

Job-Level Batching for Software-Defined Radio on Multi-core

Abby Eisenklam
University of Pennsylvania*

Will Hedgecock
Vanderbilt University

Bryan C. Ward Vanderbilt University



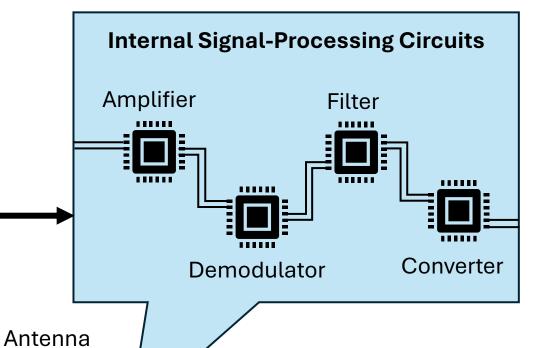


*work done while at Vanderbilt

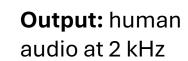
Traditional Hardware Radio

Application-specific hardware: efficient and predictable, but rigid

Input: data encoded on a 100 MHz radio wave



Speaker



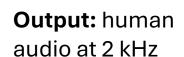


Software-Defined Radio

Software applications on GPPs: flexible, upgradeable, and adaptable

Input: data encoded on a 100 MHz radio wave

Antenna Dongle



Signal Processing Algorithms

amplify()

demod()

convert()

filter()





Why SDR on General Purpose Processors?

Flexibility	 Support multiple frequency bands and communication protocols More complex algorithms Spectrum sensing, signal classification, etc.
Upgradeability	 Efficient and cost-effective updates when technology/protocols change No need to buy new hardware, reconfigure FPGA, etc.
Adaptability	 Detect and adapt in real-time to congestion, signal interference, or malicious jamming detected in environment* Dynamic spectrum sensing Cognitive (autonomous) radio

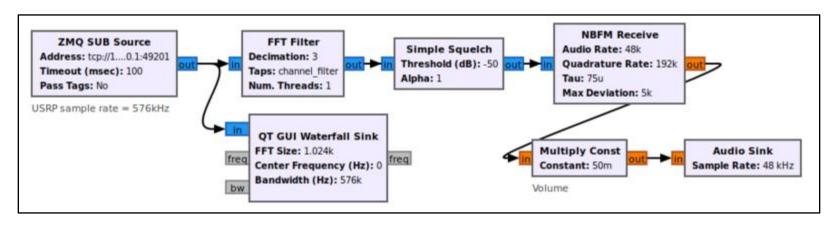
Application Domains for SDR

- Wireless networks (e.g. WiFi, Bluetooth, IoT)
- Cellular technologies (e.g. 5G)
- Satellite communications
- Spectrum sensing

Multi-core SDR is not yet widely used in these domains due to **unpredictable** latency and resource inefficiency.

GNU Radio

- Developers can build applications by combining signalprocessing "blocks"
 - filters, amplifiers, etc.
- Blocks are connected by FIFO buffers



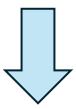
GNU Radio "Scheduler"

- Each block is assigned its own thread
- Threads are woken up when their input buffers are at least half full
 - PRO: Reduces context switching, prioritizes throughput
 - CON: Nondeterministic input size, variable execution times
- Actual scheduling decisions left to underlying OS

GNU Radio provides no control over factors that affect real-time performance*

What can we do?

 Goal: make SDR applications more predictable and hardwareefficient on general-purpose multiprocessors



- Challenge: traditional real-time scheduling models target applications with slower sampling rates
 - Period of 10ms corresponds to frequency of 10 Hz
 - Input signal frequency can be on the order of 10 GHz, 10⁹ times faster
- Context switch overhead is large compared to computational cost
- Batch processing samples is fundamental to supporting SDR on multi-core

Research Questions

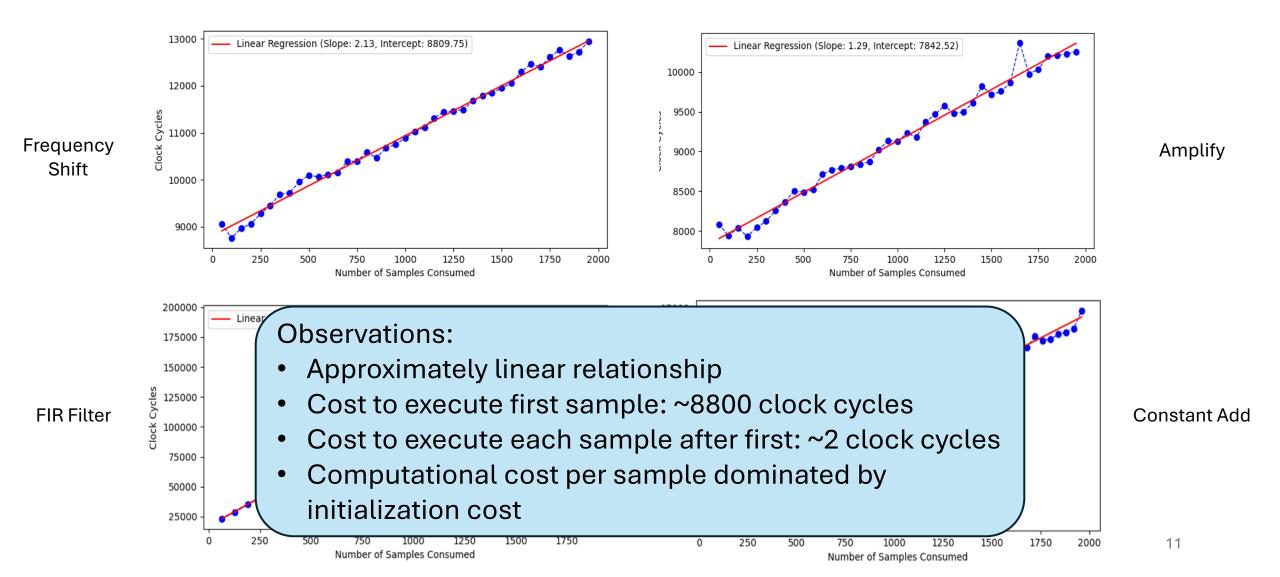
1. What is the relationship between batch size and execution time?

2. How do we choose batch sizes such that an application's latency requirements and resource constraints are met?

Outline

- 1. Motivation
- 2. GNU Radio Experiments
- 3. The Marginal Cost Model
- 4. Batching Techniques
- 5. Latency Analysis
- 6. Evaluation
- 7. GNU Radio Case Study

Batch Size vs. Execution Time in GNU Radio



Outline

- 1. Motivation
- 2. GNU Radio Experiments
- 3. The Marginal Cost Model
- 4. Batching Techniques
- 5. Latency Analysis
- 6. Evaluation
- 7. GNU Radio Case Study

The Marginal Cost Model

- The relationship between batch size (N) and execution time (e) is linear
- Y-intercept represents a fixed initialization cost (I)
- Slope represents the marginal cost (Δ) to process one additional sample
- The execution time of a task τ is given by:

$$e_{\tau} = I_{\tau} + N \cdot \Delta_{\tau}$$

Batching Jobs with the Marginal Cost Model

- GNU Radio experiments showed that the initialization cost, I >>> marginal cost, Δ
 - by a factor of 100x to over 1000x
- We can batch jobs to reduce the number of initialization costs needed to do the same amount of work



Execution time before batching

Execution time after batching

Research Questions

1. What is the relationship between batch size and execution time?

2. How do we choose batch sizes such that an application's latency requirements and resource constraints are met?

To answer this question, we need a real-time framework.

Model and System Requirements

Model of Computation

- DAG-based
- Dataflow
- Fixed source frequency
- Blocks have some consume-to-produce ratio

System Requirements

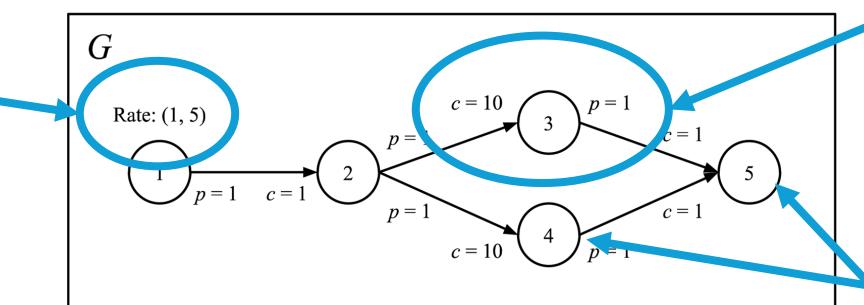
- Sufficient cores for application frequency/throughput
- No strict deadlines, but bounded latency

Even better:

- Efficiently utilized cores
- End-to-end latency bounds

Processing Graph Method (PGM)

Source node executes once every 5 time units



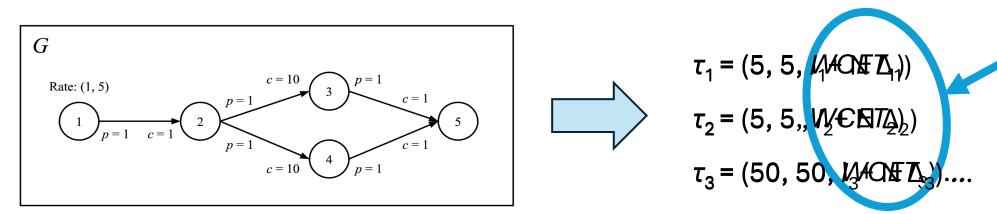
Node 3 needs to consume 10 data samples to produce one output

Rates of nonsource nodes calculated using source rate and produce-consume ratios*

^{*}S. Goddard. On the management of latency in the synthesis of real-time signal processing systems from processing graphs. PhD thesis, UNC, Chapel Hill, NC, 1998.

PGM Graphs on Multi-core

- PGM graphs can be soft real-time scheduled on multi-core with no utilization loss*
 - Calculate rates for each node of a PGM graph → transform to independent sporadic tasks
- Tardiness w.r.t. individual deadlines is bounded under G-EDF**



Replace with marginal cost model

^{*}C. Liu and J. Anderson. Supporting soft real-time DAG-based systems on multiprocessors with no utilization loss. In RTSS '10.

Outline

- 1. Motivation
- 2. GNU Radio Experiments
- 3. The Marginal Cost Model
- 4. Batching Techniques
- 5. Latency Analysis
- 6. Evaluation
- 7. GNU Radio Case Study

Uniform Batching

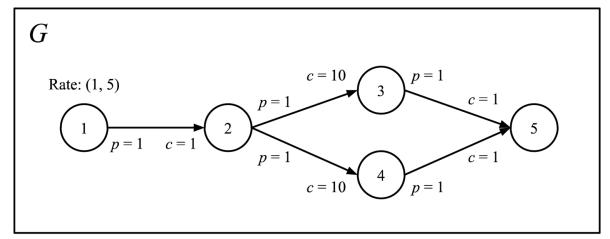
- Suppose we have a PGM graph G with batch size N
- Then the sporadic parameters for each $\tau \in G$ are given by:

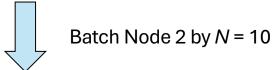
$$\tau = (Np_{\tau}, Nd_{\tau}, I_{\tau} + N\Delta_{\tau})$$

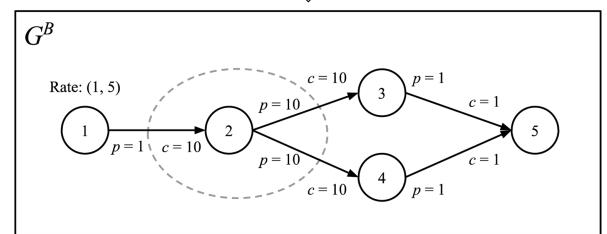
where p_{τ} and d_{τ} are calculated using source rate and unit input sizes.

Invocation of each τ is delayed until N data samples arrive, effectively batching N jobs of τ into a single job.

Rate-Exploiting Batching







Consequences:

- Rate of node 2 decreases by factor of 10
- Rates of successor nodes are unaffected
- Reduced utilization without impacting output rate

Outline

- 1. Motivation
- 2. GNU Radio Experiments
- 3. The Marginal Cost Model
- 4. Batching Techniques
- 5. Utilization and Latency Analysis
- 6. Evaluation
- 7. GNU Radio Case Study

Intuition for Utilization Decrease



Execution time before batching

Execution time after batching

Taking a conservative estimate, suppose I = 100, $\Delta = 1$, and p = 300. Then,

Utilization before batching:

•
$$U = (I + 2 \Delta) / p$$

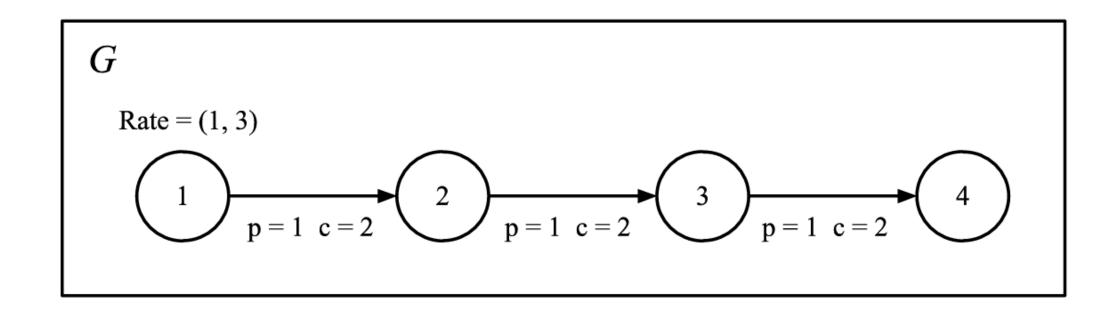
= $(100 + 2) / 300$
= 0.340

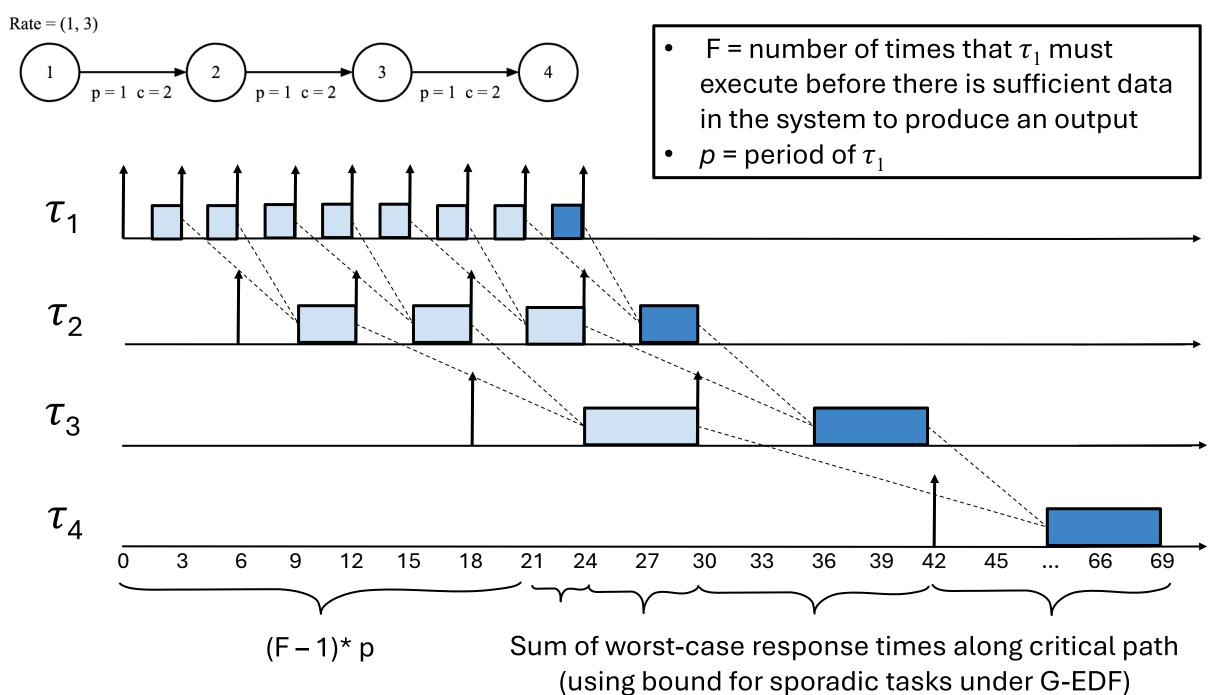
Utilization after batching:

• U =
$$(I + 6 \Delta) / 3p$$

= $(100 + 6) / 900$
= 0.118

Intuition for End-to-End Latency

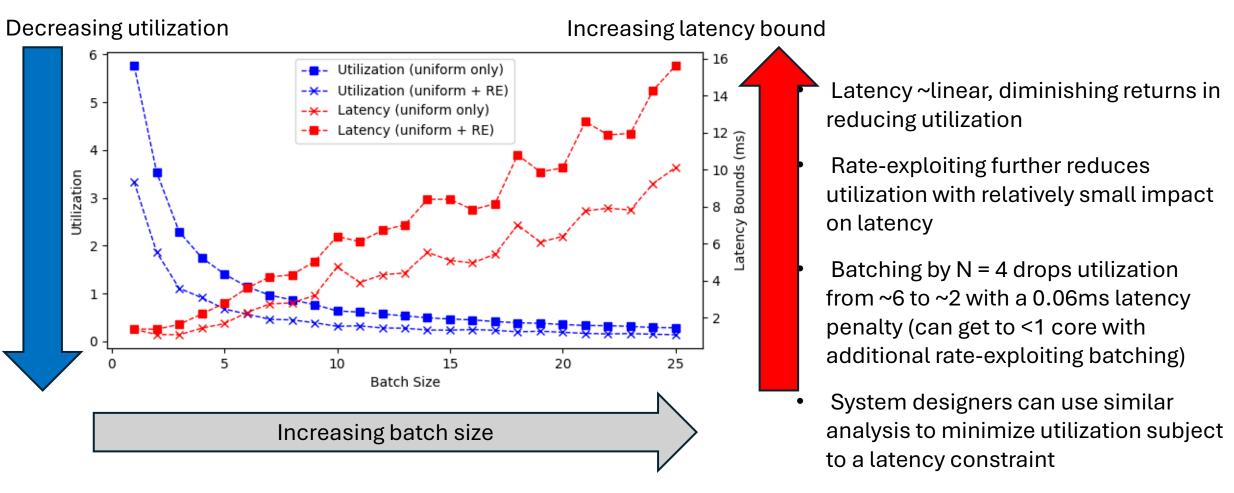




Outline

- 1. Motivation
- 2. GNU Radio Experiments
- 3. The Marginal Cost Model
- 4. Batching Techniques
- 5. Latency Analysis
- 6. Evaluation
- 7. GNU Radio Case Study

Latency and Utilization vs. Batch Size

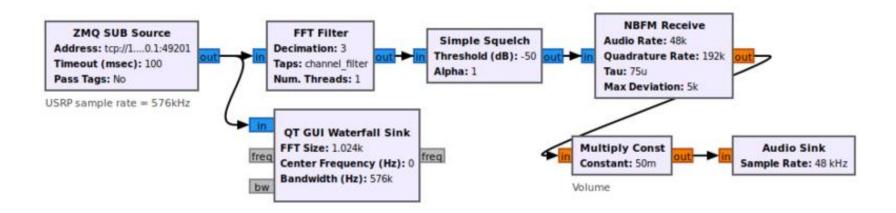


Each point is an average over 1000 synthetic PGMs, using execution times from GNU Radio experiments

Outline

- 1. Motivation
- 2. GNU Radio Experiments
- 3. The Marginal Cost Model
- 4. Batching Techniques
- 5. Latency Analysis
- 6. Evaluation
- 7. GNU Radio Case Study

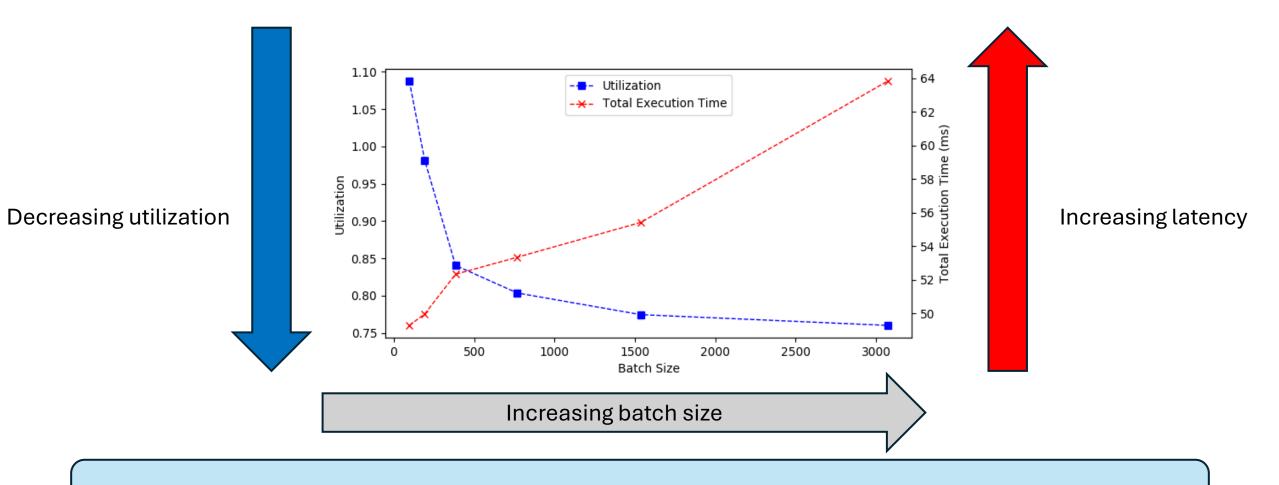
Case Study: GNU Radio NBFM Receiver



Experiment Setup

- Added an EDF API for GNU Radio that uses Linux SCHED_DEADLINE
- Isolated 3 out of 4 Raspberry Pi cores for GNU Radio tasks
 - Disabled kernel preemptions and IRQ requests from these cores
- Derived sporadic parameters for each block using source rate and produce-consume relationships
- Controlled for batch size using GNU Radio "Head" blocks and buffer sizes
- Measured max. latency using GNU Radio performance counters

Utilization and Latency vs. Batch Size for NBFM



The trends in latency and utilization as functions of batch size hold in a real SDR system.

Conclusion

- Validated the marginal cost model for batched sample processing in GNU Radio
- Presented uniform and rate-exploiting batching techniques
- Derived end-to-end latency bounds for multi-core PGM graphs scheduled under G-EDF as a function of batch size
- Evaluated randomly generated synthetic task systems to demonstrate the utilization-latency trade-off
- Extended GNU Radio to support Linux SCHED_DEADLINE and presented a case study applying the marginal cost model and batching to a real signal-processing application

