Constructing PineAPPL grids on hardware accelerators

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In this proceedings we demonstrate how to implement and construct the PineAPPL grids, designed for fast-interpolation of Monte Carlo simulation with electroweak and QCD corrections, using the VegasFlow framework for Monte Carlo simulation on hardware accelerators. We provide an example of synchronous and asynchronous filling operations of PineAPPL grids from Monte Carlo events generated by VegasFlow. We compare the performance of this procedure on multithreading CPU and GPU.

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1. Introduction and implementation

The fast evaluation of theoretical predictions for a generic set of parton distribution functions [1] and scale variation choices is a common request that has been addressed by generic tools such as APPLGRID [2], FastNLO [3] and more recently PineAPPL [4, 5]. In particular, the technology developed by the PineAPPL library provides the possibility to produce fast-interpolation grids of physical cross sections, computed with a general-purpose Monte Carlo generator, accurate to fixed order in the strong, electroweak, and combined strong-electroweak couplings.

In order to construct and use PineAPPL grids we need to interface the code to a Monte Carlo simulation library that generates event weights and kinematics configurations for the grid filling procedure. In this proceedings we use VegasFlow [6, 7], a Monte Carlo simulation framework with support for multithreading CPU, single-GPU and multi-GPU setups. The choice of VegasFlow relies on high efficiency when performing simulation thanks to its flexibility to distribute event generation across multiple hardware accelerators. Furthermore, when combined to PDFFlow [8, 9], it is possible to perform a full simulation of physical processes with quite competitive performance in comparison to specialized codes.

However, the integration of PineAPPL, or any other external library that expects input from VegasFlow, may generate a natural performance deterioration of the Monte Carlo simulation thanks to operations that may not benefit from the multithreading paradigm of VegasFlow on hardware accelerators. PineAPPL is designed for CPUs and provides to the developer the possibility to be distributed in a multithreading configuration, however it does not provide a GPU implementation and methods for asynchronous grid filling.

In order to provide a solution for such a problem, in Figure 1 we represent schematically the approach proposed here. The Monte Carlo simulation is driven by the VegasFlow framework, which takes care of generating events and distributing them among available devices such as CPUs and/or GPUs. Each batch of events evaluates a large number of matrix element weights and phase space configurations which are subsequently sent to the PineAPPL library for the construction of fast-evaluation grids. In a multithreading CPU and GPU environment, a sequential synchronous fill operation may reduce drastically the performance of the Monte Carlo simulation, by increasing the total amount of time needed to achieve precise predictions. In order to avoid such a performance deterioration, we propose to detach the computation between the VegasFlow Monte Carlo event simulation from the PineAPPL grid filling operation by creating a CPU thread pool that asynchronously queues and executes the operations required by PineAPPL.

From a technical perspective all steps presented in the previous paragraph can be achieved using the Python interfaces of VegasFlow and PineAPPL. In particular, we can generate Monte Carlo events using the eager mode feature in VegasFlow and including an asynchronous job execution CPU thread pool using the standard Python multiprocessing module.

2. Measuring performance

In order to test the asynchronous approach proposed in the previous section, we use VegasFlow with a simplified simulation of Drell–Yan, which only calculates the leading-order matrix element

\[\text{https://docs.python.org/3/library/multiprocessing.html#module-multiprocessing}\]
for the photon-induced process, $\gamma\gamma \rightarrow \ell\bar{\ell}$. The simple structure of the corresponding matrix element and phase space allows to easily test a wide range of phase-space points. Furthermore, the evaluation is cheap enough to highlight the overhead of filling a PineAPPL grid, which is basically a constant of the number of partonic processes; when simulating more complex processes, we therefore expect the relative overhead to be much lower.

At the end of each batch of events, we compute a PineAPPL grid for the $|y_{\ell\bar{\ell}}|$ observable. In terms of physical cuts we consider a single invariant-mass slice of a CMS 8 TeV analysis [10], which requires $60 < m_{\ell\bar{\ell}} < 120$ GeV, $p_T^\ell > 14$ GeV, $|y_{\ell\bar{\ell}}| < 2.4$ and $|y_\ell| < 2.4$. From the VegasFlow point of view we tested the performance by increasing number of events before cuts from $10^3$ to $10^9$, using the default maximum number of events per device of $10^6$.

In the left plot of Figure 2 we show the total computing time of the Monte Carlo simulation for an increasing number of events. We compare runs without PineAPPL (green) to the synchronous (blue) and asynchronous (orange) approaches with PineAPPL. Similarly, on the right plot of Figure 2, we present performance results on GPU. The synchronous approach produces a performance deterioration up to $\approx 30\%$ on CPU and $\approx 80\%$ on GPU, while the asynchronous approach reduces the overall deterioration by $\approx 10\%$. The main advantage of the asynchronous approach is the possibility to detach the Monte Carlo event simulation from the operations related to the construction of PineAPPL grids. From the Monte Carlo simulation perspective we observe a negligible overhead due to the submission of asynchronous jobs. On the other hand, the CPU thread pool, used by the asynchronous job, can be further optimized depending on the available system resources. Note that the system used for the performance measurement in Figure 2 has a fast clock CPU thus the expect larger performance deteriorations on average professional grade hardware.
Figure 2: Performance results for the Monte Carlo simulation of photon-induced Drell–Yan channel using VegasFlow and PineAPPL on CPU (left) and GPU (right). The synchronous approach produces a performance deterioration up to $\approx 30\%$ on CPU and $\approx 80\%$ on GPU, while the asynchronous approach reduces the overall deterioration by $\approx 10\%$. Note that the asynchronous approach detaches the Monte Carlo integration from the grid filling operations providing further options for optimization.

3. Outlook

The example presented in this proceedings shows that an asynchronous approach to filling fast-interpolation grids using PineAPPL is feasible and provides a useful interface to detach Monte Carlo event simulation from sequential operations without introducing a strong performance deterioration.

We have compared the performance of synchronous and asynchronous approaches on CPU and GPU for a leading-order Drell–Yan photon-induced simulation. As expected, the performance from an asynchronous approach outperforms the sequential synchronous mechanism and does not introduce overheads to the Monte Carlo event simulation.

The code for the exercise presented in this manuscript is public available at the VegasFlow repository [6, 7] inside the examples folder.

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