# **Process Cooperativity as a Feedback Metric** in Concurrent Message-Passing Languages

# Process Cooperativity as a Feedback Metric in Concurrent Message-Passing Languages

by

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#### **Abstract**

# Process Cooperativity as a Feedback Metric in Concurrent Message-Passing Languages

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Runtime systems for concurrent languages have begun to utilize feedback mechanisms to influence their scheduling behavior as the application proceeds. These feedback mechanisms rely on metrics by which to grade any alterations made to the schedule of the multi-threaded application. As the application's phase shifts, the feedback mechanism is tasked with modifying the scheduler to reduce its overhead and increase the application's efficiency.

Cooperativity is a novel possible metric by which to grade a system. In biochemistry the term cooperativity is defined as the increase or decrease in the rate of interaction between a reactant and a protein as the reactant concentration increases. This definition translates well as an information theoretic definition as: the increase or decrease in the rate of interaction between a process and a communication method as the number of processes increase.

This work proposes several unique takes on feedback mechanisms and scheduling algorithms which take advantage of cooperative behavior. It further compares these algorithms to other common mechanisms via a custom extensible runtime system developed to support swappable scheduling mechanisms. A minimalistic language with interesting characteristics, which lend themselves to easier statistical metric accumulation and simulated application implementation, is also introduced.

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#### Introduction

Runtime systems can be broken up into multiple distinct parts: the garbage collector, dynamic type-checker, resource allocator, and much more. One sub-system of a language's run-time is the task-scheduler. The scheduler is responsible for order of task evaluation and the distribution of these tasks across the available processing units.

Tasks are typically spawned when there is a chance for parallelism, either explicitly through spawn, or fork commands or implicitly through calls to parallel built-in functions like pmap. In either case it is assumed that the job of a task is to perform some action concurrent to the parent task because it would be quicker if given the chance to be parallel.

It is up to the scheduler of these tasks to try and optimize for where there is opportunity for parallelism. However, it's not as simple as evenly distributing the tasks over the set of processing units. Sometimes, these tasks need particular resources which other tasks are currently using, or maybe some tasks are waiting for user input and don't have anything to do. Still worse, some tasks may be trying to work together to complete an objective, like in the pmap example above.

Tasks however, in functional language verbiage, are typically called *processes* due to the inherent isolation this term brings and the language paradigm calls for. So how do these processes share information or return a value back to their parent? Message passing is a common abstraction to shared memory. Message passing is akin to emailing a colleague a question. You operate independently, and your colleague can check her mailbox when they want to and respond when they want to. Meanwhile you are free to operate on an assumption until proven wrong, wait until she gets back to you, or even ask someone else.

While message passing is a good method for inter-process communication, it is also a nice mechanism for catching when two processes are working together. Let's, for example, consider a purely functional pmap, where all workers are copied subsections of the list. Each worker thread will have no need to cooperate and thus no messages will need to be passed amongst them. However, suppose the function being mapped uses several processes

accessing a piece of shared memory to update it's particular cell in the list. These processes would do little good if treated like the course-grained parallelism the pmap workers create.

Bad example, should be more real world, and doesn't flow well into the next paragraph.

There are a large number of mechanics that scheduling systems can use in an attempt to improve work-load across all processing units. Some of these mechanics use what's called a feedback system. Namely, they observe the running behaviour of the application as a whole, (i.e. collect *metrics*), and modify themselves to operate better.

Process cooperativity is an interesting metric by which to grade a system. In biochemistry the term cooperativity can be defined as an increase or decrease in the rate of interaction between a reactant and a protein as the reactant concentration increases. We can translate this into an information theoretic definition:

**Definition 1.** The **degree of cooperativity** of a system is the increase or decrease in the rate of interaction between processes and an inter-process communication method as the concentration of processes fluctuate.

Thus, when a process attempts to pass a message to another we know it's trying to cooperate on some level. When this frequency of interaction is high, it may indicate a tight coupling of processes or fine-grained parallelism. If it is low, this could indicate course-grained parallelism. In either event, a scheduler able to recognize these clusters of cooperative and non-cooperative processes should have an edge over those that don't.

Chapter 2 will look first at the background of classical scheduling systems as well as the recent feedback-enabled approaches. Then we will also examine the types of message passing implementations and how these effect scheduling decisions, now that we are looking at process cooperativity. Chapter 3 introduces our work on a language and compiler, built to easily simulate system cooperativity and visualize the effects of scheduling mechanisms on these systems. We also discuss a few example mechanics which take advantage of cooperativity. Some example applications which demonstrate different degrees of cooperativity and phase changes are also explained. In Chapter 4 we run our cooperativity-enabled schedulers along with a few common non-feedback-enabled schedulers on the example applications and discuss the results. Finally, in Chapter 5 we give some concluding remarks and avenues of future research we believe would be fruitful.

## **Background**

#### 2.1 A Note on Control Theory

Since the formalization of feedback driven systems and the advent of Cybernetics, multiple fields have attempted to mold these principles to their own models; and run-time schedulers are no exception. This is due, in part, from process scheduling in parallel systems being fundamentally an NP-Complete problem [1].

Note that the simple case of runtime scheduling is called the Multiprocessor Scheduling Problem which states:

#### **Definition 2. MULTIPROCESSOR SCHEDULING PROBLEM**

Given a set of jobs  $\mathcal{J}=(J_1,J_2,\ldots,J_n)$ , a directed acyclic graph (lattice)  $L=(\mathcal{J},C)$  (indicating job dependence, and thus precedence constraints), an integer P (the number of processors) and an integer D (the deadline), is there a function S (the schedule) mapping  $\mathcal{J} \to \{1,2,\ldots D\}$  such that:

- 1. For all  $t \leq D$ ,  $|\{J_i : S(J_i) = t\}| \leq P$ . (# of jobs scheduled per time slice is no more than # of processors)
- 2. If  $(J_i, J_j) \in C$ , then  $S(J_i) < S(J_j)$ . (# jobs cannot be scheduled before their dependencies)

In runtime scheduling, the deadline, D, is incremented for each timeslice we pass. As such, it is possible for  $|\mathscr{J}|$  and thus C to fluctuate causing a need to re-find S. The continuous nature of this problem complicates the scheduling problem substantially. Instead, focus has been more fruitful when pursuing the optimization of various measurements using some particular objective function [2] to tune for particular edge cases. As such, scheduling based on such feedback metrics is not a new practice [3].

There is a big distinction though, which can be made between the effects of control theory in classical cybernetic applications versus that of run-time systems. This is primarily in the adaptation of the controller in the generic feedback loop (figure 2.1).

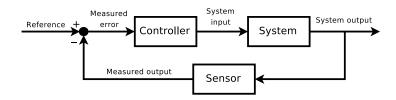


Figure 2.1: A classical feedback loop representation.

In typical physical feedback loops there are two scenarios which need to be avoided: resonance and rapid compensation. It can be seen that most controller models will attempt to damp the adjustments to reduce oscillation which could cause resonance or sharp spikes in behavior based on its output. This is due to the limitations of the physical space in which they are having to deal with.

However, in run-time scheduling systems we would very much like to do the opposite. We would prefer tight oscillations or consistent behavior of our runtime so as to achieve minimal overhead from our modifications. We can also compensate, to reach our reference signal, as quickly as we need to as there are no physical restrictions for our modifications.

As such these feedback systems are closely coupled with the design of the scheduling algorithm, rather than being an interchangable sensor, and controller modules. As such we make an effort to trace the feedback optimizations during our evaluation and explaination of the scheduler designs.

Another distinction must be made as far as the level of foresight the scheduling systems have, at least, within this paper. There is a spectrum of clairvoyance in classical job-scheduling, in that on one end, job-schedulers have full foresight over the jobs which will enter the queue and their order (i.e. the full  $\mathscr J$  set will always be known). These schedulers have the opportunity to optimize for future events (by constructing a valid lattice L based on the current time  $t \leq D$ ), which is a luxury the scheduling systems that this paper discusses do not have.

However, as it is a spectrum, there is a single point of knowledge this subrange of schedulers can assume. Namely, that the first job will always be the last, and all other jobs will spawn from it. Thus there will always be a single process in the queue at the beginning. This is true as the runtime will always require an initial primary process (*e.g.* the 'main' function), and once that function is completed, the system is terminated (despite the cases

Note: So main is required to terminate?

of unjoined children). Apart from this, all other insights will need to be gleaned from the evaluation of this initial process.

We mention this due to the implications on our scheduling problem (dynamicly evolving lattice which is altered based on evolving cooperation networks and the forking/joinging of child processes), but can we tie this back to the above definition and possibly cooperation?

#### 2.2 Classical Runtime Scheduling

I would like to discuss current methods of distributing processes across processors, as I am assuming later that readers know of terms like:

- work-stealing & sharing
- round-robin
- global/local process queues
- spawning and joining methods

Then somehow lead into process dependencies and thus process synchronization. This will tie it over to the message-passing section.

#### 2.3 Feedback-Enabled Scheduling

This is the "related work" section, I would like to give the two big examples i've been considering: CML, and Occam- $\pi$ .

#### 2.3.1 Cooperativity as a Metric

This should give the primary definitions of a *cooperative* process and the differences between focusing on cooperativity rather than interactivity, etc. It would also be a good idea to talk about its relationship with application phasses here.

#### 2.4 Message-Passing

In concurrent systems, there are a number of methods for inter-process communication. Arguably though, one of the more popular abstractions is the idea of message passing. This is expecially true in functional languages as the language assumes shared-nothing by default. Just as compilers can optimize using the language constraints, so can the run-time using the implementation.

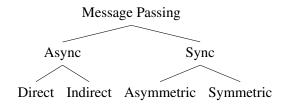


Figure 2.2: A High-Level Message-Passing Taxonomy

Message passing in general can be broken down into two types based on the language's implementation; asynchronous or synchronous. In asynchronous message passing a process can either be provided a rendezvous point or an identifier for another process. To send a message in either case requires pushing/copying the message into a shared or global memory space for another process to access (possibly) at a later time. This push/copy can be done in a lock free manner with some lower level atomic data structures such as a double-ended queue. But in either a locked or lock-free manner, the process performing the send still blocks on the operation to push the message.

In terms of scheduling, a language with asynchronous message passing can ignore the effects of these blocking operations, but will need to look closer at process placement to take advantage of possible gains in cache affinity [4]. For example, the effects of the cache on direct message passing (*e.g.* a process mailbox) can be substantial as two processes on different cores will need to copy stack frames to memory and back, rather than just a pointer to the message in the originator's stack frame. In indirect message passing the task is even worse as multiple processes may need access to the same data.

In synchronous message passing, a process must meet another at a provided rendezvous point but can either be symmetrical or asymmetrical. Note that the rendezvous point is not a requirement in the sense that direct synchronous messaging isn't possible. Instead we think of a rendezvous point in synchronous communication to be time bound rather than location bound (*i.e.* two processes are blocked until communication occurs, the implementation of this passing is irrelevant to this classification).

Asymmetrical message passing is synonymous with Milner's Calculus of Communicating Systems [5] or standard  $\Pi$ -Calculus [6], in that you have a sender, and then a receiver which will both block on their respective functions around an anonymous channel

until the pass has been completed. This differs from a symmetrical message passing in that the only operation on the channel is a blocking function which swaps values with the process at the other end.

Note, it is possible to simulate symmetrical message passing on asymmetrical message channels, but in terms of scheduling of synchronizing processes, order is now a factor that needs to be considered. On top of this, directionality can also be a factor which complicates the channel implementation. Namely, the internal queuing of senders or receivers may not percolate hints up to the scheduler regarding their queue position.

For the alternative, symmetrical message passing or swap channels, the order is directly handled by the scheduling of the system (*i.e.* the order at which the channels evaluate the *swap* command can be directly governed). And it is for this purpose along with simplifying our base language we have chosen to base our semantics on symmetric synchronous message-passing.

I feel like this section would benefit from some more figures describing the language primitives involved and their semantics. The key point to make is how much is avaliable to the scheduler in each instance:

- Async: Potentially nothing, but if so:
  - Direct: The particular process which it communicated with,
  - Indirect: The channel/rendezvous it communicated with,
- Sync: If it blocks, a communication happened, potentially can know order in queue -Asymm: Directionality.

But noting that for cooperativity purposes, directionality is not meaningful.

Should i talk more about possible implementations and their considerations? I could go into the process absorption that SML and Occam- $\pi$  does?

## Methodology

#### 3.1 Overview

To examine the effects of cooperativity conscious schedulers we needed to have a method for comparing several scheduler implementations without needing to modify the underlying implementation of processes, channels, or application source code. It would be also beneficial if our solution were able to visualize these differences similar to Haskell's ThreadScope [7].

Our solution, ErLam, is a compiler for an experimental version of Lambda Calculus with Swap Channels and a runtime system which allows for swappable scheduler mechanisms and an optional logging system which can be fed into a custom report generator. We break up our solution description into three parts; Section 3.2 will duscuss our language syntax and semantics. It will also demonstrate our Runtime Scheduler API by breaking down the CML Interactivity scheduler. Section 3.3 will go more into depth about our testing environment which involves our logging system, the report generator, and the set of example applications we used to represent different cooperativity levels. Finally, Section 3.4 will go over our example schedulers we wrote which demonstrate cooperative-conscious behavior. These will be the schedulers we provide our results against.

#### 3.2 ErLam

The ErLam toolkit is itself broken down into three parts, the language and its semantics, the Runtime System, and the Scheduler API. We will first lay out the language and it's basic semantics, as the finer-details are reliant on the exact selected scheduling solution as well as the chosen swap-channel implementation. We will then examine the possible channel implementations and how they effect the given semantics. Finally we will discuss the Scheduler API using an example scheduler implementation.

Figure 3.1: The ErLam language grammar, without syntax sugar or types.

#### 3.2.1 The ErLam Language

The ErLam Language is based on Lambda Calculus, with first-class single variable functions, but deviates somewhat in that it provides other first-class entities. It deviates from Church representation to provide Integers, this is purely for ease of use. It also provides a symmetric synchronous Channel type for interprocess communication. As a note, this language can also be classified as a Simply-Typed Lambda Calculus.

ErLam also makes a number of ease-of-use decisions like providing a default branch operator and has some useful syntactic sugar such as SML style *let* expressions and multivariable function definitions. There is also a set of built in functions for numeric operations, type checking, and standard functional behaviors (*e.g.* combinators, *etc.*) which are ignored in this document.

Figure 3.1 expresses ErLam in its simplified BN-Form. The semantics for the language is fairly straight forward, but it's operational semantics are layed out in appendix 1. All expressions reduce to one of the terminal types: Integer, Channel, or Function. To spawn for instance, if any terminal is passed other than a function, it returns a 0 (e.g. false). When the function is passed, it is applied with nil to initialize the internal expression.

Also, note the possible steps **swap** can take: either returning the null set or another expression and a set of functions. In the former's case, the channel has blocked and the only course of action is to get another expression to work on from the scheduler. In the later's case, we have an expression to work on, but we also may have unblocked other processes in the process so we need to reschedule them.

Note however, that return set may be null and the expression returned may be another attempt at swapping (i.e.  $e = \mathbf{swap} \, ev$ )). This would let the scheduler choose whether

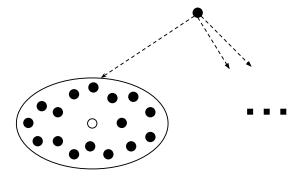


Figure 3.2: Graphical representation of PTree, N Parallel work groups.

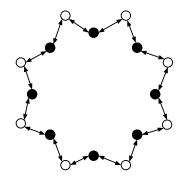


Figure 3.3: Graphical representation of PRing, full system predictable cooperation.

to retry immediately or reschedule it for a later time and work on something in the mean time.

- 3.2.2 Channel Implementations
- 3.2.3 The Scheduler API
- 3.2.4 Example: The CML Scheduler
- 3.3 Simulation & Visualization
- 3.3.1 Runtime Log Reports
- 3.3.2 Cooperativity Testing

As part of the thought experiment, we needed to implement a set of test cases which would give a decent coverage of applications which

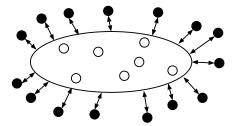


Figure 3.4: Graphical representation of ClusterComm, N processes to M channels for unpredictable full system cooperation.

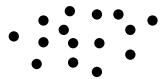


Figure 3.5: Graphical representation of ChugMachine, N worker processes without cooperation.

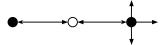


Figure 3.6: Graphical representation of UserInput, single randomly hanging process.

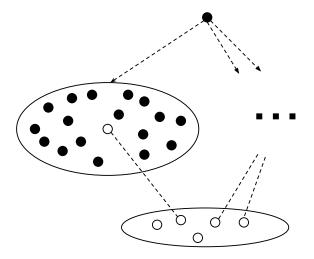


Figure 3.7: Graphical representation of JumpShip, N Parallel phase shifting work groups.

- 3.4 Cooperativity Mechanics
- 3.4.1 Overview
- 3.4.2 Longevity-Based Batching
- 3.4.3 Channel Pinning
- 3.4.4 Bipartite Graph Aided Shuffling

# **Results and Discussion**

# **Conclusion and Future Work**

Appendix

# Appendix 1

## **ErLam Operational Semantics**

$$\begin{aligned} & \frac{E(x) \Rightarrow v}{S,C,E:x \rightarrow S,C,E:v} & \text{Integer} \frac{-}{S,C,E:n \rightarrow S,C,E:n} \\ & \text{Fun} \frac{-}{S,C,E:\text{fun } x.e \rightarrow S,C,E:\text{fun } x.e} & \text{Unwrap} \frac{-}{S,C,E:(e) \rightarrow S,C,E:e} \\ & \frac{|C|+1=n-C\downarrow n \Rightarrow chan_n}{S,C,E:\text{newchan} \rightarrow S,C;\{chan_n\},E:chan_n} \\ & \frac{S,C,E:\text{newchan} \rightarrow S,C;\{chan_n\},E:chan_n}{S,C,E:e_1 \rightarrow S',C',E':e_1'e_2} \\ & \frac{S,C,E:e_1 \rightarrow S',C',E':e_1'e_2}{S,C,E:\text{fun } x.e_1e_2 \rightarrow S',C',E':\text{fun } x.e_1e_2'} \\ & \frac{S,C,E:e_2 \rightarrow S',C',E':\text{fun } x.e_1e_2'}{S,C,E:\text{fun } x.e_1e_2 \rightarrow S',C',E':\text{fun } x.e_1e_2'} \\ & \frac{S,C,E:e_1 \rightarrow S',C',E':\text{fun } x.e_1e_2'}{S,C,E:\text{fun } x.e_1v \rightarrow S,C,E;(x,v):e_1} \\ & \frac{S,C,E:e_1 \rightarrow S',C',E':\text{fir } e_1'e_2e_3}{S,C,E:\text{if } ve_2e_3 \rightarrow S,C,E:e_2} \\ & \text{If}(3) \frac{v \leq 0}{S,C,E:\text{if } ve_2e_3 \rightarrow S,C,E:e_3} \\ & \frac{S,C,E:e_1 \rightarrow S',C',E':\text{e}_1'}{S,C,E:\text{swap } e_1e_2 \rightarrow S',C',E':\text{swap } e_1'e_2} \\ & \frac{S,C,E:e_1 \rightarrow S',C',E':e_1'}{S,C,E:\text{swap } e_2 \rightarrow S',C',E':\text{swap } e_2'e_2} \\ & \frac{S_0}{S,C,E:\text{swap } e_2 \rightarrow S',C',E':\text{swap } e_2'e_2} \\ & \frac{S_0}{S,C,E:\text{swap } e_2 \rightarrow S',C',E':\text{swap } e_2'e_2} \\ & \frac{S_0}{S_0,C,E:\text{swap } e_2 \rightarrow S',C',E':\text{swap } e_2'e_2} \\ & \frac{S_0}{S_0,C,E:\text{swap } e_2 \rightarrow S',C',E':\text{swap } e_2'e_2} \\ & \frac{S_0}{S_0,C,E:\text{swap } e_2 \rightarrow S',C',E':\text{swap } e_2'e_2} \\ & \frac{S_0}{S_0,C,E:\text{swap } e_2 \rightarrow S',C',E':\text{swap } e_2'e_2} \\ & \frac{S_0}{S_0,C,E:\text{swap } e_2 \rightarrow S',C',E':\text{swap } e_2'e_2} \\ & \frac{S_0}{S_0,C,E:\text{swap } e_2 \rightarrow S',C',E':\text{swap } e_2'e_2} \\ & \frac{S_0}{S_0,C,E:\text{swap } e_2 \rightarrow S',C',E':\text{swap } e_2'e_2} \\ & \frac{S_0}{S_0,C,E:\text{swap } e_2 \rightarrow S',C',E':\text{spawn } e_2'e_2} \\ & \frac{S_0}{S_0,C,E:\text{swap } e_2 \rightarrow S',C',E':\text{spawn } e_2'e_2} \\ & \frac{S_0}{S_0,C,E:\text{spawn } e_2 \rightarrow S',C',E':\text{spawn } e_2'e_2} \\ & \frac{S_0}{S_0,C,E:\text{spawn } e_2 \rightarrow S',C',E':\text{spawn } e_2'e_2} \\ & \frac{S_0}{S_0,C,E:\text{spawn } e_2 \rightarrow S',C',E':\text{spawn } e_2'e_2} \\ & \frac{S_0}{S_0,C,E:\text{spawn } e_2 \rightarrow S',C',E':\text{spawn } e_2'e_2} \\ & \frac{S_0}{S_0,C,E:\text{spawn } e_2 \rightarrow S',C',E':\text{spawn } e_2'e_2} \\ & \frac{S_0}{S_0,C,E:\text{spawn } e_2 \rightarrow S',C',E':\text{spawn } e_2'e_2} \\ & \frac{S_0}{S_0,C,E:\text{spawn } e_2 \rightarrow S',C',E':\text{spawn } e_2'e_2} \\ & \frac{S_0}{S_0,C,E:\text{spawn } e_2 \rightarrow S',C'$$

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