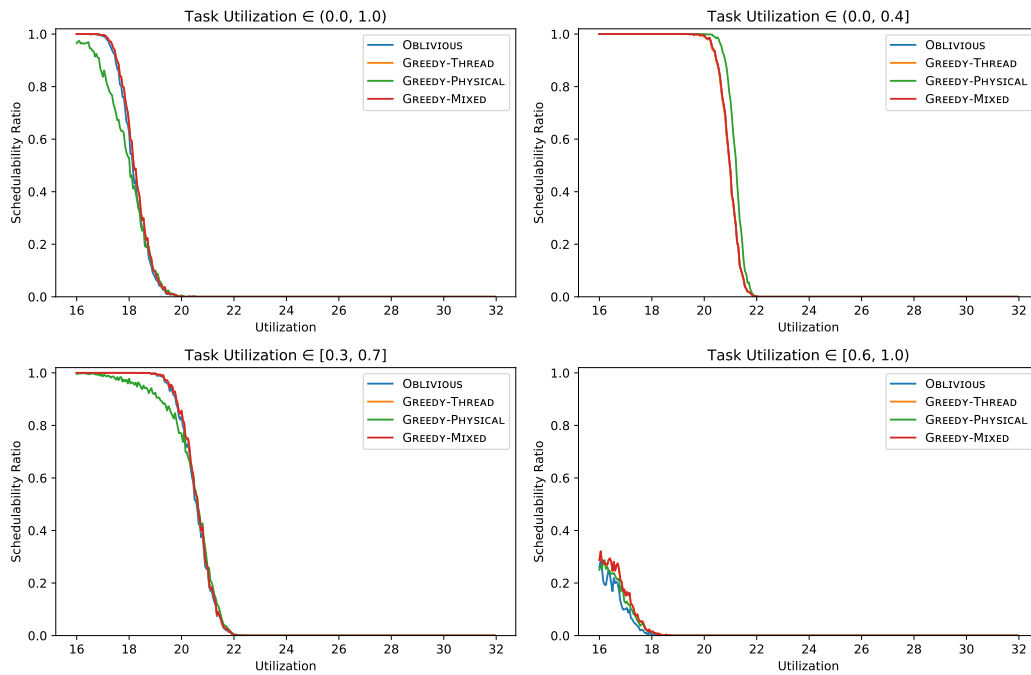


22:30 Simultaneous Multithreading Applied to Real Time

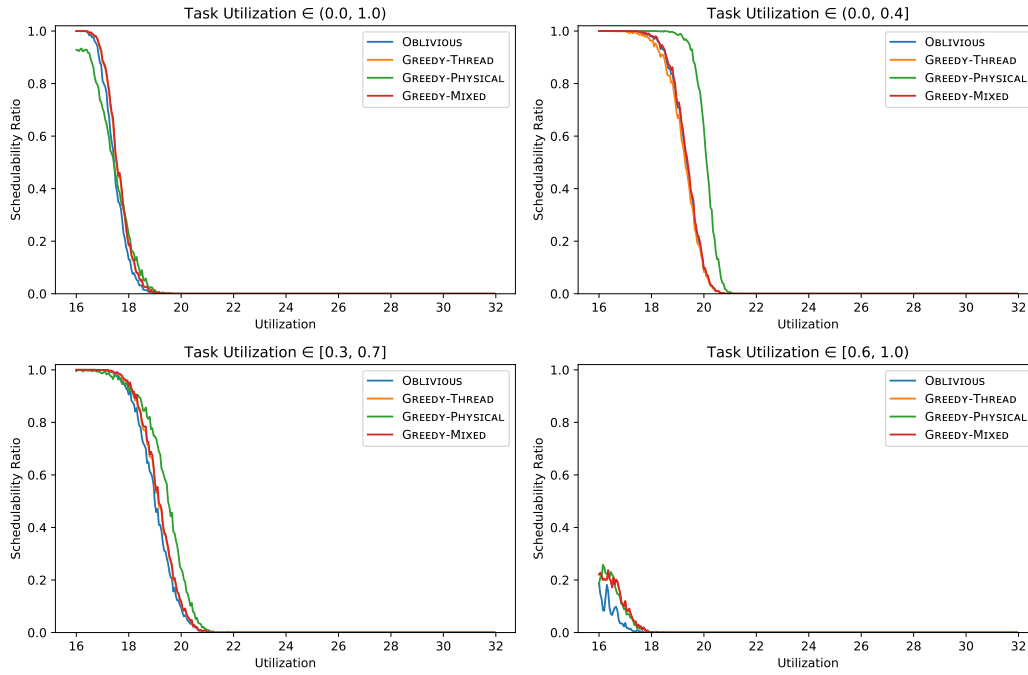
8 Cores, $s \in \text{Gauss}(0.72, 0.39)$, $f \in \text{Gauss}(0.72, 0.13)$



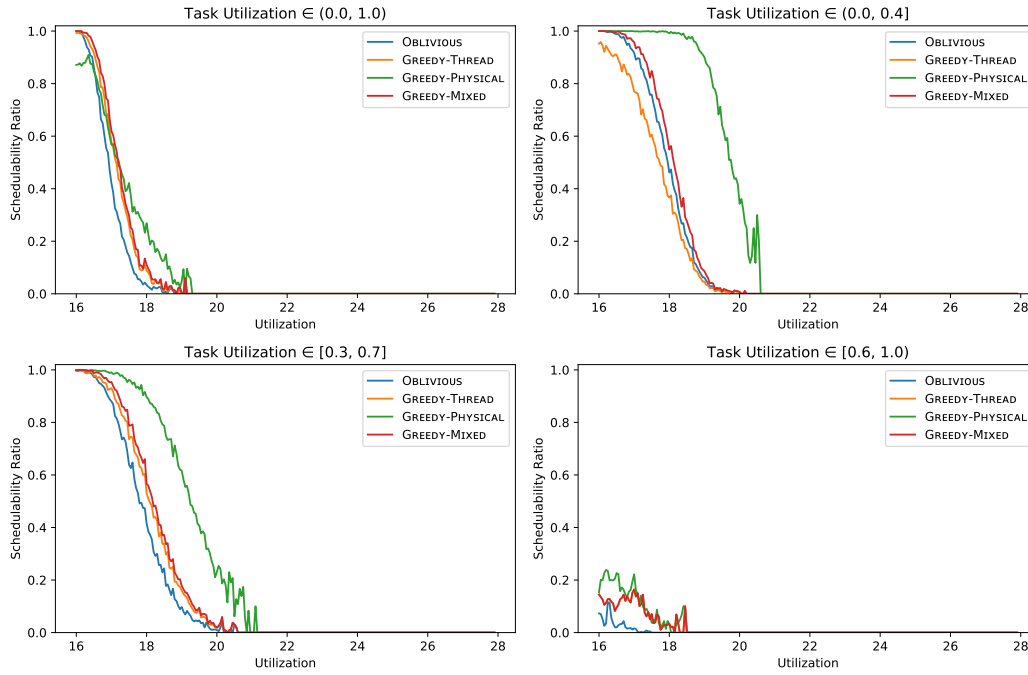
16 Cores, $s \in \text{Gauss}(0.72, 0.13)$, $f \in \text{Gauss}(0.72, 0.04)$



16 Cores, $s \in \text{Gauss}(0.72, 0.13)$, $f \in \text{Gauss}(0.72, 0.09)$

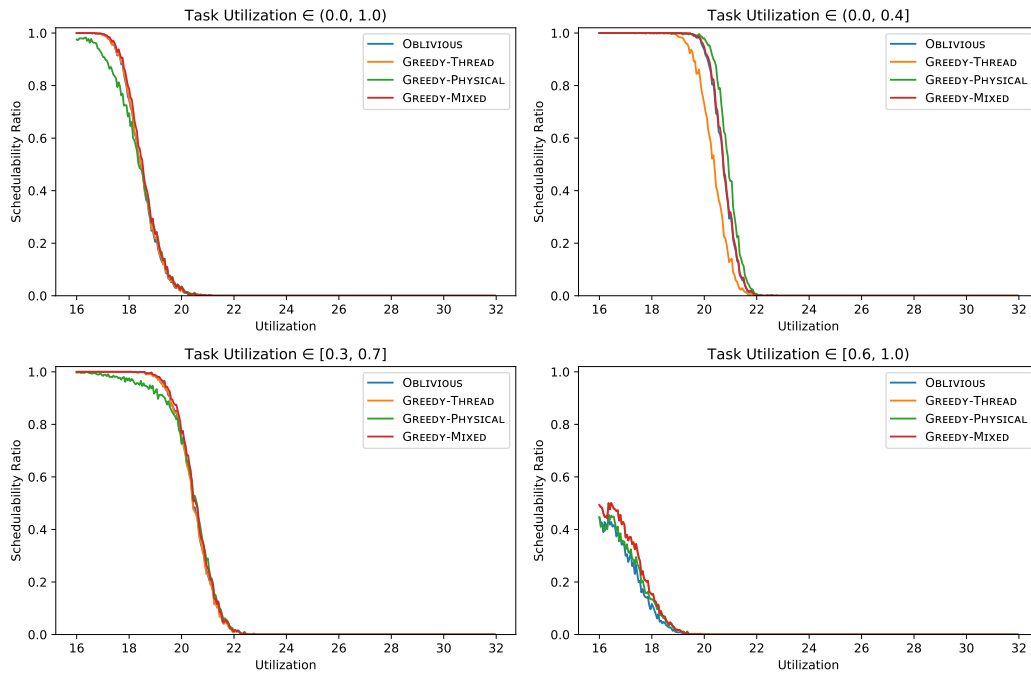


16 Cores, $s \in \text{Gauss}(0.72, 0.13)$, $f \in \text{Gauss}(0.72, 0.13)$

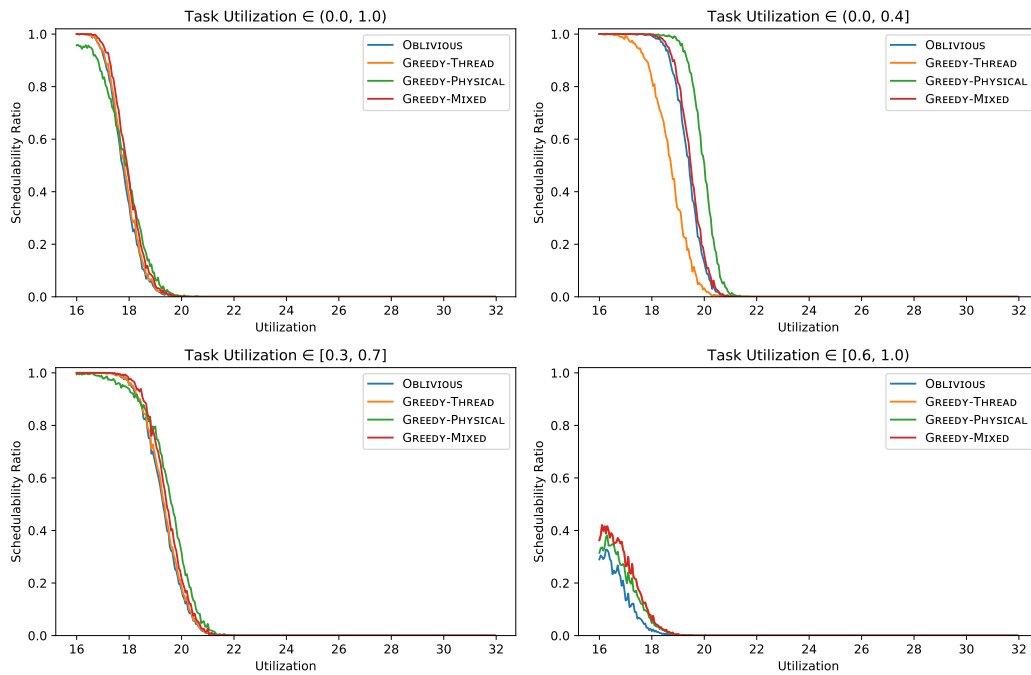


22:32 Simultaneous Multithreading Applied to Real Time

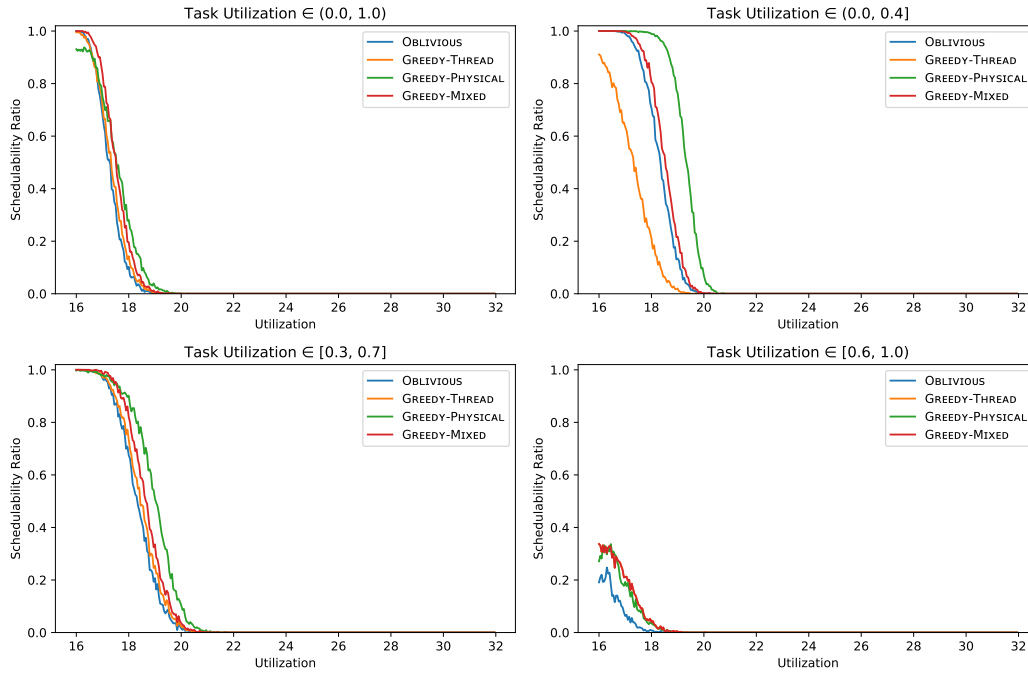
16 Cores, $s \in \text{Gauss}(0.72, 0.26)$, $f \in \text{Gauss}(0.72, 0.04)$



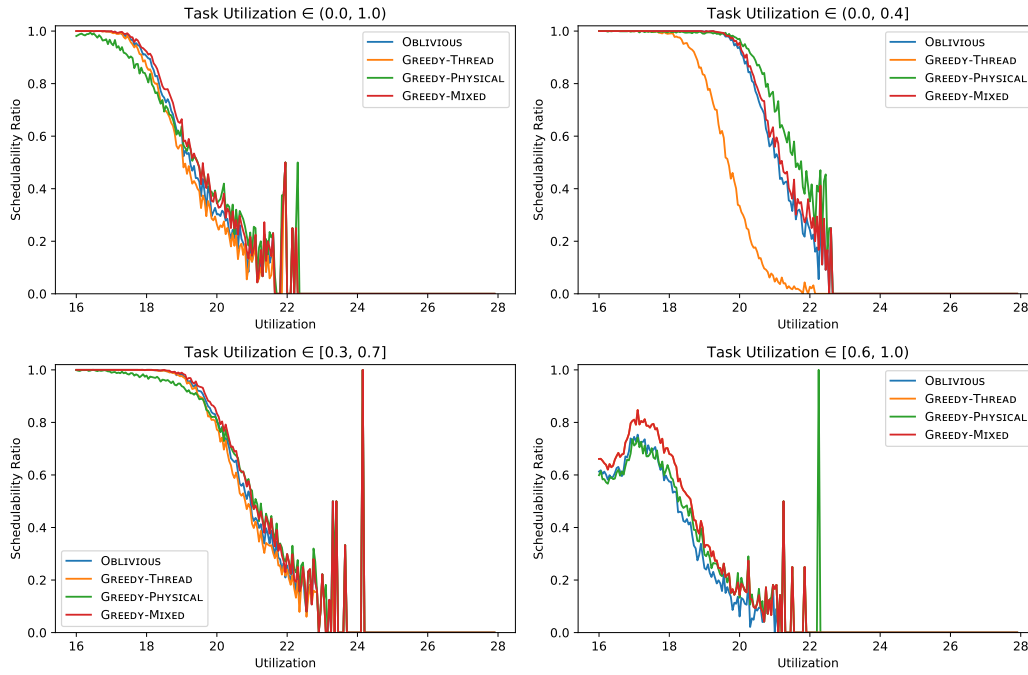
16 Cores, $s \in \text{Gauss}(0.72, 0.26)$, $f \in \text{Gauss}(0.72, 0.09)$



16 Cores, $s \in \text{Gauss}(0.72, 0.26)$, $f \in \text{Gauss}(0.72, 0.13)$

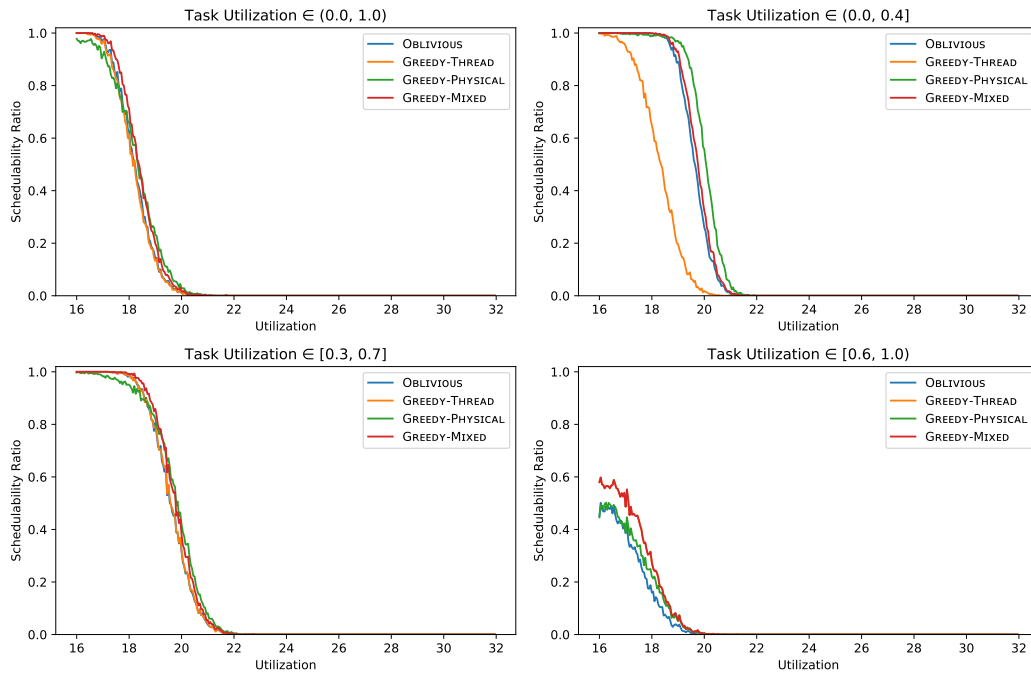


16 Cores, $s \in \text{Gauss}(0.72, 0.39)$, $f \in \text{Gauss}(0.72, 0.04)$

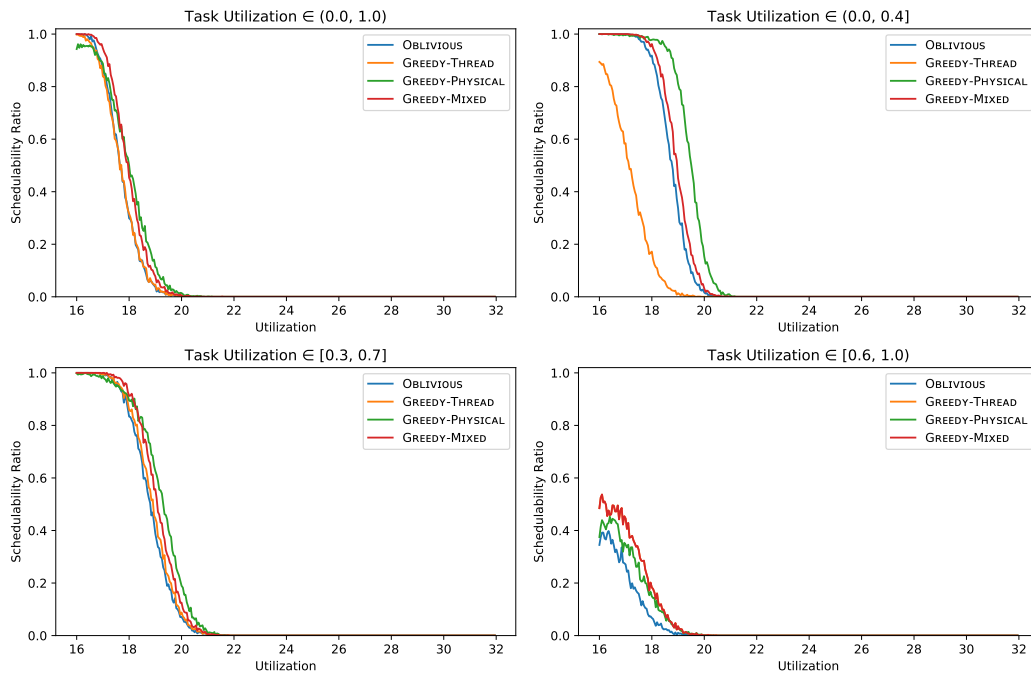


22:34 Simultaneous Multithreading Applied to Real Time

16 Cores, $s \in \text{Gauss}(0.72, 0.39)$, $f \in \text{Gauss}(0.72, 0.09)$



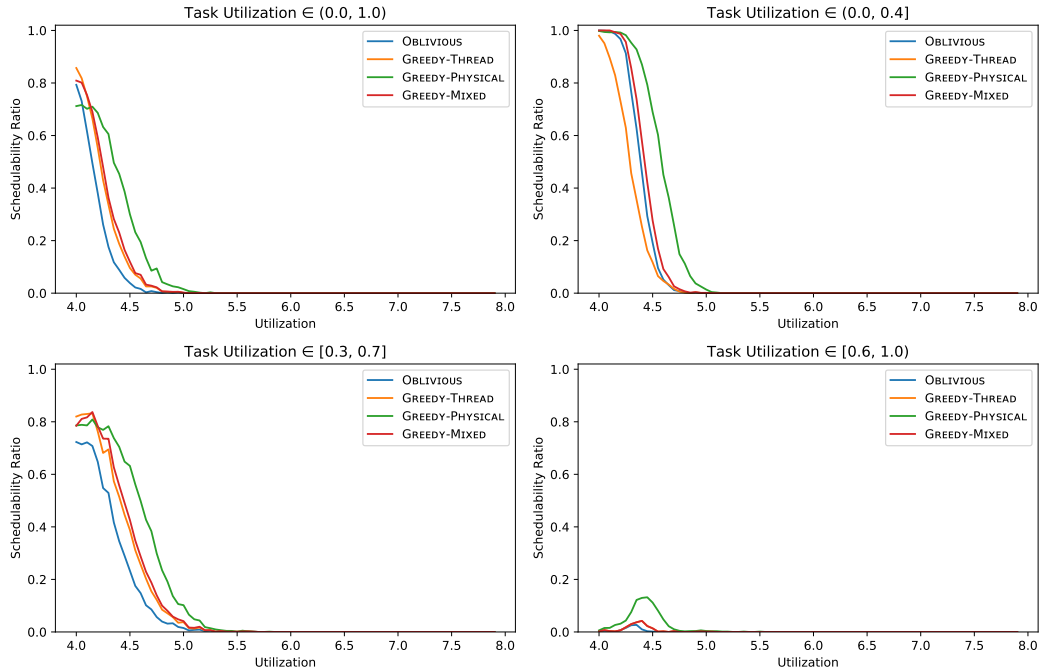
16 Cores, $s \in \text{Gauss}(0.72, 0.39)$, $f \in \text{Gauss}(0.72, 0.13)$



A.2 Schedulability Studies with Uniform-Normal Interference

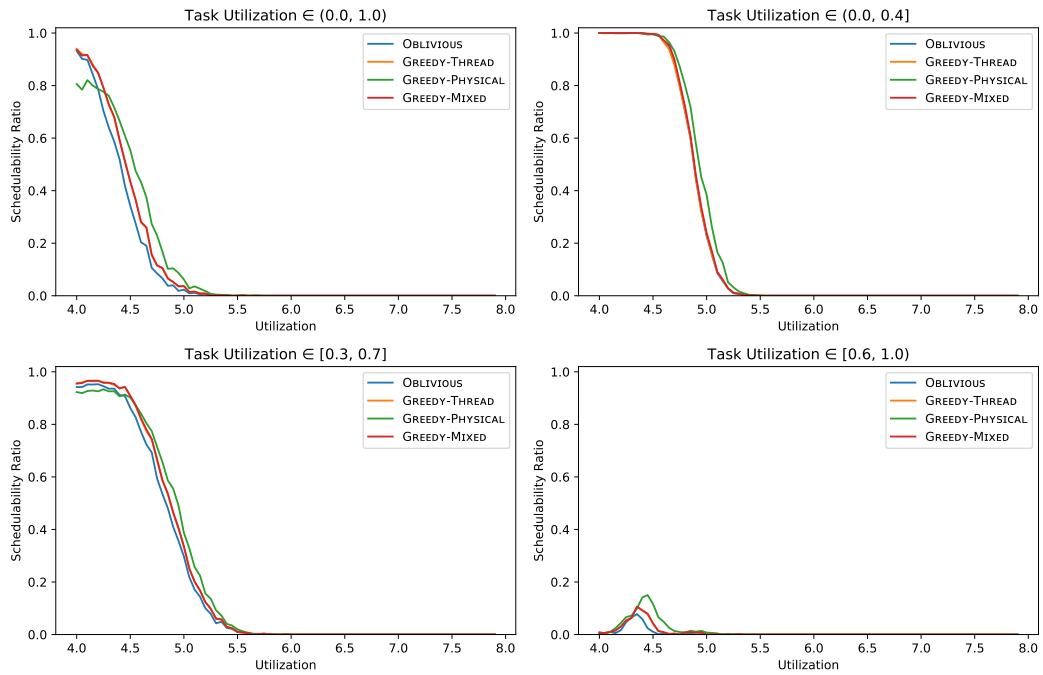
In this subsection we show graphs depicting task systems with $r_{i:j}$ values determined by first selecting uniformly-distributed strength and friendliness scores and then selecting $r_{i:j}$ from a normal distribution with mean $s_i f_j$ and standard deviation of either 0.01, 0.055, or 0.1. All combinations of strength distributions, friendliness distributions, standard deviations, per-task utilization distributions, and core counts are shown.

4 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$

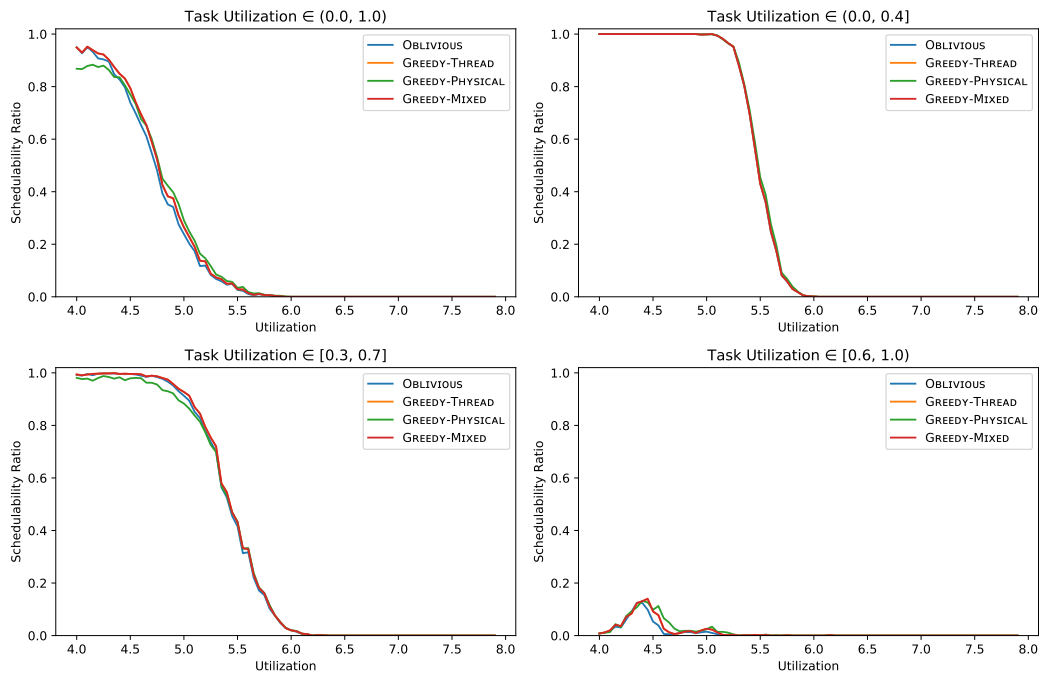


22:36 Simultaneous Multithreading Applied to Real Time

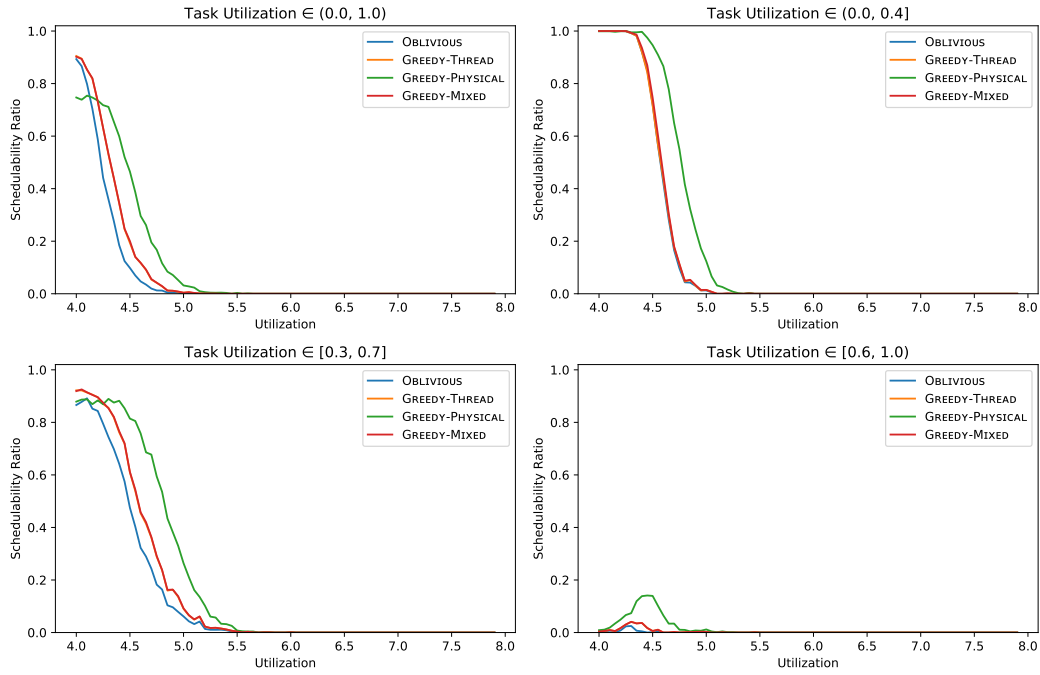
4 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$



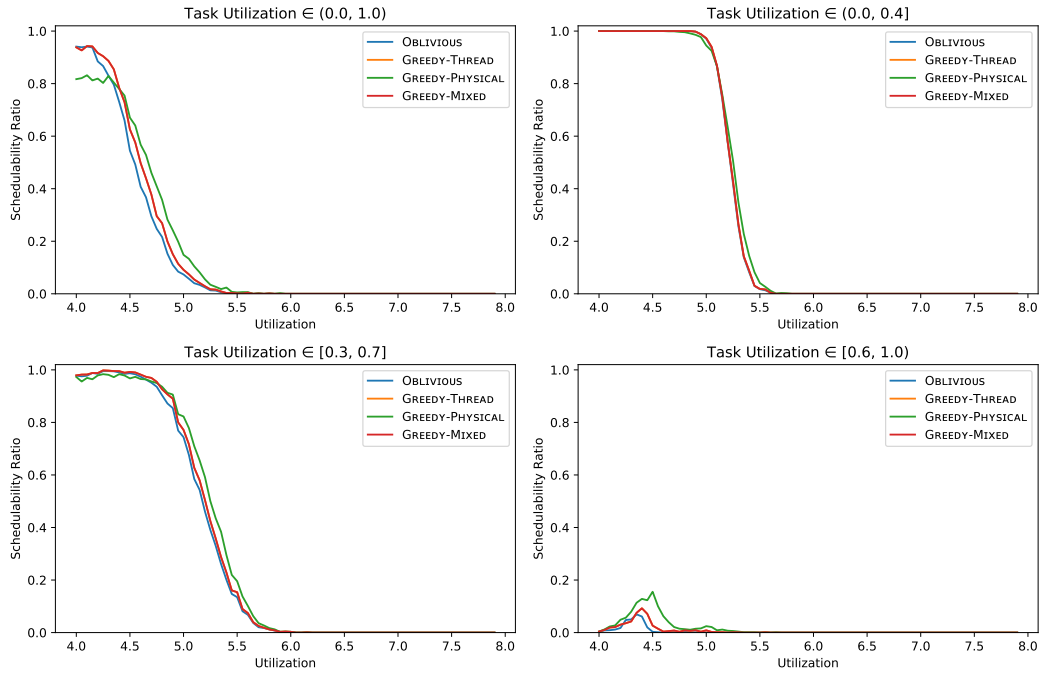
4 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$



4 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$

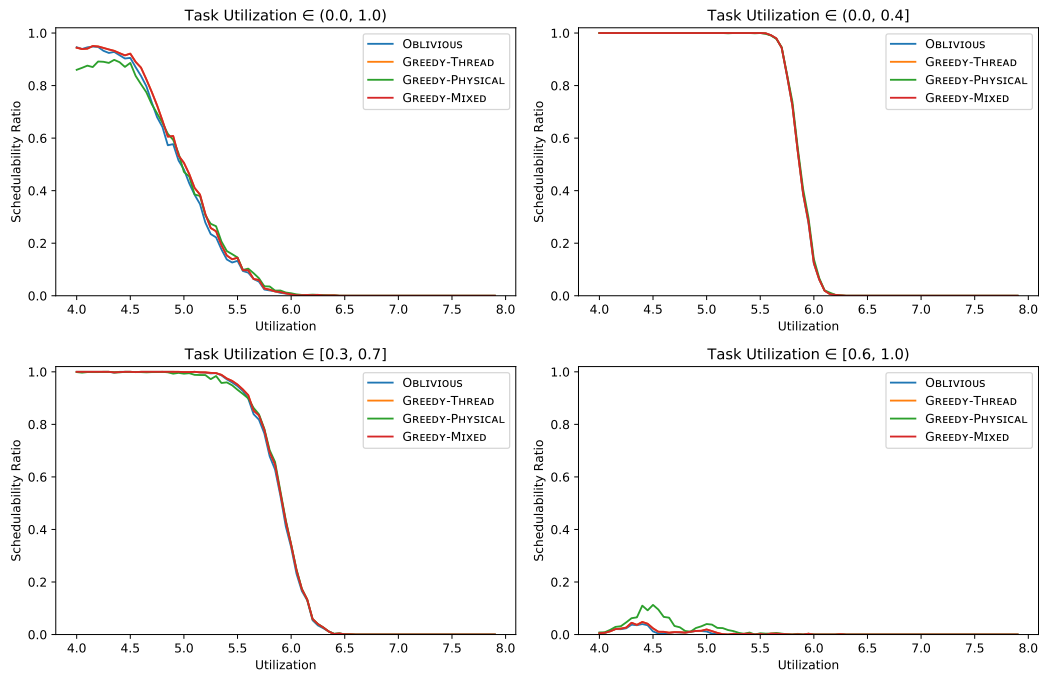


4 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$

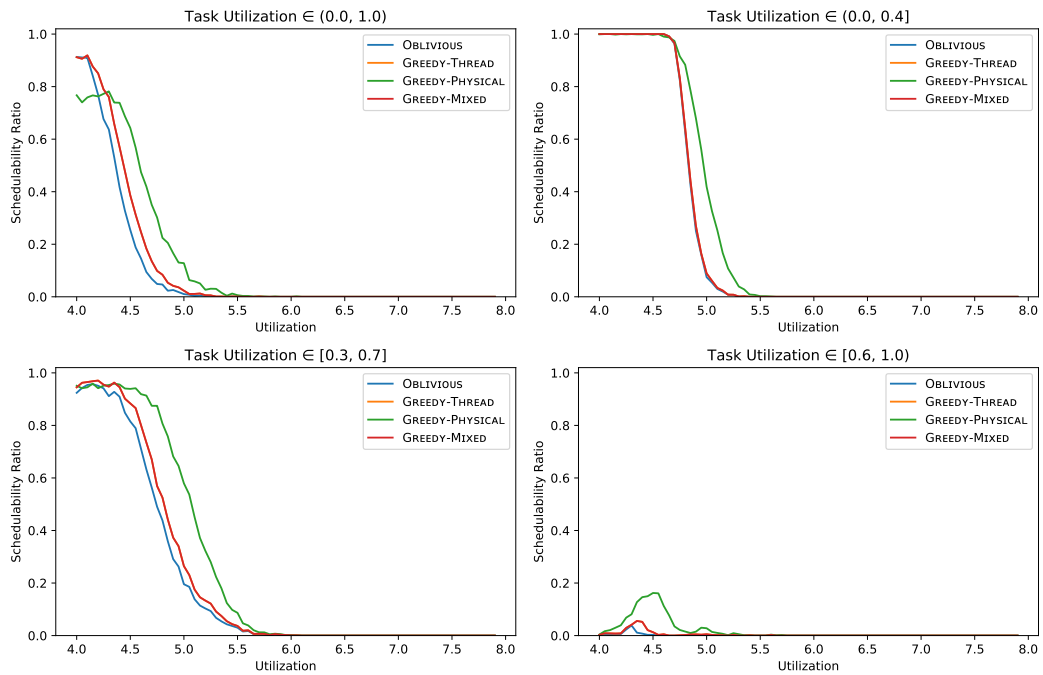


22:38 Simultaneous Multithreading Applied to Real Time

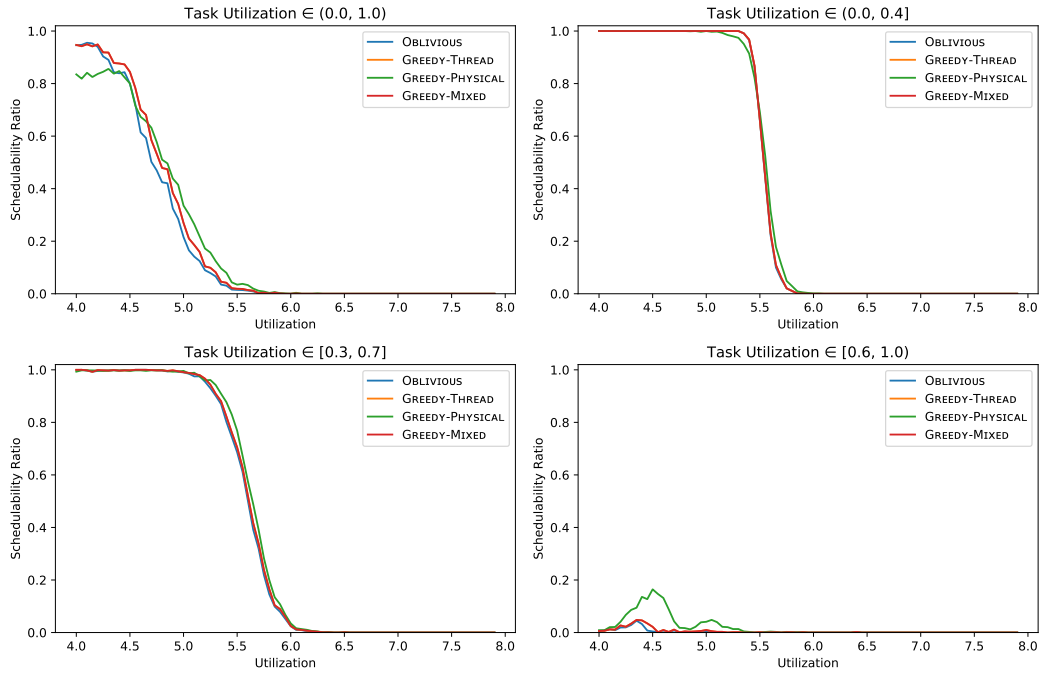
4 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$



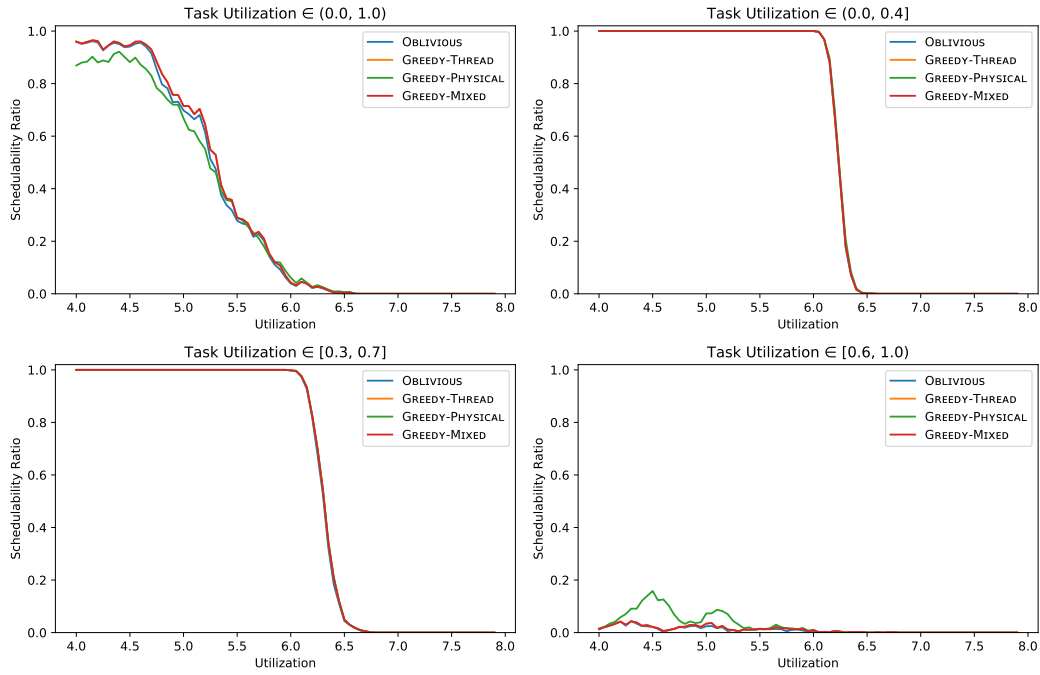
4 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$



4 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$

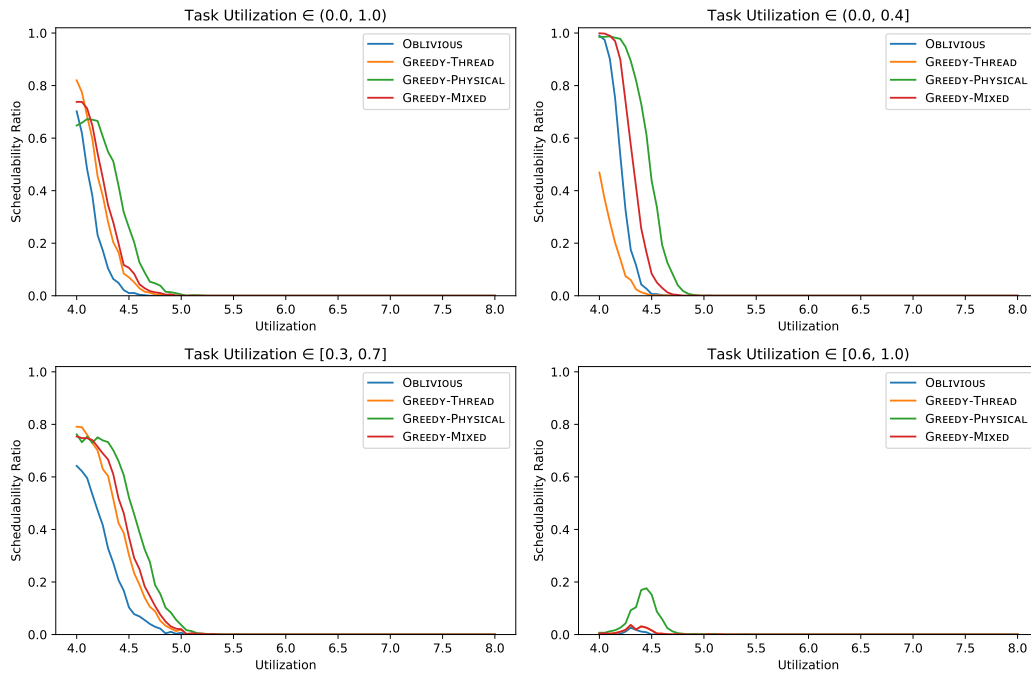


4 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$

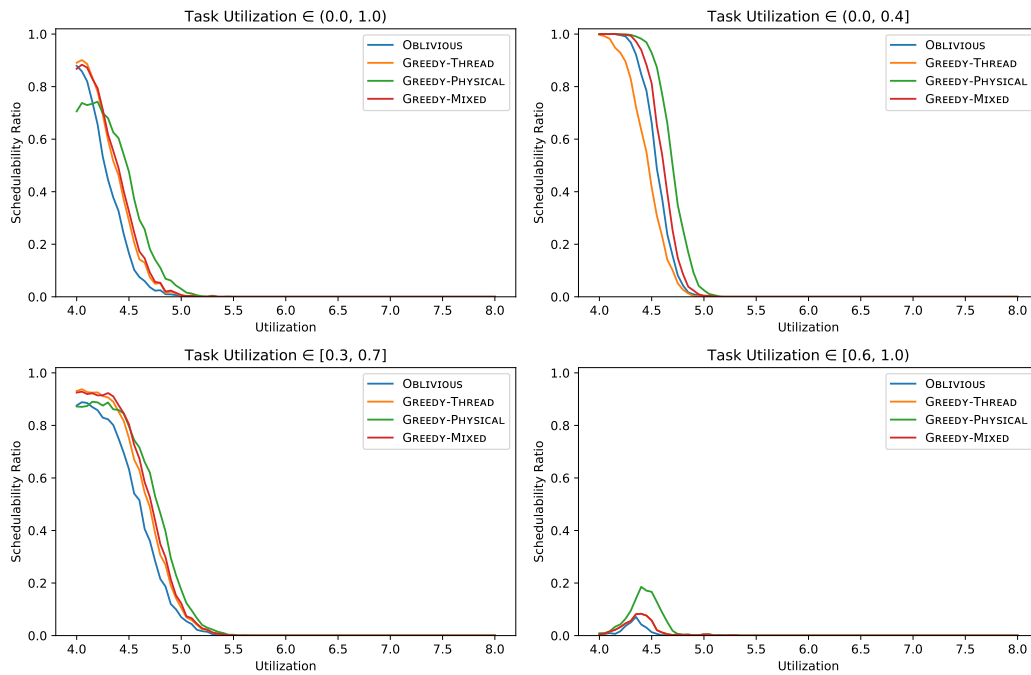


22:40 Simultaneous Multithreading Applied to Real Time

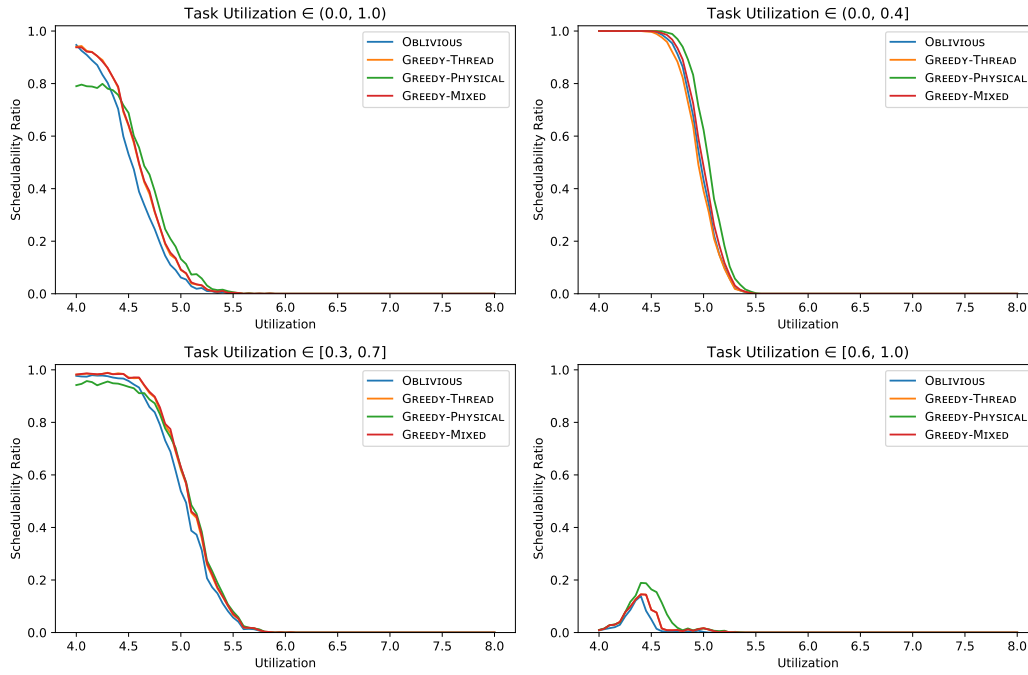
4 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$



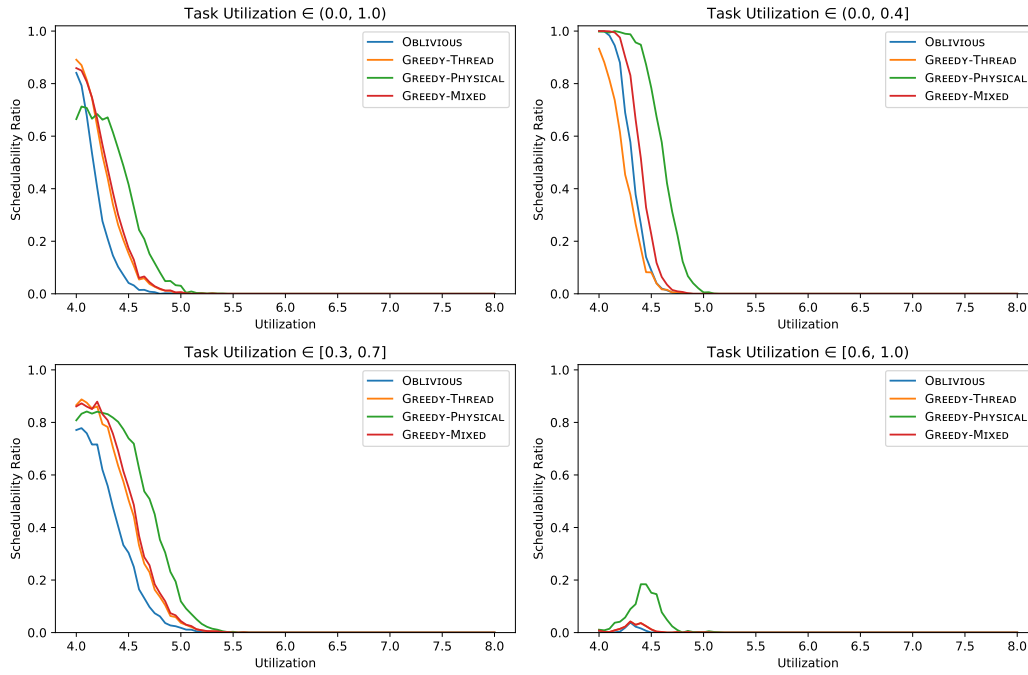
4 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$



4 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$

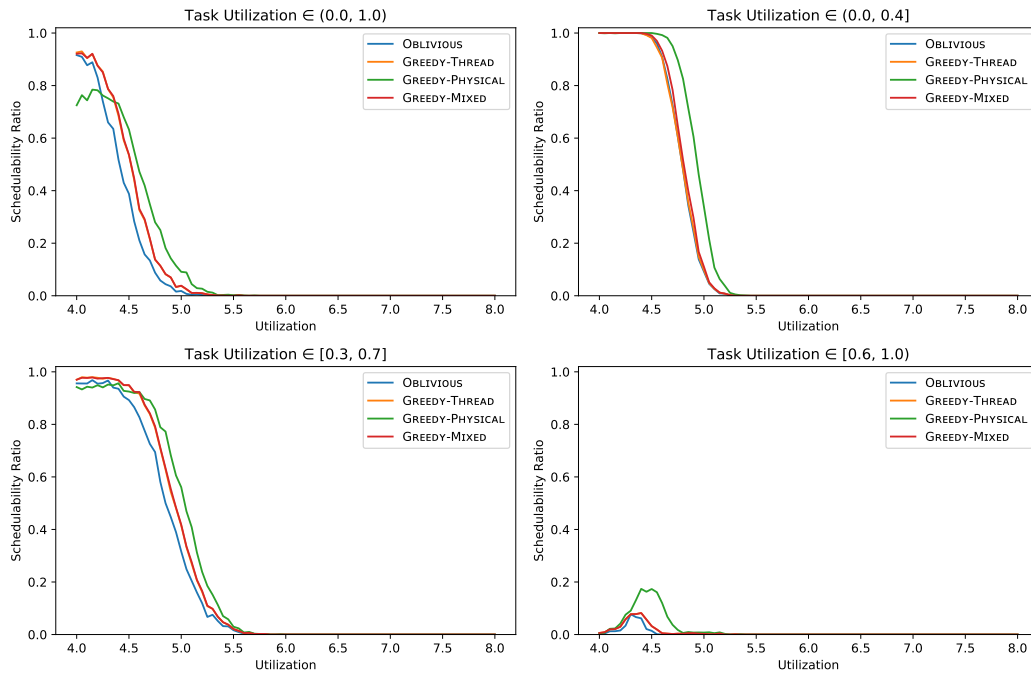


4 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$

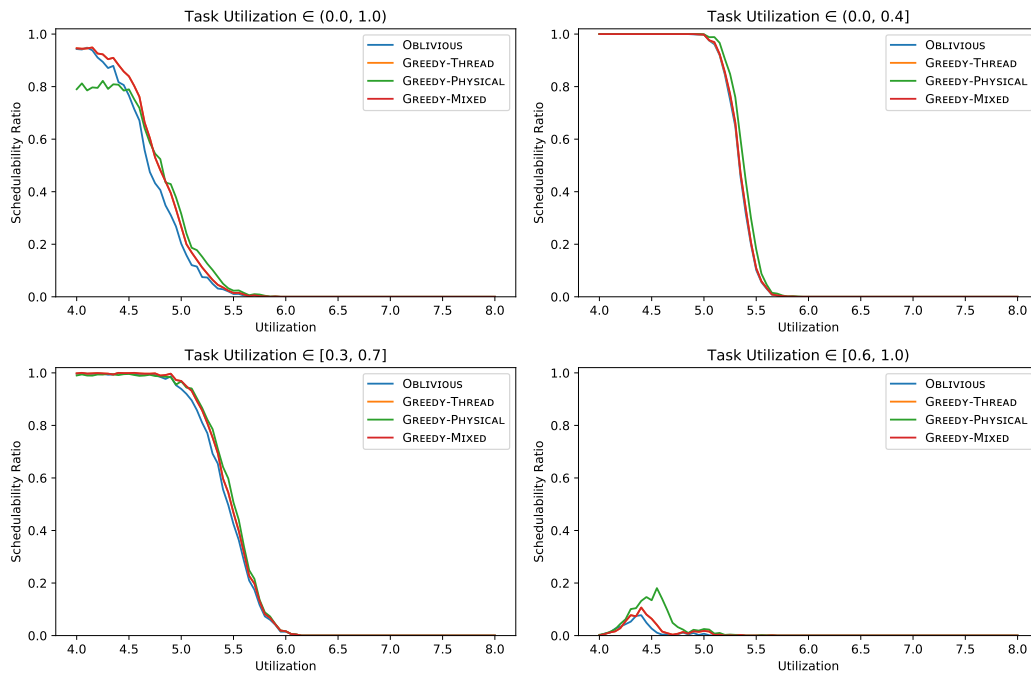


22:42 Simultaneous Multithreading Applied to Real Time

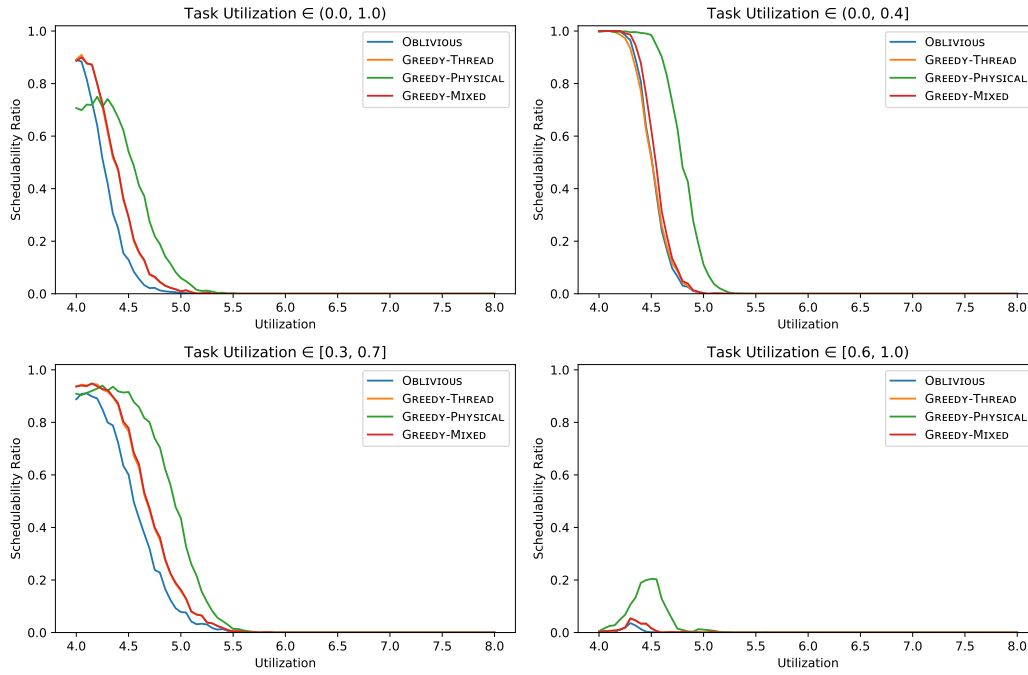
4 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$



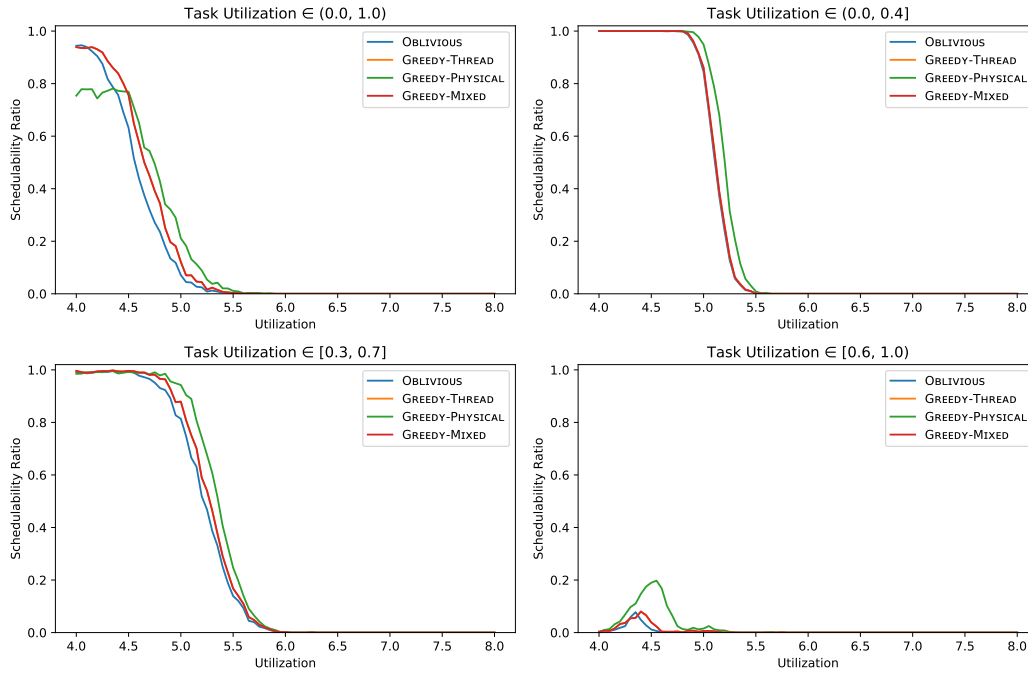
4 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$



4 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$

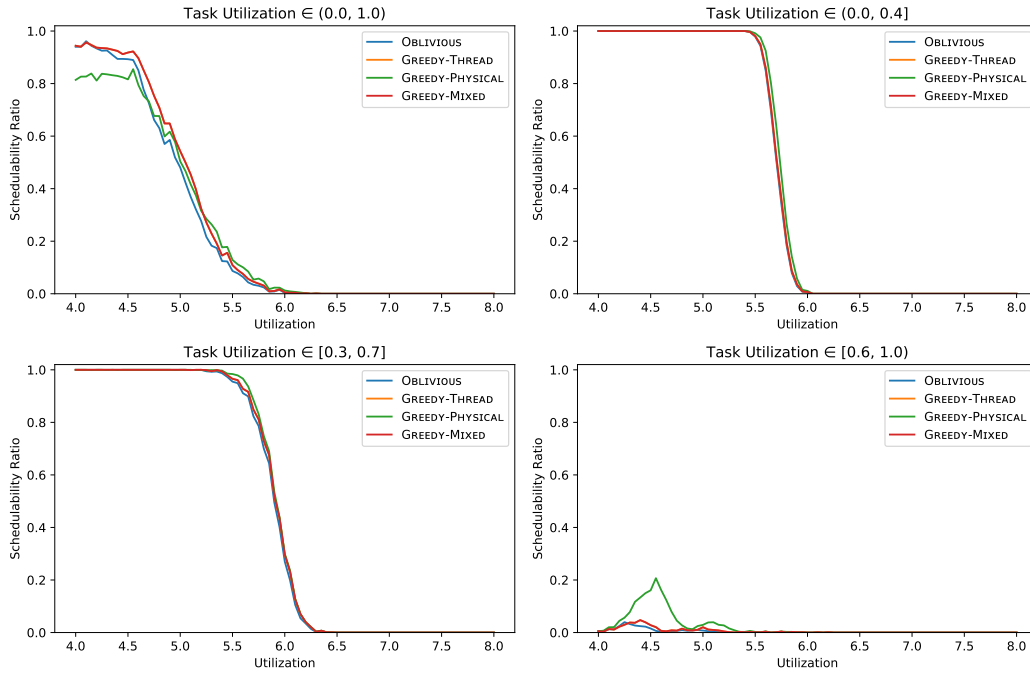


4 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$

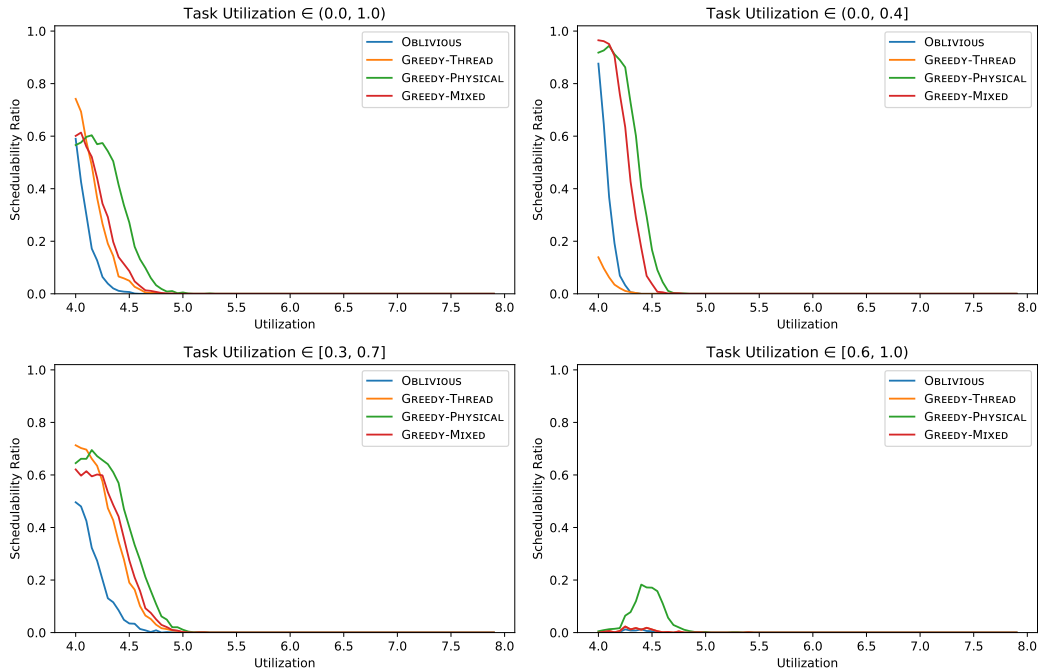


22:44 Simultaneous Multithreading Applied to Real Time

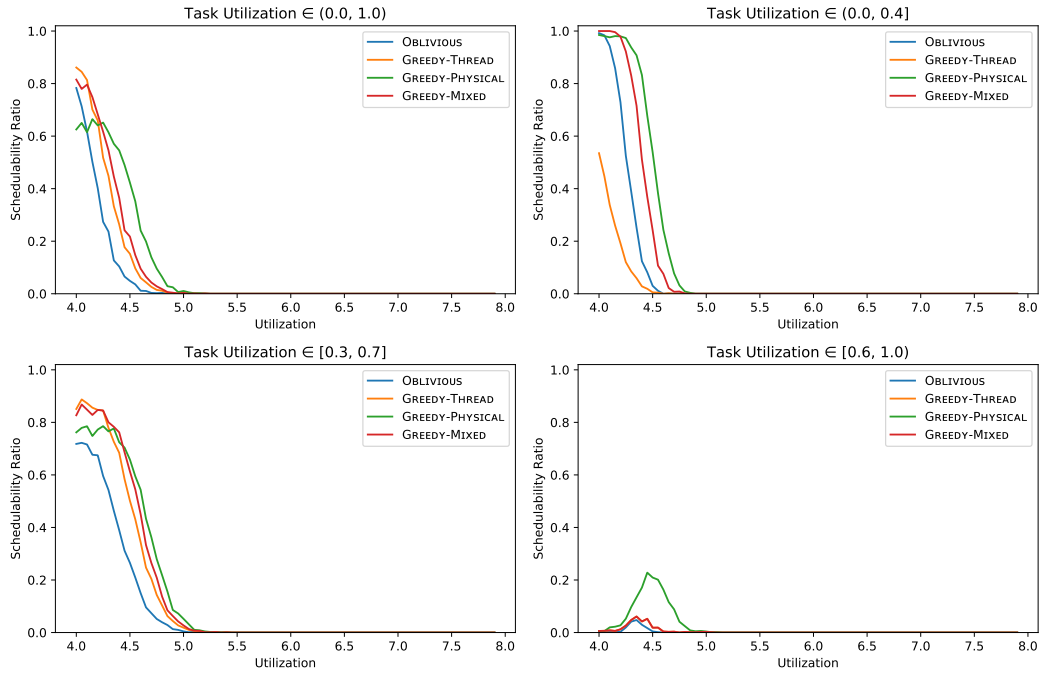
4 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$



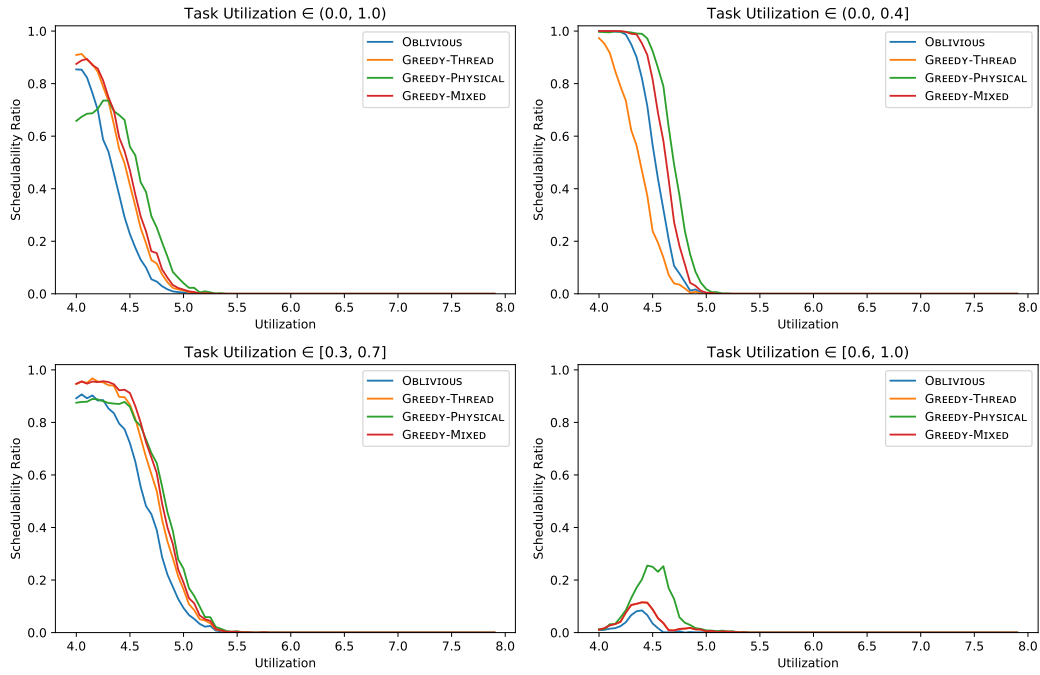
4 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$



4 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$

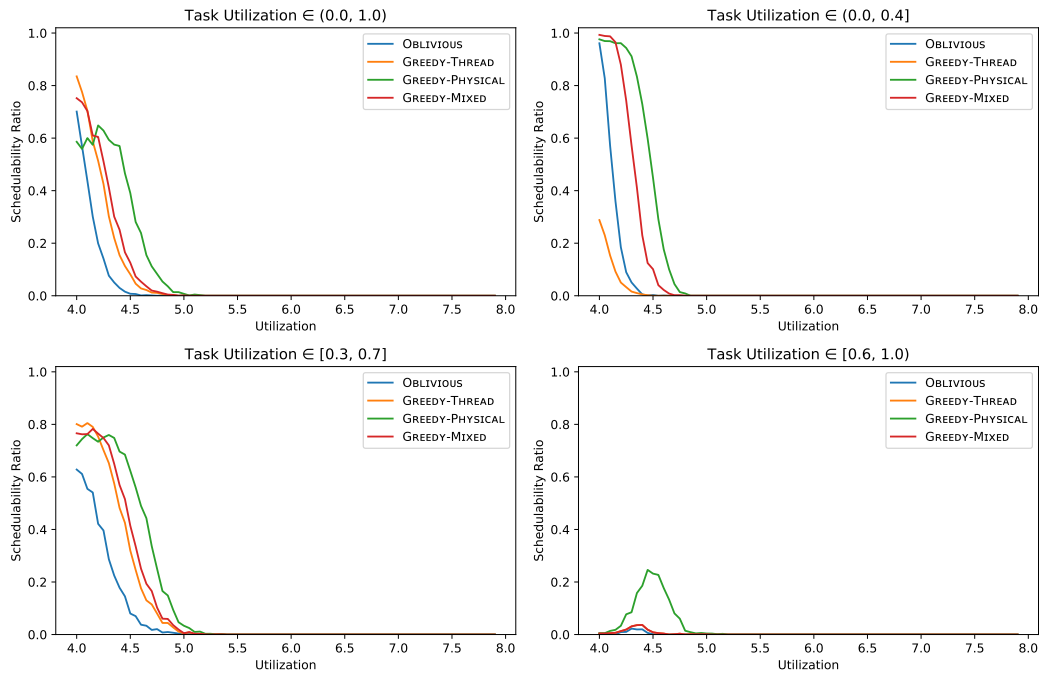


4 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$

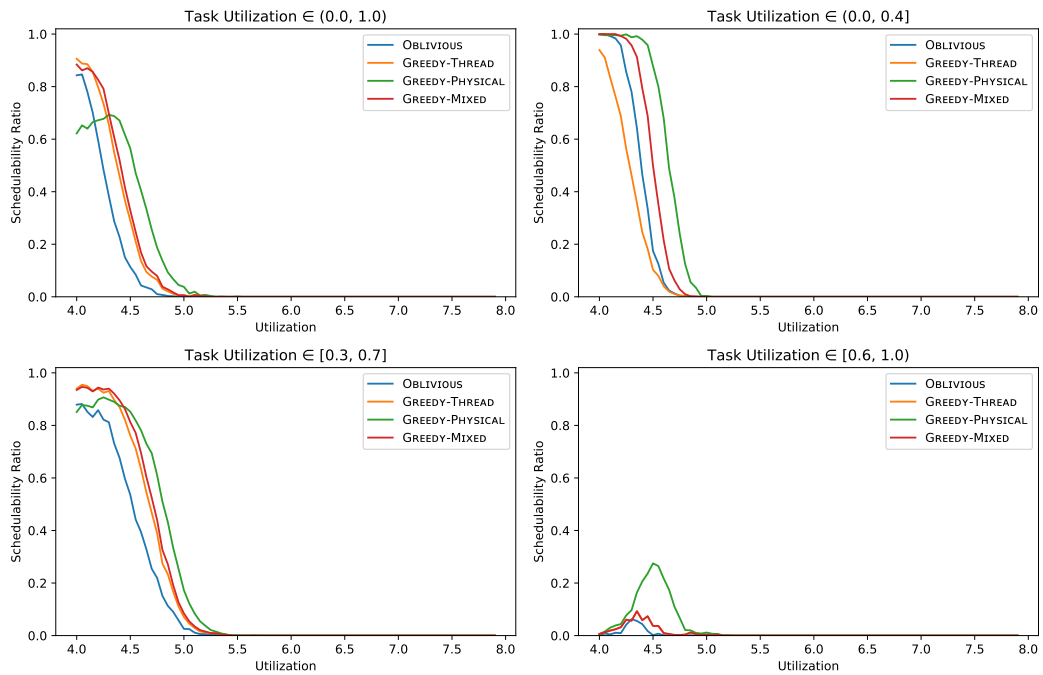


22:46 Simultaneous Multithreading Applied to Real Time

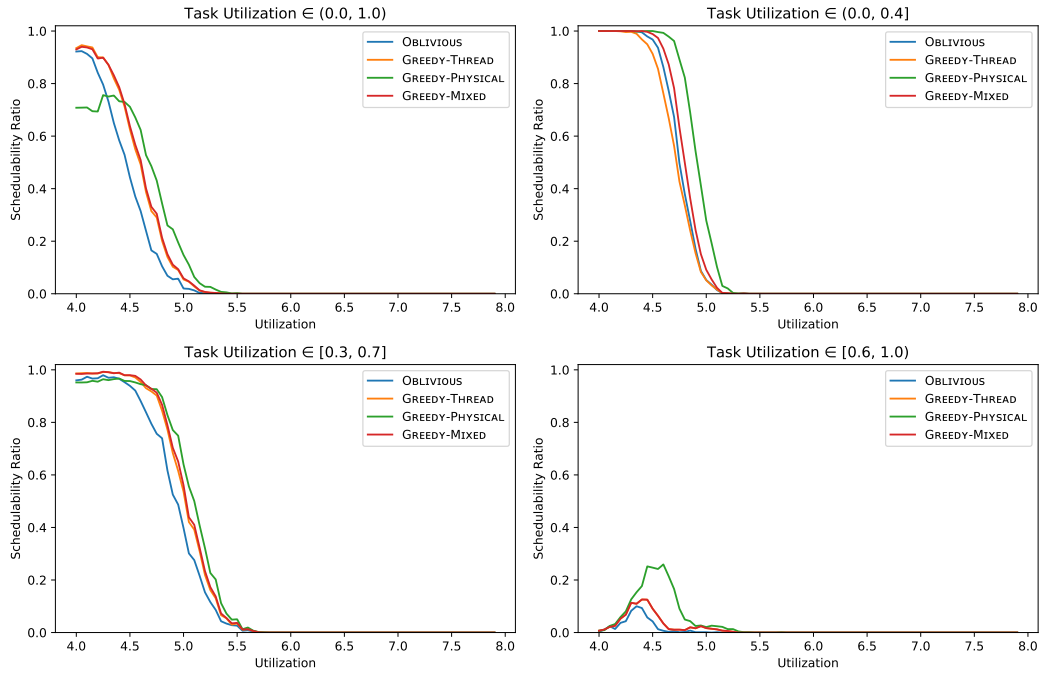
4 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$



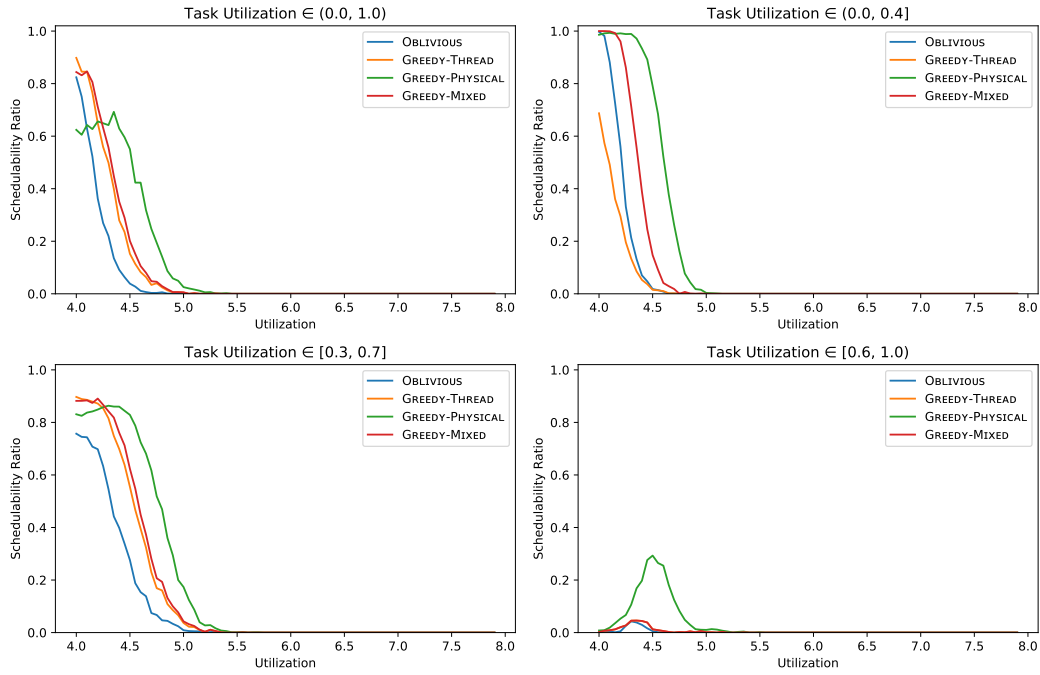
4 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$



4 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$

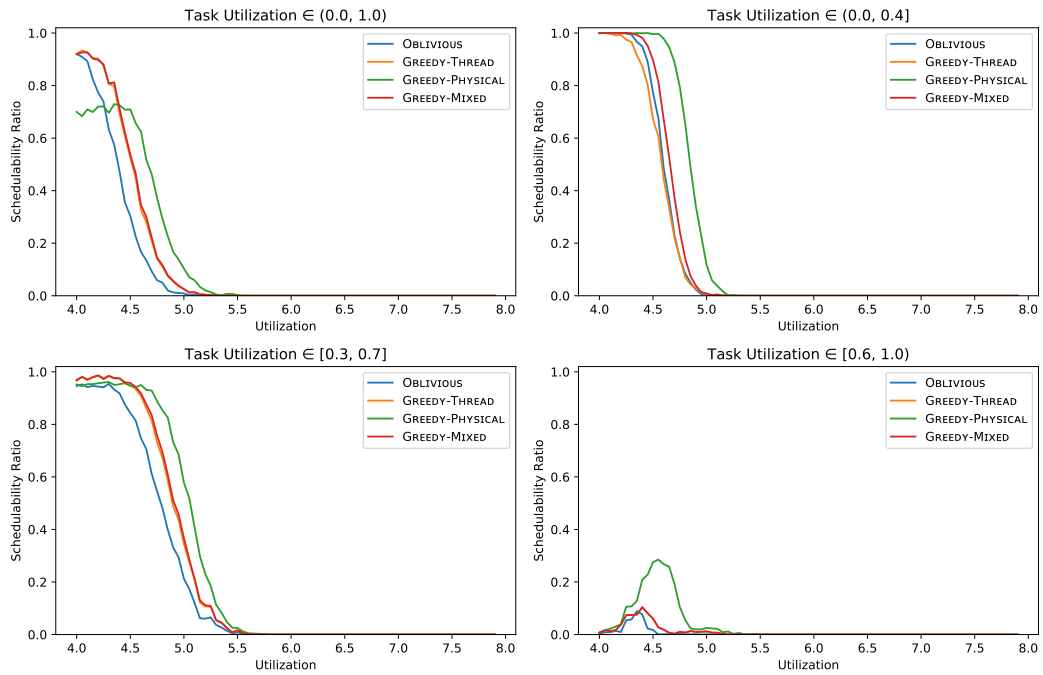


4 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$

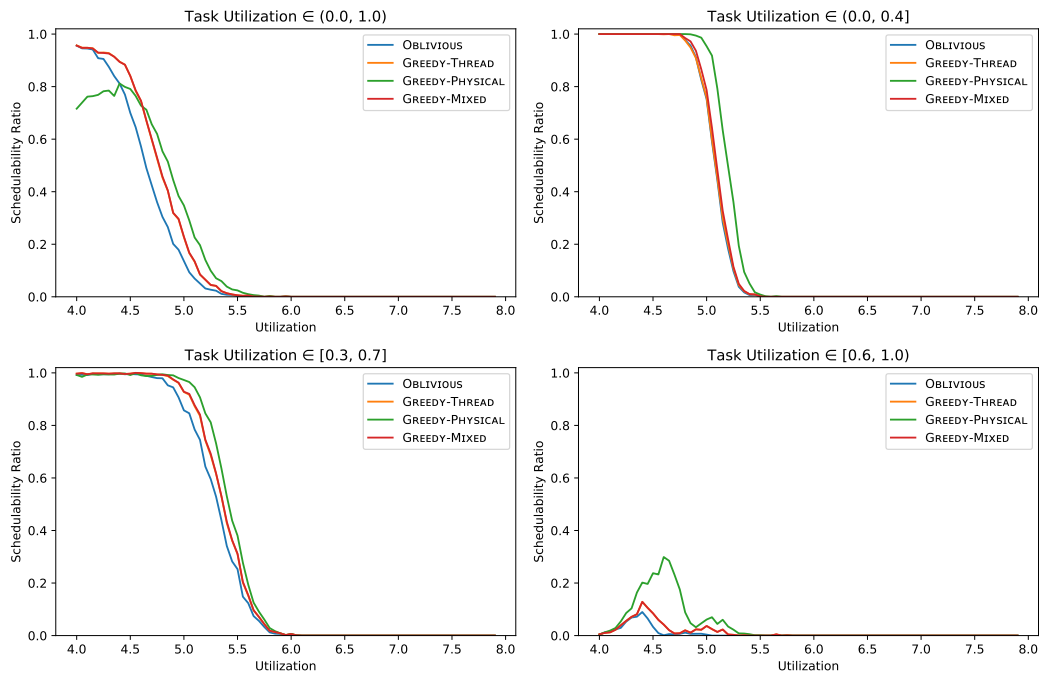


22:48 Simultaneous Multithreading Applied to Real Time

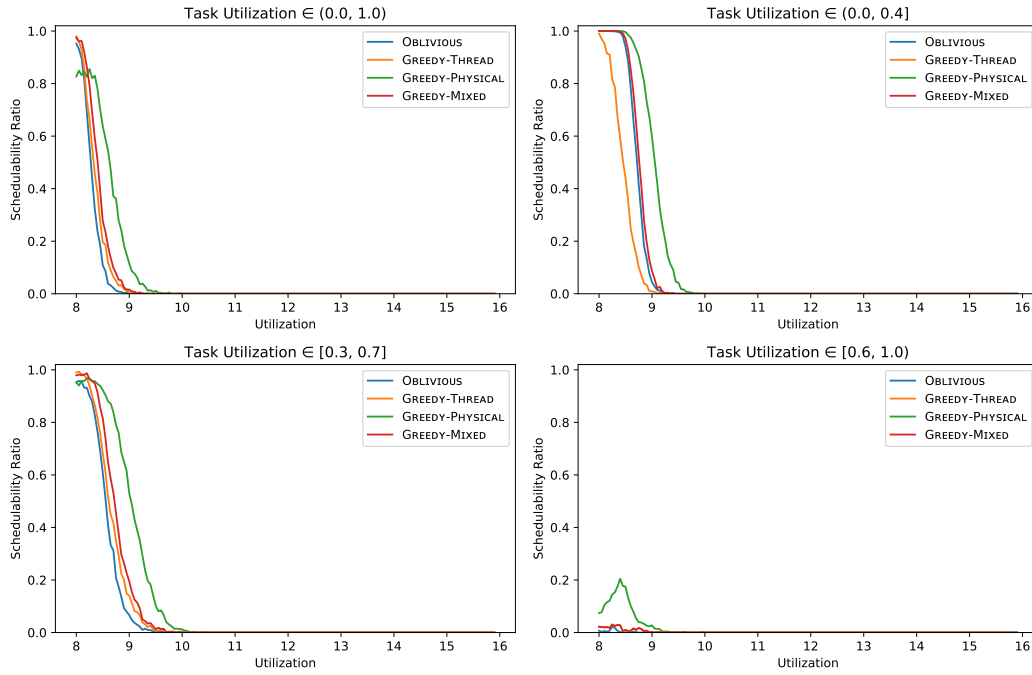
4 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$



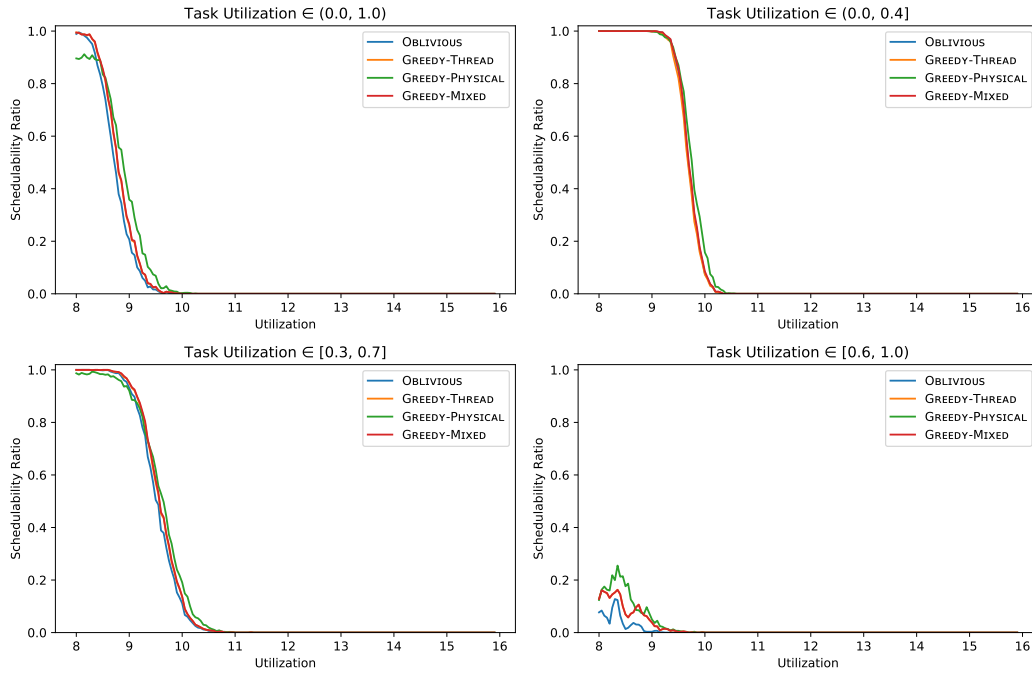
4 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$



8 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$

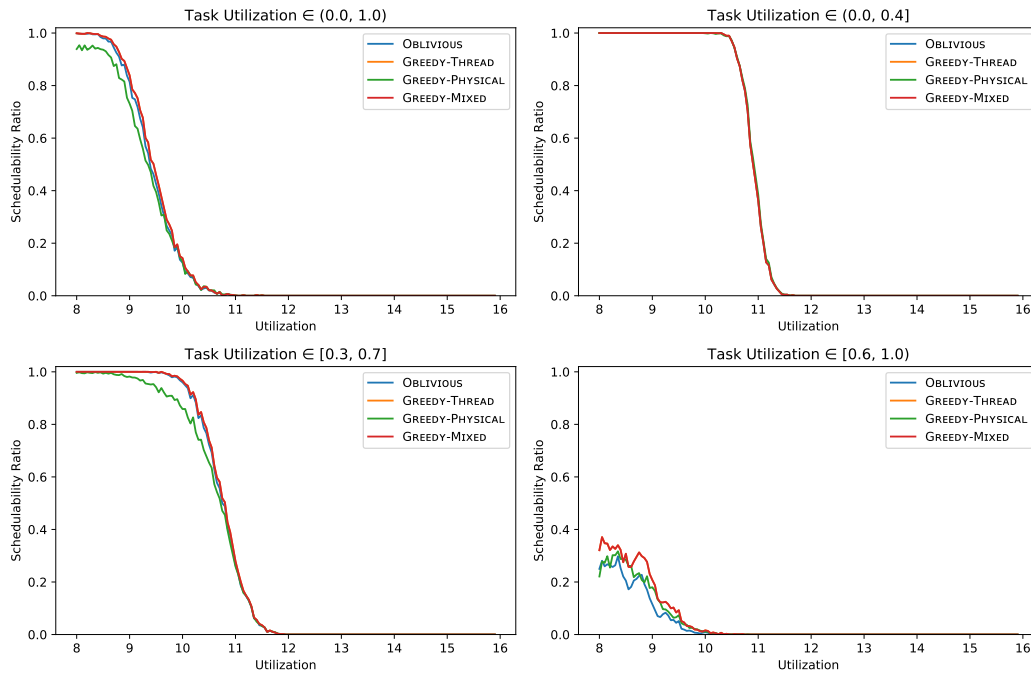


8 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$

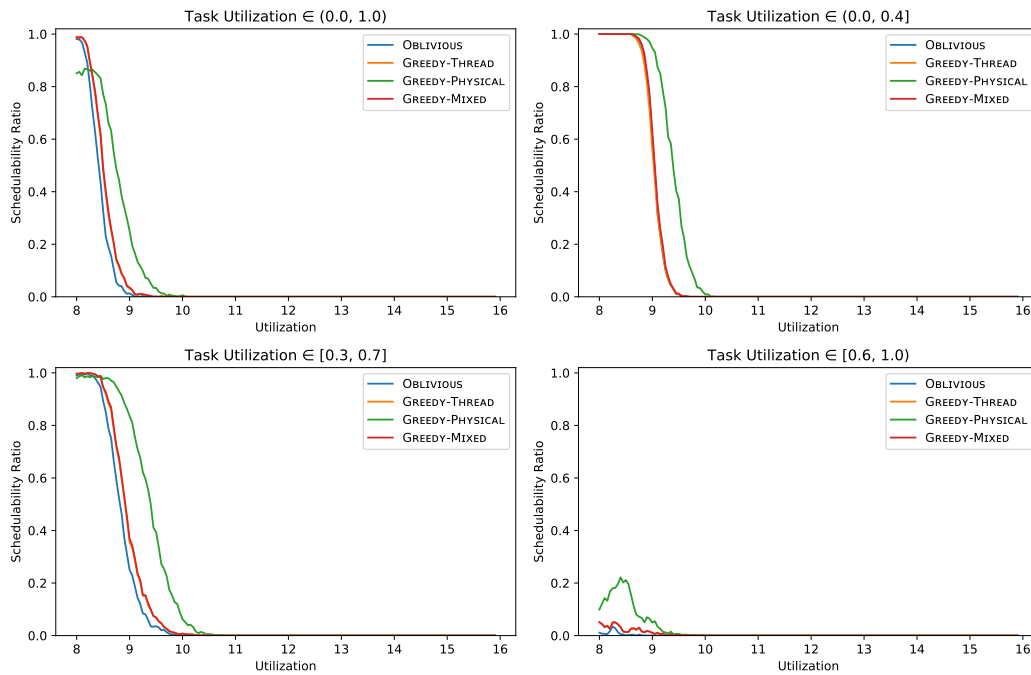


22:50 Simultaneous Multithreading Applied to Real Time

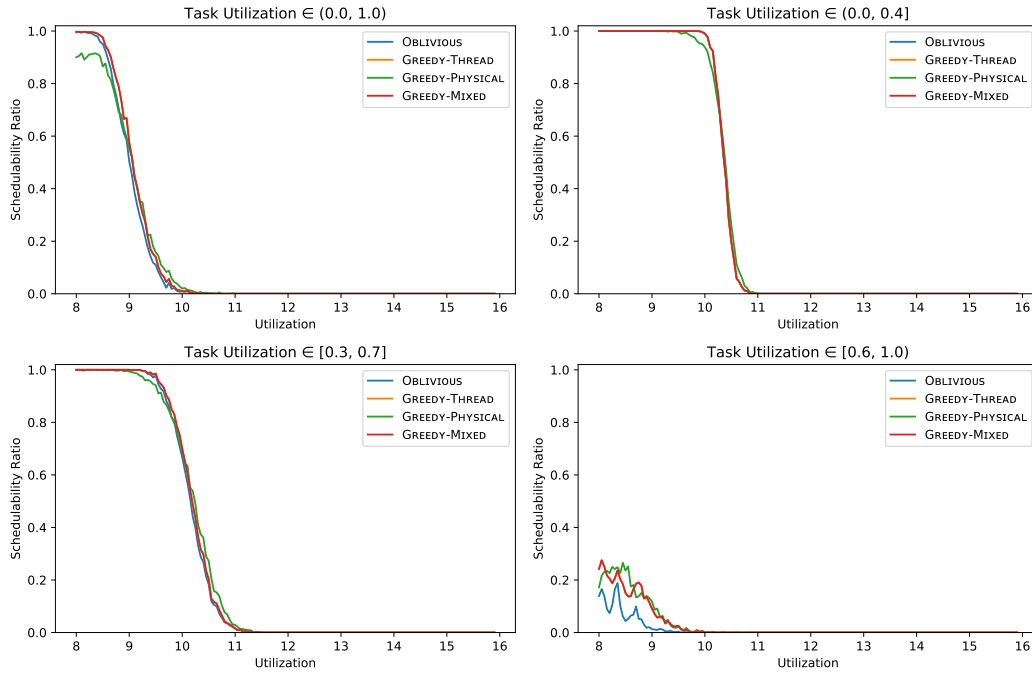
8 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$



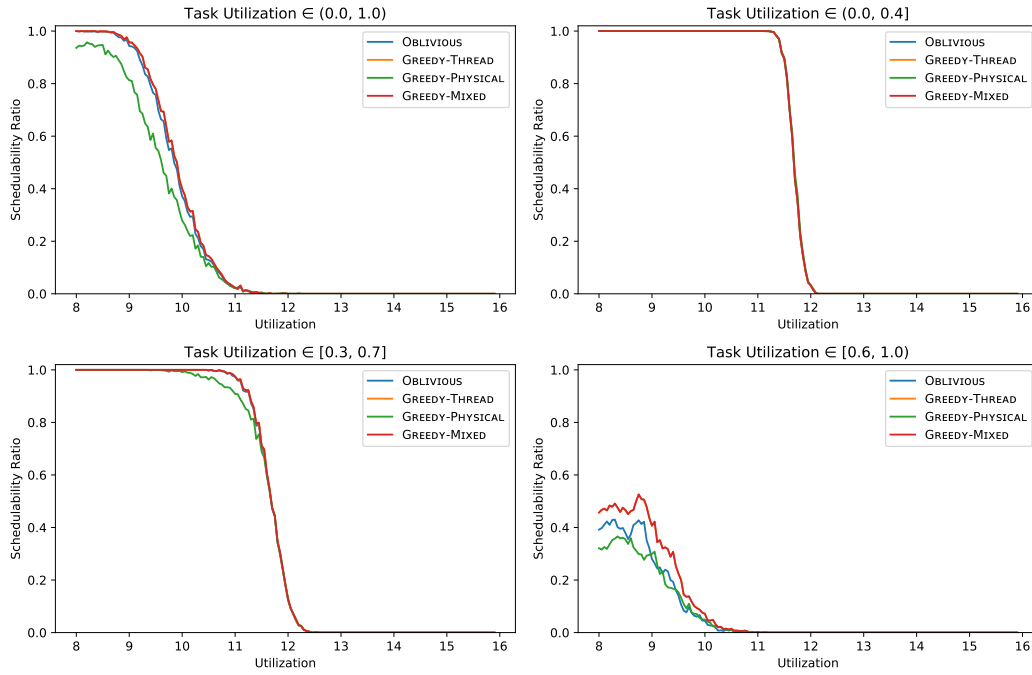
8 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$



8 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$

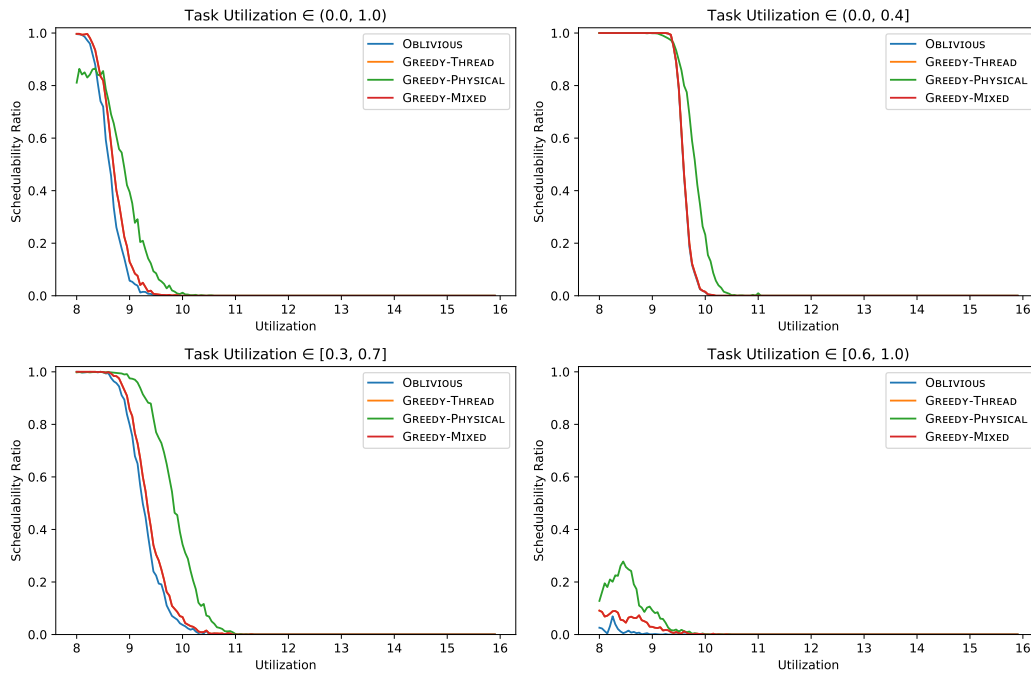


8 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$

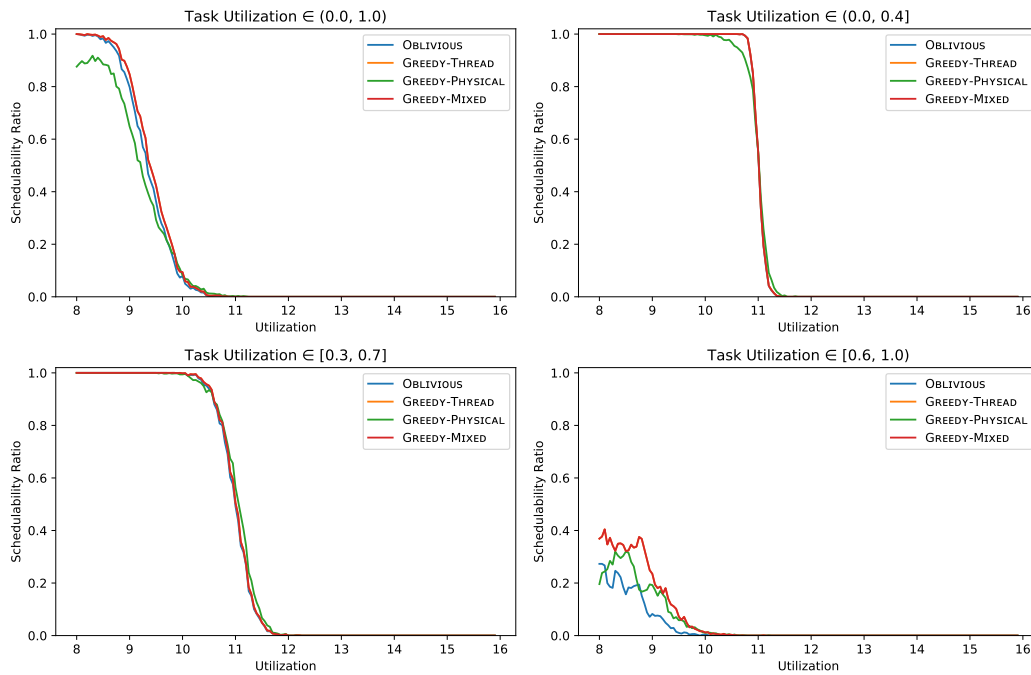


22:52 Simultaneous Multithreading Applied to Real Time

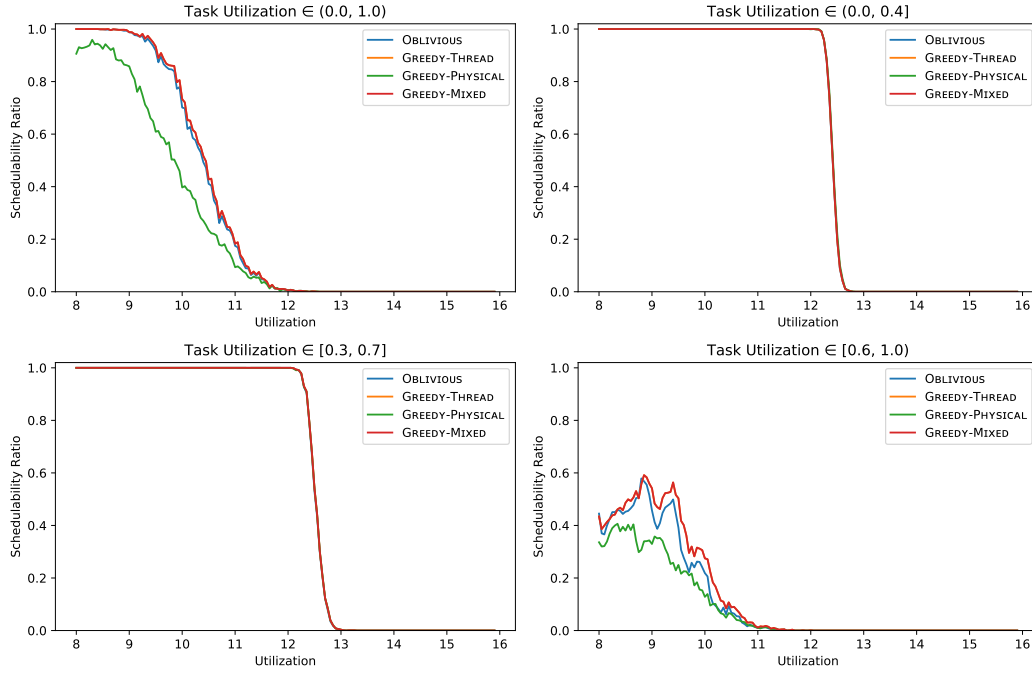
8 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$



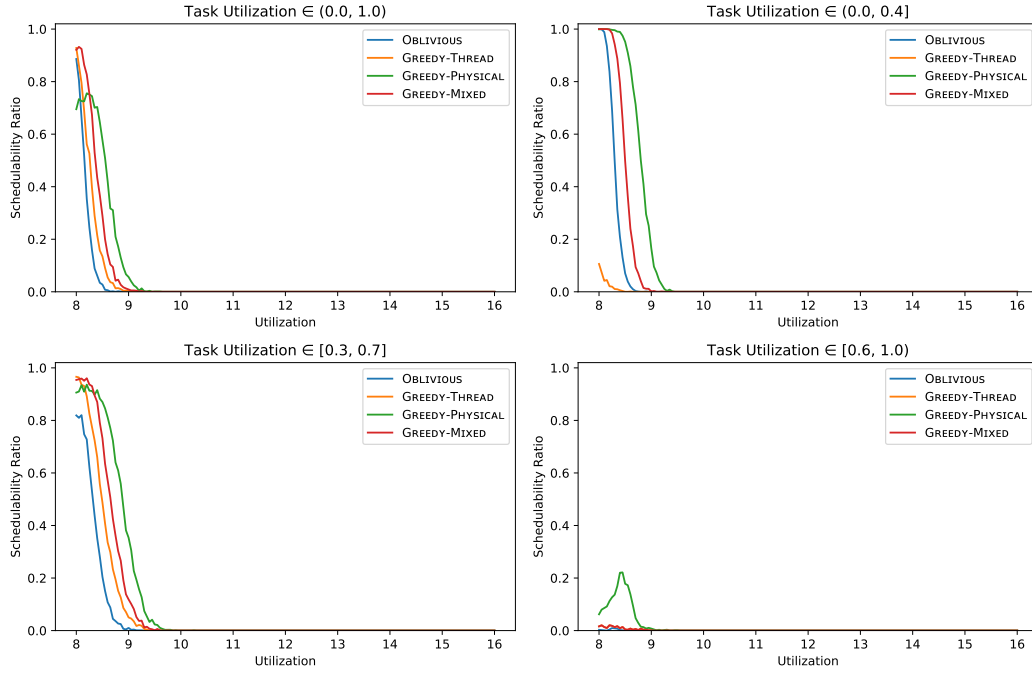
8 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$



8 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$

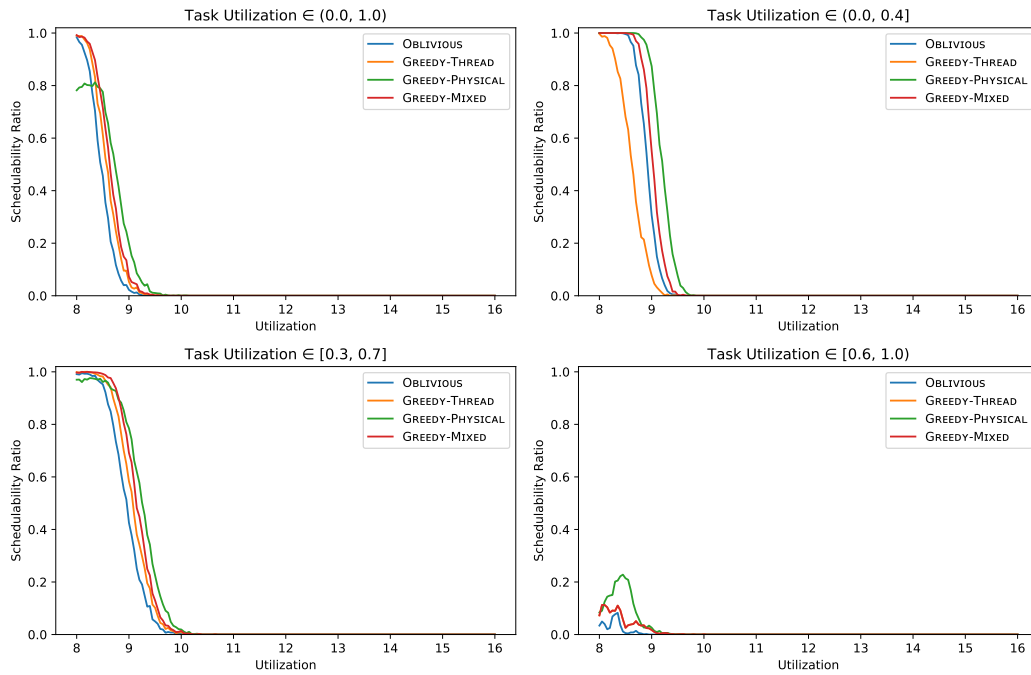


8 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$

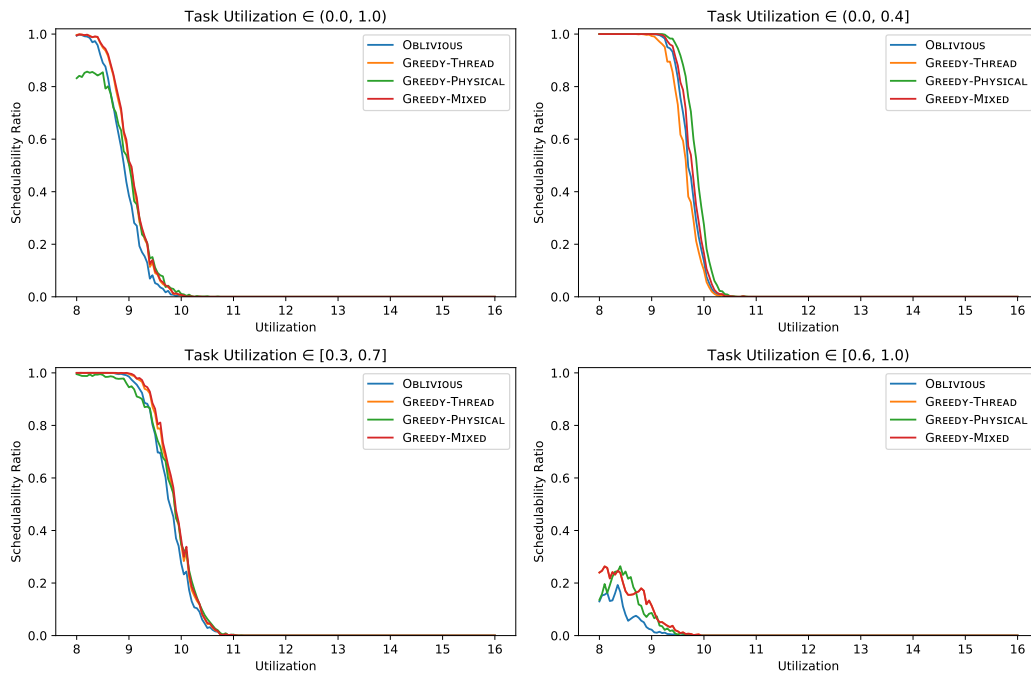


22:54 Simultaneous Multithreading Applied to Real Time

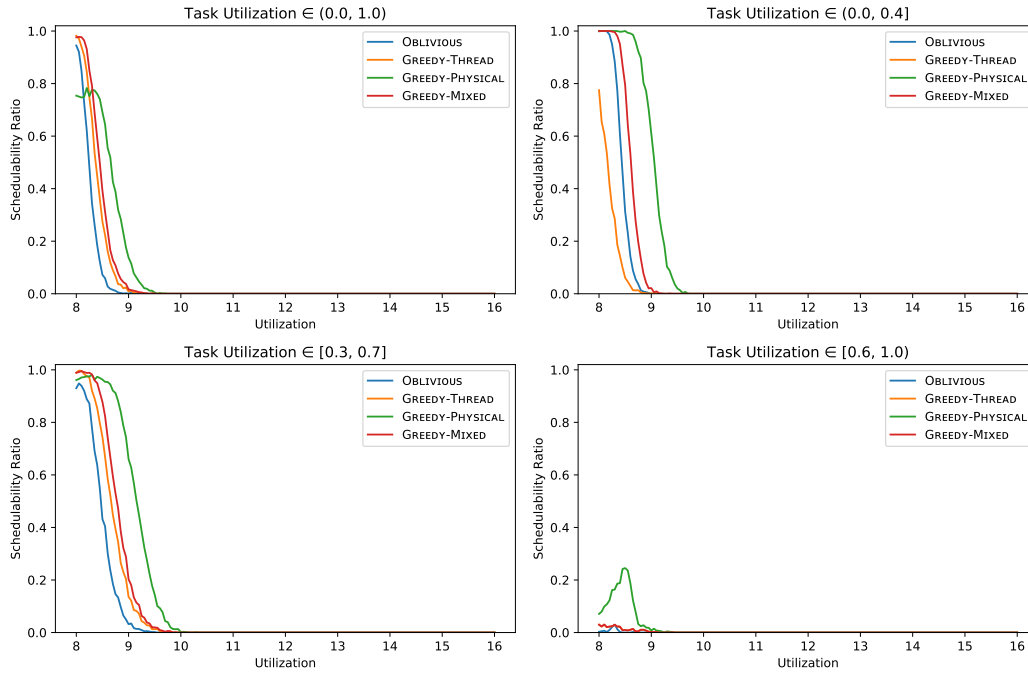
8 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$



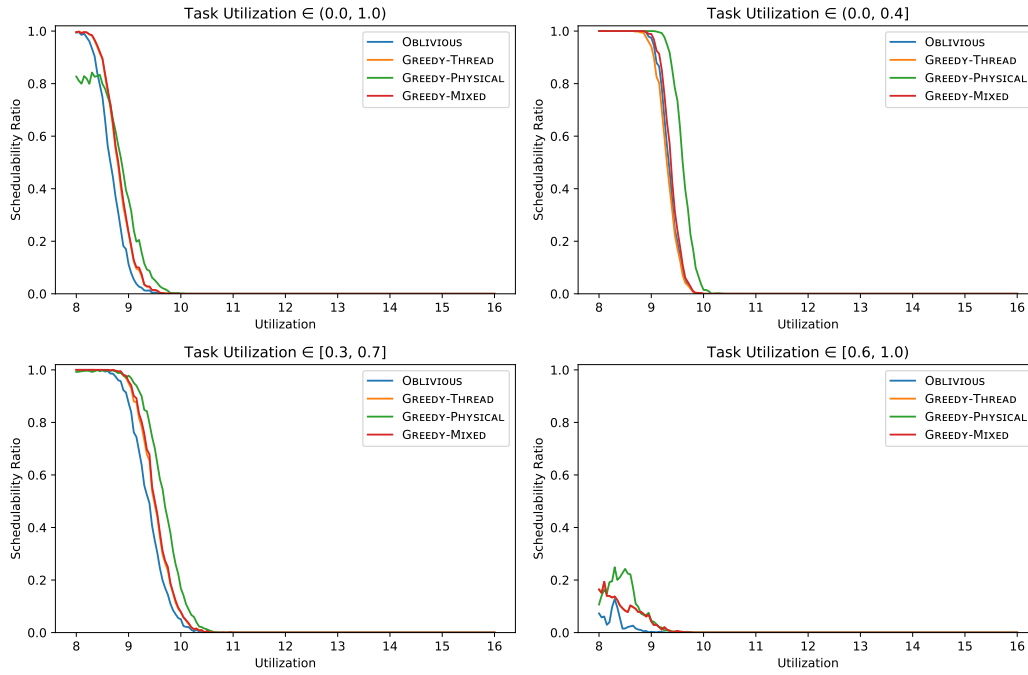
8 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$



8 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$

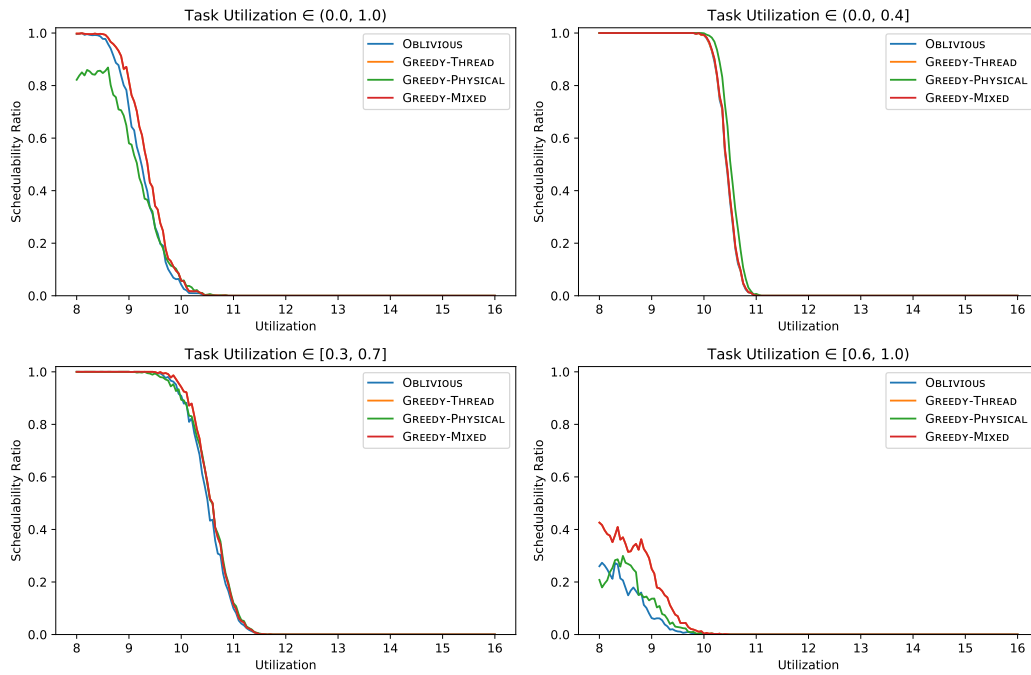


8 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$

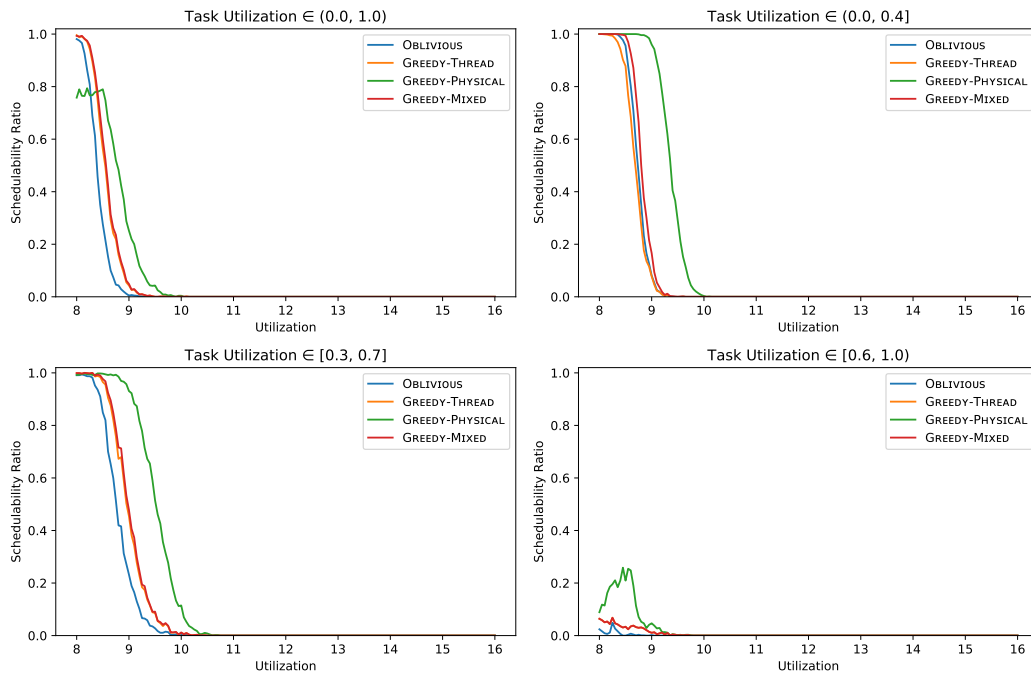


22:56 Simultaneous Multithreading Applied to Real Time

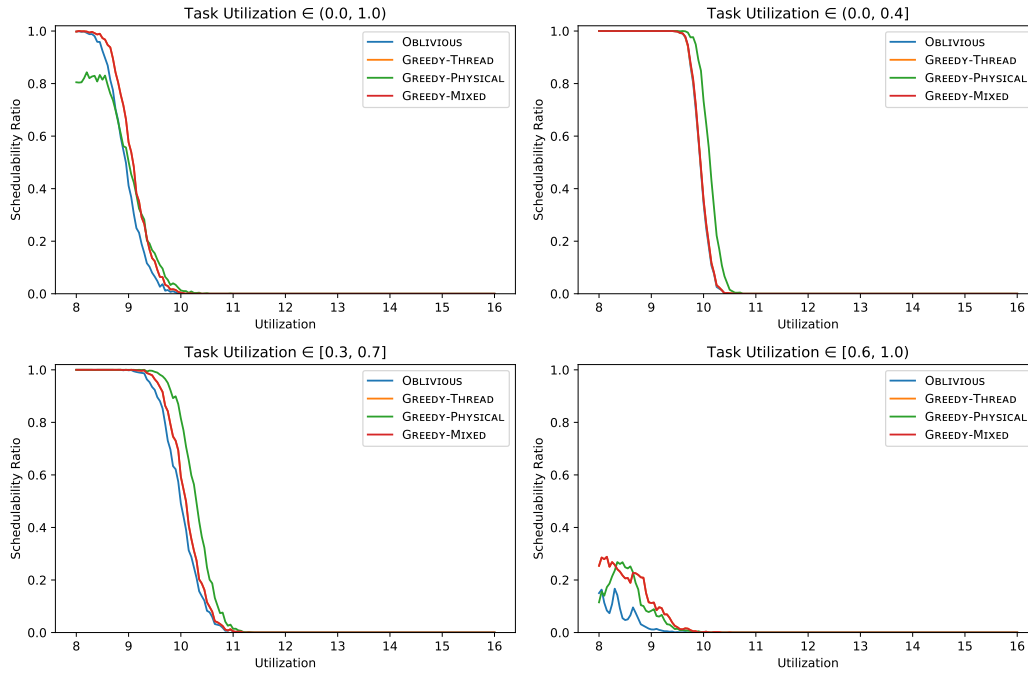
8 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$



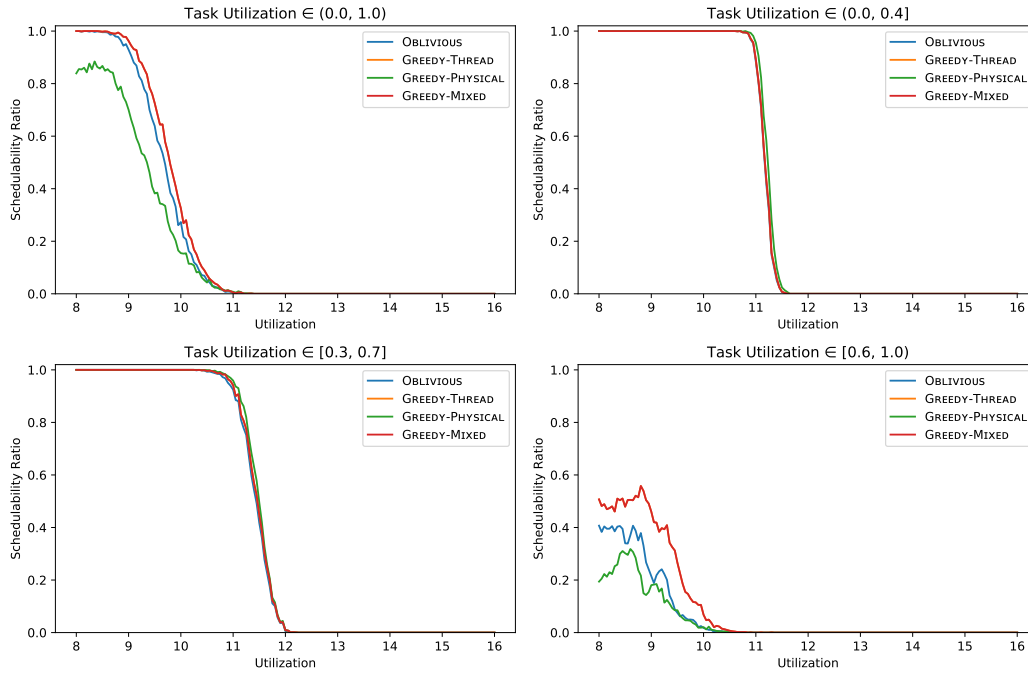
8 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$



8 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$

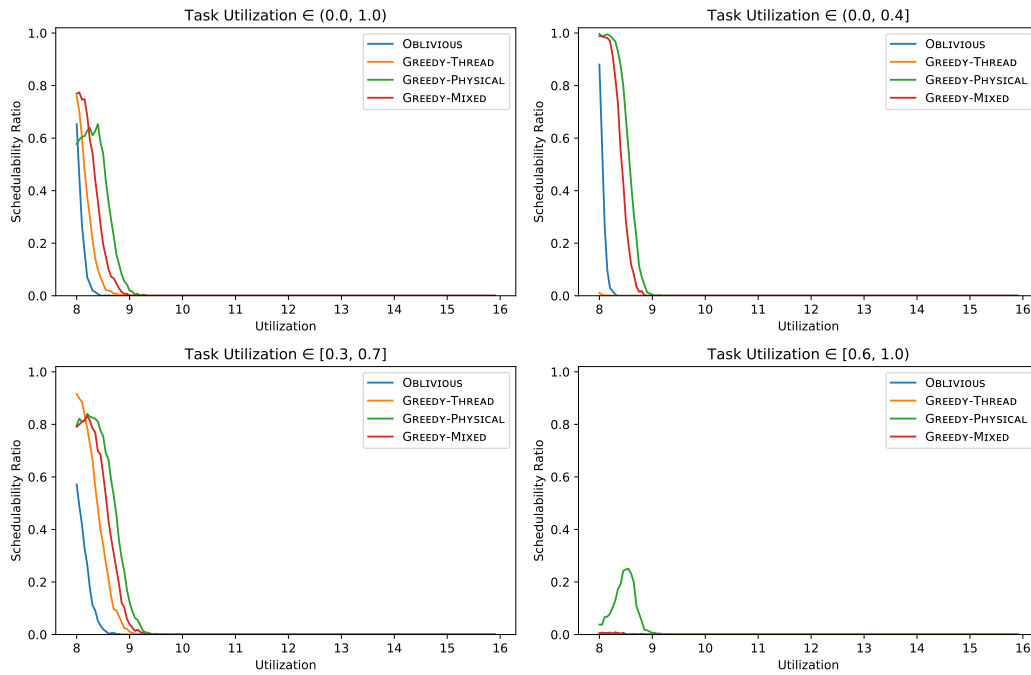


8 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$

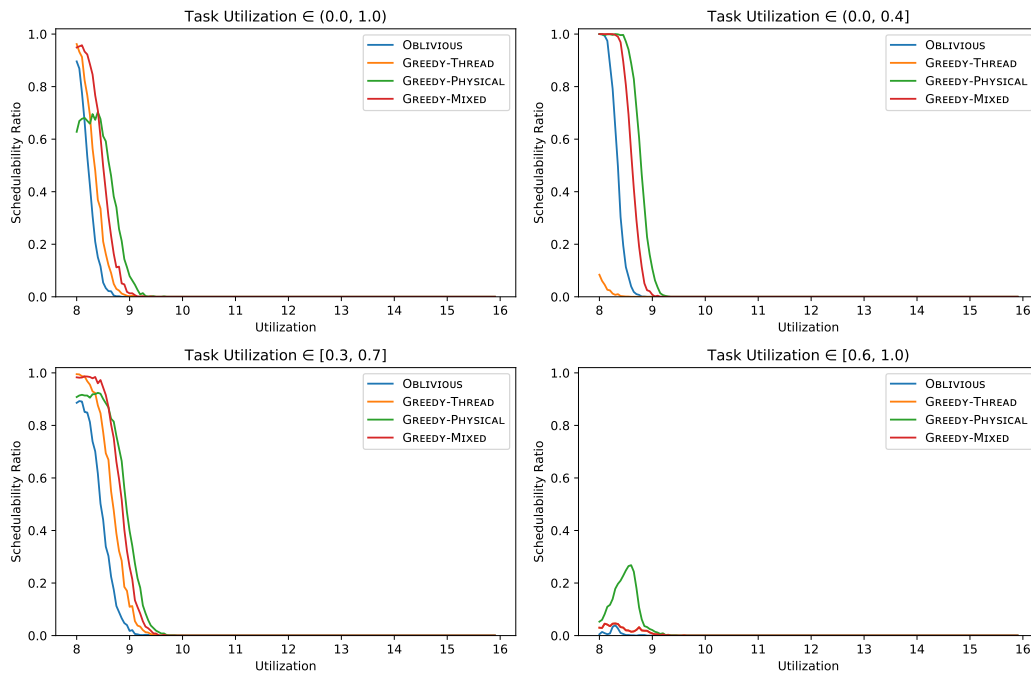


22:58 Simultaneous Multithreading Applied to Real Time

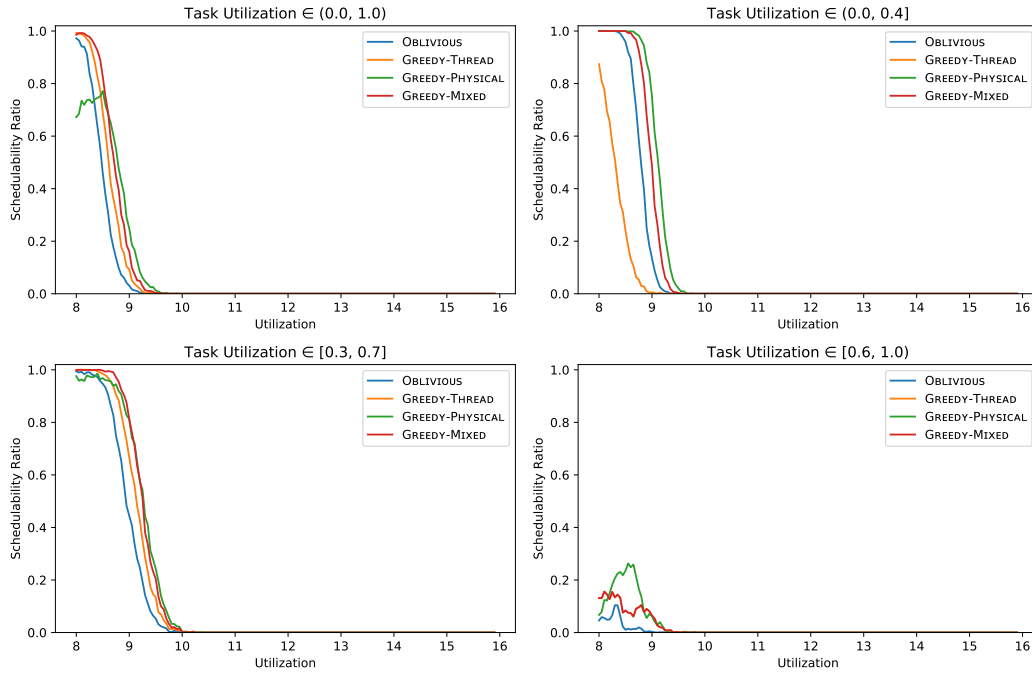
8 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$



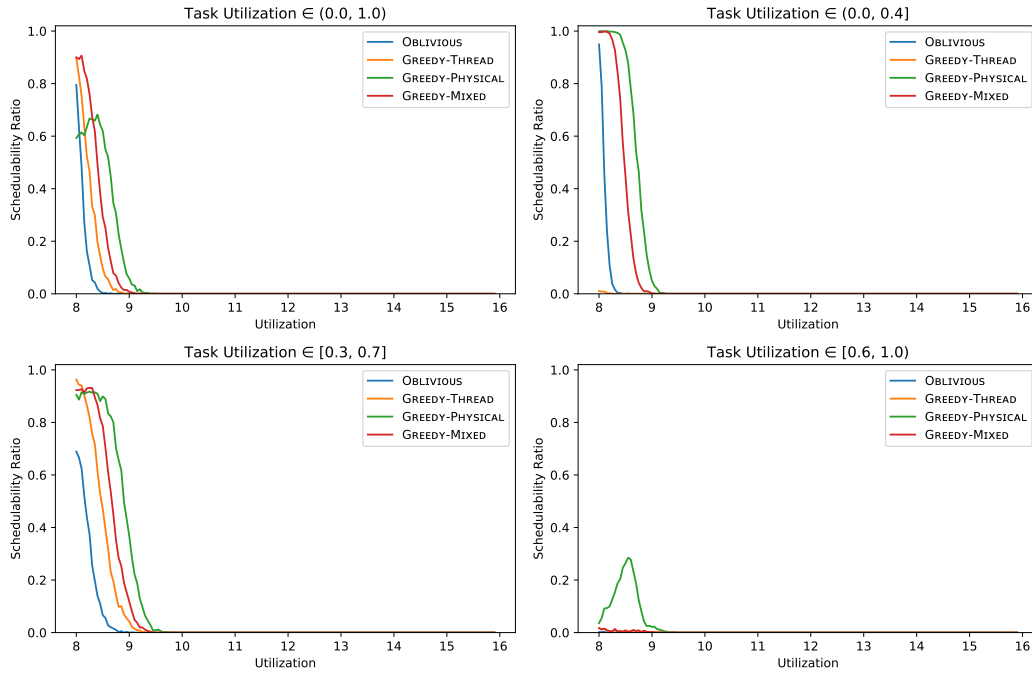
8 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$



8 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$

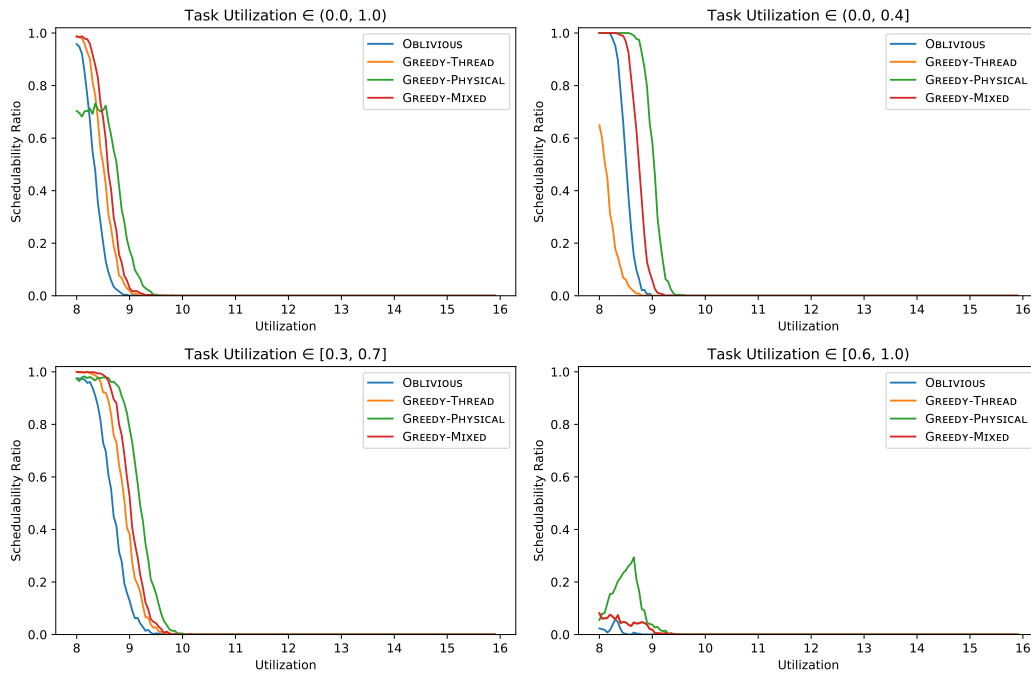


8 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$

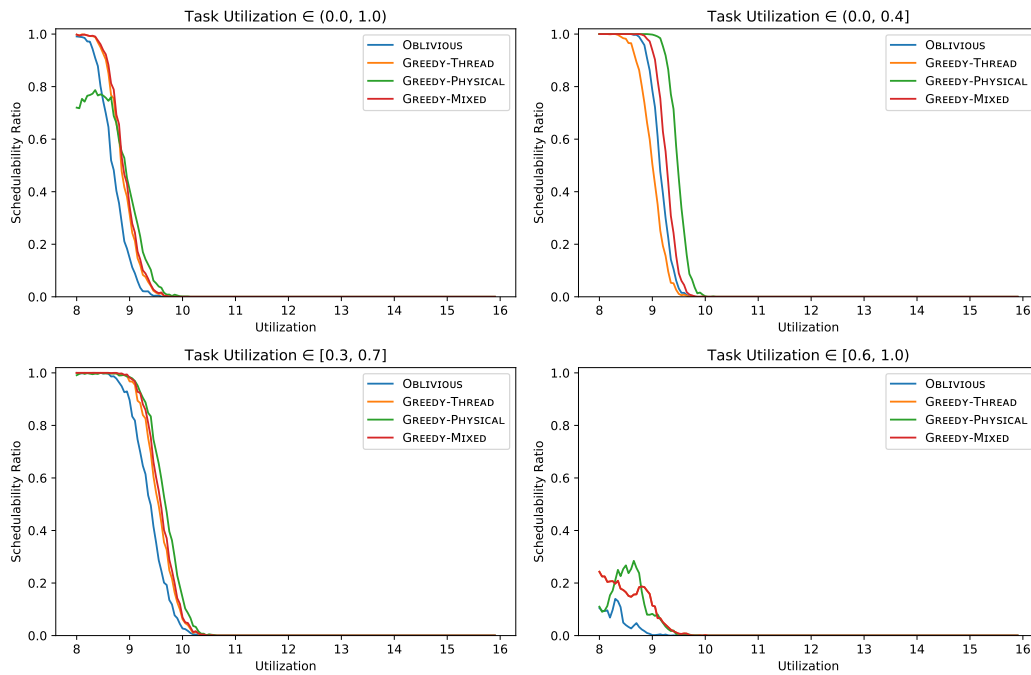


22:60 Simultaneous Multithreading Applied to Real Time

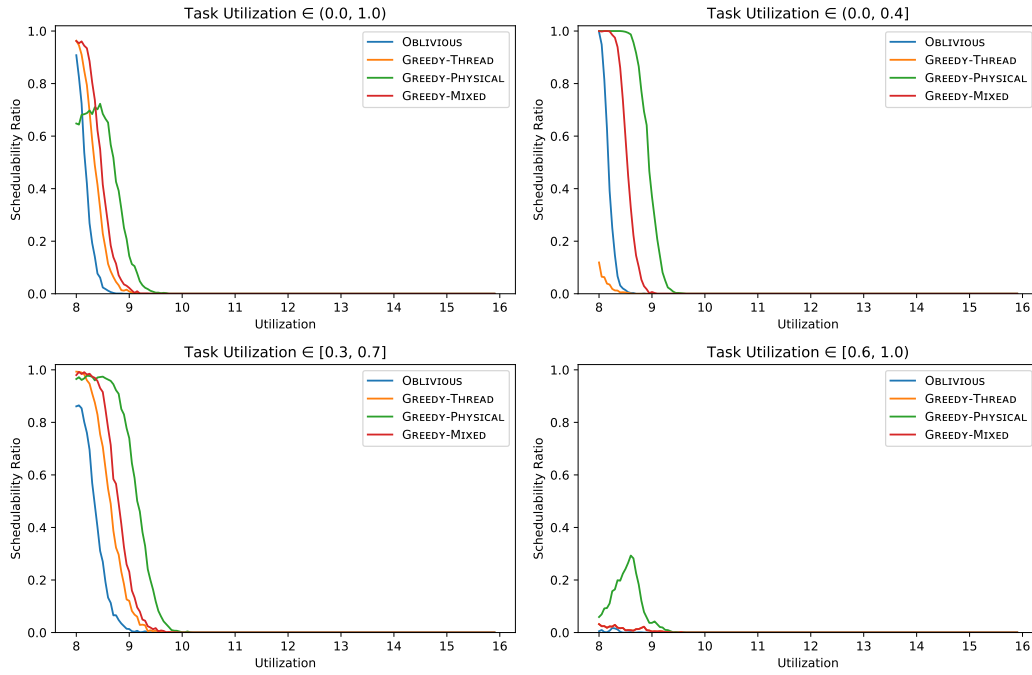
8 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$



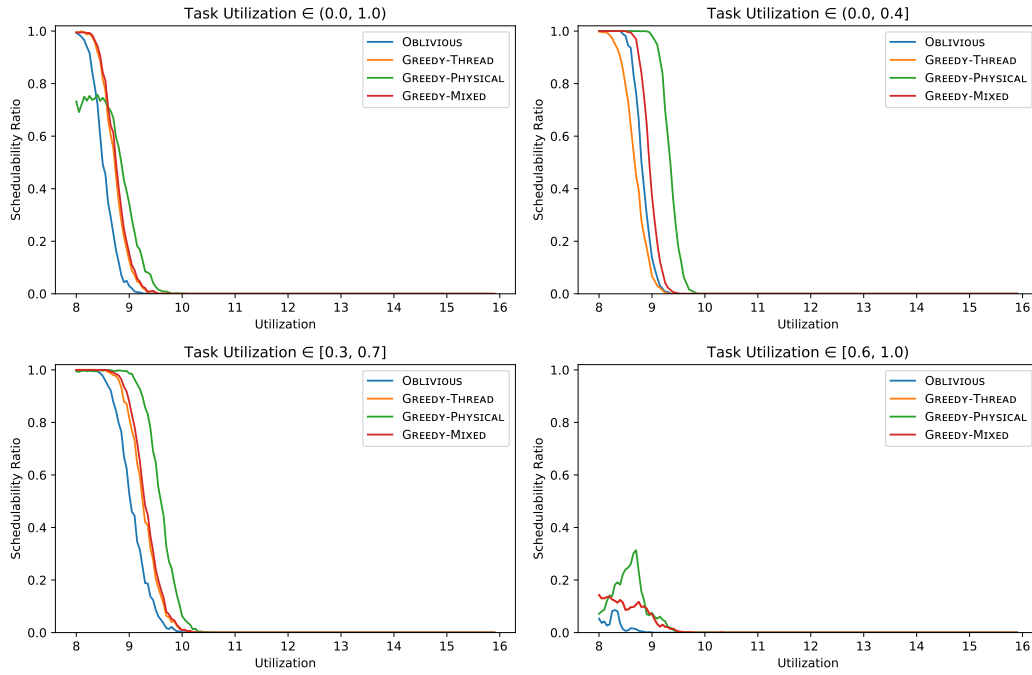
8 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$



8 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$

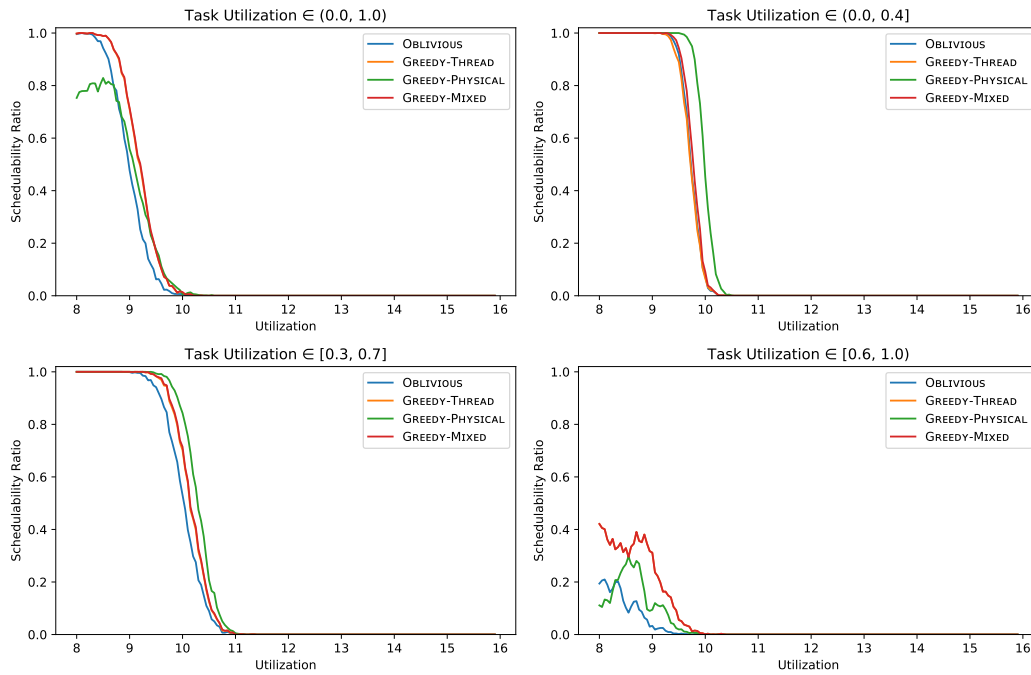


8 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$

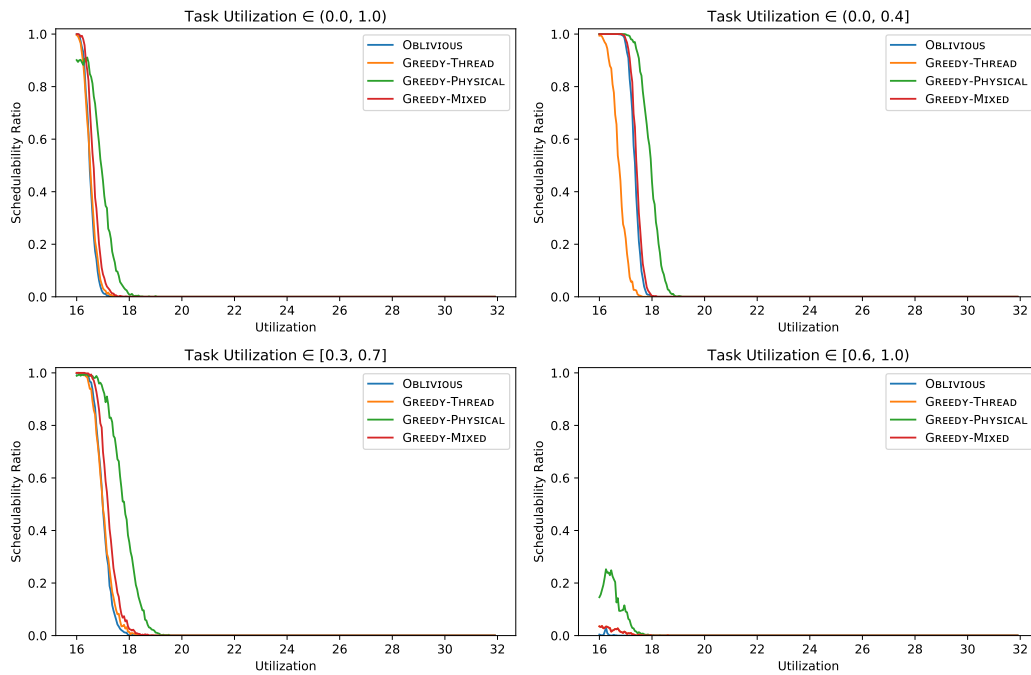


22:62 Simultaneous Multithreading Applied to Real Time

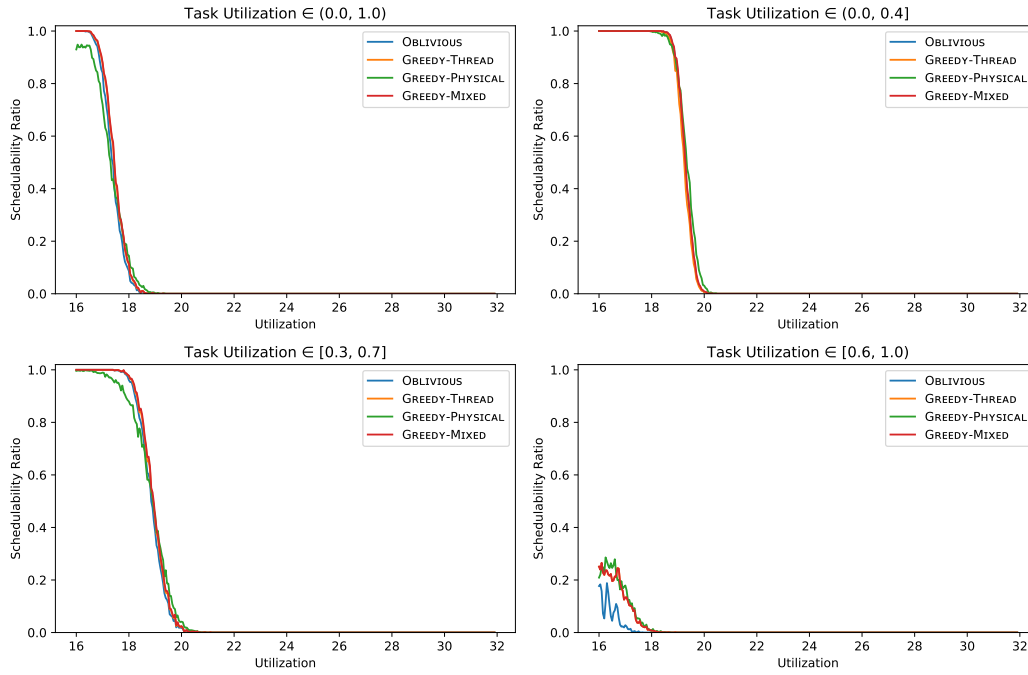
8 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$



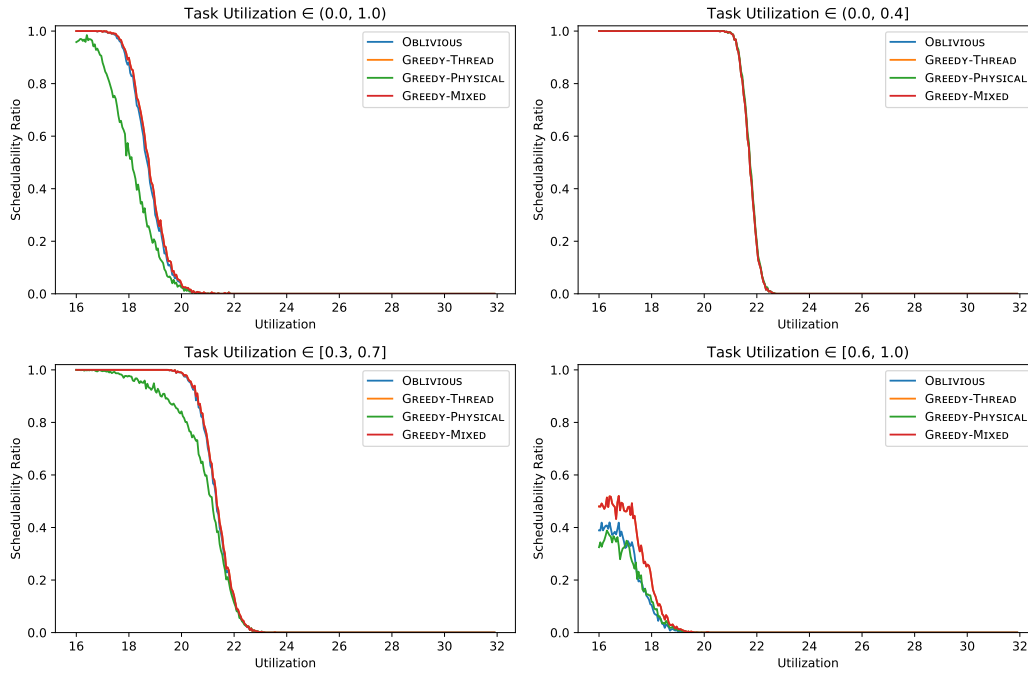
16 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$



16 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$

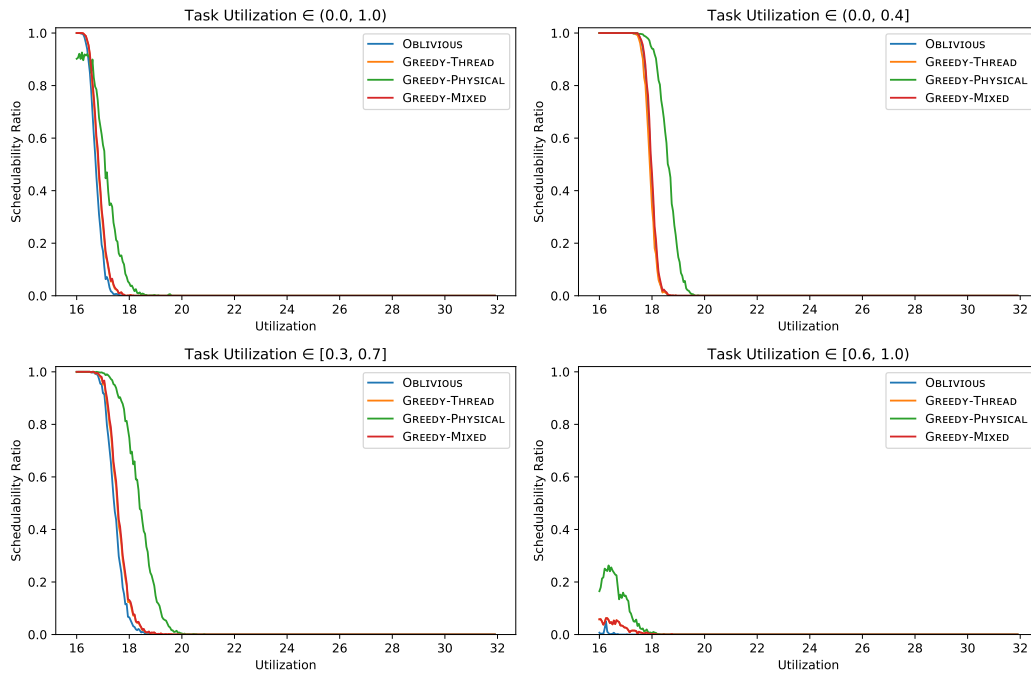


16 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$

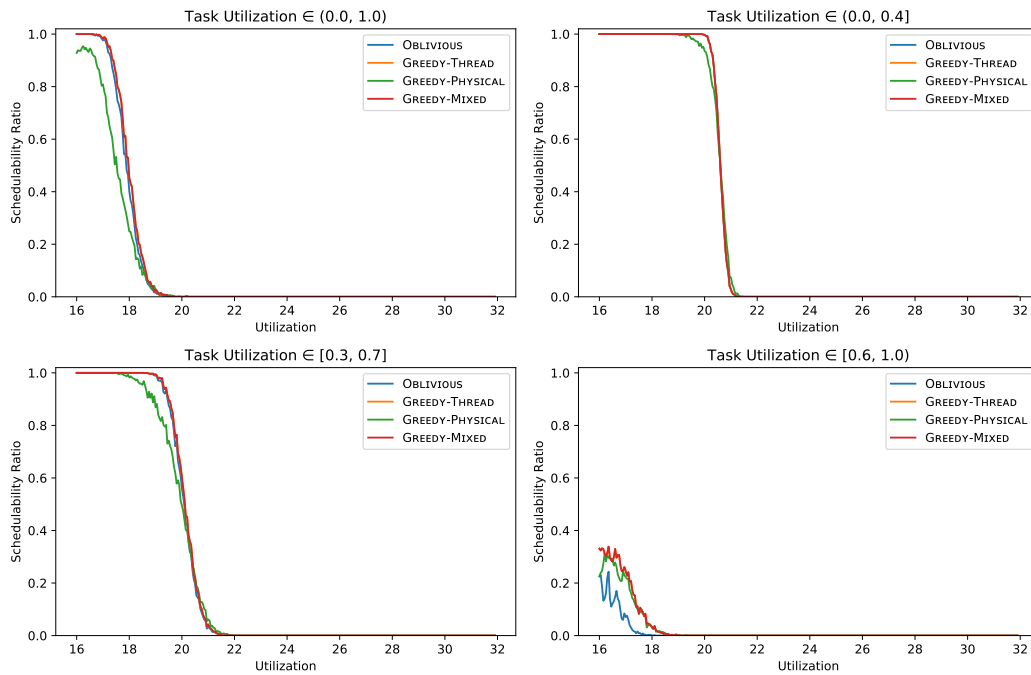


22:64 Simultaneous Multithreading Applied to Real Time

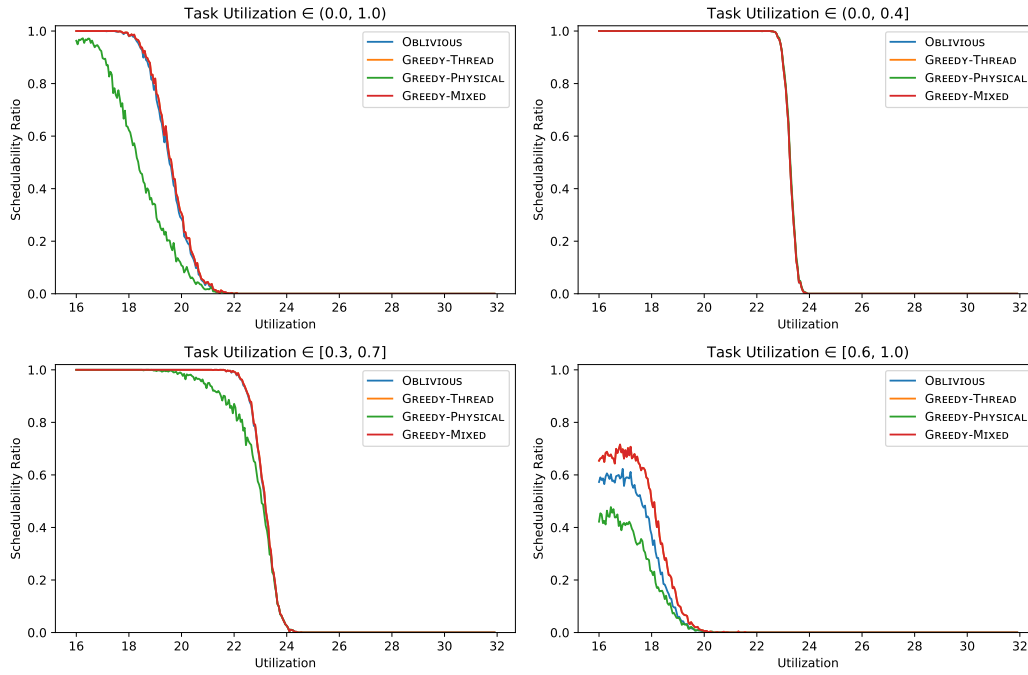
16 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$



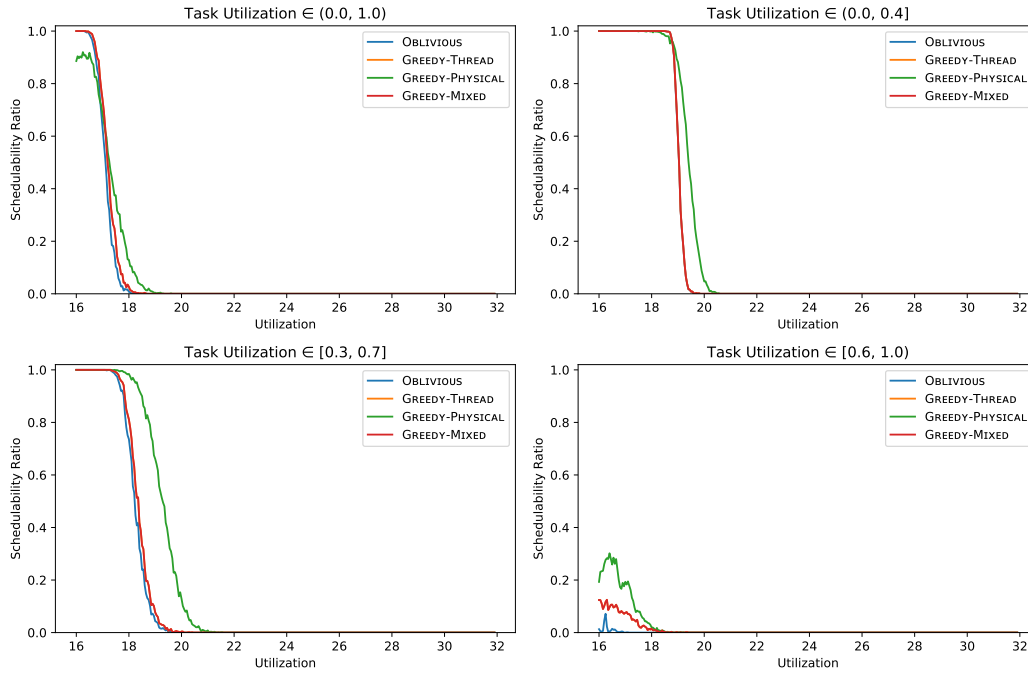
16 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$



16 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$

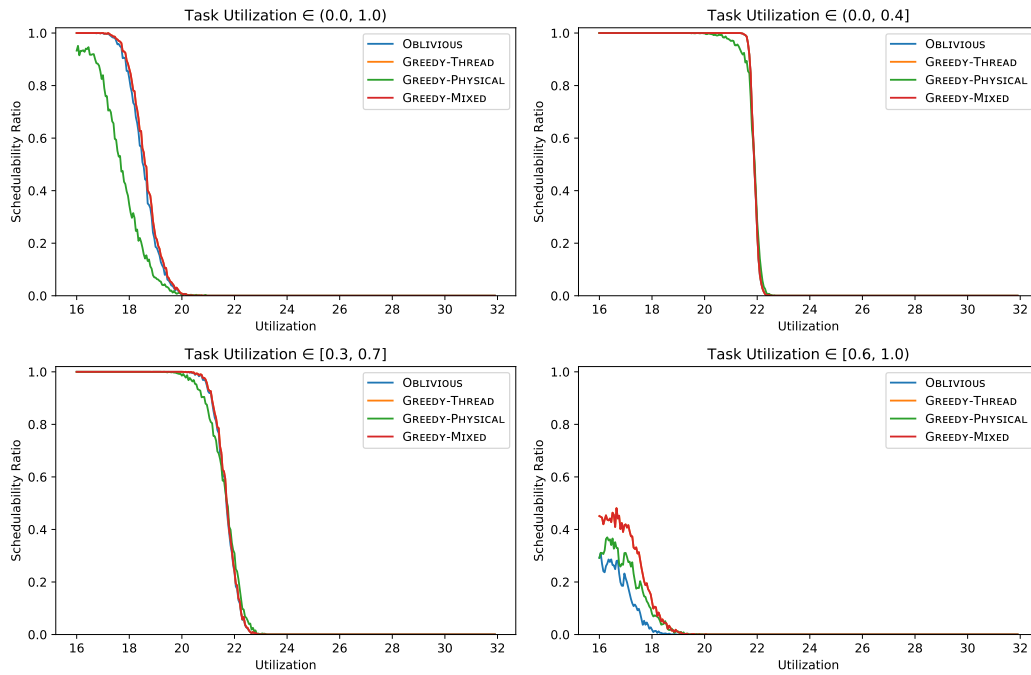


16 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$

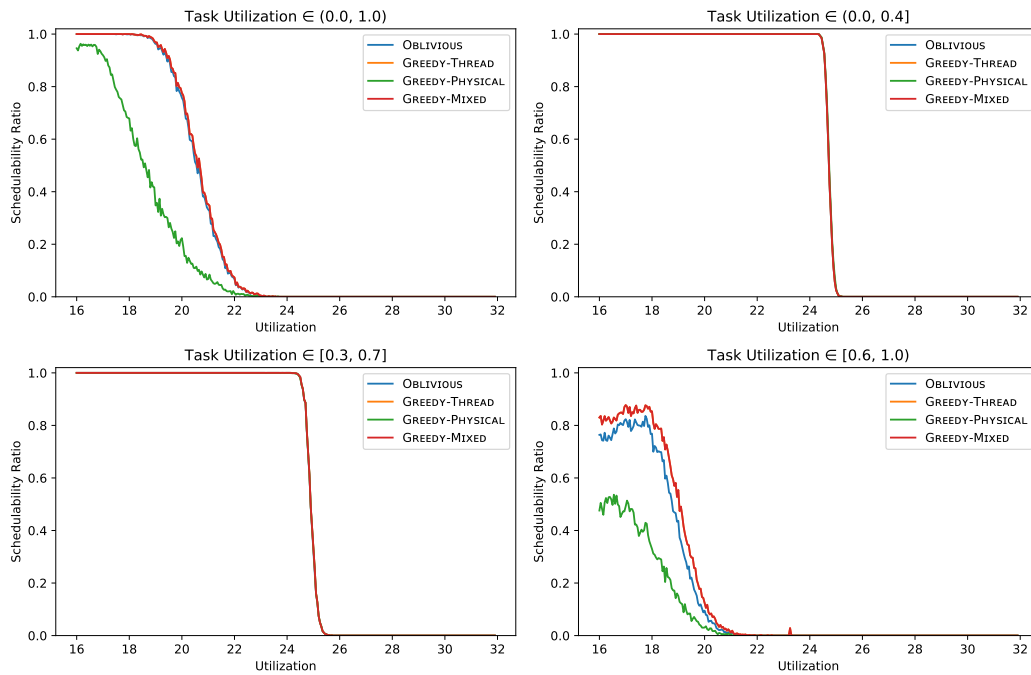


22:66 Simultaneous Multithreading Applied to Real Time

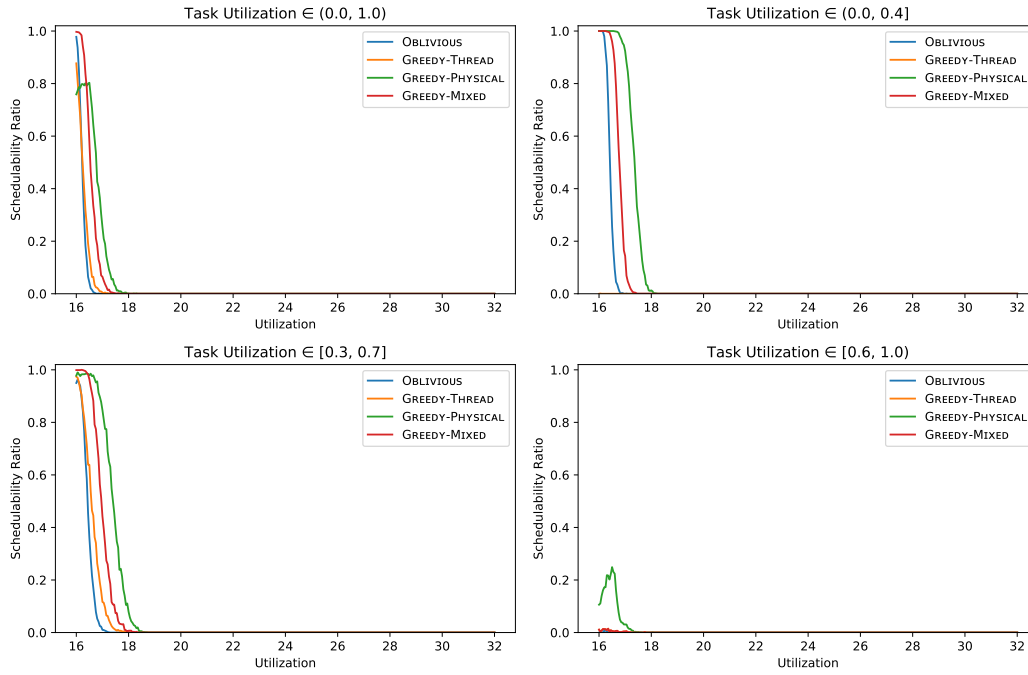
16 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$



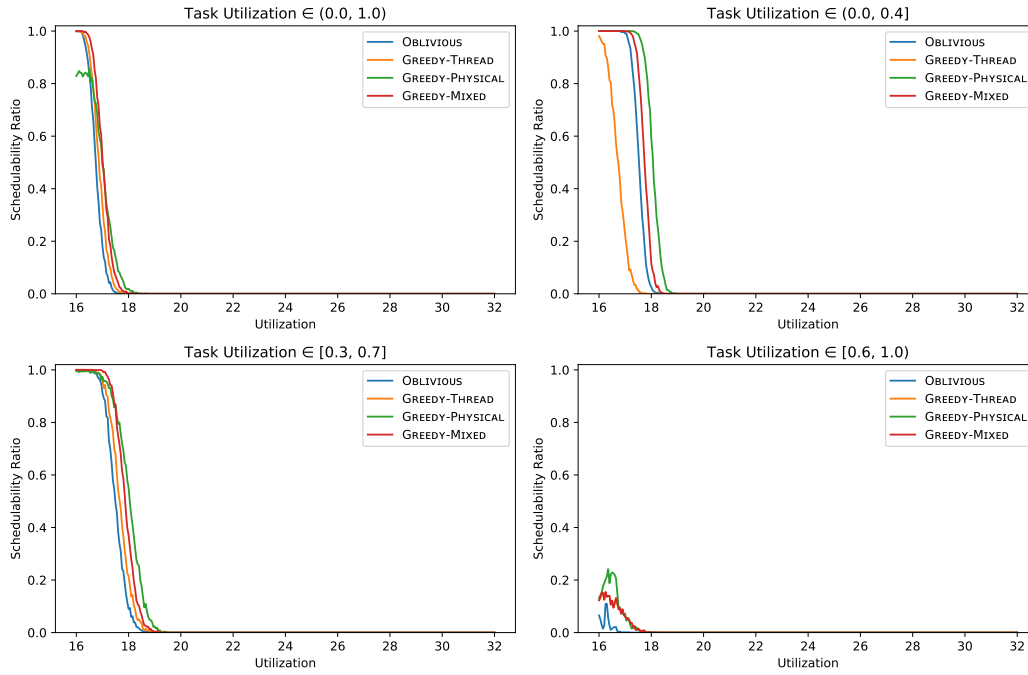
16 Cores, $\epsilon = 0.010$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$



16 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$

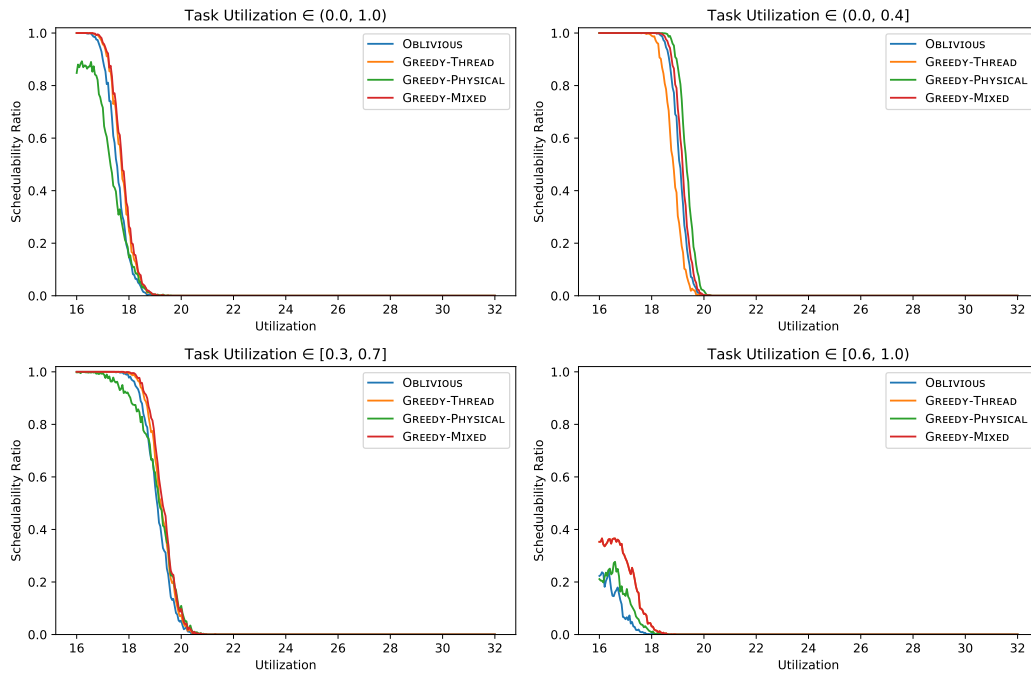


16 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$

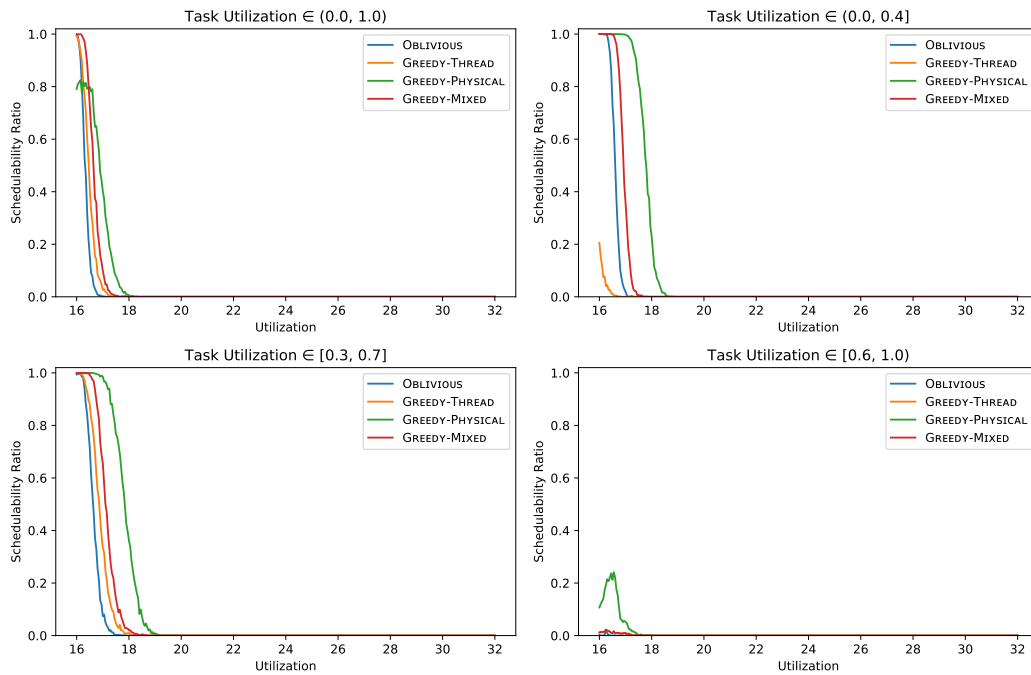


22:68 Simultaneous Multithreading Applied to Real Time

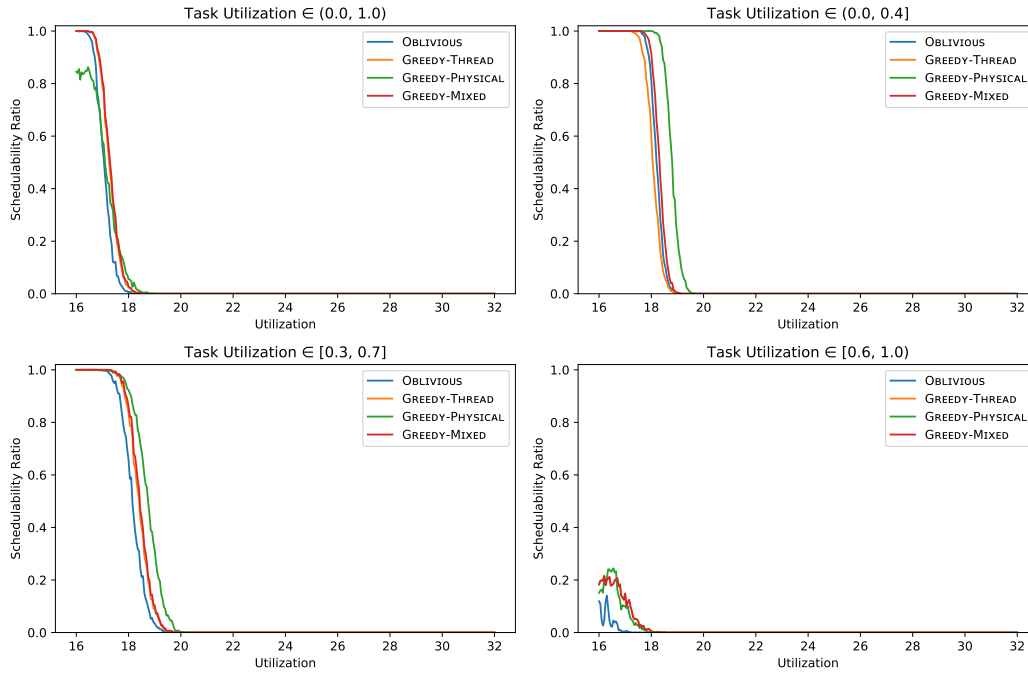
16 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$



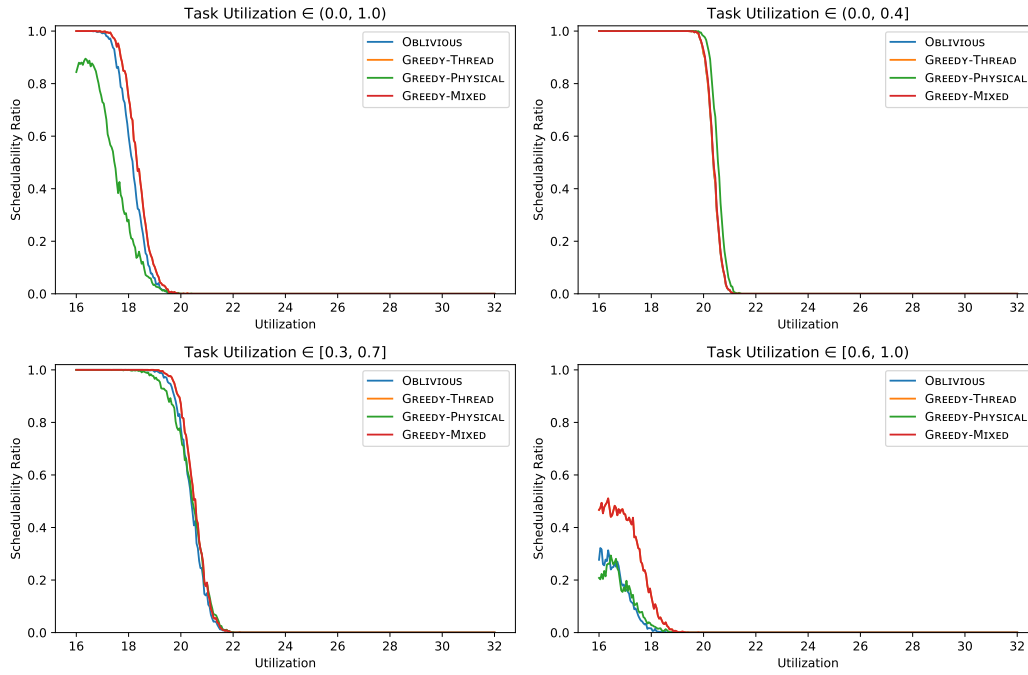
16 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$



16 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$

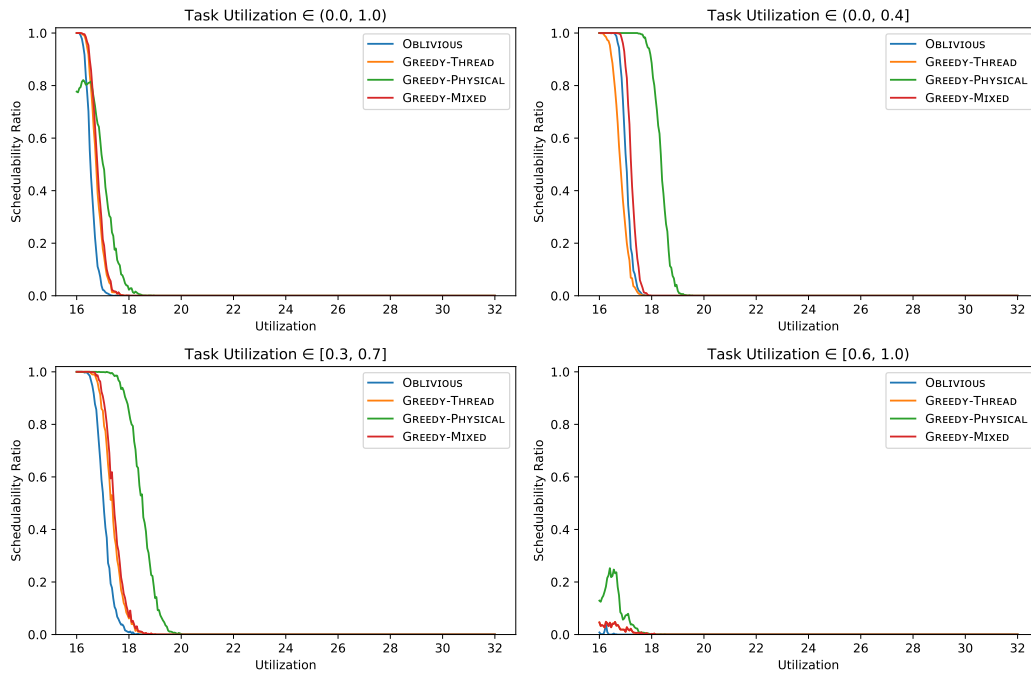


16 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$

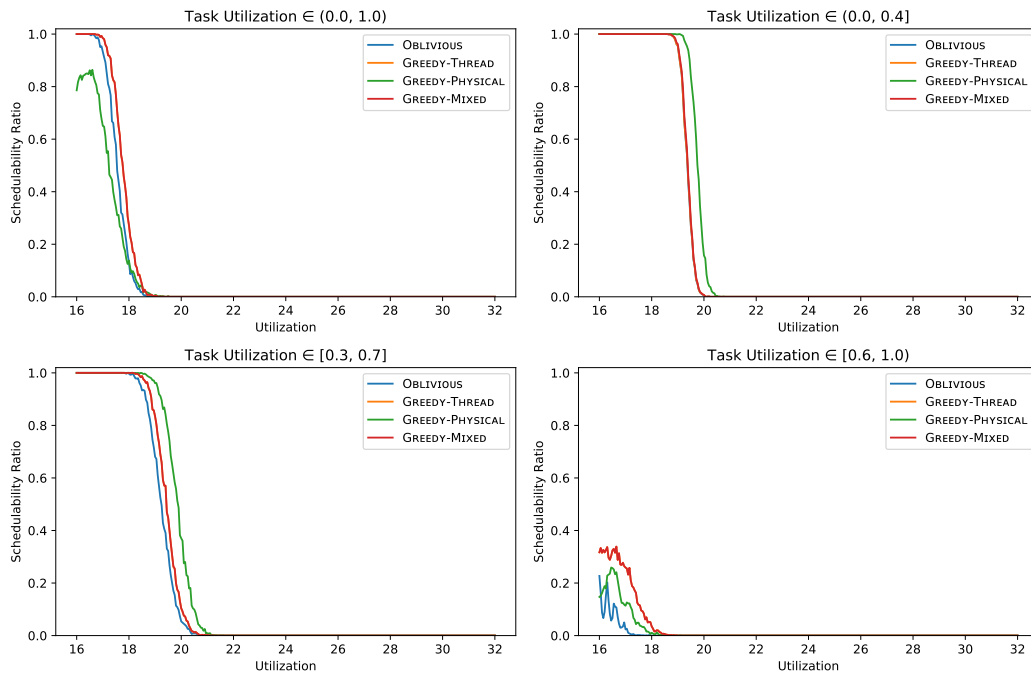


22:70 Simultaneous Multithreading Applied to Real Time

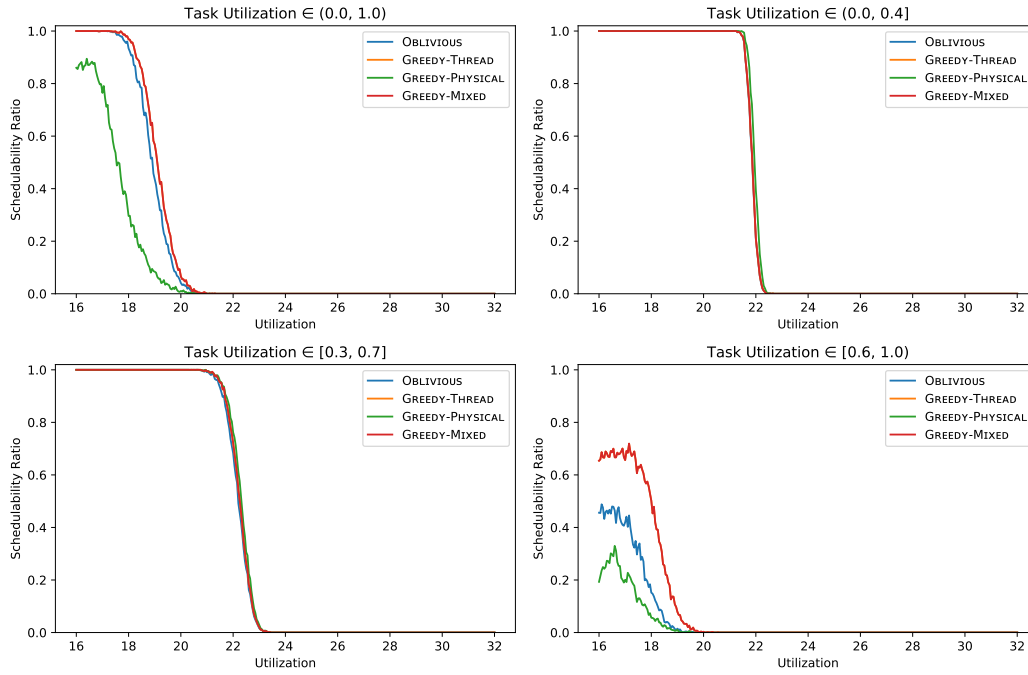
16 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$



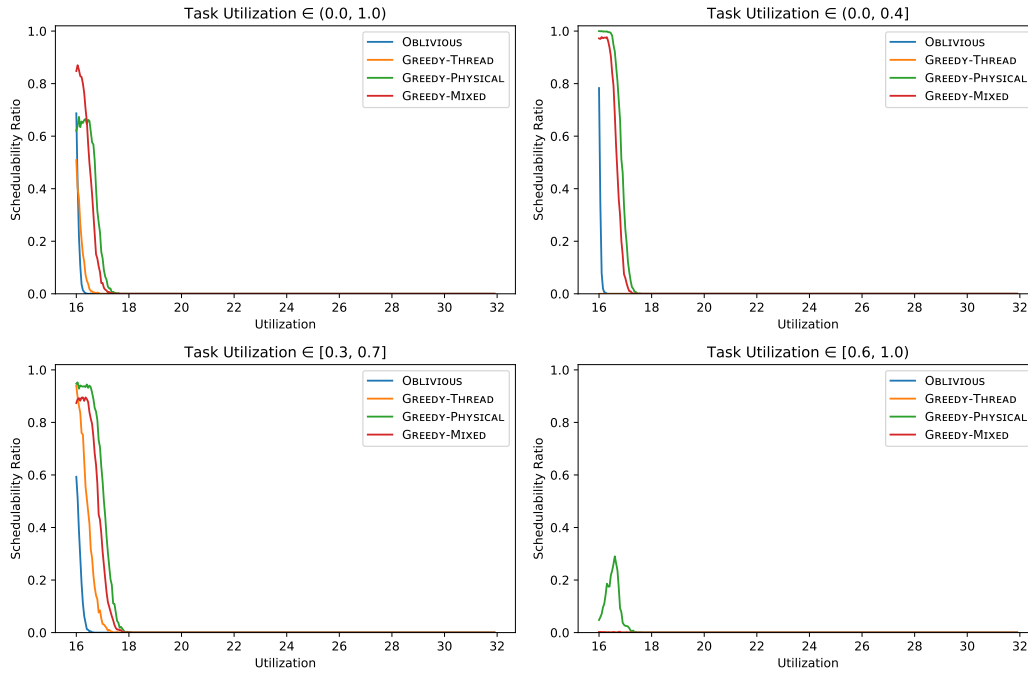
16 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$



16 Cores, $\epsilon = 0.055$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$

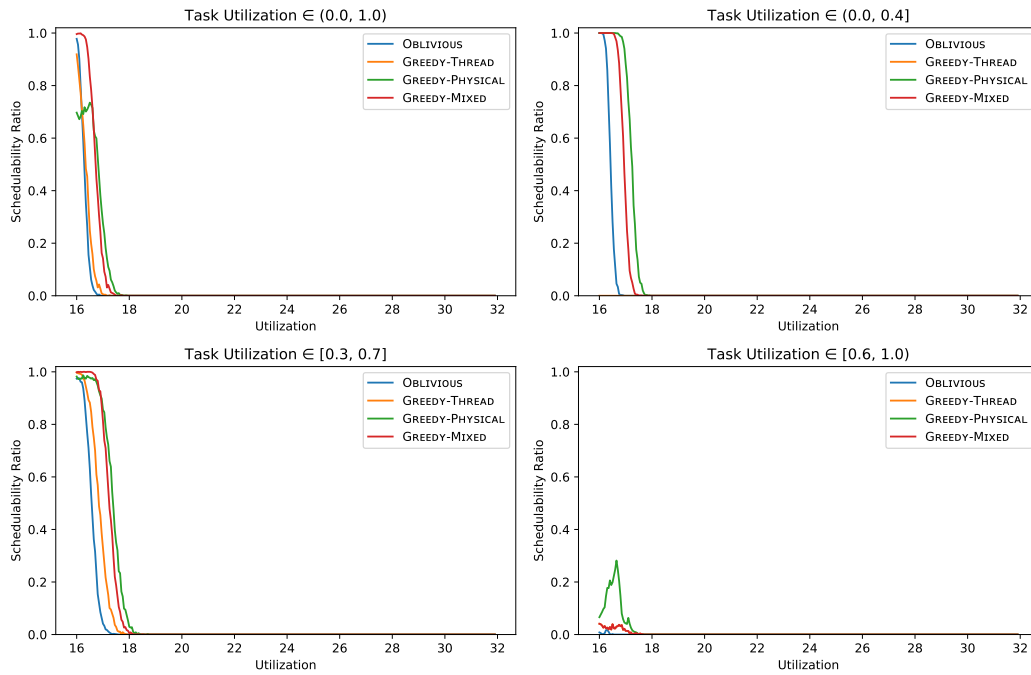


16 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$

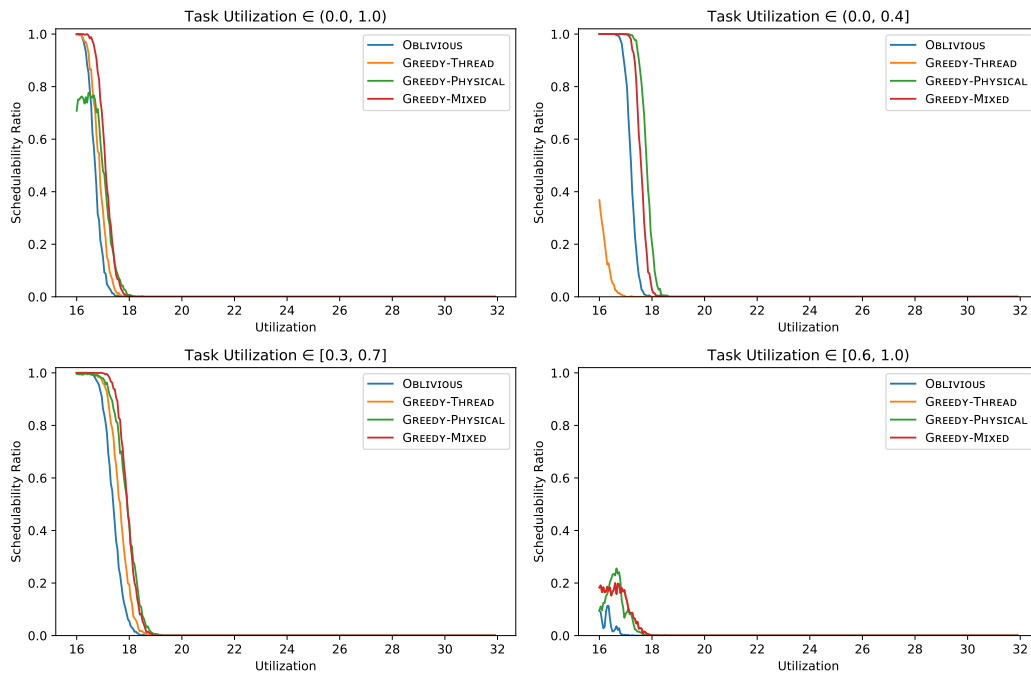


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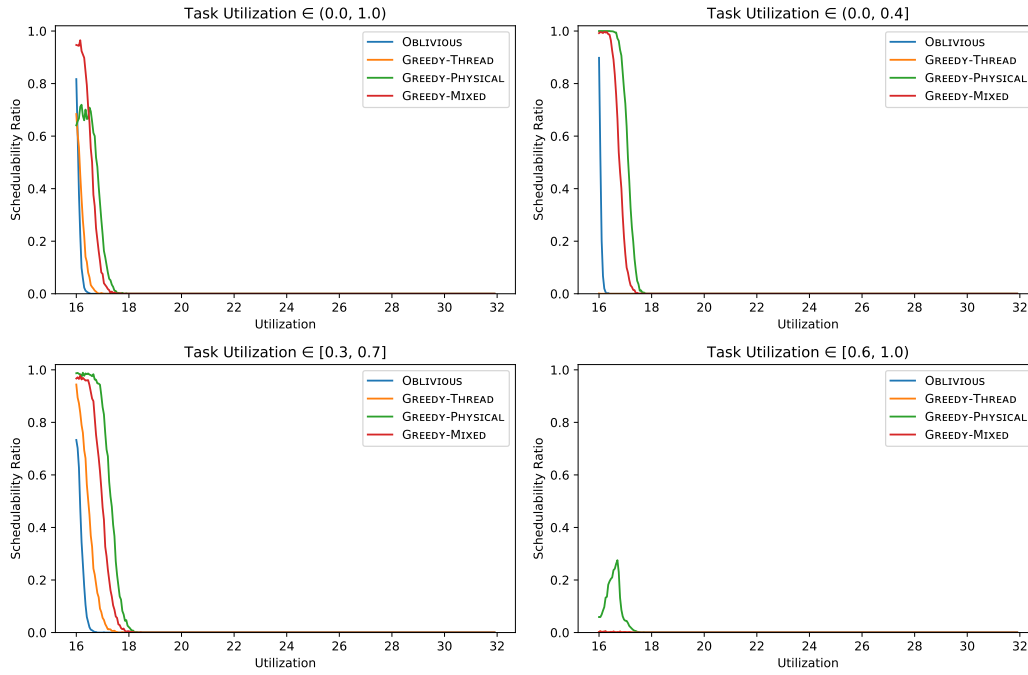
16 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$



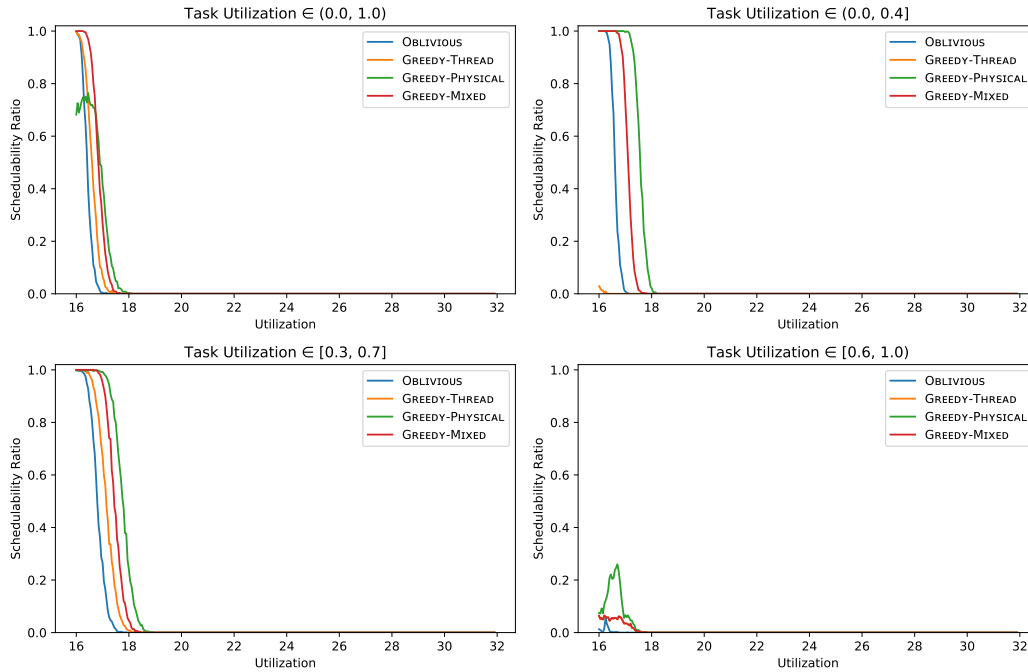
16 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.65, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$



16 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$

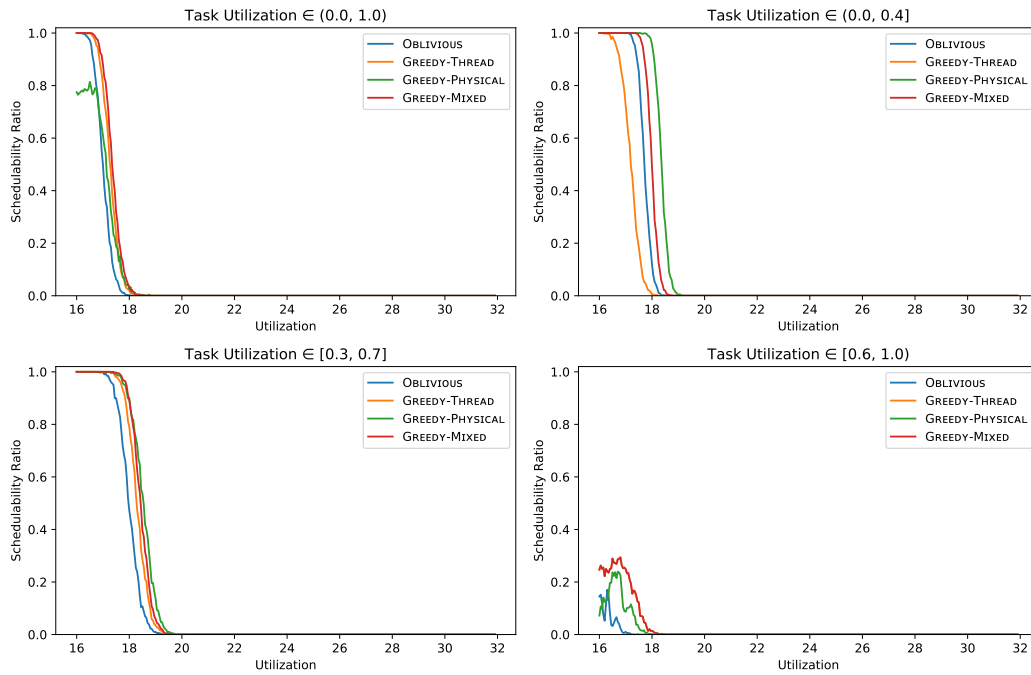


16 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$

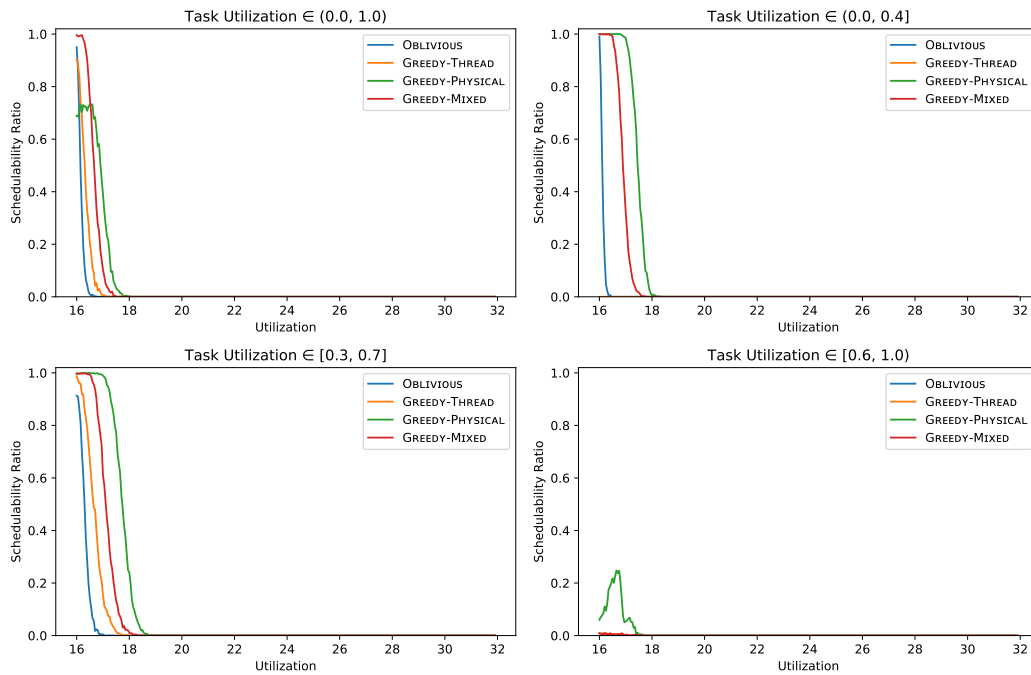


22:74 Simultaneous Multithreading Applied to Real Time

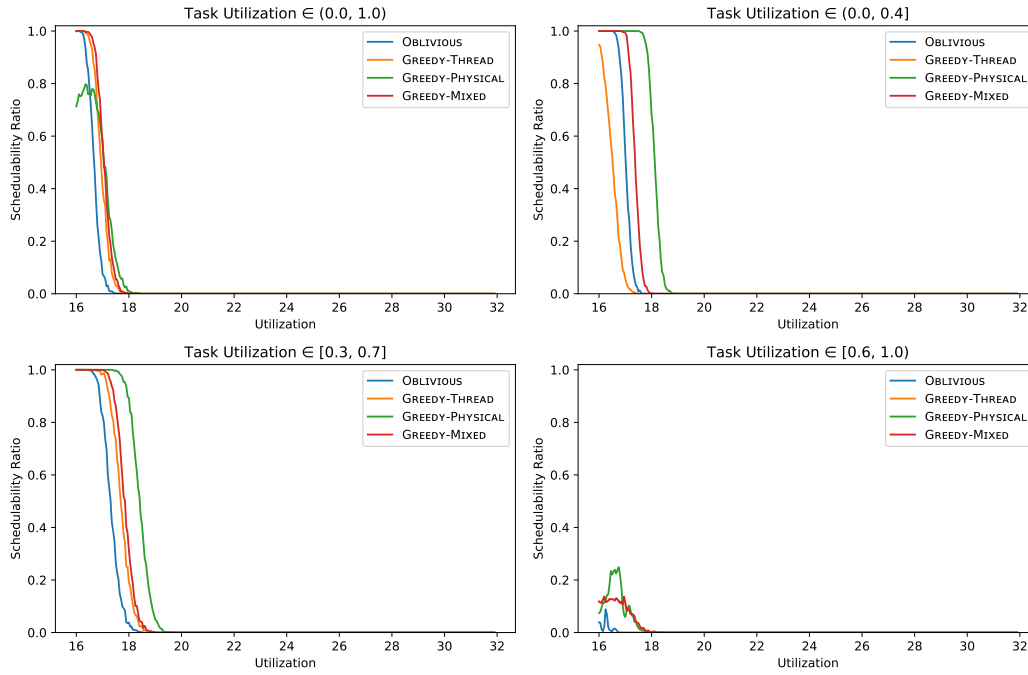
16 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.75, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$



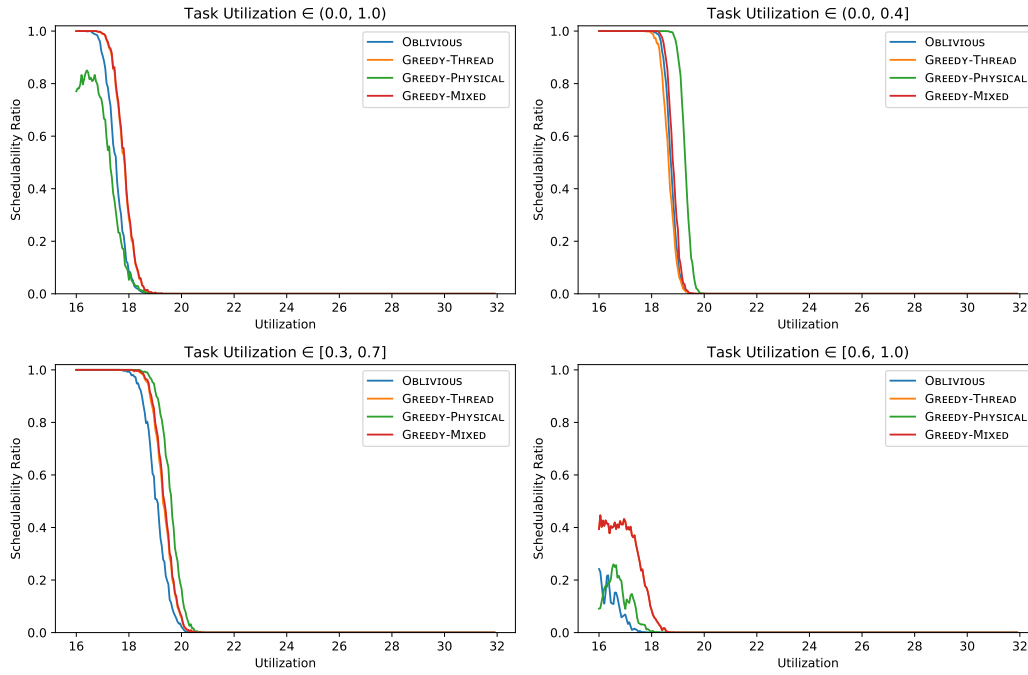
16 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.65, 1.00)$



16 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.75, 1.00)$



16 Cores, $\epsilon = 0.100$, $s \in \text{Uniform}(0.85, 1.00)$, $f \in \text{Uniform}(0.85, 1.00)$



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