# Analysis of SMP Kernel Scheduling

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#### Abstract

As algorithms become larger and more resource intensive, computer hardware must become increasingly efficient to effectively run such programs. In particular the capability for multiprocessing allows systems to distribute work across multiple CPUs. Modern operating systems and their respective schedulers have the opportunity to make policy decisions, such as load balancing and locality of access to shared memory, when distributing programs that can influence the performance of both the kernel and user programs.

# 1 Introduction

Hardware has evolved to become increasingly more complex and efficient, through areas such as processor frequency and memory latency. Amdhal's law gives diminishing returns on the impact of parallelizing computation. However, with respect to symmetric multiprocessing (SMP) in the kernel, Gustafon's law proves as now as a **scheduler?** can run "in the same time with a larger workload" [3]. Thus as new system architectures take advantage of these processors. They also require new approaches for system software and applications that run on top of this hardware [1].

Within an operating system, the task scheduler is responsible for running user tasks and determining an execution order. Yet with multiprocessing the scheduler offers an opportunity to distribute work across multiple cores. In a preemptive system, the scheduler may migrate tasks between different cores which generally will incur a local penalty in cache performance yet may improve global performance. Thus for this report several variations on the classical task scheduler will be compared to their impact on system performance and task scheduling.

A run-queue will consist of a FIFO list of tasks and will be protected by a run-queue specific lock. The performance of several tasks will be measured using a single global run-queue, a per-core run-queue without pull migration, and a per-core run-queue with pull migration. Pull migration refers to periodically re-balancing the load on two run-queues. In addition, some tasks will be constrained to a smaller set of CPUs and thus may be fixed to a single process or not undergo push migration. This paper will demonstrate the performance implications that various scheduling algorithms have on both local and global task execution.

# 2 Environment?

The OS will be compiled for the ARMv8 (aarch64) instruction set and will be executed on the Raspberry Pi 3 B. The device includes a 1.2 GHz 64-bit quad-core ARM Cortex-A53 processor and consists of a 16KB L1 data cache, a 16KB L1 instruction cache, and a 128KB L2 cache. Intra-core cache coherency is enabled and uses the MOESI protocol [2]. The memory address space is configured using a linear two level translation table with accessible memory configured as normal outer and inner write-back, write-allocate and device register memory configured as device nGnRnE. Performance will be measured using the ARM Performance Monitor Unit (PMU) and the core timer. The timer executes at a frequency of 19.2 MHz [6] and each core maintains its own core timer which will preempt the current task each MS. Both the kernel and pseudo-user processes will execute in EL1; however, the kernel will use SPx and the user processes will use SP0.

# 3 Definitions?

A ready queue can be defined as a generic data structure with two primary operations: ready and next. For its internal manipulation a ready queue is protected by a spinlock. Internally there are several queues corresponding to different priority levels allowing for processes with different priorities to be pushed onto the queues with no contention. However, for discussion we will assume that all processes run at the same priority.

A naieve implementation for a SMP scheduler involves a global ready queue which all cores pull processes from. This naieve implementation acts identical that in a single core system. On SMP systems this data structure is benificial since it guarantees equal distribution of workload across all cores in addition to implementation simplicity. Additionally, this workload distribution is not affected by a rapid stream of exiting processes. However, the approach suffers from kernel lock contention since all cores must synchronize access to the queue.

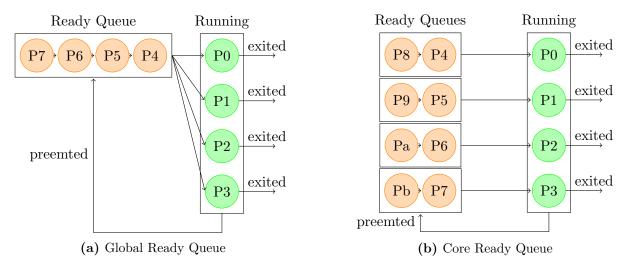


Figure 1: Scheduler Structures

Under the current scheduler, a process, if eligible, is migrated during preemption. In general the scheduler will opt to keep a process on the same CPU unless another core is sufficiently empty or its affinity set dictates otherwise. In essence this rebalancing means that as processes are rescheduled, the cores will tend towards a balanced configuration. However, it is also possible that a core will have a stream of processes terminate leading to a load imbalance between ready queues (fig. 3).

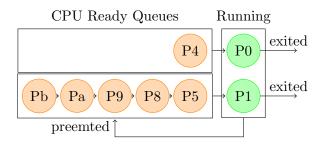


Figure 2: Unbalanced Scheduling

Pull migration is an effort to improve the re-balancing latency for queues which exhibit the behaviour in fig. 3. Periodically, an idle core will search for an overloaded core and pull eligible processes from the tail of its queue. This task requires locking the busy queue for a longer period of time; while, the idle queue tends to have have its lock acquired more frequently. This task leads to high lock contention for both queues which can negatively impact the time spent in the kernel. Specifically, while iterating, the busy core will

be unable to retrieve the next runnable process. Thus it is important to pick a large enough interval to do the re-balancing that it does not cause much strain on performance. However, in most cases, if a core is not sufficiently loaded the pull migration task will instead exit prematurely instead of locking both cores.

**TODO: Process Fairness** 

# 4 Performance

Different user programs were tested against the variations of the OS scheduler described previously. In general the tests consist of performing a scalar multiply for a specified runtime. These tests aim to simulate process cache access and utilization and provide repeatable results for user and kernel event counters. For example, some examples heavily utilize memory while others operate on smaller subsets of data. Additionally other processes are designed to run for longer or shorter periods of time to represent variability in task runtime.

While the actual measured runtime is presented in jiffies, these measurements should rarely be compared as they are rough measurements based on the CPU timer. Additionally, tasks which may have a variable amount of IO with the FTDI output device are likely to see unrepresentative results compared to the actual execution of tasks. Finally it is also important to note that all data fits into the L2 cache and thus pipeline stalls involved when fetching from main memory are limited the initial cold start. **TODO: talk about the L2 cache in conclusion as it amplifies performance penalty** 

To describe the behaviour for execution some terms should be defined.

- Work: the local size of data available for execution in a single thread, specifically related to the dimensions of the L1 data cache.
  - $\text{ uint} 64 \text{\_t} [20][25] = 25\% \text{ L1 data cache}$
  - $\text{ uint} 64\_t[40][50] = 100\% \text{ L1 data cache}$
  - $\text{ uint} 64 \text{\_t} [80][100] = 400\% \text{ L1 data cache}$
- Samples: the global number of work entries available for execution by all threads
  - 4 independent work entries = high spatial locality
  - 16 independent work entries = low spatial locality
- Difficulty: a relative measure of bounded execution time by a single thread
  - $-10^2$  repeated computations = easy
  - $-10^3$  repeated computations = normal
  - $-10^4$  repeated computations = hard

Figure 3: single core baseline comparison

$\operatorname{Type}$	Total	User	Kernel
Instrs	467617788	466380022	1237766
Cycles	693267321	691411106	1856215
Access	264291472	263665636	625836
Refill	1539	1340	199
Runtime	45374884	-	

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Type	Total	User	Kernel
Instrs	1858506765	1853566311	4940454
Cycles	2768987855	2760678046	8309809
Access	1050488228	1047990323	2497905
Refill	1389770	1293188	96582
Runtime	178608294	-	-

(b) single core - matrix 4x utilization

These initial examples were designed to demonstrate the performance penalty of poor cache utilization in user programs. **TODO: reword intro** The first example will operate on 100% of the cache while the second

will use 400% thus requiring a full cache refill for each iteration (with some degree of randomness relating to the replacment policy). These examples serve as important baselines for demonstrating predicable behaviour when scaling a problem up to multiple cores; having a 4 times larger cycle count and nearly 900% larger refill rate when the L1 cache is consistently overfilled. This results in frequent thrashing of kernel memory leading to roughly 11% less IPC. For the next example, the highly utilized ...

#### TODO: note about how to improve global runtime in userspace programs

Figure 4: strided matrix multiplication running against different schedulers

Type	Total	User	Kernel	Type	Total	User	Kernel
Instrs	1316124843	1313462560	2662283	Instrs	1168350039	1164789764	3560275
Cycles	1960863928	1956137419	4726509	Cycles	1734415365	1728844371	5570994
Access	744181293	742708916	1472377	Access	660416972	658632902	1784070
Refill	537874	509212	28662	Refill	69361	47662	21699
Runtime	41175374	-	-	Runtime	40603738	-	-

<sup>(</sup>a) multi core - global queue

(b) multi core - cpu queue (affinity)

Extending the previous examples to a multicore system serves as a demonstration to the benefit of well thought out parallelization. A single sample executing on 400% of a processors cache at hard difficulty was executed against the global scheduler and the core scheduler with fixed process affinity to separate computation onto each of the four cores. Even without comparing the difference between the two schedulers, the global queue outperforms the single core example working on a matrix of the same size. This was particularly interesting since it is not inherently obvious what may be causing this superliner speedup.

However, this is likely a result of data being precached in the L2 cache by the other processors. As a result of striding the computation, the innermost edges of data will exhibit high spatial locality despite having low temporal locality. However, having low temporal locality effectively acts as a form of prefetching for subsequent accesses. This prefetching results in data being present in the L2 cache well before its equivalent access to main memory would be made. Since the single core example does not perform inner striding, these performance gains cannot be made. Although, the stride length was chosen to be quite large for these examples, decreasing the stride length would allow more data to be prefetched in the L2 cache depending on scheduling and process execution order.

Aside from locality improvements from the single core example, there is a notable difference between multicore scheduling types. Primarily there is a 13% decrease in cycles which can be attributed to the **TODO:** 700% difference in cache refills and the 12% decrease in total instructions, primarily in userspsace. calculate better...

To test the behaviour of the system under a rapid stream of terminating processes difficulty was split into 12 hard threads and 60 easy threads designed to quickly terminate. Each matrix would operate on 25% of the cache and the examples were under each of the sample data locality cases.

For these examples, the impact of pull migration is immediately evident in the reduced cycle count of 75-80% and a maximum reduction to 25% of the original l1 refill rate to the same tests running without push migration. Despite the apparent cache benefit it is important to note that the cycles spent in userspace tend to remain around the same meaning that the more important improvement ends up in the kernel. When investigated more, it is noticable that the kernel cycle count and number of instructions executed drops significantly with pull migration. Since the change in the kernel refill rate more insignificant this decrease in kernel overhead seems to be highly related to decreased lock contention between the cores.

This contention is derived from the scenario of multiple busy queues attempting to rapidly push work onto an idle queue. In these cases, the added lock contention will inhibit processes from being scheduled on all cores since they are all effectively being serialized by the idle queue runqueue lock. With pull migration, instead of having many repeated requests to the idle cores runqueue, a busy core elects to offload a larger amount of work until the idle core is no longer considered idle. This prevents other cores from offloading work unnecessarily leading to improved userspace cache behaviour for those cores as well as reduced lock

Figure 5: 60 threads consisting of 48 hard processes and 12 easy processes

Type	Total	User	Kernel	Type	Total	User	Kernel	
Instrs	1543135939	1396850683	146285256	Instrs	1468299596	1398454567	69845029	
Cycles	3146576481	2076522276	1070054205	Cycles	2552848279	2079318879	473529400	
Access	845516444	789349384	56167060	Access	816022596	790256656	25765940	
Refill	7653514	112360	7541154	Refill	3126952	140213	2986739	
Runtime	152657484	-	-	Runtime	156804356	-	-	
(a) without pull migration (high locality)				(b)	(b) with pull migration (high locality)			

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(b) with pair inigration (ingli locality)							
рe	Total	User	Kernel				
strs	1520575475	1445743126	74832349				
1	2574252547	0147070620	496490017				

Type	Total	User	Kernel
Instrs	1613944771	1404552990	209391781
Cycles	3491698429	2086636583	1405061846
Access	873214573	793699999	79514574
Refill	9792188	28820	9763368
Runtime	152321428	_	-

Ins 49 Cycles 426480917 25743535472147872630 Access 816978128 26665609 843643737 Refill 2283806 34930 2248876 Runtime 161631896

contention and minimal serialization. Additionally, pulling processes from the end of the runqueue; instead of migrating recently executed processes, provides the opportunity to exhibit temporal locality and less cache pollution from more processes being executed.

TODO: cleanup these paragraphs

Figure 6: even distribution of easy, normal and hard threads

Type	Total	User	Kernel	Type	Total	User	Kernel
Instrs	2945961318	2829092653	116868665	Instrs	2727027846	2625810066	101217780
Cycles	4881259967	4206238780	675021187	Cycles	4421745433	3903078399	518667034
Access	1642115209	1598721424	43393785	Access	1520521397	1483845439	36675958
Refill	4570467	298213	4272254	Refill	3226996	222090	3004906
Runtime	175864992	-	-	Runtime	171909144	-	

<sup>(</sup>a) without pull migration

While the previous example demonstrated the immediate benefit of pull migration, this case utilizes an even difficulty distribution with each thread using 25% of the L1 data cache and low spatial locality across all cores to demonstrate long term scheduling benefits. Specifically, despite lower lock contention with pull scheduling, the processors are able to minimize the migration overhead relating to cache invalidation. Since tasks are migrated from the end of the runqueue, it is highly probable that a tasks would have its cache fully polluted on a busy core by the time it was next executed. Thus migrating to an idle queue helps distribute evenly distribute memory access across all cores reducing the impact on any one individual core. Additionally, the higher cache utilization leads to fewer stall cycles and thus more opportunity to complete work earlier in userspace leading to fewer preemption context switching.

In Fig. 7, a work sized at 25% of the L1 data cache is distributed to 64 threads scheduled into independent threads assigned work or 16 worker threads pulling work from a protected queue. They run under the same difficulty; however, locality between the two classes of samples is evaluated separately.

TODO: try this with semaphore TODO: remeasure runtimes? repeat tests with different MD case? TODO: mention that these were unexpected, was expecting much better performance with the pooled - md This example has ipc larger in both cases. However, one noticeable factor is that for larger sample size there is less 11 thrashing relevant in the APR (contributing factor for larger runtimes). Instead for the smaller case the caches are able to be fully utilized.

<sup>(</sup>a) without pull migration (high locality)

<sup>(</sup>c) without pull migration (low locality)

<sup>(</sup>d) with pull migration (low locality)

<sup>(</sup>b) with pull migration

Figure 7: comparison between thread pool and fully threaded execution

Type	Total	User	Kernel	Type	Total	User	Kernel
Instrs	836416320	753346040	83070280	Instrs	811243405	741653152	69590253
Cycles	1724034904	1123196400	600838504	Cycles	1789217832	1113471653	675746179
Access	456170314	426134109	30036205	Access	444450115	419398612	25051503
Refill	4027452	231590	3795862	Refill	4540858	379326	4161532
Runtime	165276806	-	-	Runtime	66534692	-	

<sup>(</sup>a) threads (high locality)

(b)	pooled	(high	locality)
(D)	pooled	(111811	10 Carrey

Type	Total	User	Kernel	Type	Total	User	Kernel
Instrs	819345284	755845609	63499675	Instrs	840132648	753613514	86519134
Cycles	1672284752	1125235476	547049276	Cycles	1849770029	1127098953	722671076
Access	450878090	427546852	23331238	Access	457128942	426166739	30962203
Refill	3740754	148469	3592285	Refill	4435166	143616	4291550
Runtime	167143162	-	-	Runtime	66158114	-	-

<sup>(</sup>a) threads (low locality)

# 5 Conclusion

Conclusions here [4] [5].

# 6 Lessons Learned

Since the development of the operating system was done from the ground up there were plenty of learning opportunities along the way. **TODO** 

Measuring success in a bare metal environment can prove to be quite difficult; thus it became important to progressively escalate to better debugging utilities. Initially, bare minimum assembly required to toggle an LED gave limited success or failure output. The jump to FTDI was helpful in debugging simple conditions and CPU registers. Finally, for tracing complex execution; having a proper debugging protocol with JLink, OpenOCD and GDB enabled better debugging for context switching, exception handling and multicore execution.

The initial JLink setup proved to be quite tricky due to incompatability with the current version of OpenOCD, mismatched toolchain and microcontroller architecture/configuration. However, the end result was incredibly important to establishing a sucessfull development workflow. Instead of having to perform a careful SD card dance between my computer and the Pi, having the capability to flash to RAM allowed for much quicker development cycles. Initially, I had looked into alternative bootloaders (UBoot, GRUB) which can work over ethernet, WIFI, or serial connections; however, in general the end result of having proper host debugging utilities was invaluable.

On the topic of toolchain incompatability, code was initially compiled using arm-none-eabi which is used for cross compiling to 32-bit ARM and while aarch64 offeres compatability with older architecture, the generid debug interface is much less compatible. Thus it is important to recognize that despite older systems often having more documentation; the complexity of developing on a compatability layer is not worth the extra effort. Additionally, the aarch64 architecture proved to have clear and better structured semantics especially related to exception handling. The exception model for aarch64 is standardized across four types; whereas, the ARMv7-A architecture had various subsets of banked registers and return instructions depedent on the interrupt source. Additionally, the addition of a unified exception status register (ESR) was much more helpful with determing the cause of a processor fault.

Newlib was designed to provide the C library on embedded systems [7] with minimal system support handled by low level functions. For example, the overhead of implementing the malloc and printf routines

**<sup>(</sup>b)** pooled (low locality)

was reduced to simply stubbing out sbrk and establishing a minimal device descriptor table for stdout. As I built the system with the design of having POSIX compliancy, I additionally included the <sys> headers from Newlib; which often imposed lower level structure restrictions. For example, when implenting CPU sets I wanted to follow the same type of implementation as embedding the affinity set in the pthread\_attr\_t which was hidden around several macro guards; which when enabled led to conflicts in other areas. Thus for long term kernel development; having custom POSIX headers is beneficial to customizablility of the system and allows the system to not be constrianed by "immutable" or static headers.

# 7 Future Work

- exception higherarchy from EL0 up to EL1
- better POSIX compliancy thus port more complex programs to run tests against. eg. real programs; what would the necessary components be?, networked bounded programs, file bounded programs
- device drivers in multicore domain, network drivers would likely be interesting to explore. also sockets and their implementation
  - user virtual address space and compare the performance penalty for TLB reloads
  - slab allocation algorithms for a more complex system
- implementation of the linux rb-tree scheduler with implications for migration, variable quantum size + penalizing bad programs
  - better migration policies with book-keeping.
  - microkernel architecture + performance
  - run to idle then pull, with idle sleeping

# References

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