# HPC: Implementing and benchmarking CPU algorithms for Nvidia GPUs

Author: Daniel Álvarez Conde Supervisors: Jonas Rembser / Lorenzo Moneta

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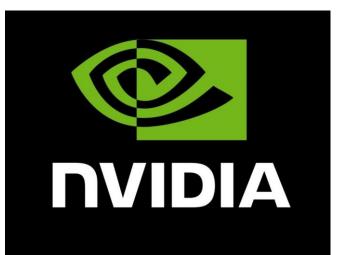




Introduction



Framework for data processing, focused on High-Energy Physics

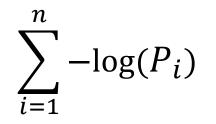


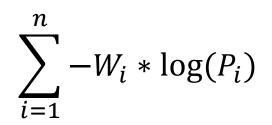
CUDA (Compute Unified Device Arquitecture)





# Negative Log Likelihood (NLL)





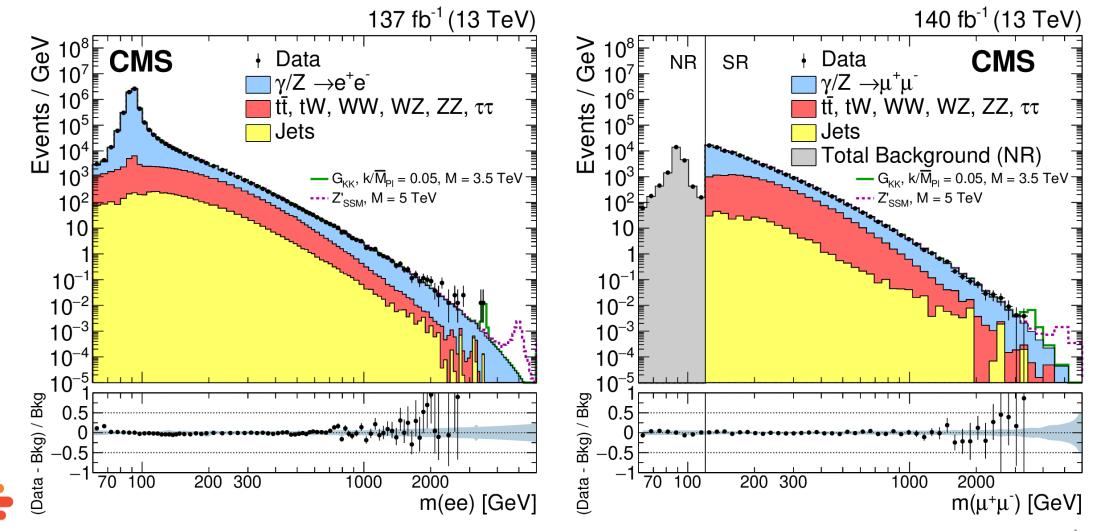
Unbinned Fits

**Binned Fits** 

- **n** -> the size of the array
- W -> Weight. Number of events of each sample *i*
- **P** -> Probability for each sample *i*



### Negative Log Likelihood: Binned Fits





# Optimizing time execution

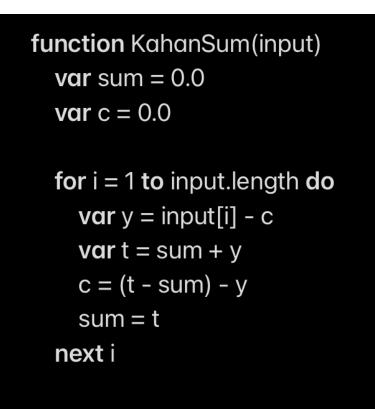
- Initial situation: The single and double precision variables can store either very large numbers or very small numbers.
- **Problem**: When you try to add (or subtract) them, the information of the decimal part will be lost:





### Kahan Summation

Helps to **minimise the error** when adding large data sets, by keeping track of the error made in each addition (variable c):



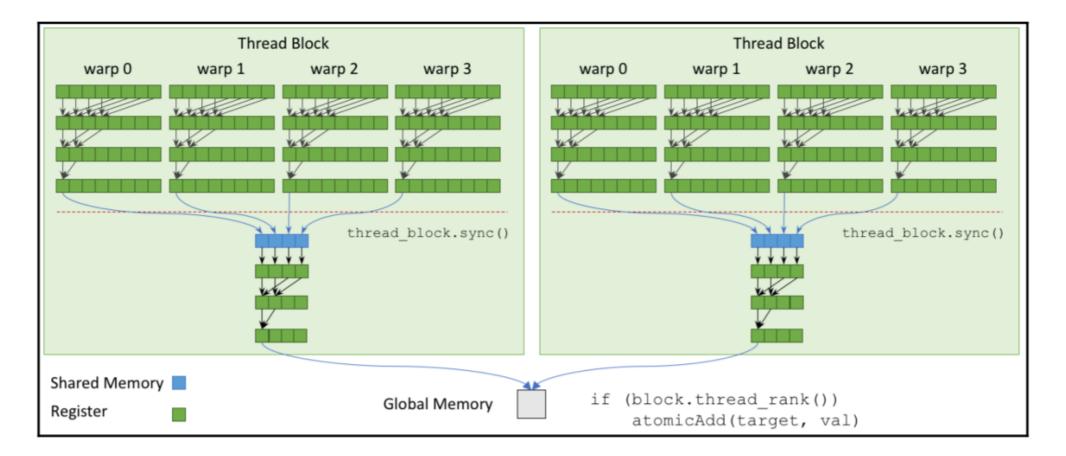
return sum

6





### Kahan Summation in the GPU

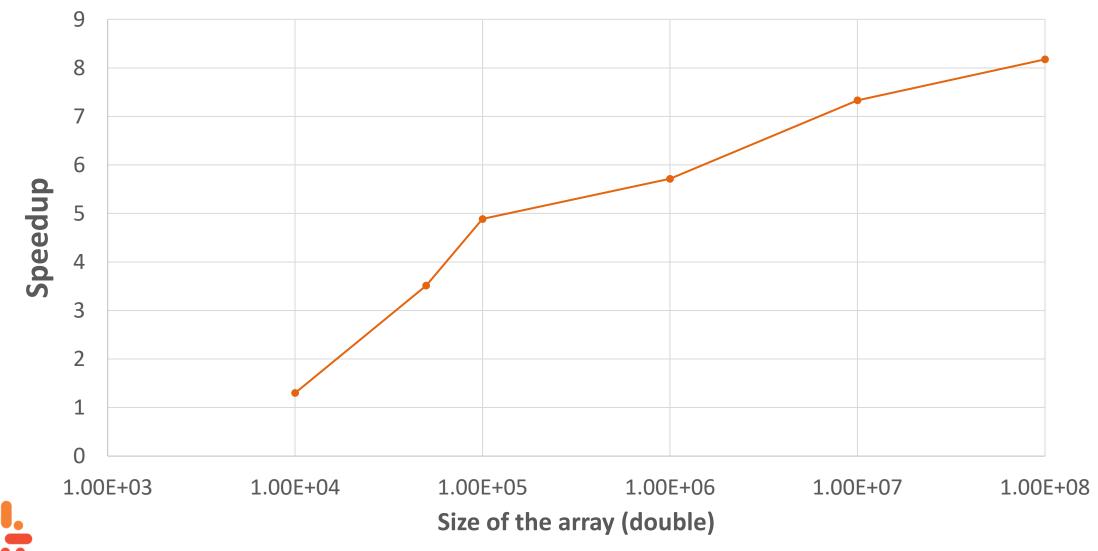




Ilustration: <u>PL00, Parallel computing (userdyk-github.github.io)</u>

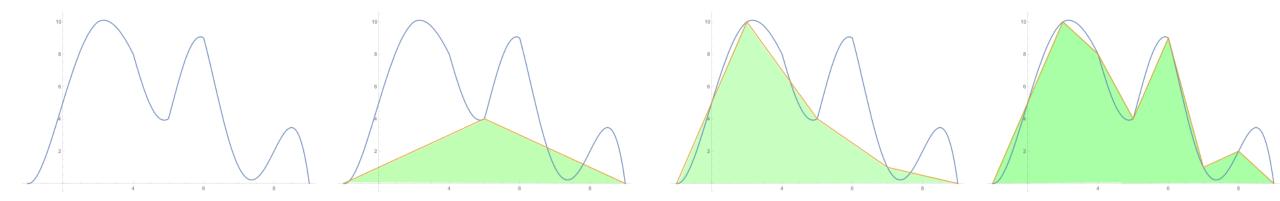


Kahan GPU vs CPU



# Compute numeric integrals in the GPU

It is what is limiting most costly realistic uses cases. **Romberg's method** makes a good approximation:





### Conclusions

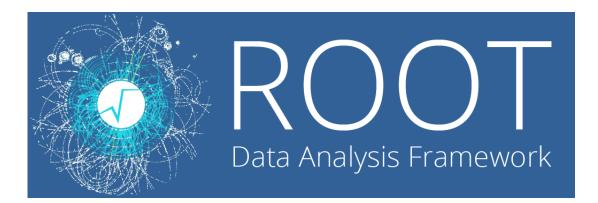
- Don't use GPUs for less than 10e7 data.
- Be very careful with the synchronisation between threads inside each block.
- Try to have all the cores at **100% work**.
- **Use memory:** 70% of the programs are "compute bound", meanwhile 30% are "memory bound".





# Thank you for your attention

Questions?



root-project/root

danielalvcon@gmail.com



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