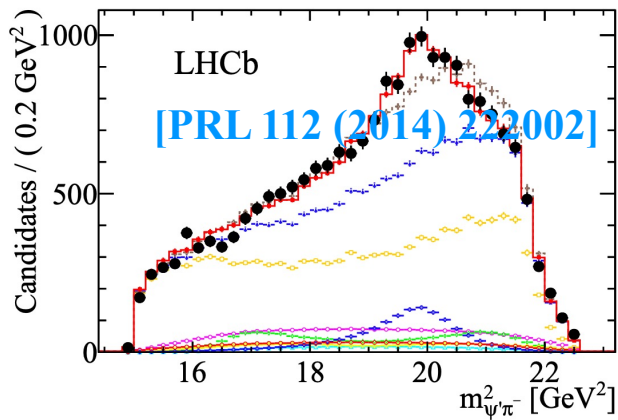
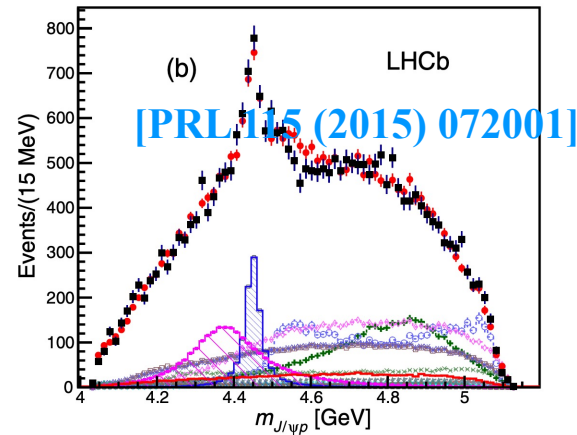


Amplitude analyses with RooFit fitter

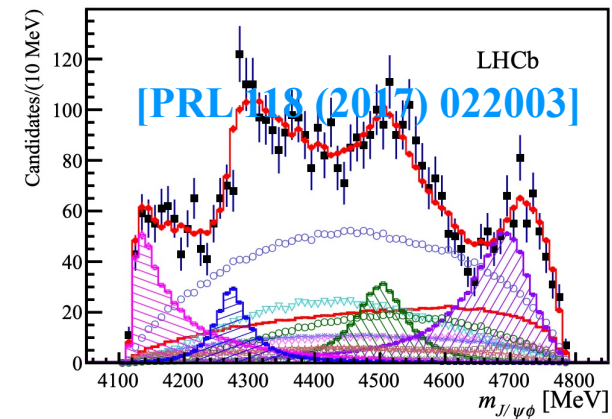
Many amplitude analyses in spectroscopy studies using **RooFit fitter**, e.g. :



Observation of
 $Z_c(4430)^-$



Observation of
pentaquark



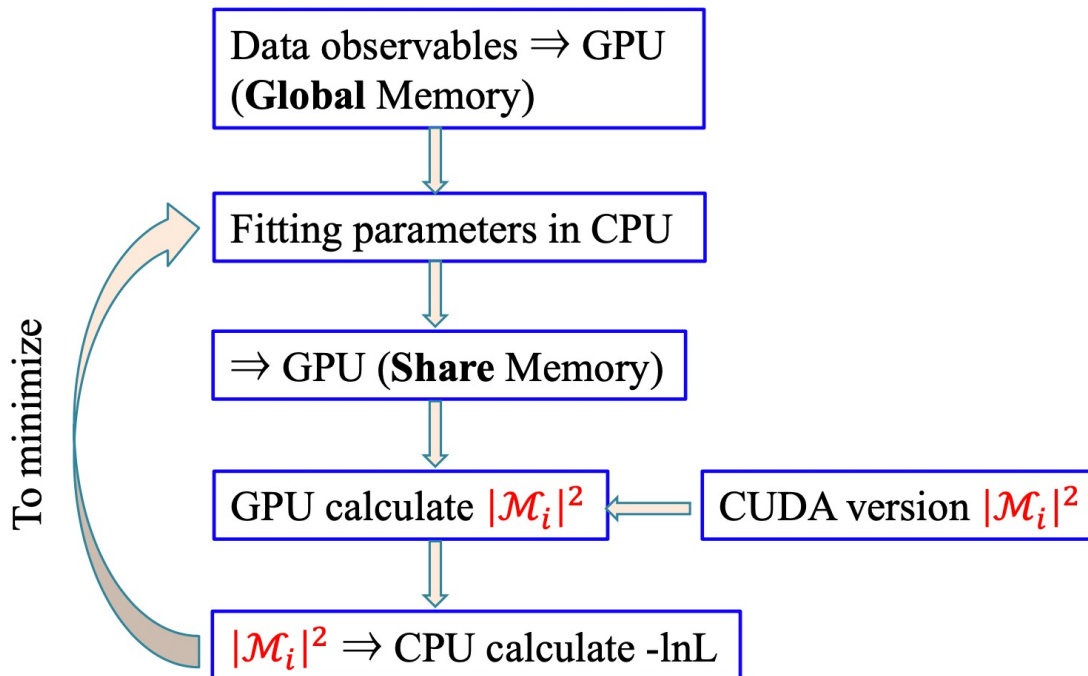
Observations of $X(4270)$
 $X(4450)$ & $X(4700)$

- RooFit widely used in HEP community
- CPU based codes could handle Run 1 statistic
 - ✓ e.g., $\sim 30,000$ signal in P_c study, multi-CPU fit
- Computation time related to many factors:
 - ✓ Data sample size
 - ✓ Initial values of fitting parameters
 - ✓ Number of floating parameters

RooFit&CUDA based AmAn program

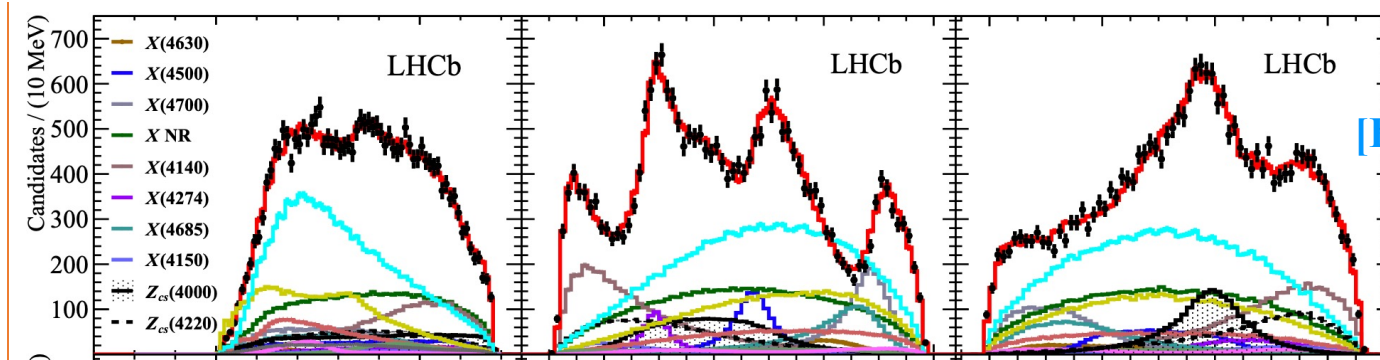
CPU based fit takes too much time with **Run 2** data sample

- Several hours → acceptable
- Several days → unacceptable
- **Idea: Let GPU to calculate $|\mathcal{M}_i|^2$ for events of both data and MC**
 - CUDA (GPU based C++) for computation of $|\mathcal{M}_i|^2$
 - CUDA memory transfers between CPU and GPU



Performance in $B \rightarrow J/\psi\phi K$ study

First publication using RooFit&CUDA fit technology at LHCb



[[PRL 127 \(2021\) 082001](#)]

Total signal number: ~ 24 k

Float parameter number: 144 (default model)

Computing time of once NLL: ~ 3 ms

Computing time of one fit: few tens of minutes

- ✓ Closely related to the initial parameters
- ✓ Related to the float parameters number
- ✓ The complexity of different resonance line-shape
- ✓

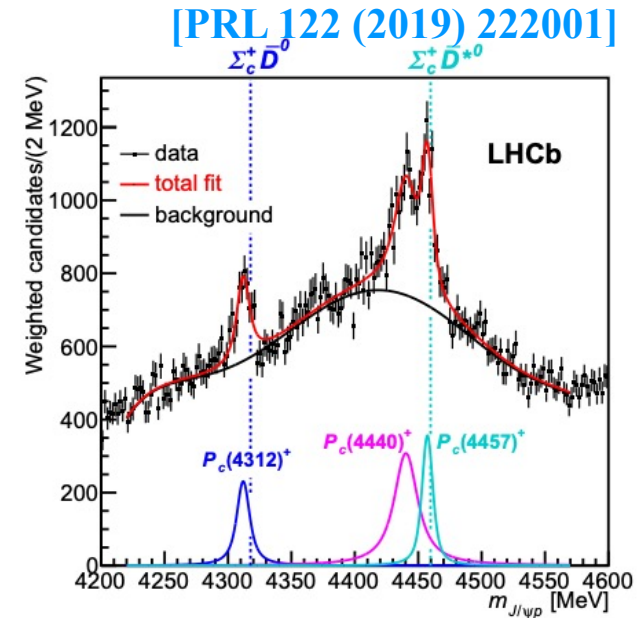
Theoretical Performance		TITAN V
Pixel Rate:	139.7 GPixel/s	
Texture Rate:	465.6 GTexel/s	
FP16 (half) performance:	29.80 TFLOPS (2:1)	Our GPU card
FP32 (float) performance:	14.90 TFLOPS	
FP64 (double) performance:	7.450 TFLOPS (1:2)	

Tips in RooFit&CUDA AmAn program

- Reducing precision, double → float
 - ✓ The final likelihood difference is negligible
 - ✓ GPU computing time reduced by half
 - ✓ Also dependent on the type of GPU card
- Malloc once and reuse memory
 - ✓ Reduce the data transfer between CPU & GPU
- Use less registers
 - ✓ Reducing the register pressure increases the number of working thread
- Split kernel function
 - ✓ Increase the GPU computing occupancy
-

Pentaquark in $\Lambda_b^0 \rightarrow J/\psi p K$

- Amplitude analysis with Run 1 & Run 2 ongoing
 - (cannot show physical results)
- Total signal numbers : $\sim 200k$
- Longer computational time:
 - ✓ Larger statistic
 - ✓ More float parameters
 - ✓ Narrow structure, mass resolution considered
- Computing time of once NLL: $\sim 0.17s$
- Computing time of one complete fit: $\sim 1-2$ days
 - ✓ With good initial parameters, < 1 day
 - ✓ Many GPU cards used at same time
 - ✓ Not very convenient, computation is not the biggest obstacle for this study
 - ✓ Due to small fraction, not an ideal channel to determine J^P of pentaquark



Future

- RooFit & CUDA is not designed for widespread use, to optimize it more readable and extensible
- Computation timing smaller than several hours is acceptable in the view of physical analysis
- Many AmAn programs based on different frameworks are developed, e.g., RooFit&CUDA, GooFit, Tensorflow, etc., analysts prefer to choose the familiar one if their performances are similar
- More efficient fitter could boost the analyst work efficiency !

Backup

Signal PDF

$$\frac{d\mathcal{P}}{dm_{\phi K}d\Omega} = \mathcal{P}_{sig}(m_{\phi K}, \Omega | \vec{\omega}) = \frac{|\mathcal{M}(m_{\phi K}, \Omega | \vec{\omega})|^2 \Phi(m_{\phi K}) \epsilon(m_{\phi K}, \Omega)}{I(\vec{\omega})}$$

Diagram illustrating the components of the Signal PDF:

- Matrix element of decay
- Input observables
- Fitted parameters (helicity couplings, M_0, Γ_0)
- Phase space factor
- Efficiency

$\vec{\omega}$: fitting parameters
 Φ : phase-space factor = pq
 ϵ : efficiency
 $I(\vec{\omega})$: normalization integral

AmAn likelihood construction

Efficiency parameterization from MC sample

$$\begin{aligned}
 -\ln L(\vec{\omega}) &= -\sum_i \ln \left[(1 - \beta) \mathcal{P}_{sig}(m_{\phi K i}, \Omega_i | \vec{\omega}) + \beta \mathcal{P}_{bkg}(m_{\phi K i}, \Omega_i) \right] \\
 &= -\sum_i \ln \left[(1 - \beta) \frac{|\mathcal{M}(m_{\phi K i}, \Omega_i | \vec{\omega})|^2 \Phi(m_{\phi K i}) \varepsilon(m_{\phi K i}, \Omega_i)}{I(\vec{\omega})} + \beta \frac{\mathcal{P}_{bkg}^u(m_{\phi K i}, \Omega_i)}{I_{bkg}} \right] \\
 &= -\sum_i \ln \left[|\mathcal{M}(m_{\phi K}, \Omega | \vec{\omega})|^2 + \frac{\beta I(\vec{\omega})}{(1 - \beta) I_{bkg}} \frac{\mathcal{P}_{bkg}^u(m_{\phi K i}, \Omega_i)}{\Phi(m_{\phi K i}) \varepsilon(m_{\phi K i}, \Omega_i)} \right] + N \ln I(\vec{\omega}) + \text{constant}
 \end{aligned}$$

Normalization factor

Background parameterization from sideband