

An Investigation into Parareal

Abhijit Chowdhary

New York University

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Problem Statement

We would like to numerically solve the autonomous ordinary differential equation:

$$\begin{cases} u'(t) = f(t, u), & t \in [t_0, t_f] \\ u(t_0) = u_0 \end{cases}$$

where $f : \mathbb{R}^d \rightarrow \mathbb{R}^d$ and $u : \mathbb{R} \rightarrow \mathbb{R}^d$.

Framing this in a HPC context, how can we solve the above while taking advantage of massively parallel hardware?

Serial Methods

- Forward and Backward Euler:

$$u_{n+1} = u_n + hf(t, u_n), u_n + hf(t, u_{n+1})$$

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- Runge Kutta Methods
- Etc.

Parallel Techniques

Problem!

Most of the previous methods are iterative, depending on the previously computed values.

Not embarrassingly parallel... We have to get creative.

Parallel-In-Time

One technique is to parallelize the problem across time, i.e. split $[t_0, t_f]$ into slices $[t_n, t_{n+1}]$ and in parallel solve each slice.

Iteration Dependence

But since each slice depends on the previous, how can we chain together these individual solutions to each slice?

Predictor-Corrector

What if we were to predict the portions that $[t_n, t_{n+1}]$ are dependent on with a cheap coarse operator, and then correct with a fine operator!

Yes!

Parareal

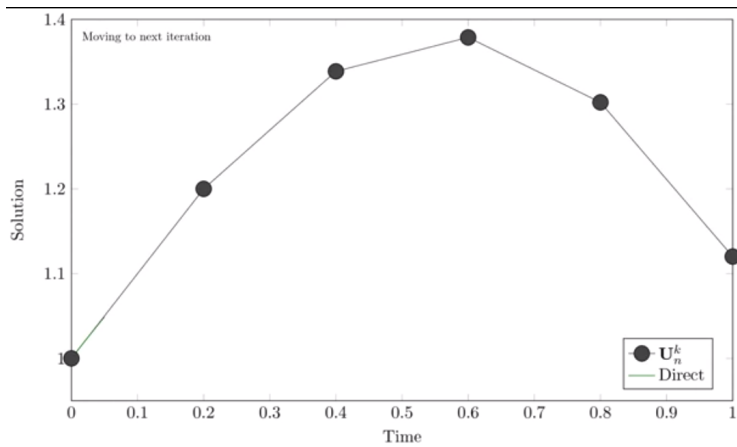
Let \mathcal{G} be a course and inexpensive operator, and let \mathcal{F} be a fine operator of high order. Then Parareal iteration is:

$$\lambda_{n+1}^{k+1} = \mathcal{G}(t_{n+1}, t_n, \lambda_n^{k+1}) + \mathcal{F}(t_{n+1}, t_n, \lambda_n^k) - \mathcal{G}(t_{n+1}, t_n, \lambda_n^k)$$

Parallel Potential

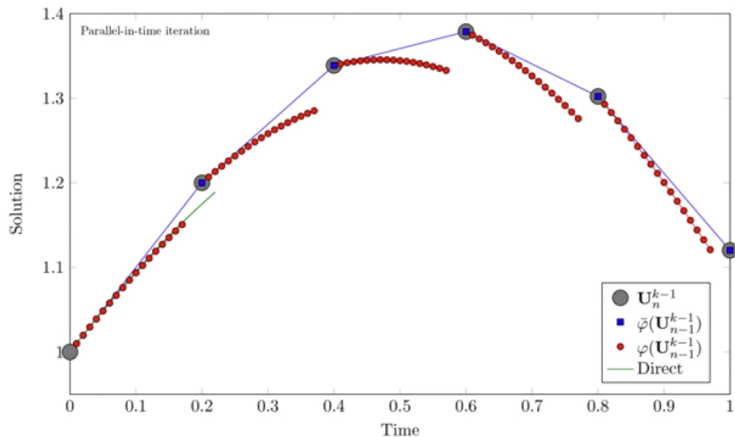
Notice! The \mathcal{F} term depends only on the previous solutions, which means it can be computed in parallel for each k step. Furthermore, $\mathcal{G}(\lambda_n^k)$ satisfies the FSAL property!

Parareal: Visual Example



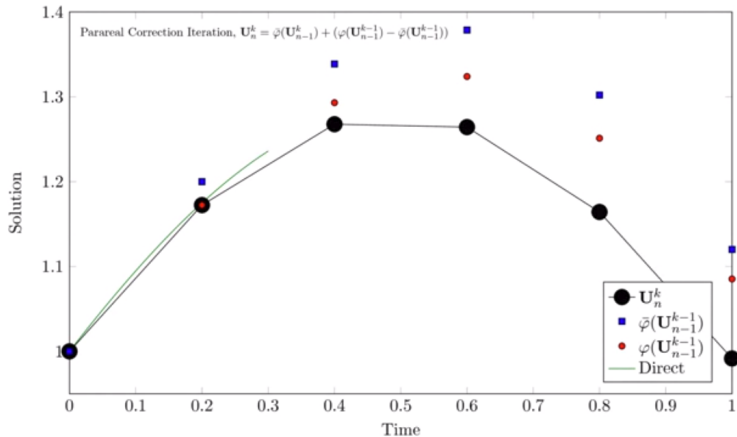
Full beautiful animation found on wikipedia's Parareal page.

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Pseudocode

Require: y_0 and coarse and fine solvers \mathcal{G} , \mathcal{F} .

$y_c \leftarrow \mathcal{G}(t_f, t_0, y_0).$ ▷ Coarsely approximate solution

$y \leftarrow y_c.$

while iter < max_iter && not converged **do**

for $n = 0 \rightarrow P$ **do** ▷ Parallel capable

$y_f(n) = \mathcal{F}(t_{n+1}, t_n, y(n)).$

$\delta y(n) = y_f(n) - y_c(n).$ ▷ corrector term. FSAL

end for

for $n = 0 \rightarrow P$ **do**

$y_c(n) = \mathcal{G}(t_{n+1}, t_n, y(n)).$ ▷ Predict.

$y(n) = y_c(n) + \delta y(n).$ ▷ Correct.

end for

end while

Parallel Pseudocode

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while iter < max_iter && not converged **do**

 #pragma omp parallel for

for $n = 0 \rightarrow P$ **do**

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$y(n) = y_c(n) + \delta y(n).$ ▷ Correct.

end for

end while

Speedup Analysis

Suppose we have P processors, and suppose our fine method takes T_f time, T_g for the course method. Furthermore, assume that our Parareal iteration needs k steps. Then the speedup it provides is:

$$S = \frac{PT_f}{PT_g + k(PT_g + T_f)} = \frac{1}{\frac{T_g}{T_f} + k(\frac{T_g}{T_f} + \frac{1}{P})} = \frac{1}{\frac{T_g}{T_f}(1+k) + \frac{k}{P}}$$

Note:

$$\lim_{P \rightarrow \infty} \frac{1}{\frac{T_g}{T_f}(1+k) + \frac{k}{P}} = \frac{T_f}{T_g(1+k)}$$

No Free Lunch

$$S_{\infty} = \frac{T_f}{T_g(1+k)}$$

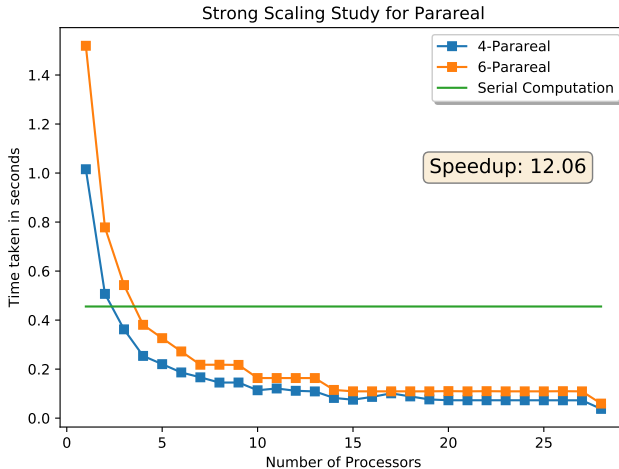
This tells us something very important, we want the ratio T_f/T_g to be as large as possible, and we want k to be as small as possible.

Theorem

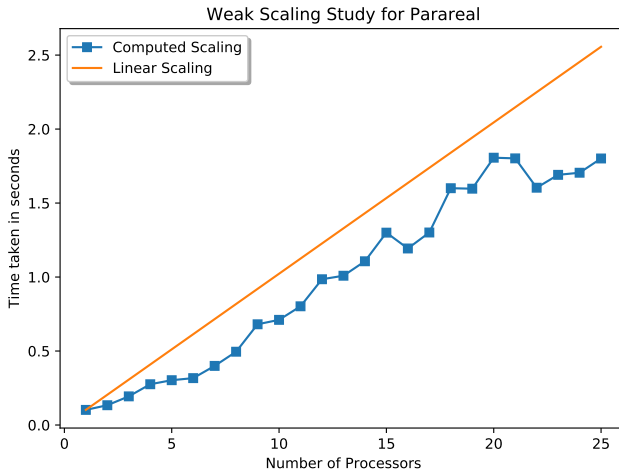
The parareal method has order of accuracy mk , where $k - 1$ is the number of parareal iterations made and m is the order of \mathcal{G} , assuming \mathcal{F} is close to truth. (Bal)

Therefore, we have this uncomfortable optimization problem. We want k small for speedup, but large for order. If we make m large instead, then the ratio T_f/T_g becomes smaller...

Strong Scaling



Weak Scaling



Improvements | Pipelined Parareal

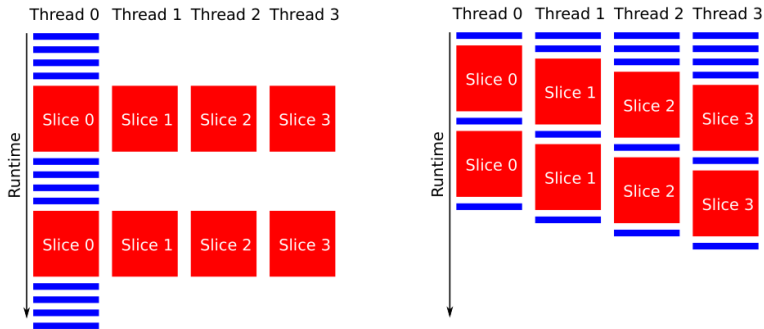


Figure borrowed from Ruprecht's *Implementing Parareal*.

Pipelined Parareal Pseudocode

It's a bit too complicated to present in a slide, but the general idea is to begin the omp parallel statement at the beginning and then pretend like were in a MPI environment.

General Idea

Thread p processes the new iteration at $p + 1$. MPI communication is mimicked with locks, and done in place.