

$$\Lambda_b \rightarrow J/\Psi p K$$

PRODUCTION CORRECTION

RUN2 {'16, '17, '18}

M. Ferrillo

UZH

NOVELTY

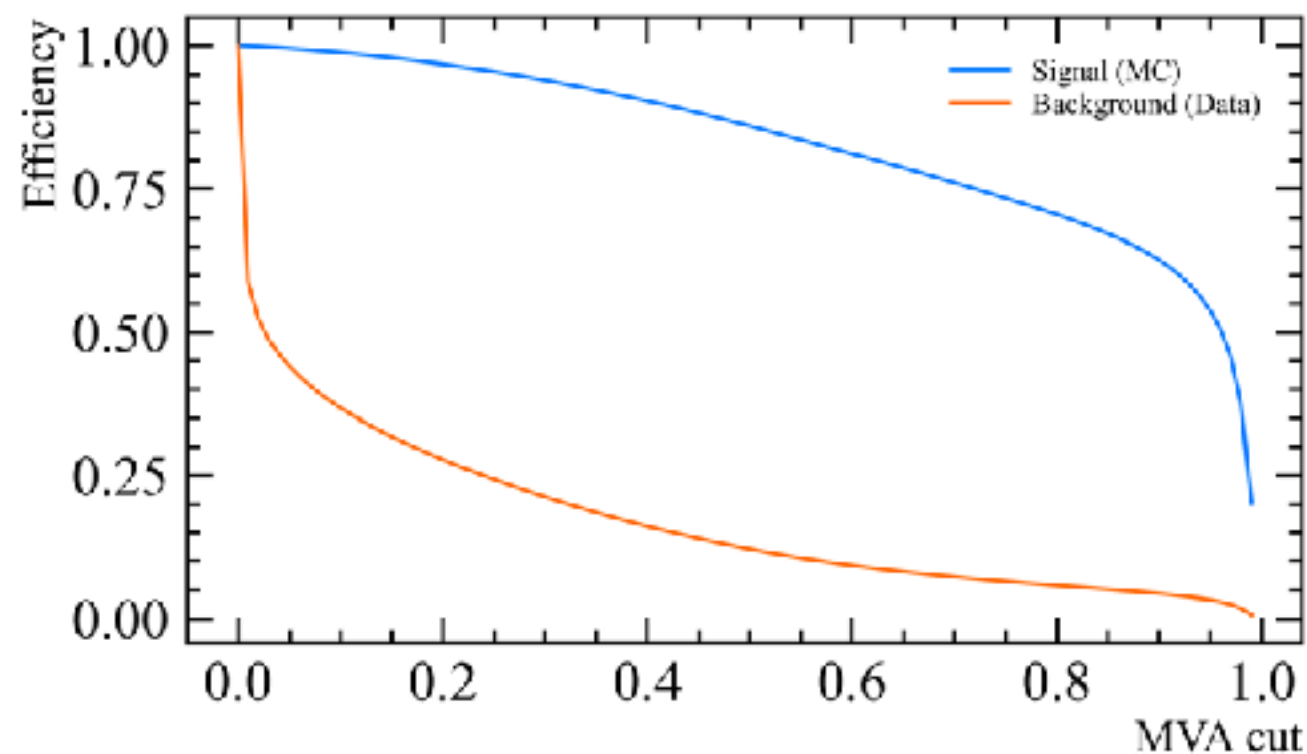
```
# Lb -> J/Psi p K
# L0 trigger requirement + enforced mu_P, mu_PT cut to get rid of L0 mis-modelling
sel = (( (rootdf['J_psi_1S_L0MuonDecision_TOS']==1) & (rootdf['muplus_P']>2000) & (rootdf['muplus_PT']>2000) )
|       | ( (rootdf['J_psi_1S_L0MuonDecision_TOS']==1) & (rootdf['muminus_P']>2000) & (rootdf['muminus_PT']>2000) ))
```

- ◇ Enforced a cut on **mu_P, mu_PT** > 2 GeV
- ◇ Repeat the analysis chain for **2016, 2017, 2018**
 - ◇ Run PIDCorr to retrieve PID vars correlations ({p, K}ProbNN, DLL)
 - ◇ Apply preselection
 - ◇ Build MC/Data samples for MVA to tackle MisID/comb bkg
 - ◇ Optimise hyperparams MVA
 - ◇ Train MVA. Post training checks on Lb_M correlation with MVA score
 - ◇ Apply MVA cut and design a MisID selection (cut on reflection backgrounds)
 - ◇ Pre-fit MC Lb_M to extract the shape parameters of signal (extended unbinned ML fit with Double Crystal Ball)
 - ◇ Extended unbinned ML fit data Lb_M with DCB (sig) + exp (bkg)
 - ◇ Extract sWeights for signal Data
 - ◇ Calculate the Data/MC correction in bins of Lb_P, Lb_PT. Check pre-post correction distributions

MVA TRAINING SET VARIABLES

Taken from the Pentaquark analysis [[PRL122\(2019\)222001](#)]

- ◆ $K_{\text{PT}} + p_{\text{PT}}$
- ◆ $\ln p_{\text{P}}$
- ◆ $\ln \text{IPCHI2}_{\Lambda_b}$
- ◆ $\Lambda_{b\text{PT}}$
- ◆ $\ln \text{FD}_{\Lambda_b}$
- ◆ $\ln \text{CHI2OWNPV}_{\Lambda_b}$
- ◆ $\ln(1 - \text{DIRA}_{\Lambda_b})$
- ◆ $\ln \text{MINIPCHI2}_h$
- ◆ $\min \text{DLL}(\mu - \pi)$ - PIDCorr sampled
- ◆ $p_{\text{probNN}p}$ - PIDCorr sampled
- ◆ $K_{\text{probNN}K}$ - PIDCorr sampled

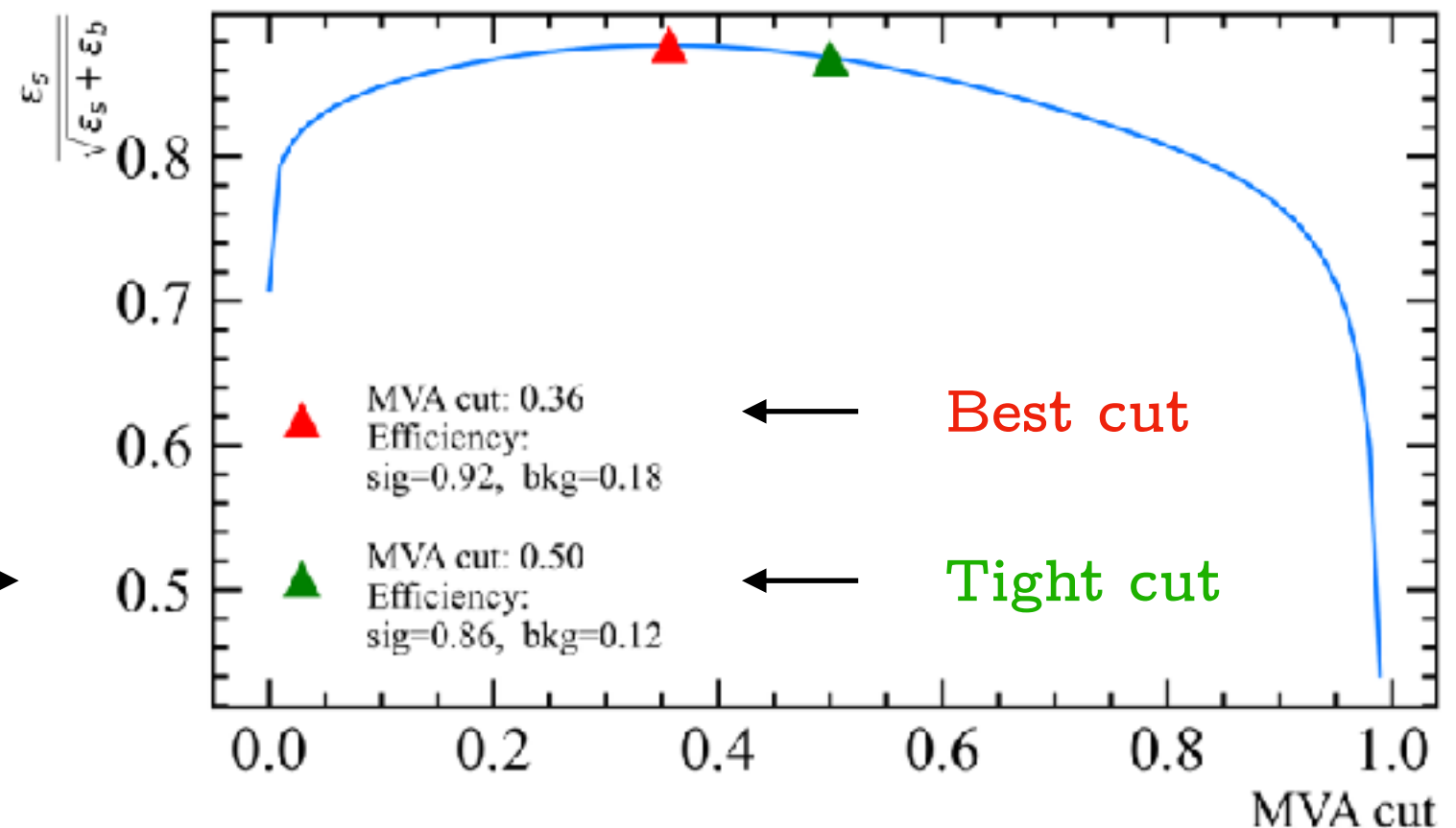


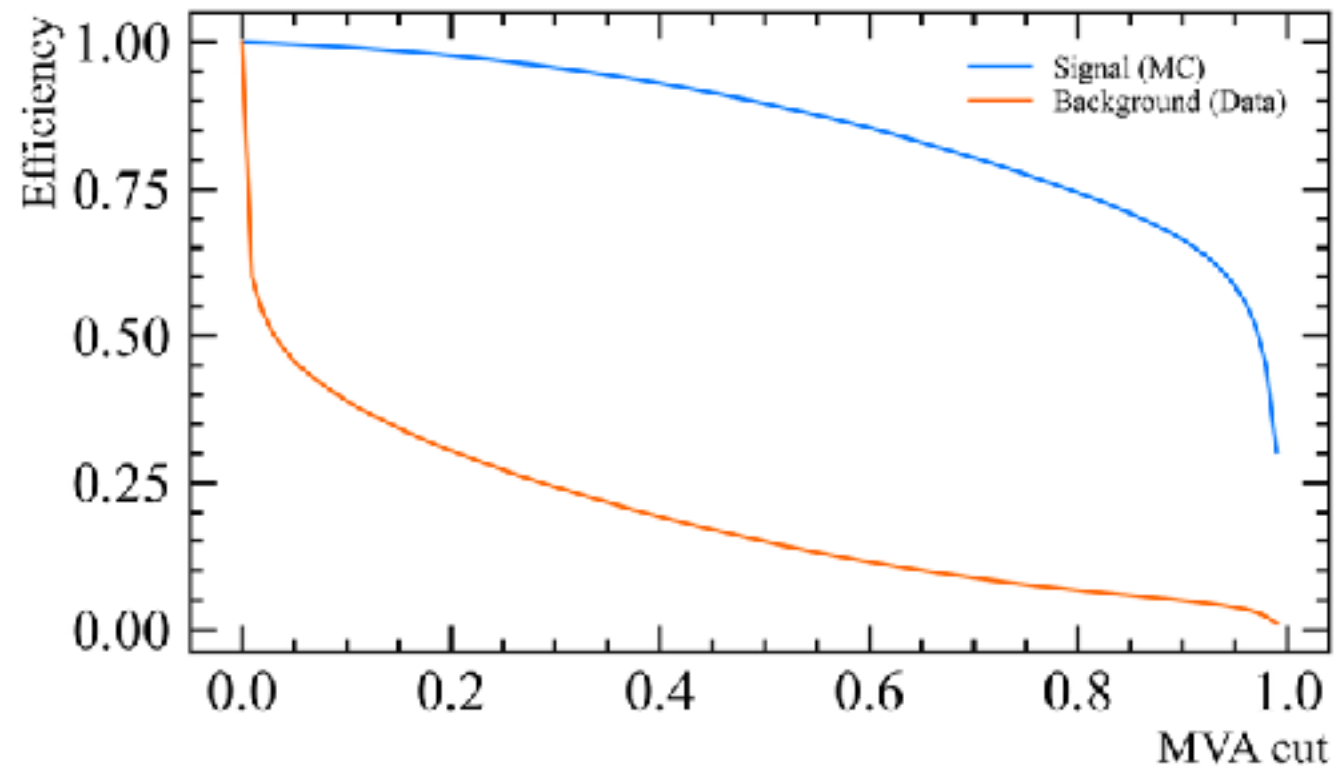
I have had a quick look at the efficiency curves. No fitting to extract the yields.

Signal: peak region

Background: lower and upper sidebands

Significance of the cut
as efficiency ratios



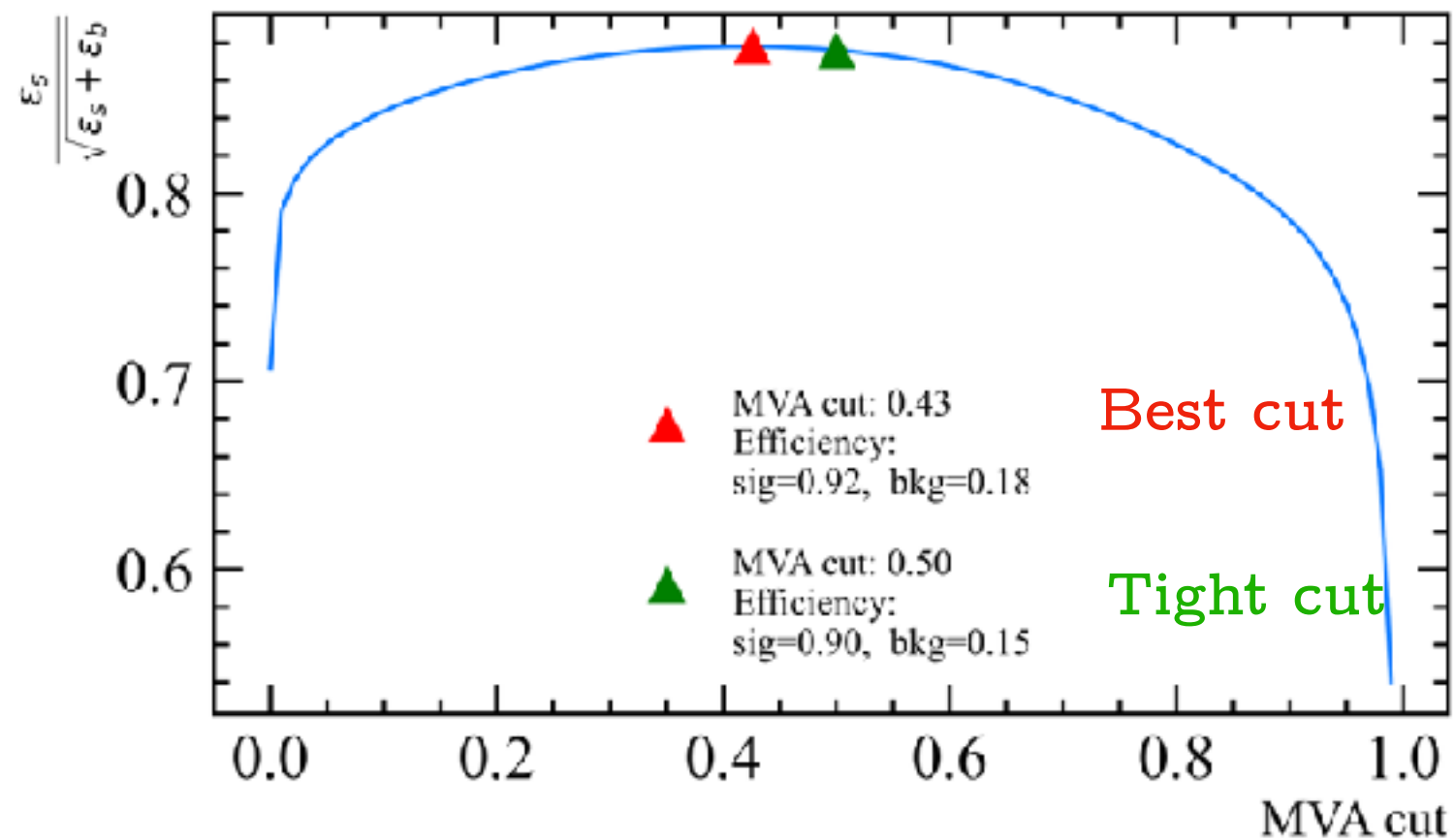


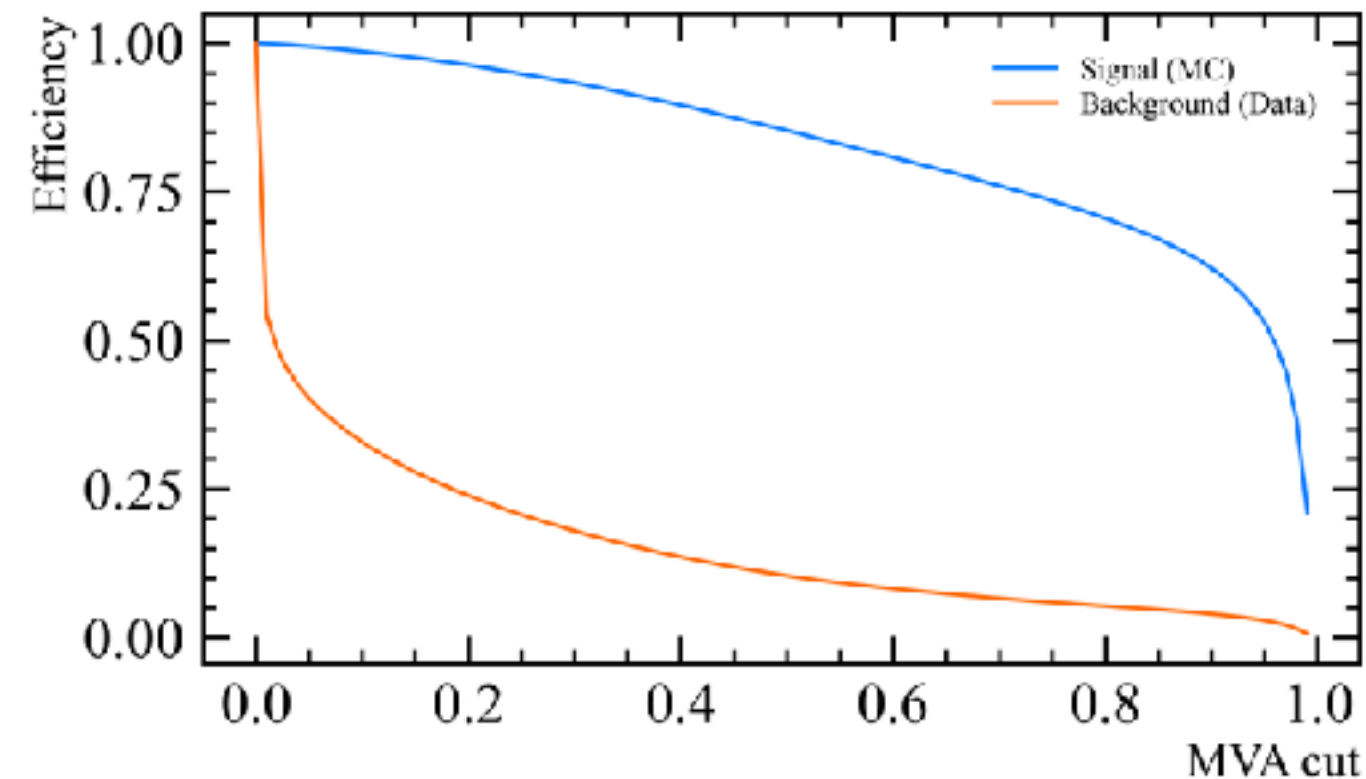
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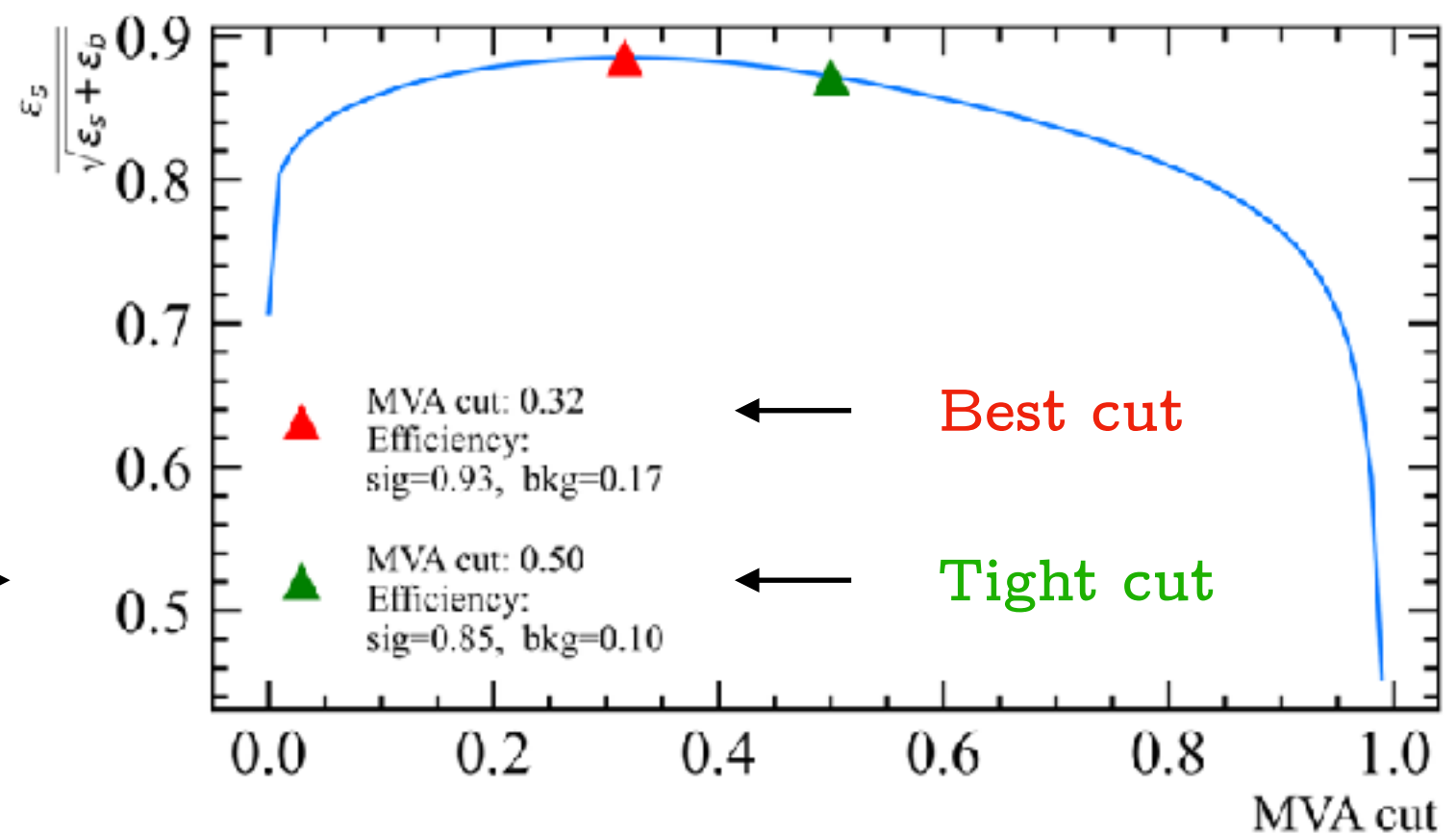


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Significance of the cut
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MisID COMPONENTS

◇ The three main contributions to MisID of $\Lambda_b^0 \rightarrow J/\Psi p K^-$

a) $\bar{B}_s^0 \rightarrow J/\Psi(\Phi \rightarrow K^+ K^-)$

b) $B^0 \rightarrow J/\Psi \pi^+ K^-$

c) $\bar{\Lambda}_b^0 \rightarrow J/\Psi \bar{p} K^+$

I have checked the following:

◇ The **reconstructed mass** distribution in the replaced mass hypothesis for each MisID channel, **after** the MVA selection

◇ Case I: MVA score > best cut

◇ Case II: MVA score > 0.50 (tighter cut)

◇ The **reconstructed Λ_b mass** after a $1\sigma, 2\sigma, 3\sigma$ Veto cut on the MisID reconstructed mass for each contribution ($\sigma = 20 \text{ MeV}$):

◇ Singularly (cut on a, b or c separately)

◇ Combined cut

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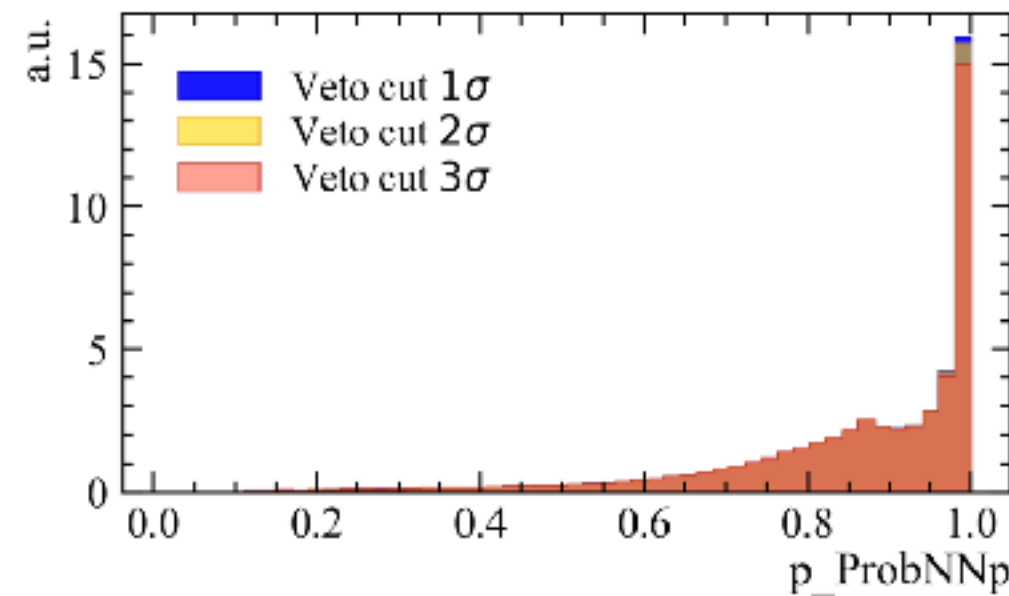
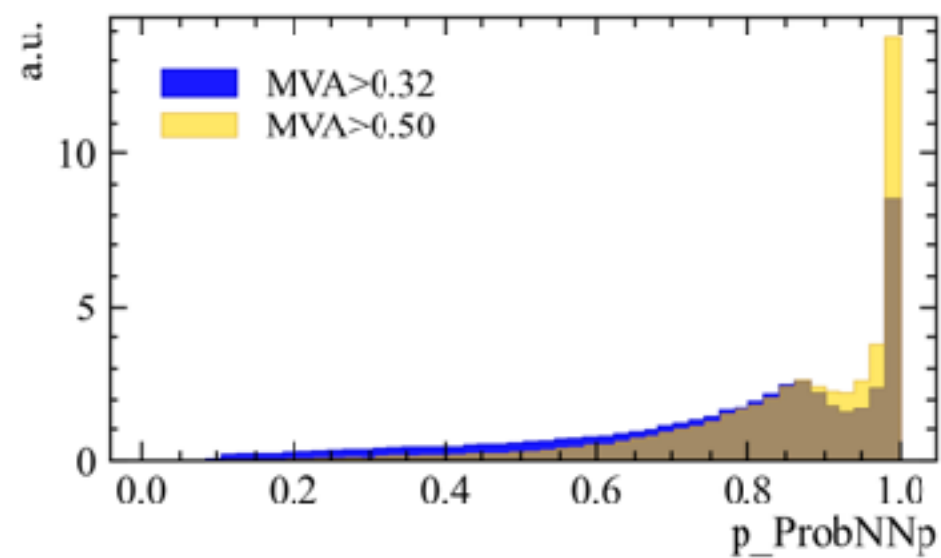
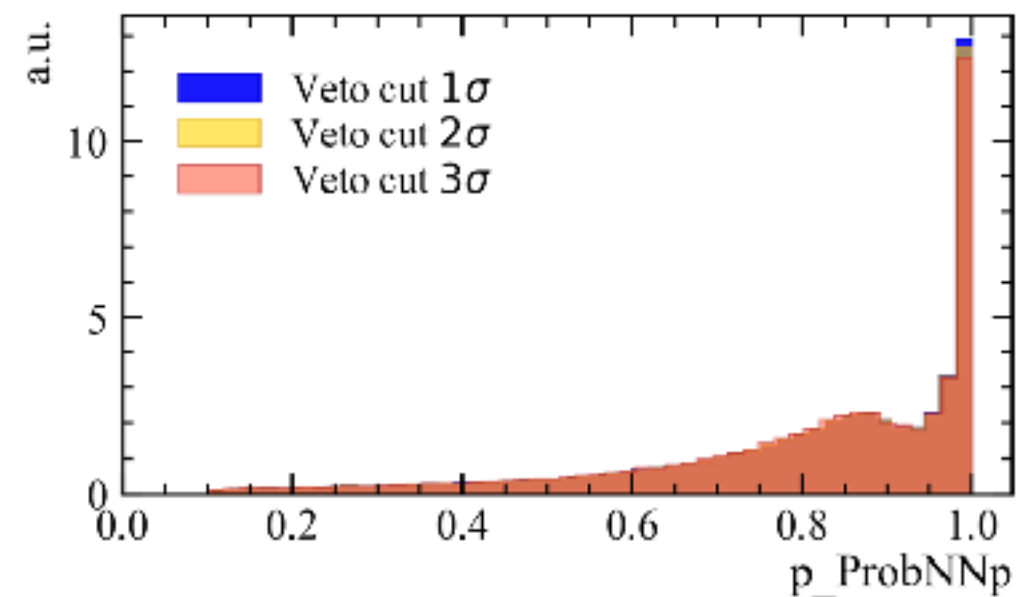
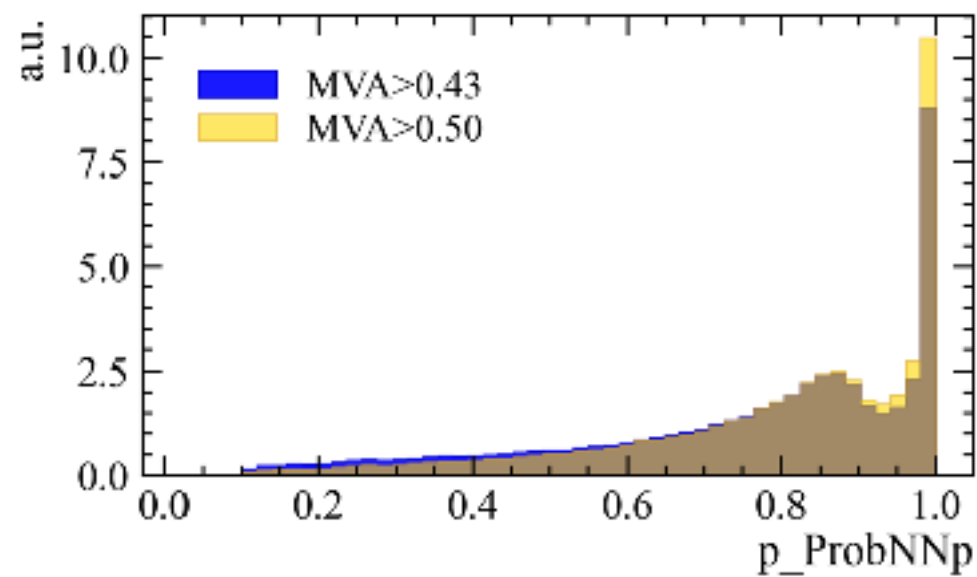
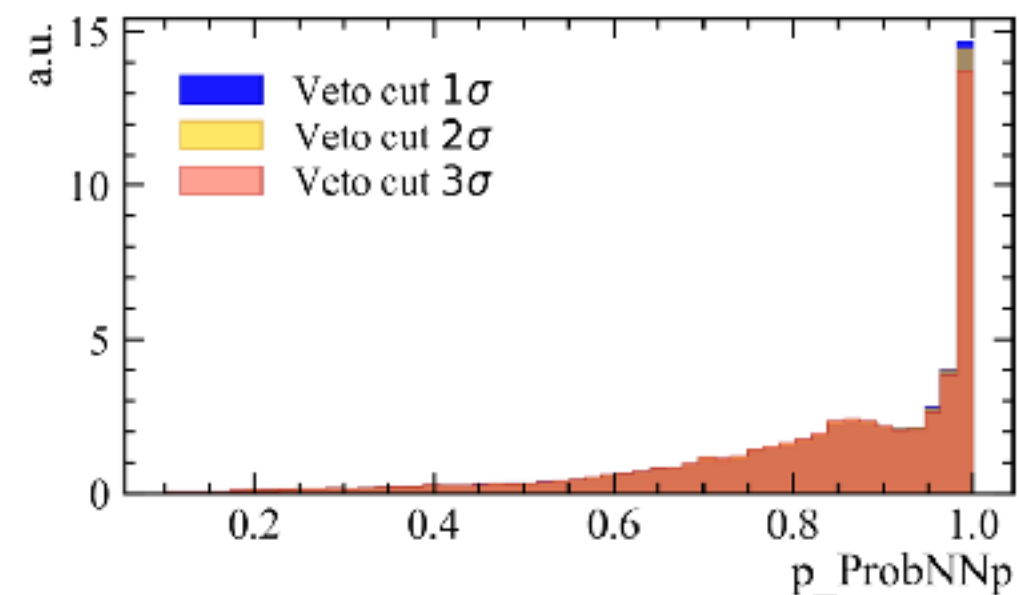
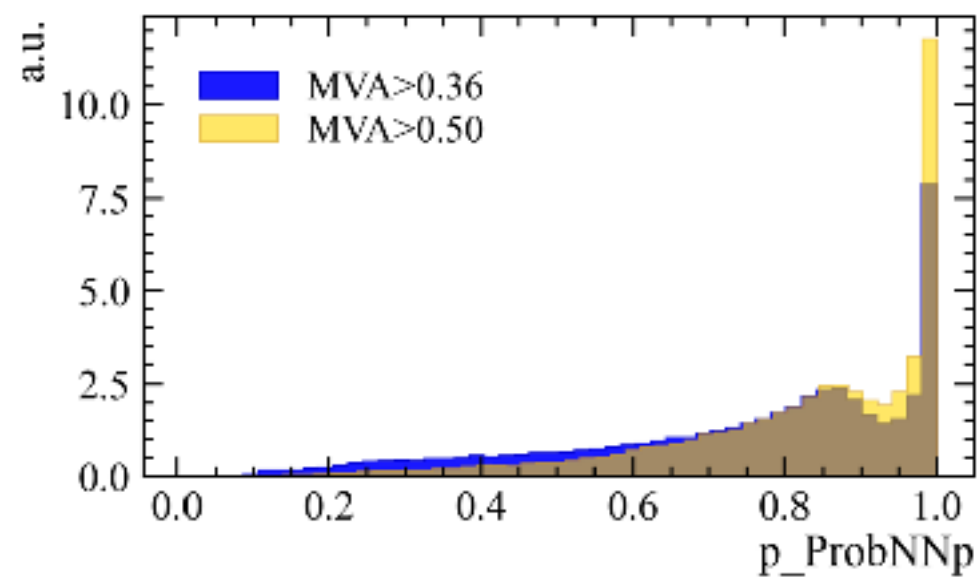
why: probNN distributions show less contamination

◇ The **reconstructed Λ_b mass** after a $1\sigma, 2\sigma, 3\sigma$ Veto cut on the MisID reconstructed mass for each contribution ($\sigma = 20$ MeV):

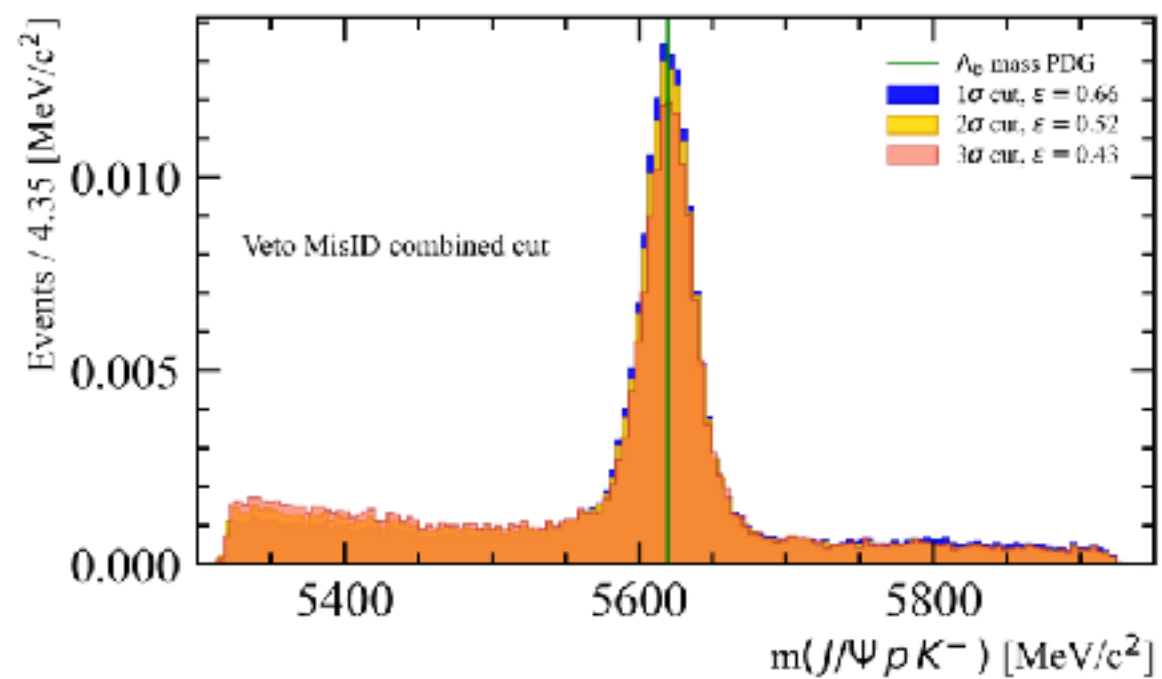
◇ Singularly (cut on a, b or c separately)

◇ Combined cut

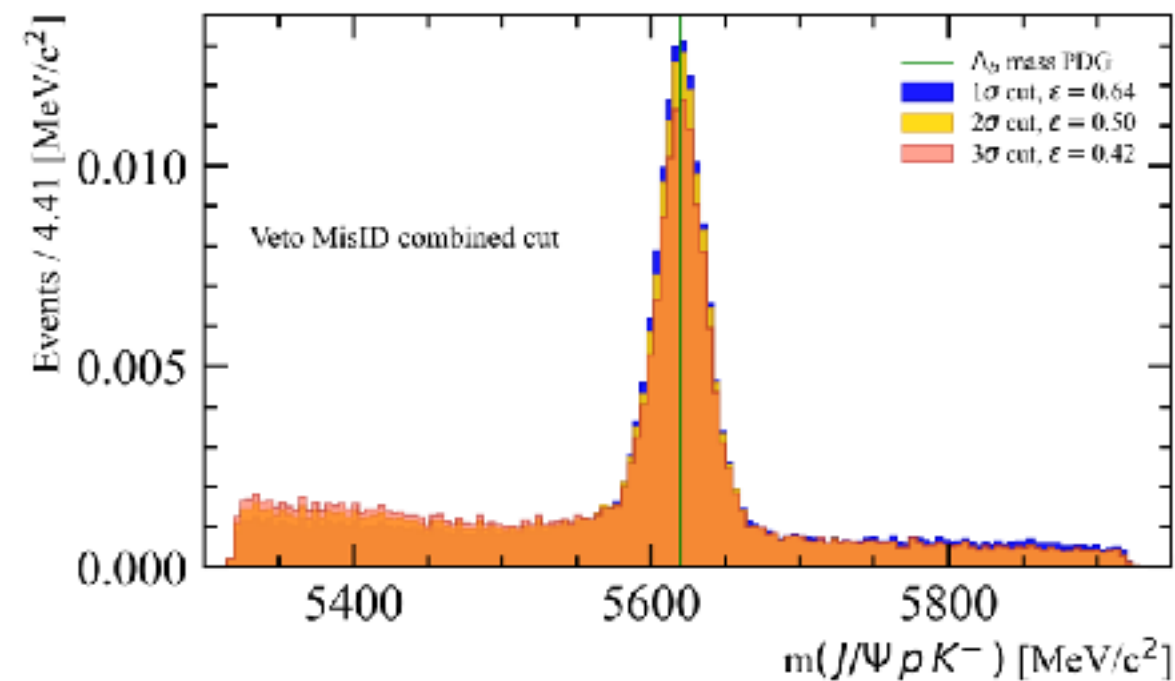
MisID COMPONENTS: PROBNN



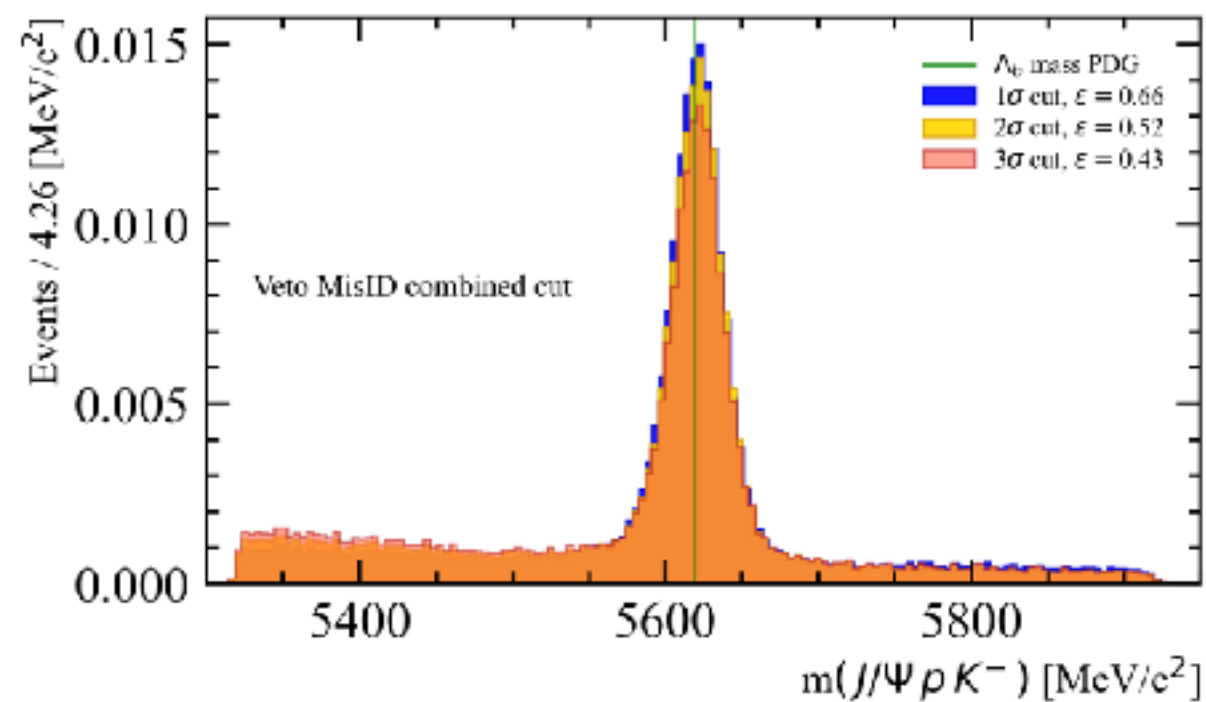
2016



2017

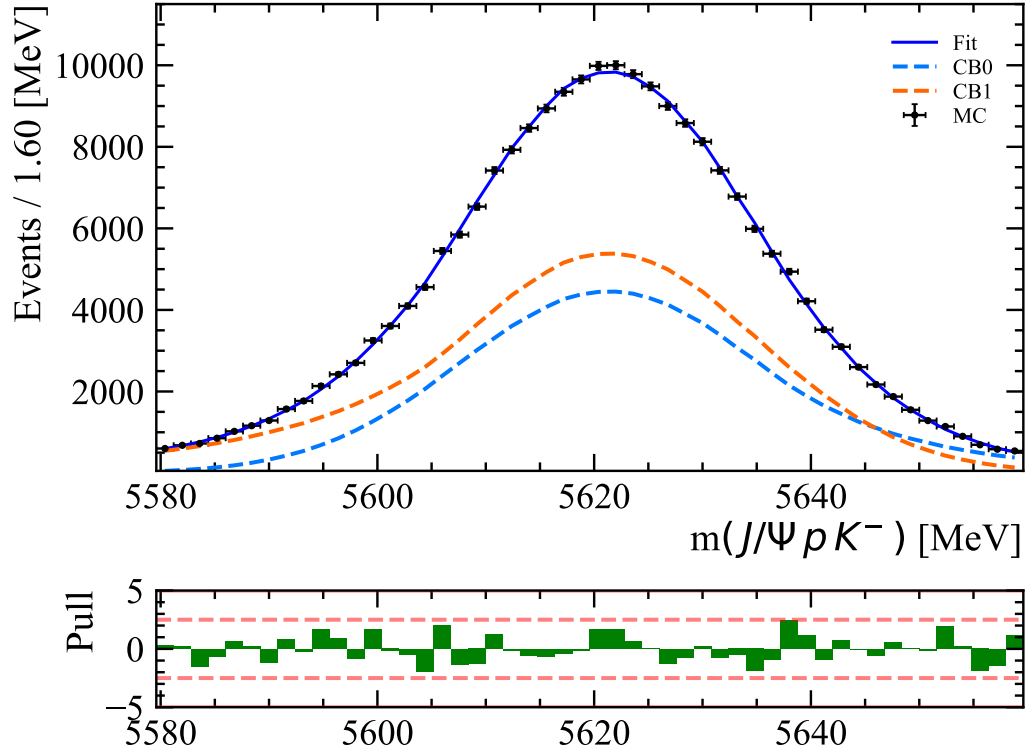


2018

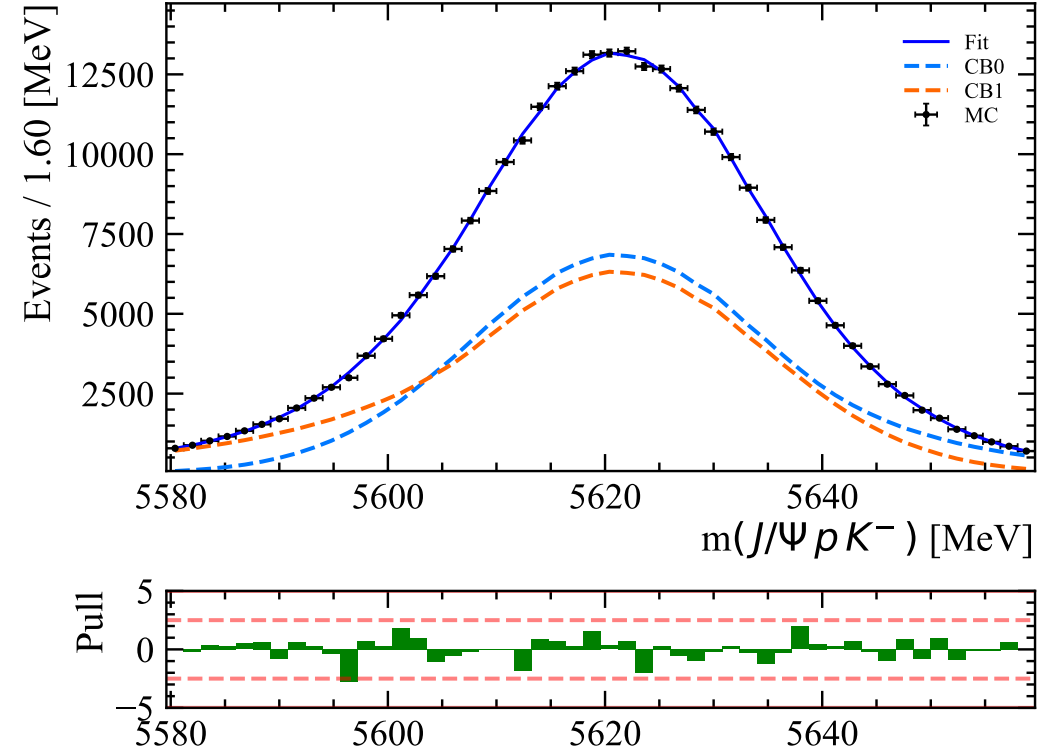


LBM FIT (MC)

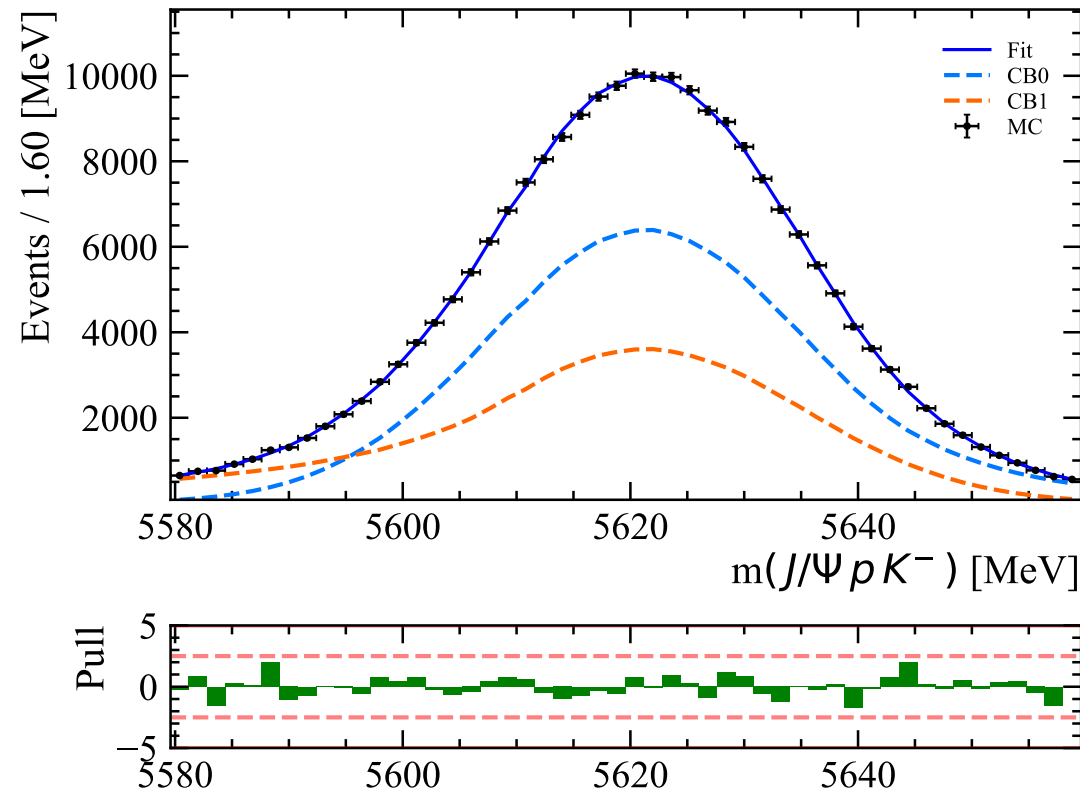
2016



2017

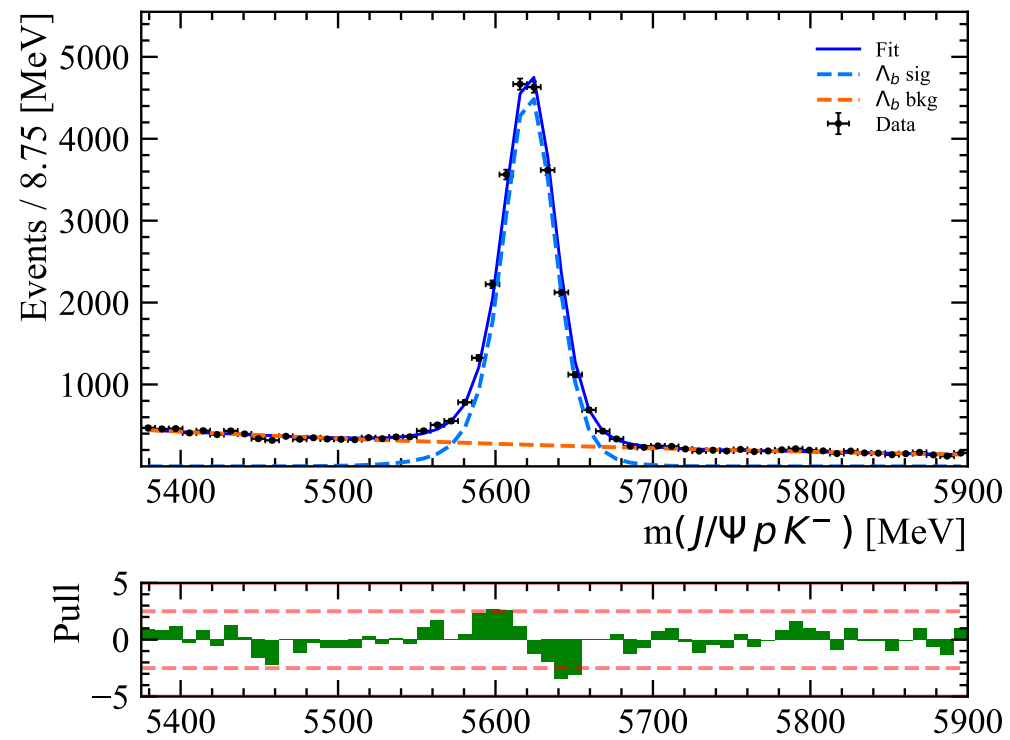


2018

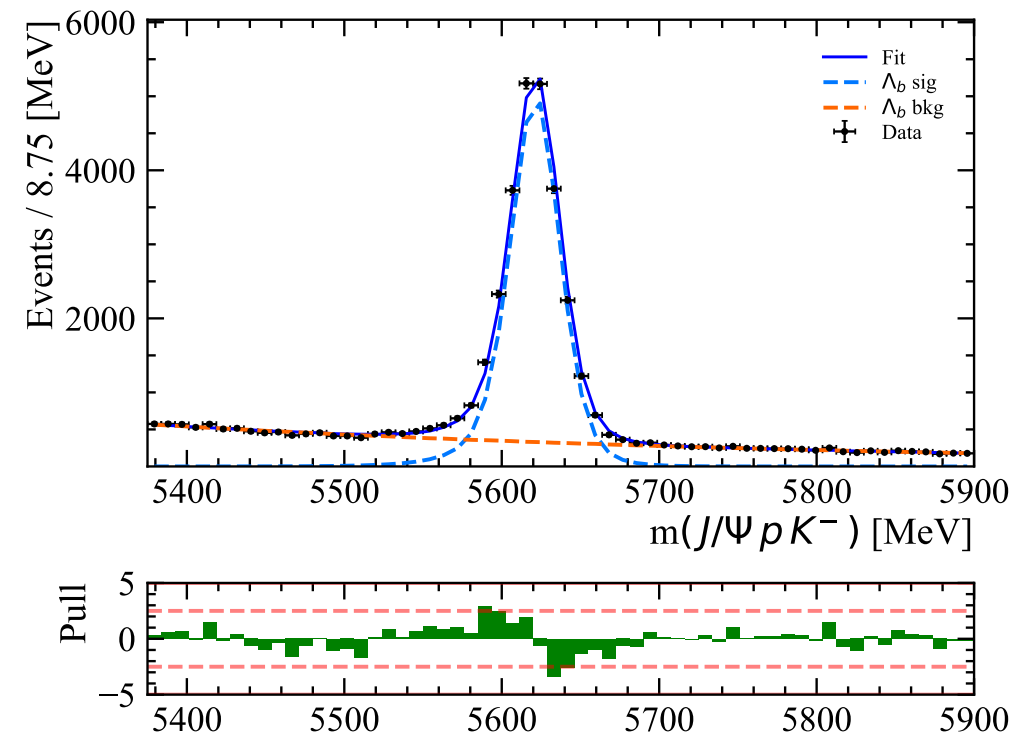


LBM FIT (DATA)

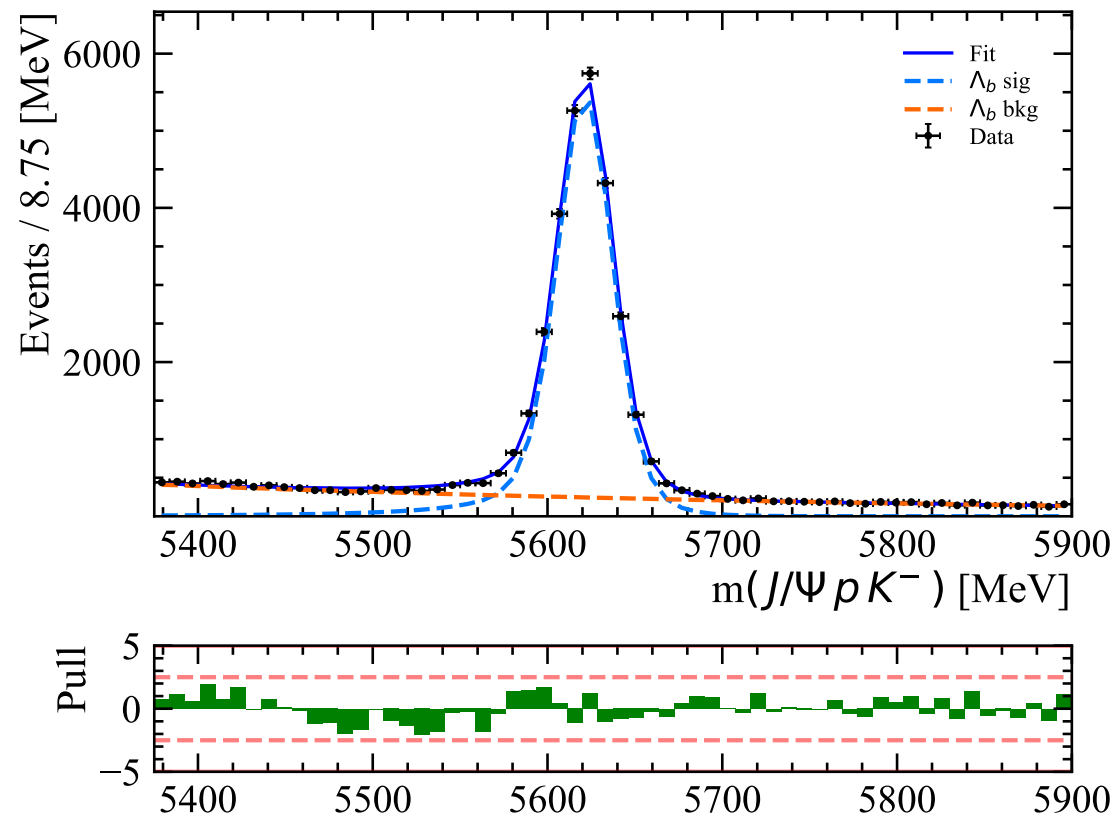
2016



2017



2018



EVALUATE THE CORRECTION: STRATEGY

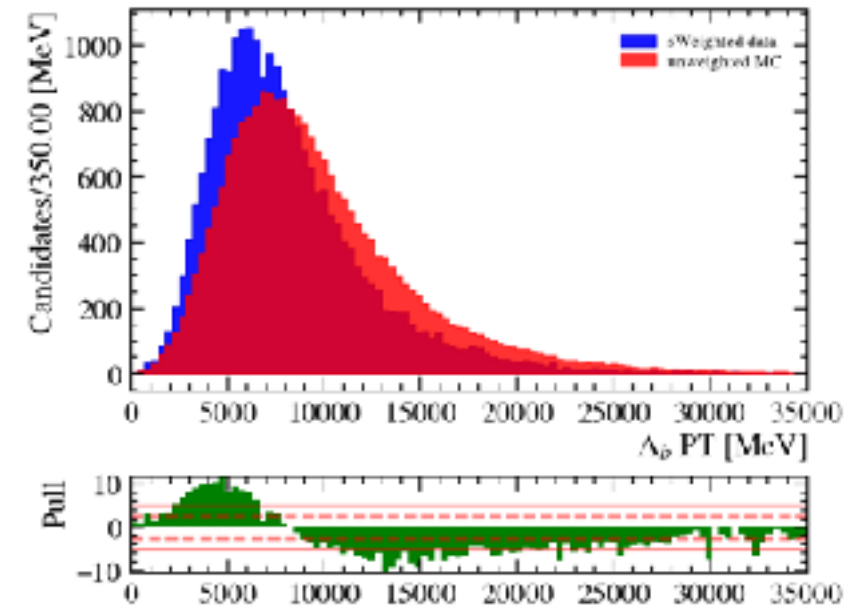
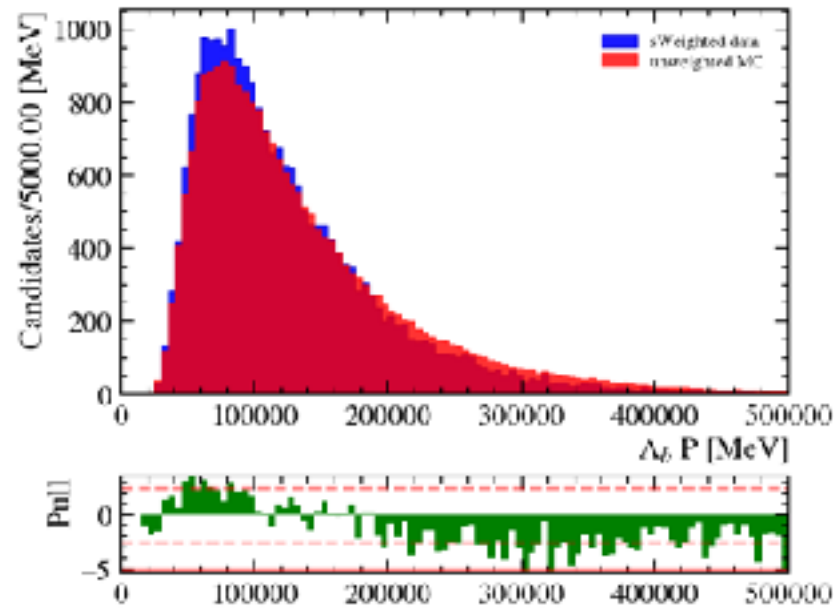
- Option 1:
 - Define a **regular-sized binning** scheme ($N_{\text{bins}} = 15 \times 15 = 225$);
 - Retrieve the corrections as the ratio of the normalised Data/MC 2D (**Lb_P**, **Lb_PT**) distributions;
 - Save to ROOT file;
 - Get the MC reweighed P and PT distributions and compare them to data.
- Option 2:
 - Define a **rectangular binning** scheme ($N_{\text{bins}} = 15 \times 30 = 450$);
 - Retrieve the corrections as the ratio of the normalised Data/MC 2D (**Lb_P**, **Lb_PT**) distributions;
 - Save to ROOT file;
 - Get the MC reweighed P and PT distributions and compare them to data.

STRATEGY

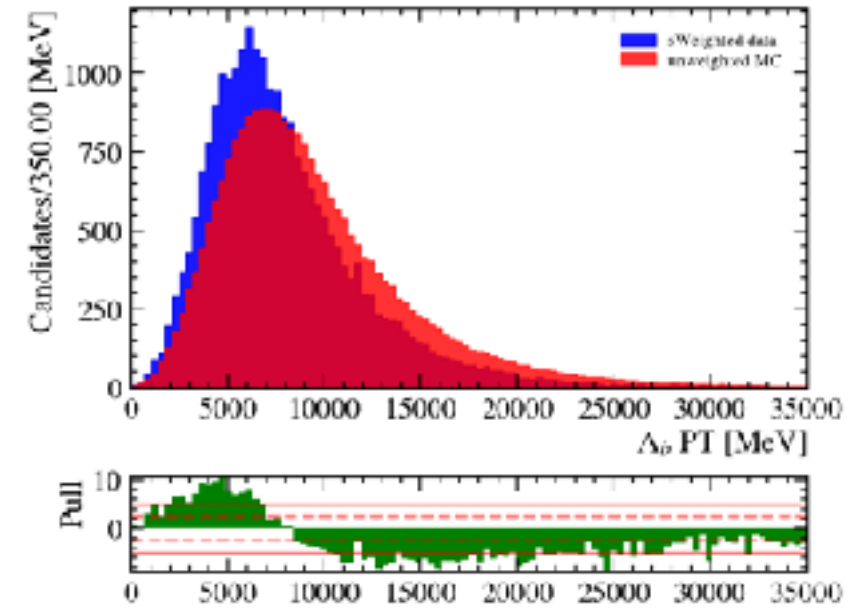
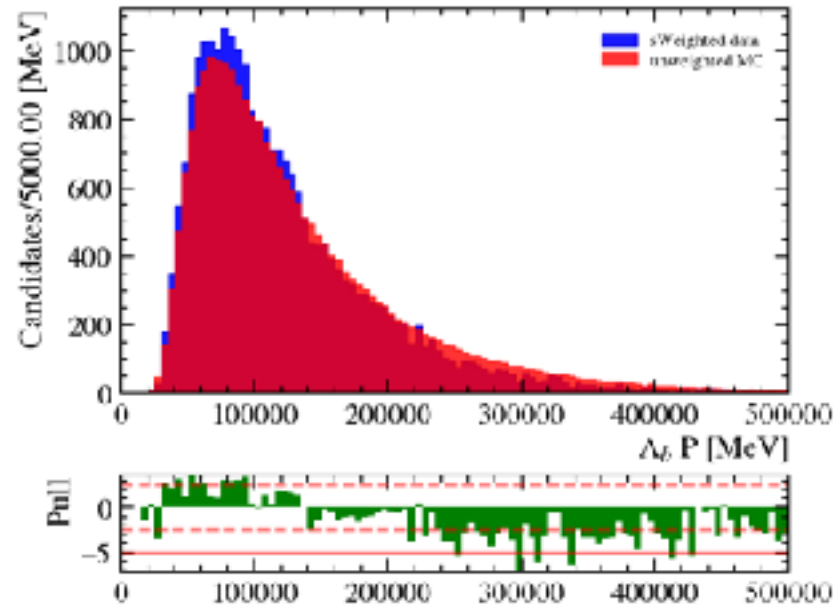
- Option 3:
 - Define a **polygonal binning scheme** by optimising for the number of MC entries per bin ($N_{\text{entries}} = 10^3 \Rightarrow N_{\text{bins}} \sim 250$);
 - Optimisation performed with a **KDTreeBinning** algorithm (ROOT);
 - The binning structure of MC is adopted also for Data. **TH2Poly** histograms are used;
 - Retrieve the corrections as the ratio of the normalised Data/MC 2D (**Lb_P**, **Lb_PT**) distributions;
 - Save to ROOT file;
 - Get the MC reweighed P and PT distributions and compare them to data.

PRE-CORRECTION DISTRIBUTIONS

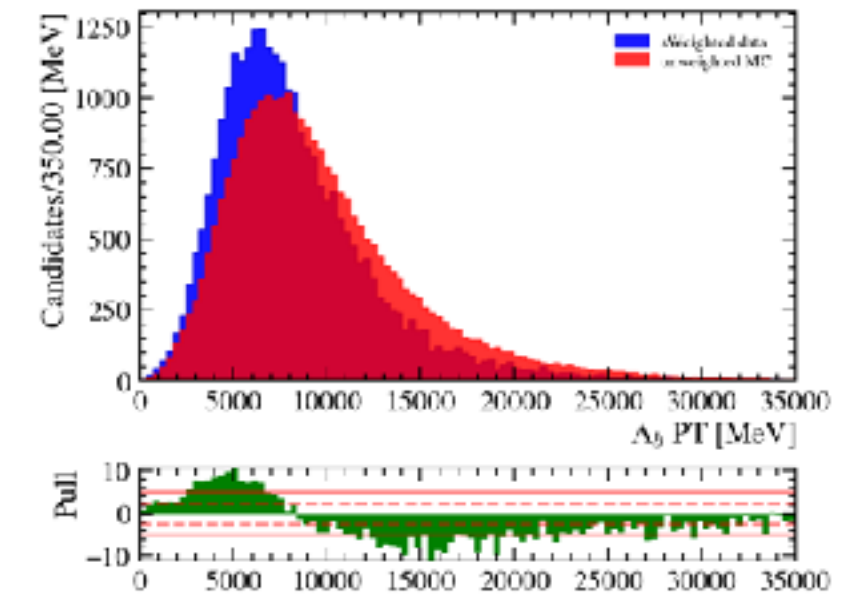
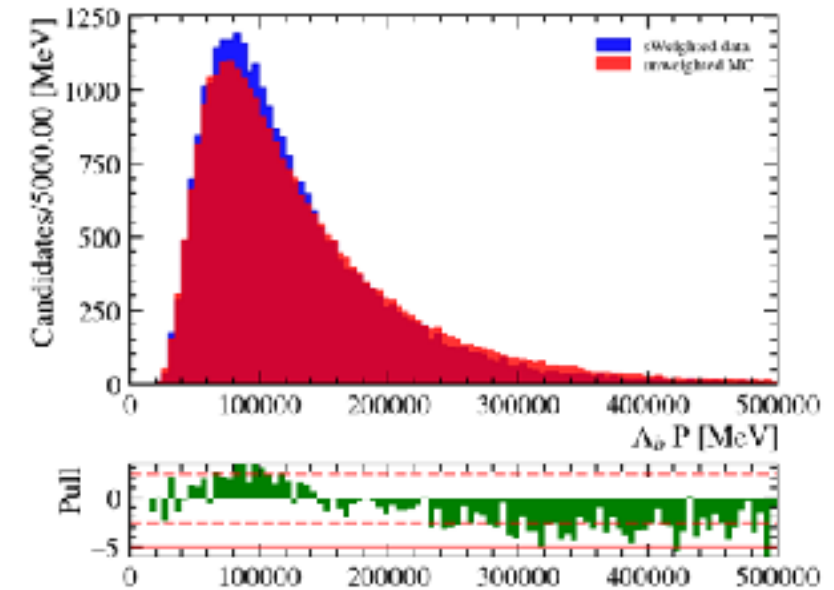
2016



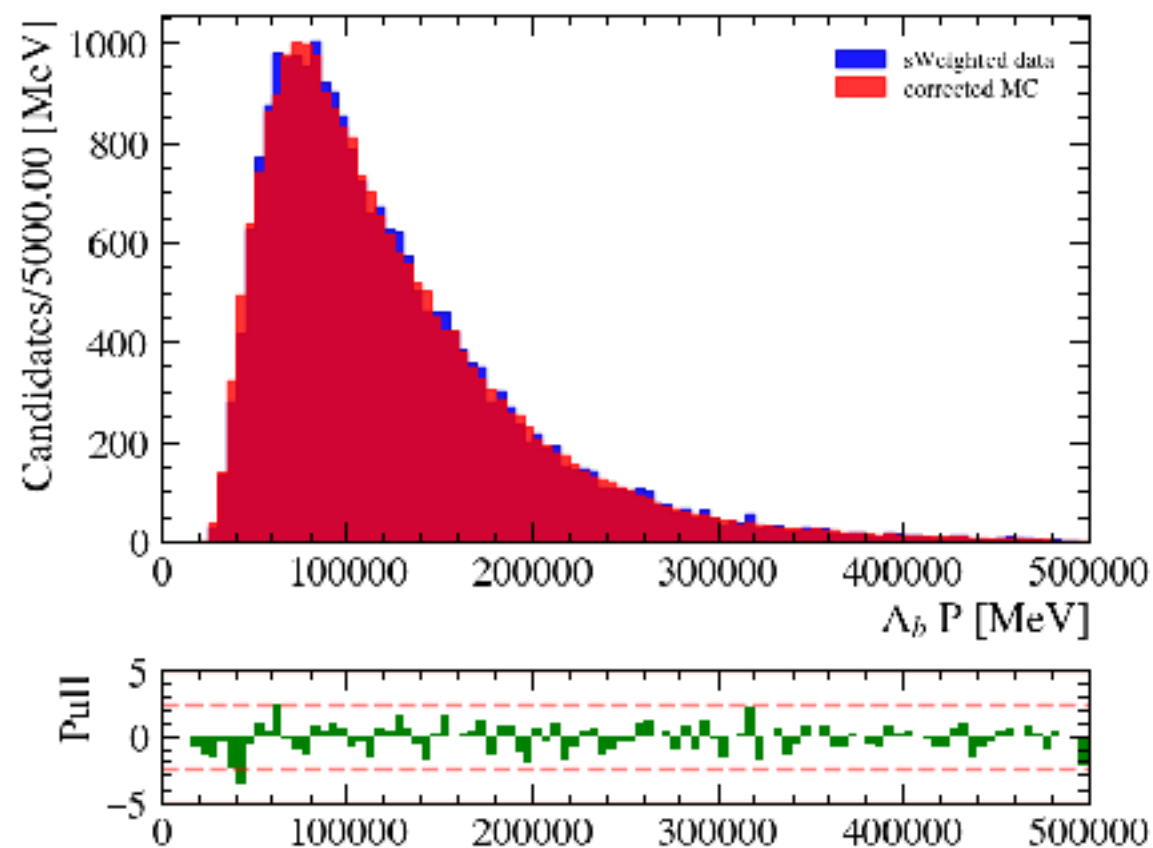
2017



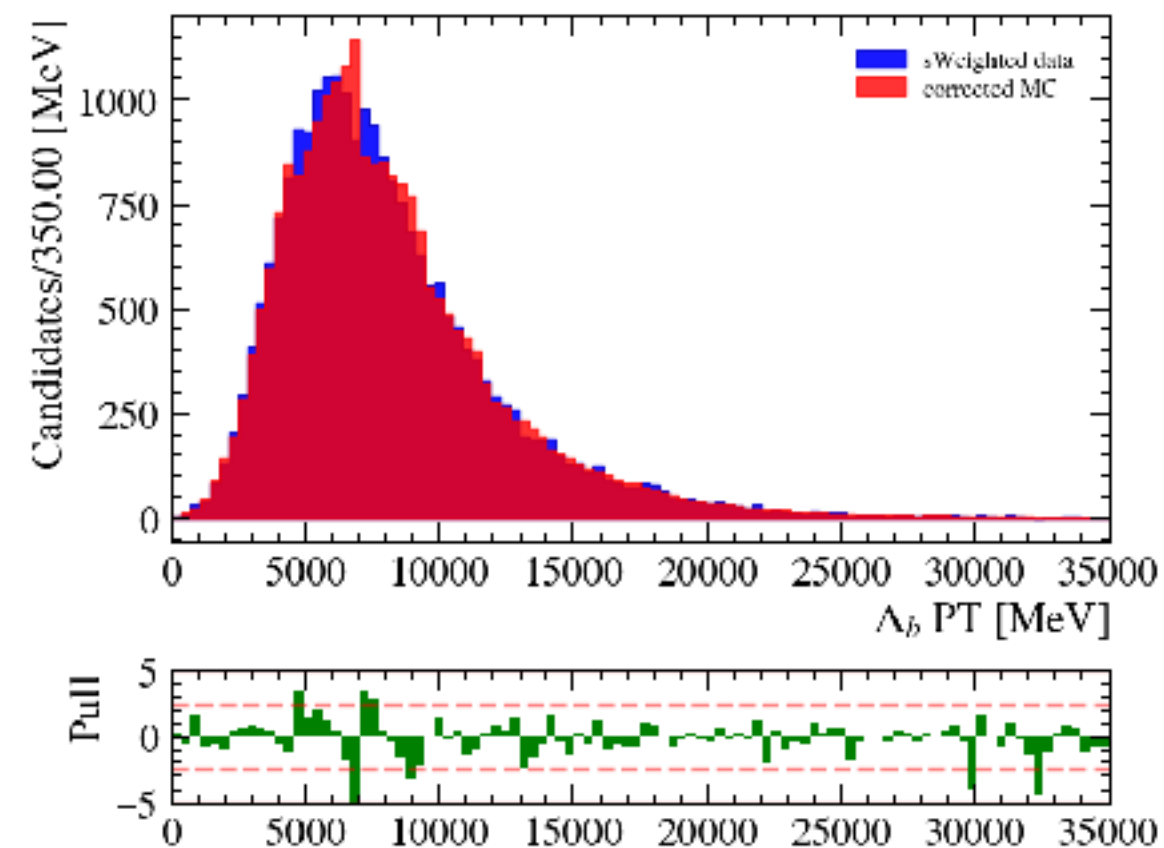
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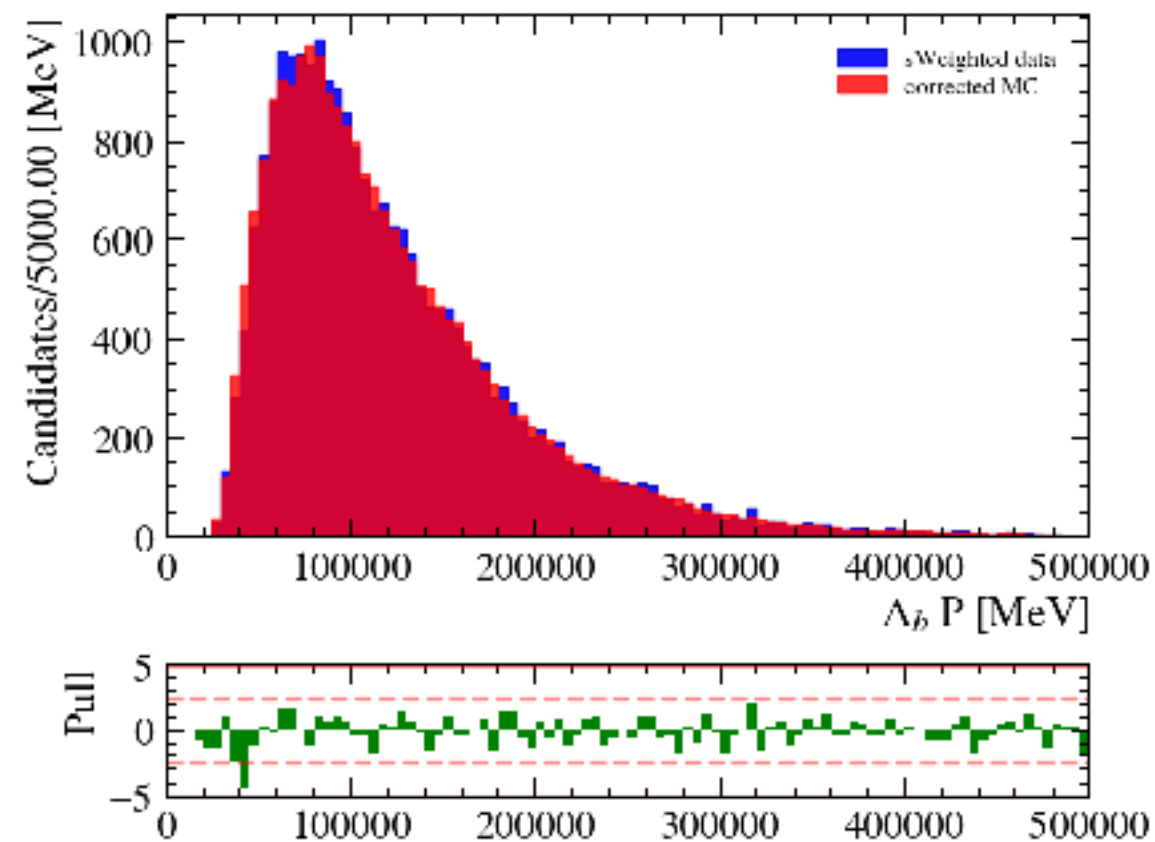
REGULAR BINNING



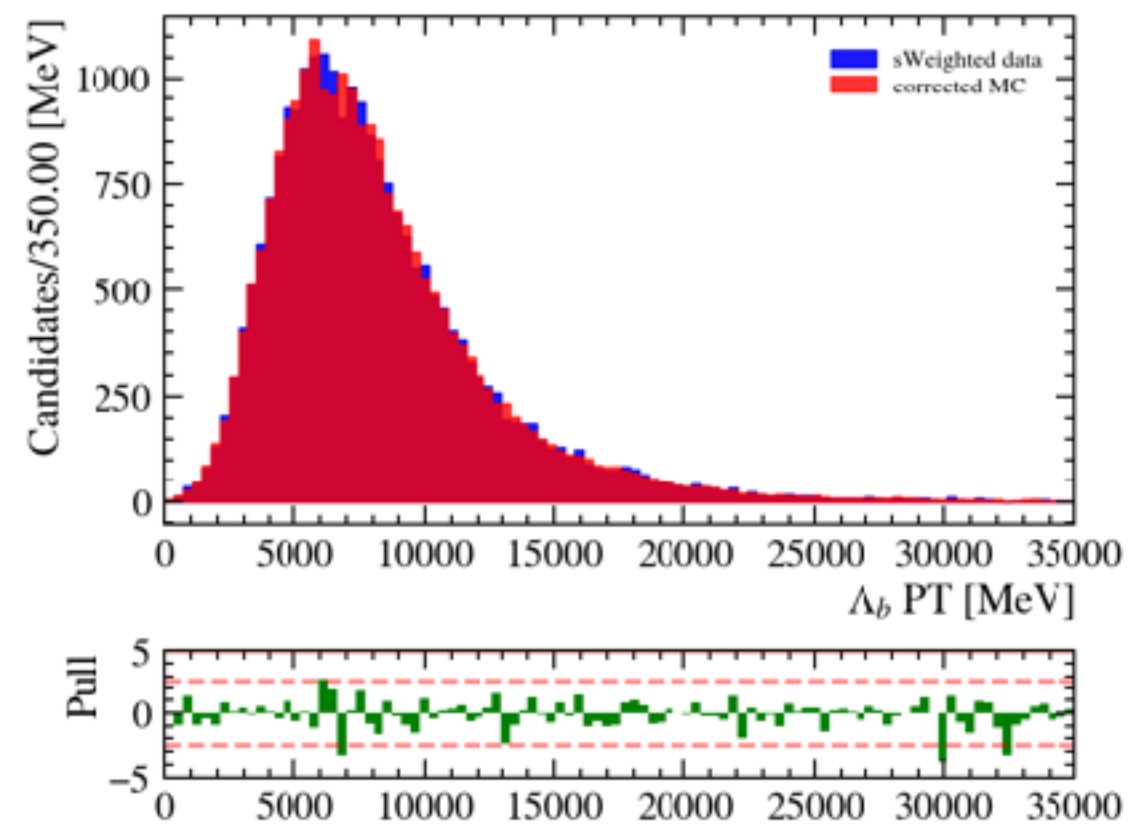
POST-CORRECTION DISTRIBUTIONS



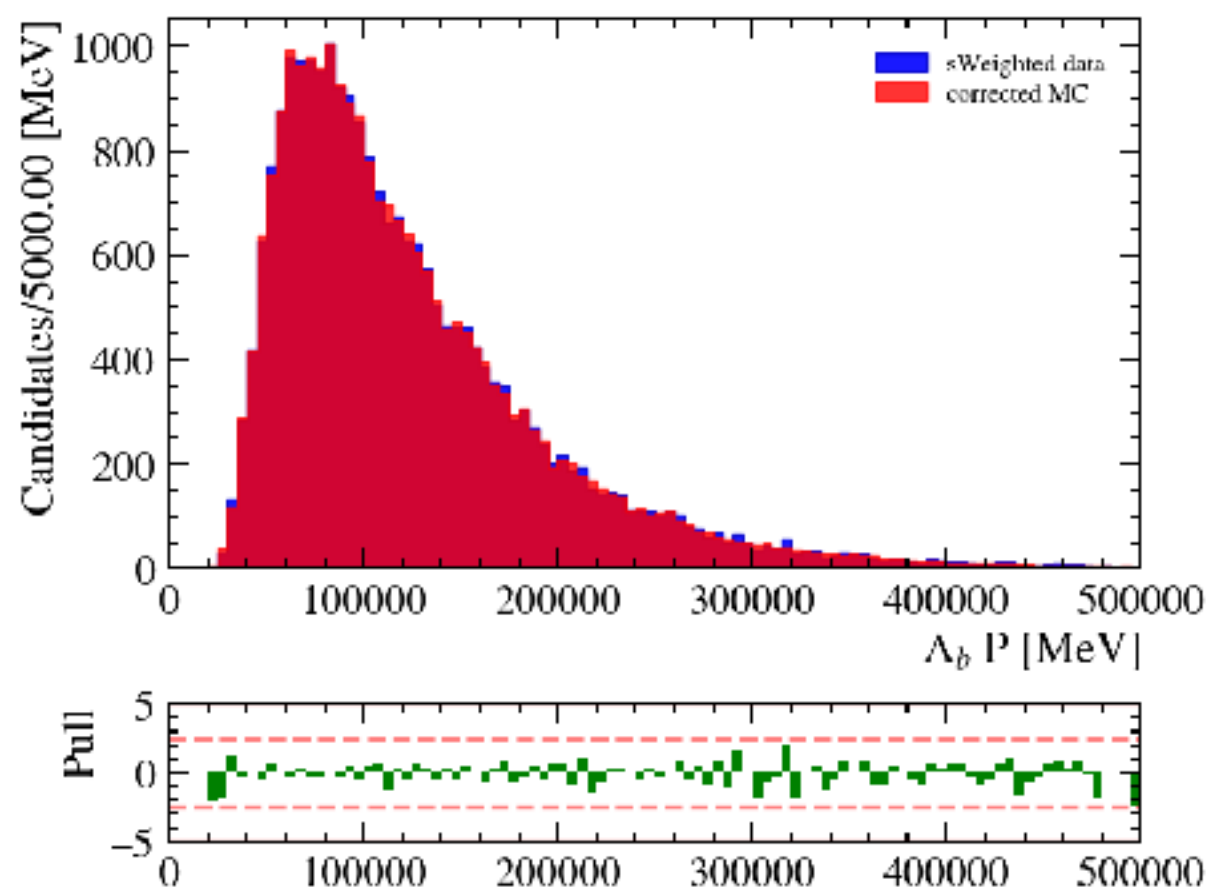
RECTANGULAR BINNING



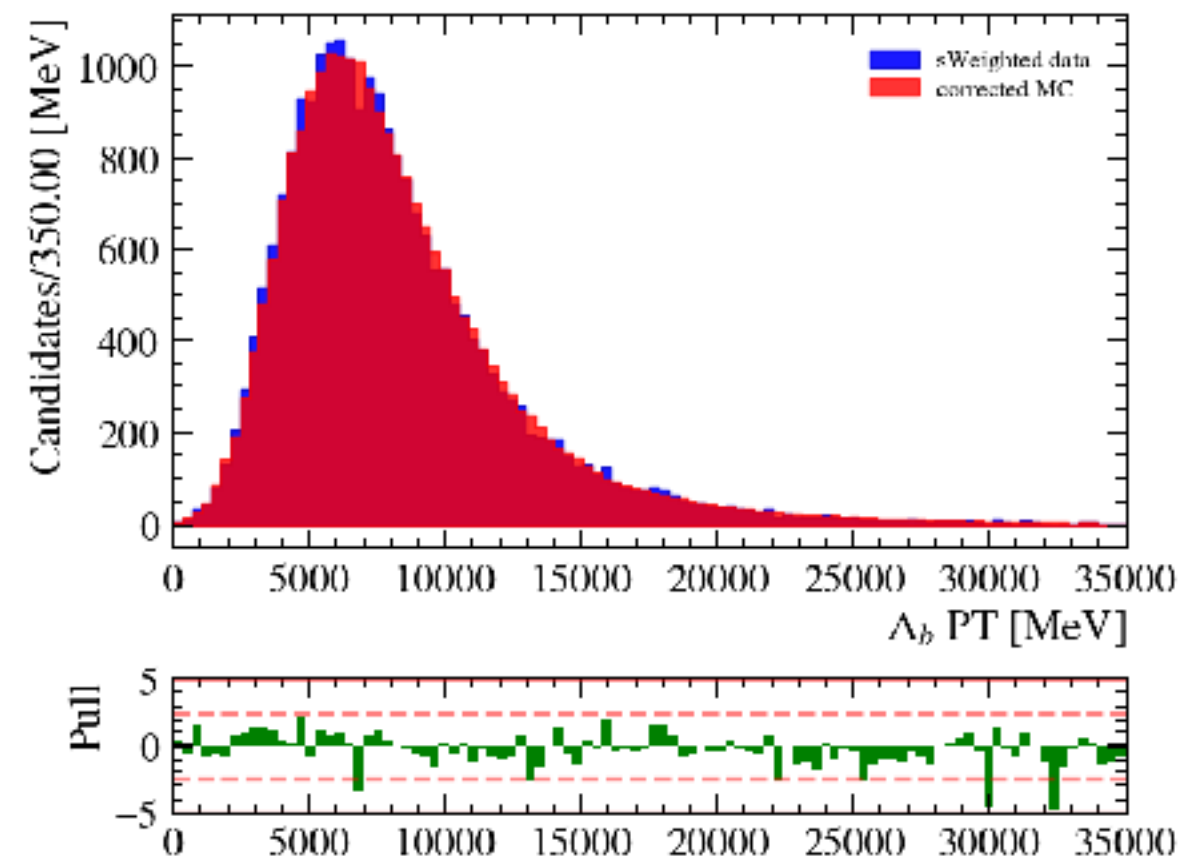
POST-CORRECTION DISTRIBUTIONS



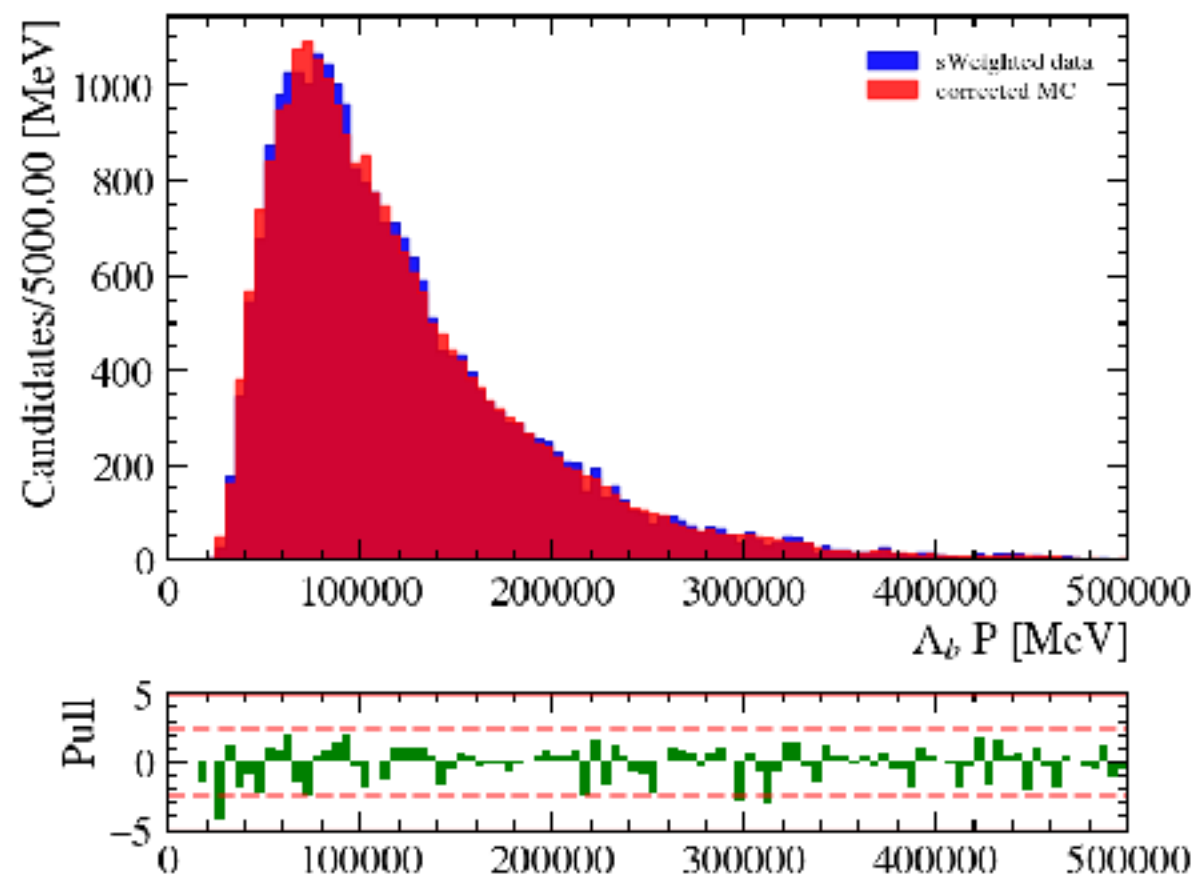
ADAPTIVE BINNING



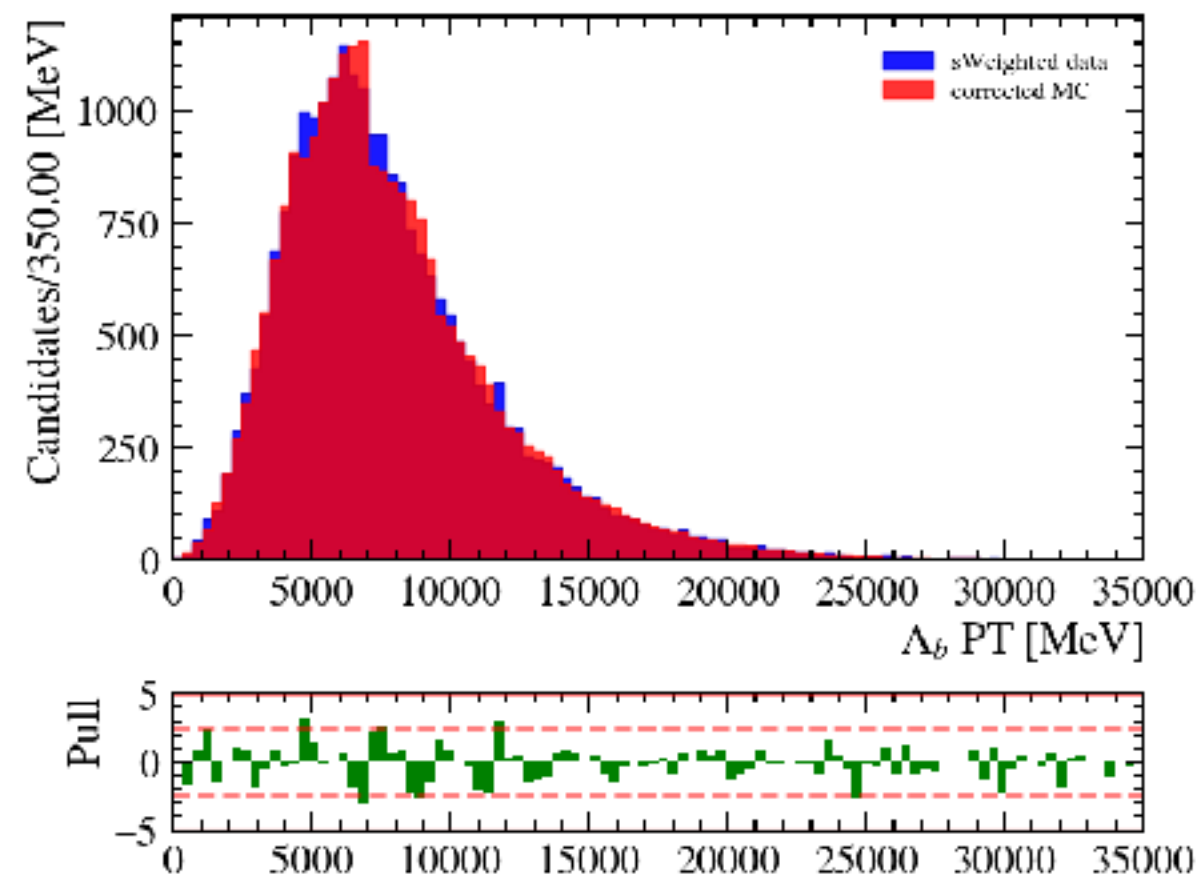
POST-CORRECTION DISTRIBUTIONS



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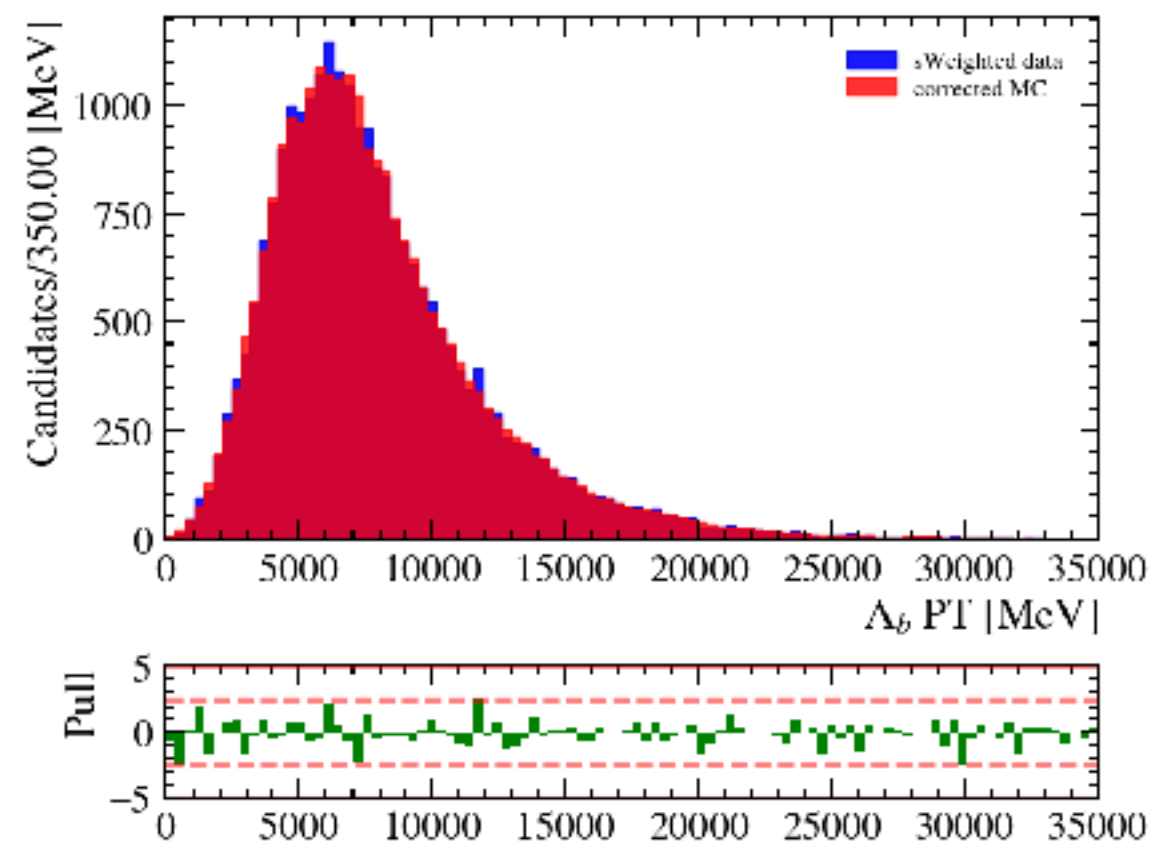
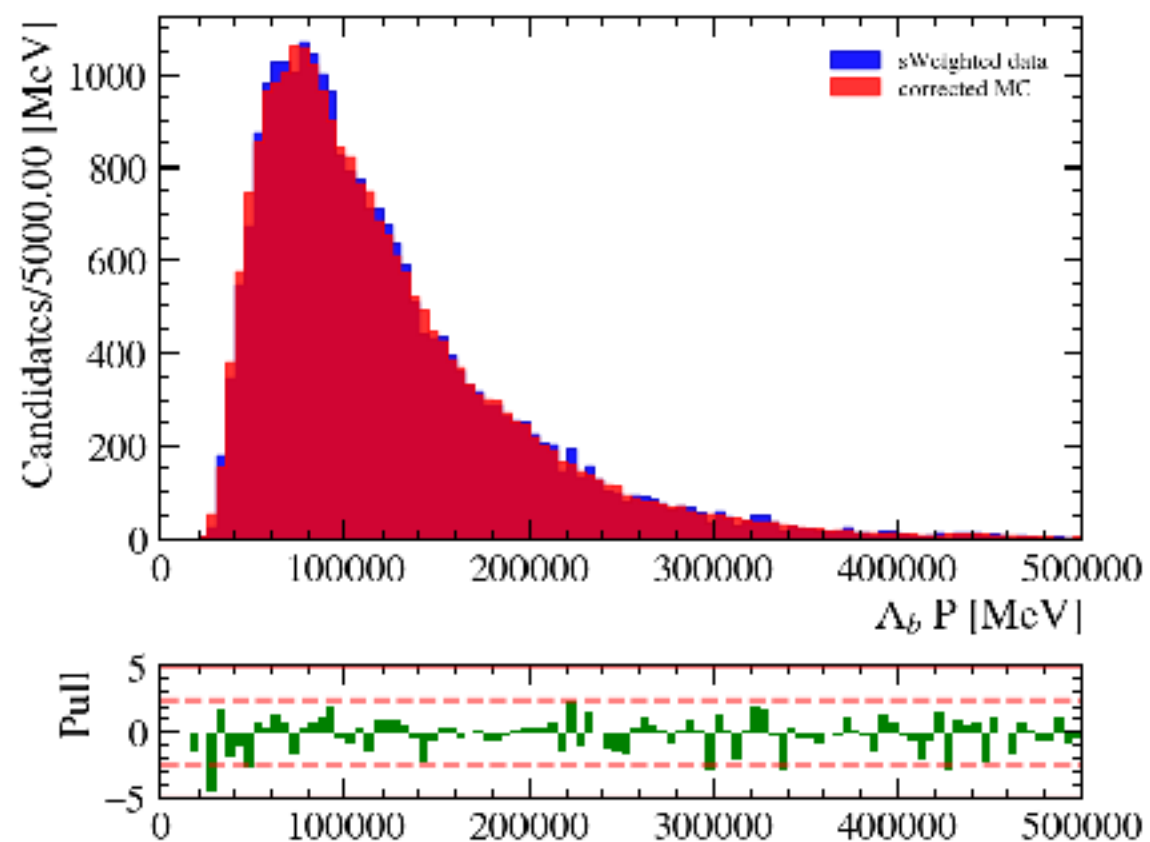


POST-CORRECTION DISTRIBUTIONS

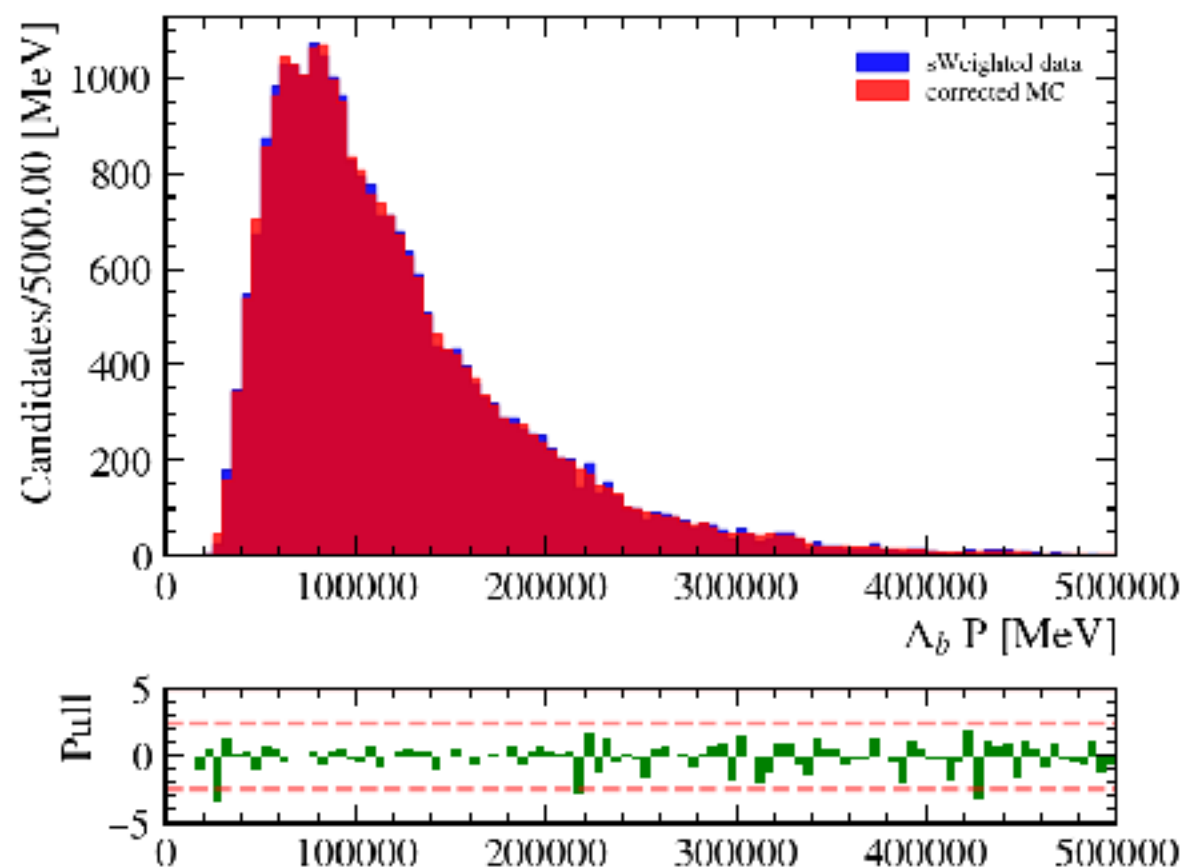


RECTANGULAR BINNING

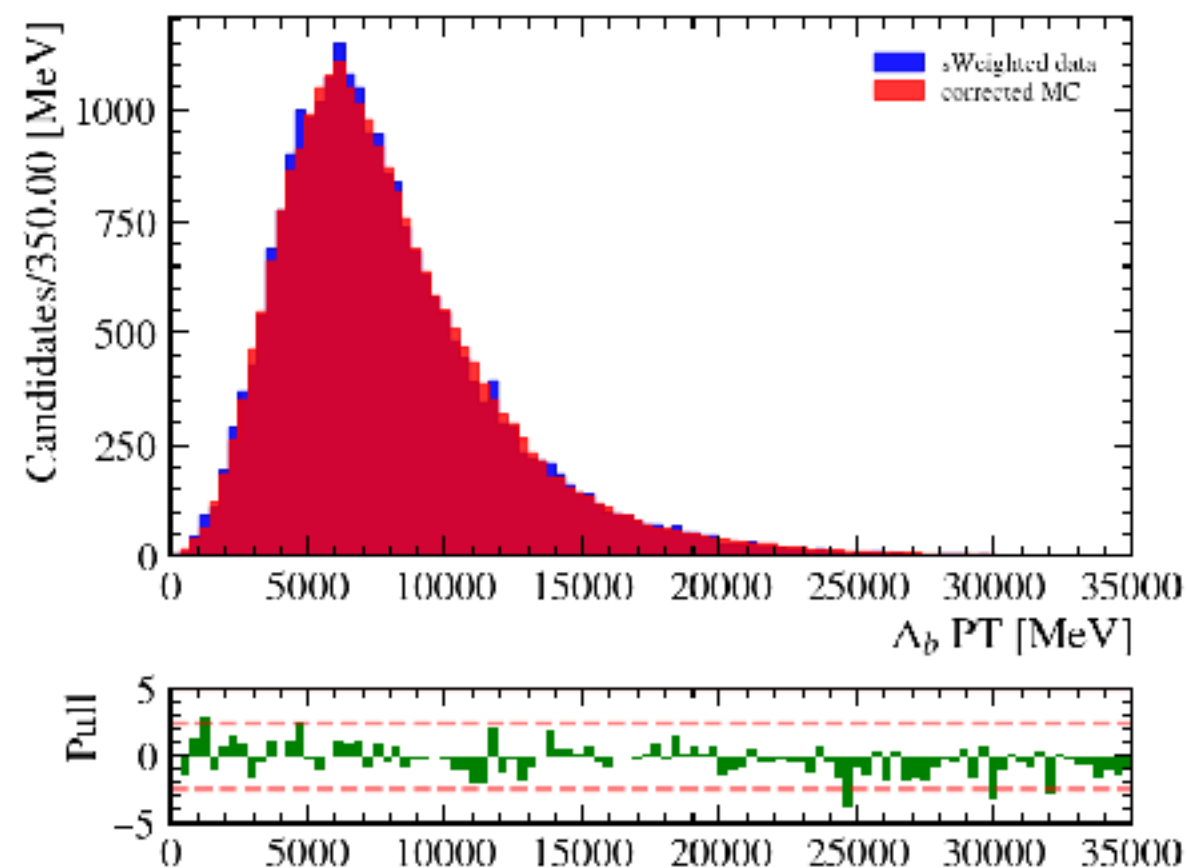
POST-CORRECTION DISTRIBUTIONS



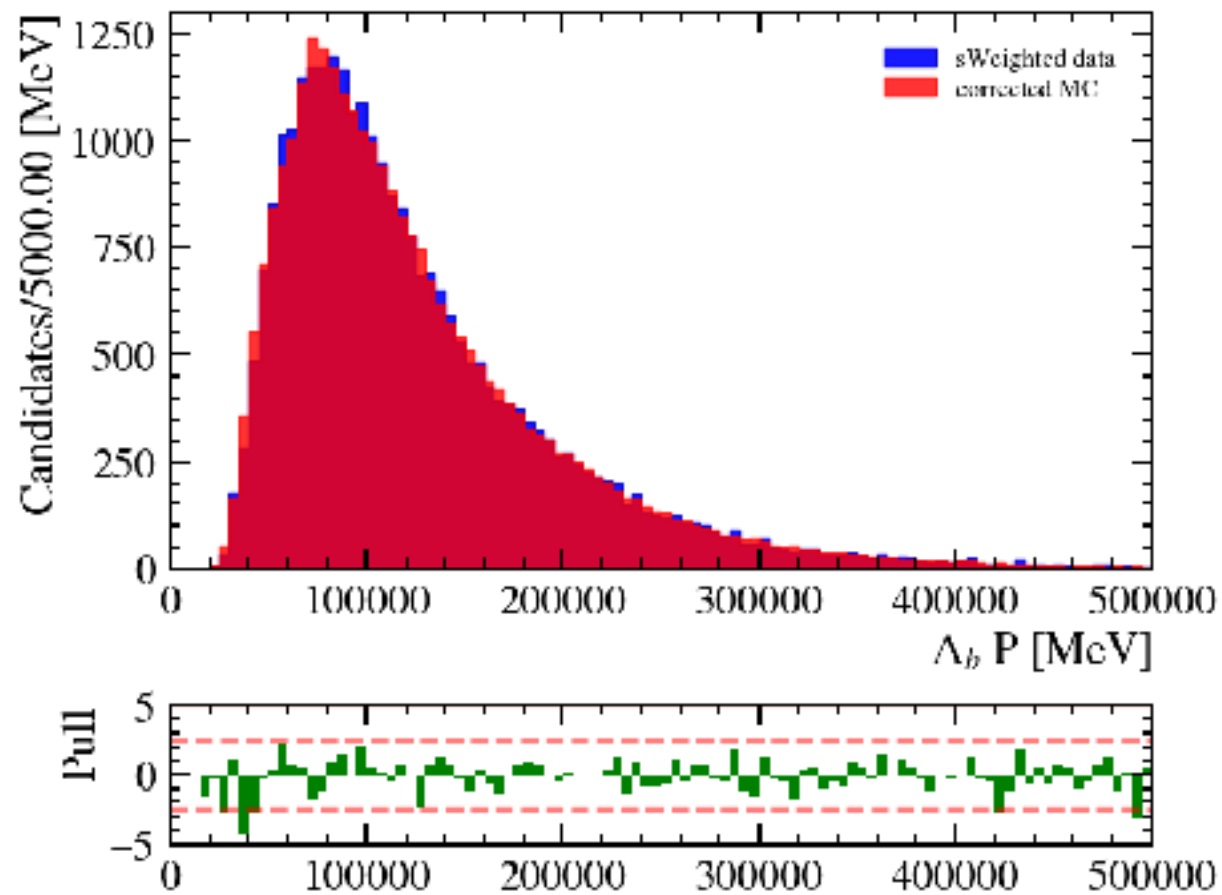
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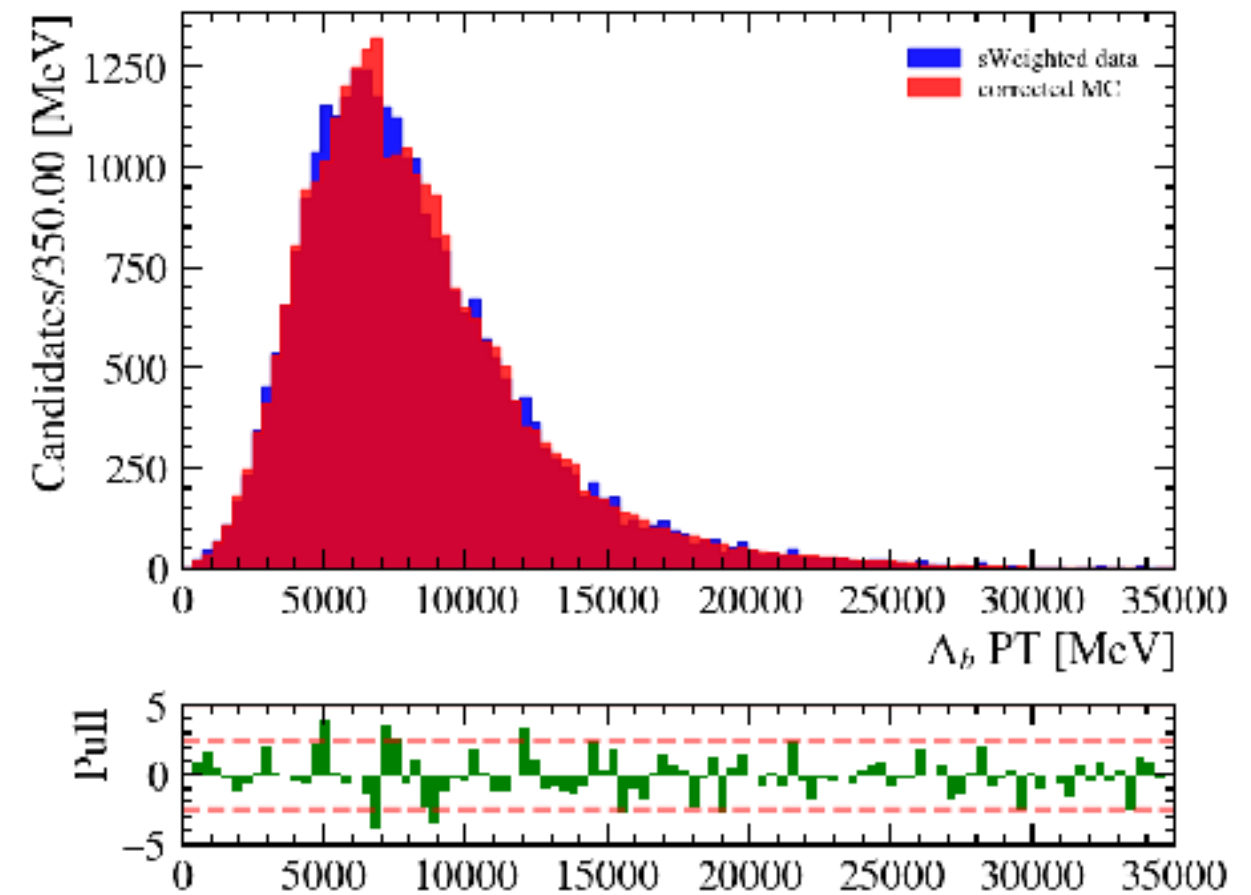
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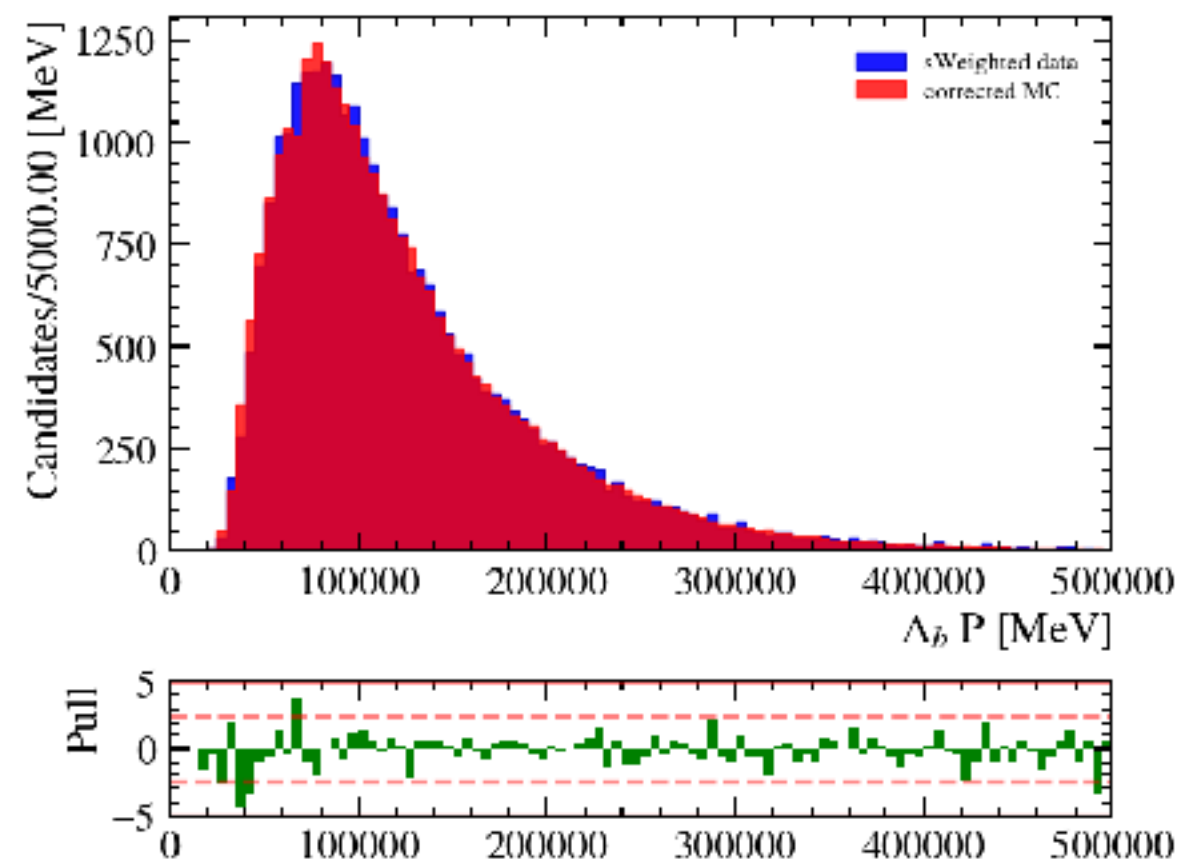
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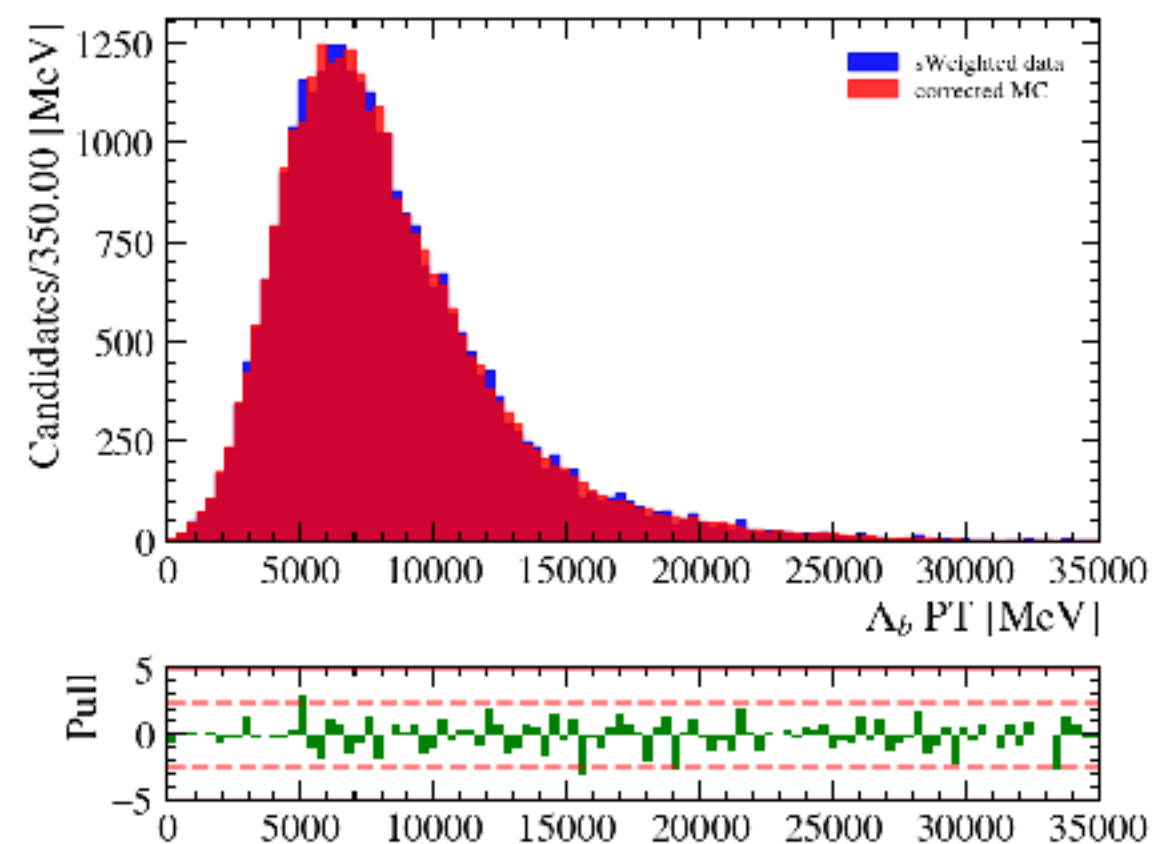
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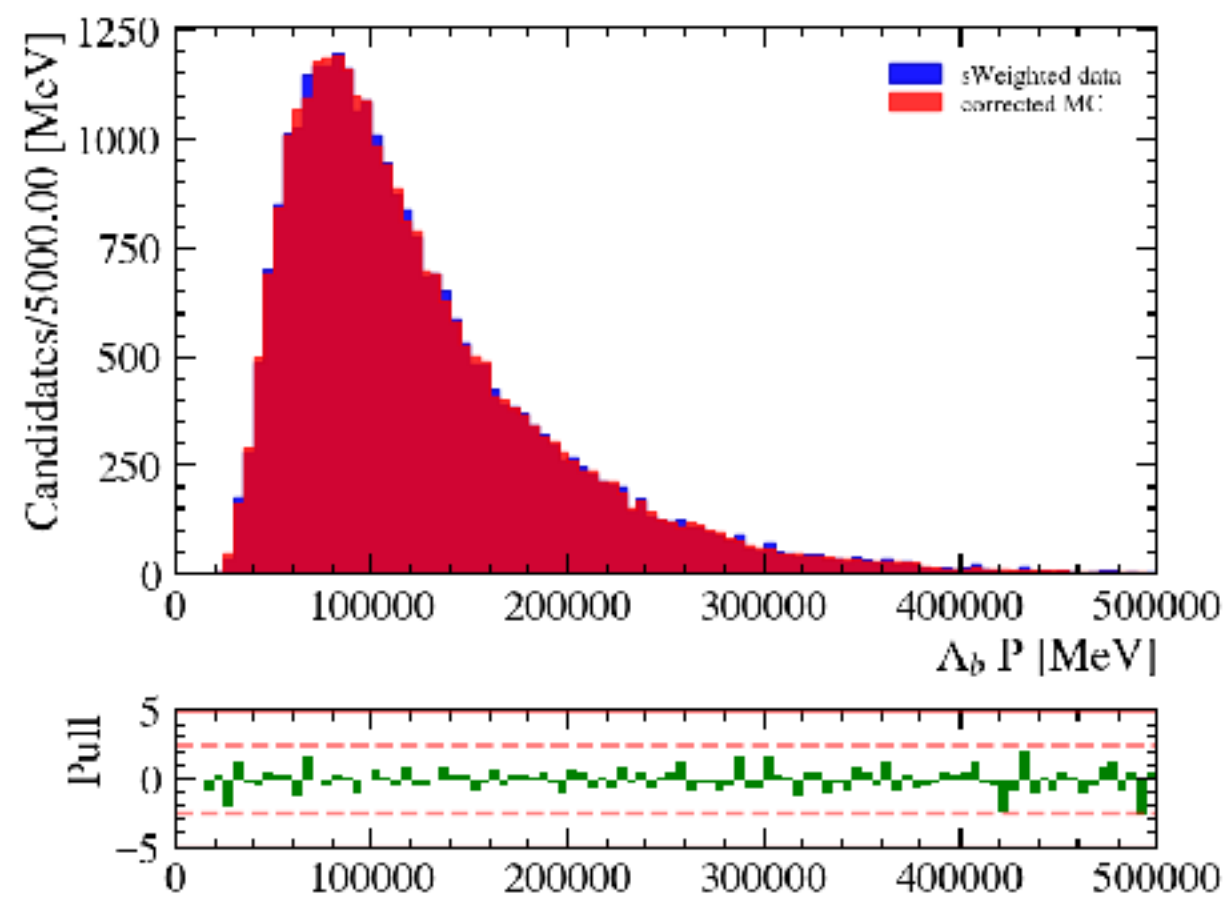
RECTANGULAR BINNING



POST-CORRECTION DISTRIBUTIONS



ADAPTIVE BINNING



POST-CORRECTION DISTRIBUTIONS

