# PARALLELISM IN PYPWA

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#### Overview

- Introduction to Amplitude Analysis
- Motivation for Parallel Computing in Amplitude Analysis
- Multiprocessing in Python
- Introduction to the Xeon Phi
- Amplitude Analysis on the Xeon Phi
- Future directions of MIC architecture
- Summary

### Introduction to Amplitude Analysis

- Amplitude Analysis is an umbrella term used to describe analyses that use amplitudes to describe the physics of decays/reactions
- Example: 3-body decay of  $\omega/\phi$ ->3 $\pi$
- These amplitudes will be used in a maximum likelihood fit and they take a set of event dependent kinematic variables and overall fit parameters
- PyPWA is the framework that is focused on ease of use and simplification of the process of Amplitude Analysis

$$I(\overrightarrow{x},\overrightarrow{a})\equiv\sum_{ext.\ spins}|\mathscr{M}|^2$$



### Why Parallel Computing?

- Fitting in Amplitude Analysis is a computationally intensive task
  - It is not uncommon that the calculation of the likelihood function would take several hours due to the complexity of the theoretical amplitudes
- The trend over the past decade has been an increase in the number of cores/threads per CPU
  - On an average workstation this allows for a significant speedup
- To keep up with an increasing amount of data and stagnant serial performance, we must use parallel computing in the future



Figures: Jim Jeffers and James Reinders

## Python Multiprocessing

- The first attempt to speed up the fit process involves some modification to the PyPWA framework
- Python multiprocessing works well for splitting up the likelihood function
  - Multiprocessing was used to skirt around the GIL which prevents more than one python interpreter running per process
- Provided excellent speedup compared to the serial version and is now included in the production build



#### Xeon Phi Development (Knight's Corner)

- Many Integrated Cores (MIC) Technology
  - ~60 cores x 4 threads/core = ~240 threads
- Little to no code modification to get up and running but modifications needed later to tune the code for higher performance
- Theory Amplitudes written in Fortran compile with **no modification**!
  - The FORTRAN code would likely need rewriting if used with any GPGPU accelerator card
- After compiling comes integration with C/C++ likelihood function and testing
- Many possible ways to utilize the card: Intel's TBB, pthreads, Cilk, and **OpenMP**



# OpenMP on the Xeon Phi (or CPU)

- OpenMP is a library developed to allow for easy to use portable shared memory multiprocessing and vectorization
- This library functions using #pragmas which contain work sharing constructs and clauses
- Allows for parallelization and vectorization without altering the meaning of the original code
  - Program will still compile serially if the compiler does not support OpenMP!



# MIC and CPU Scaling

- Xeon Phi performance normalized to CPU performance
- The results from the first successful test are encouraging for using the Xeon Phi for production analyses
- Xeon Phi code is only multiprocessed here, with vectorization the performance could be **8x** higher!



Model	#Processors	Cores/CPU	Threads/Core	Total #Threads
Xeon E5-2650	2	8	2	32
Xeon Phi 5100P	I	60	4	240

### Future directions of MIC Architecture

- Knight's Landing:
  - 72 Silvermont cores
  - Shipping in 2016!
  - Socketed and PCI-Express versions available
  - Back to homogeneous computing?



All products, computer systems, dates and figures specified are preliminary based on current expectations, and are subject to change without notice. Yoyer 3 Teraflops of peak, theoretical double-precision performance is preliminary and based on current expectations of cores, clock frequency, and floating point operations per scycle. FLOPS = cores, x clock frequency and ploating point operations per scycle. FLOPS = cores, x clock frequency and ploating point operations per scycle. FLOPS = cores, x clock frequency, and ploating point operations per scycle. FLOPS = cores, x clock frequency, and floating point operations per scycle. FLOPS = cores, x clock frequency and floating point operations per scycle. FLOPS = cores, x clock frequency, and floating point operations per scycle. FLOPS = cores, x clock frequency, and floating point operations per scycle. FLOPS = cores, x clock frequency, and floating point operations per scycle. FLOPS = cores, x clock frequency, and floating point operations per scycle. FLOPS = cores, x clock frequency, and floating point operations per scycle. FLOPS = cores, x clock frequency, and floating point operations per scycle. FLOPS = cores, x clock frequency and floating point operations per scycle. FLOPS = cores, x clock frequency and floating point operations per scycle. FLOPS = cores, x clock frequency and floating point operations per scycle. FLOPS = cores, x clock frequency and floating point operations per scycle. FLOPS = cores, x clock frequency and floating point operations per scycle. FLOPS = cores, x clock frequency and floating point operations per scycle. FLOPS = cores, x clock frequency and floating point operations per scycle. FLOPS = cores, x clock frequency and floating point operations per scycle scalar per scycle and floating point cores and floating point operations per scycle scalar point operations per scycle scalar point operations per scalar point point operations per scalar point operations per scalar point point point actions per scycle scalar point operations pe

# Summary

- Multiprocessing has been utilized to increase performance in Amplitude Analysis using Python Multiprocessing and OpenMP on the Xeon Phi
- A performance increase of ~23x on a CPU and ~30x on the Xeon Phi has been seen using OpenMP
- Performance optimizations and further benchmarks will be pursued before integration into the production build
- For more information:
  - http://pypwa.jlab.org
  - https://github.com/JeffersonLab/PyPWA