



Update for Acceptance systematics

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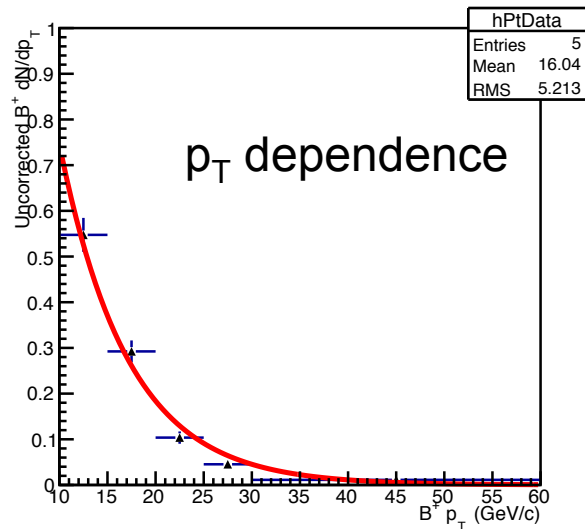
Scheme for acceptance systematics

- **Premise**
 - Difference of the p_T or y distribution from between data and MC reflects the acceptance from MC
- **Weighting to denominator and numerator respectively**
 - Weighting function is continuous (not binned)
 - Variation of acceptance with weighting can be considered as systematics for acceptance

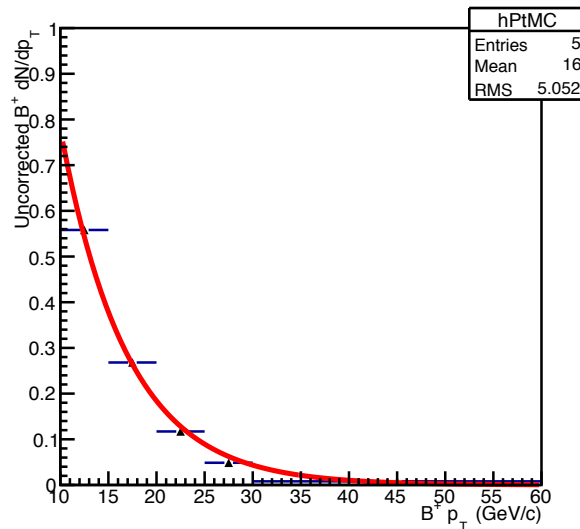
$$\alpha = \frac{N_{reconstructable, M}^{B^{signal}}(p_T, y)}{N_{-2.40 < y^{triplet} < 2.40}^{B^{signal}}(p_T, y)}$$

Data/MC ratio

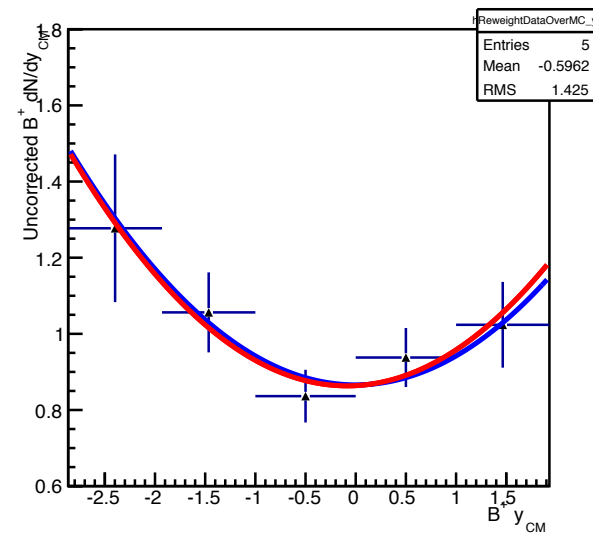
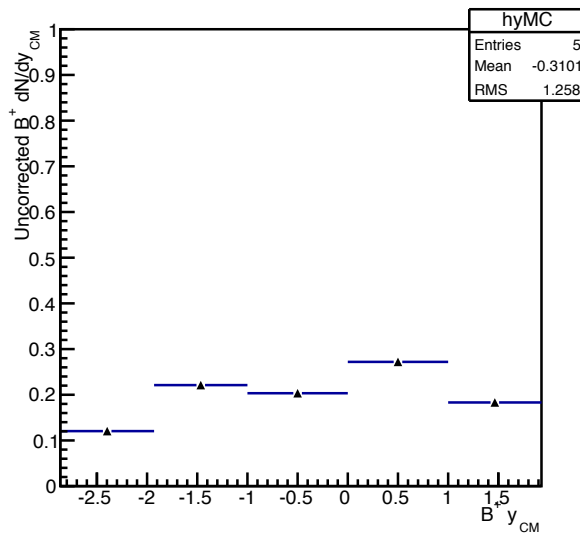
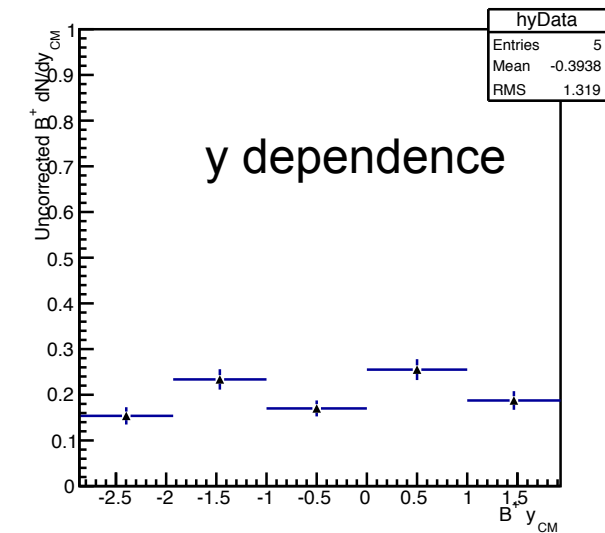
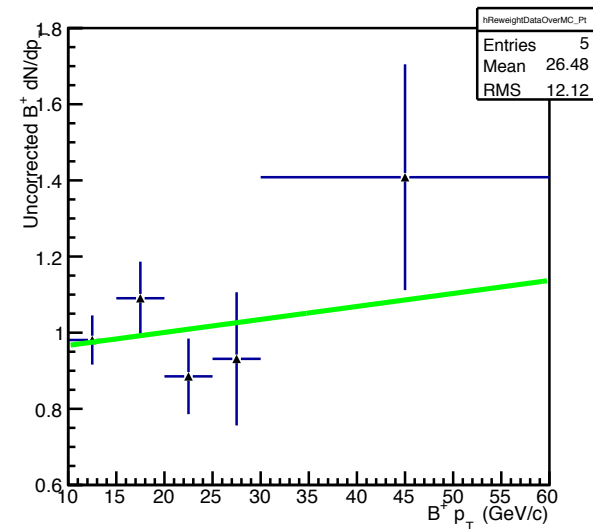
Normalized uncorrected yield from data



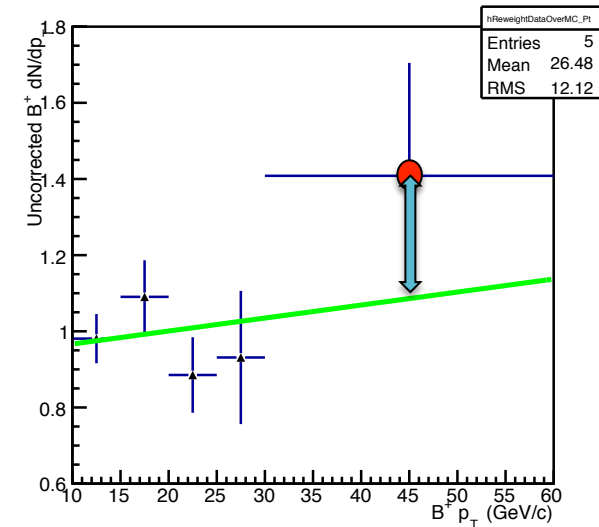
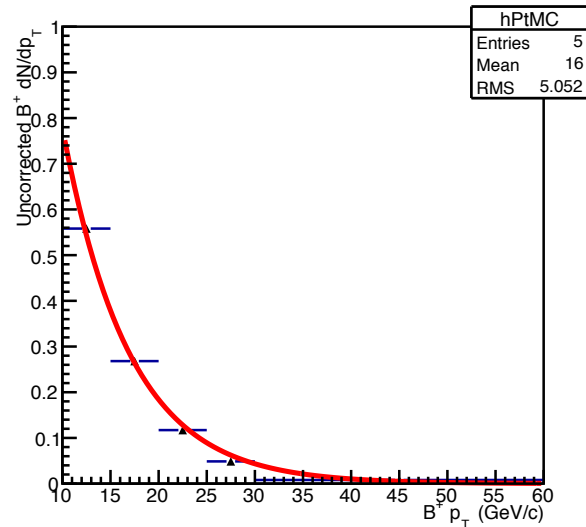
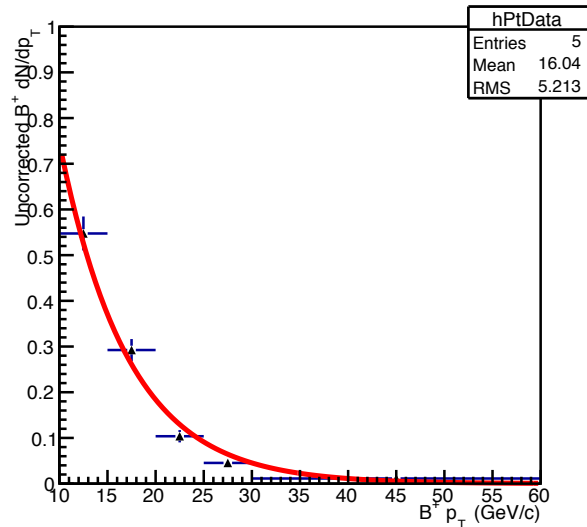
Normalized uncorrected yield from MC



Data/MC ratio of normalized uncorrected yield



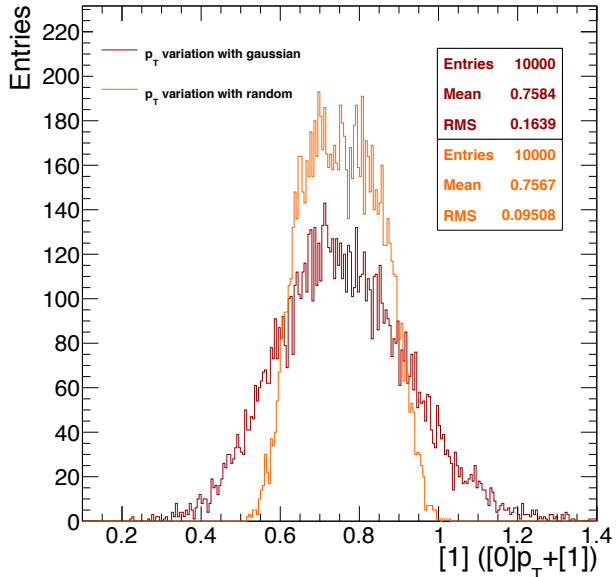
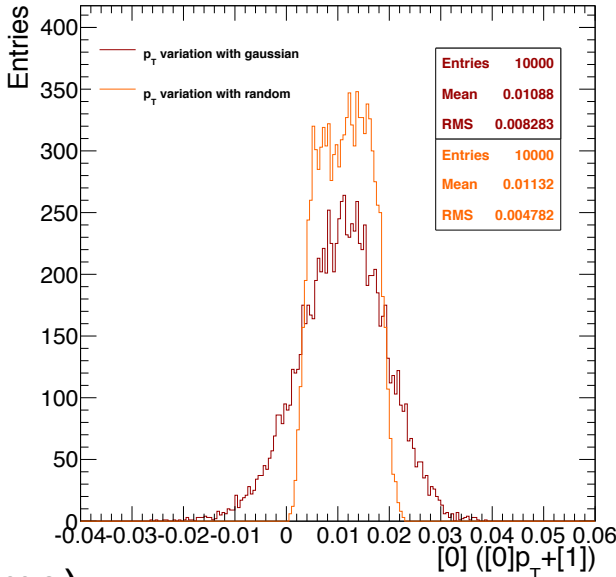
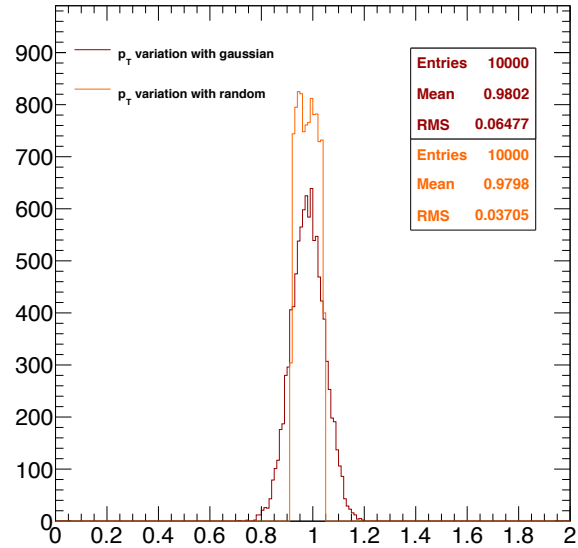
Estimate acceptance systematics



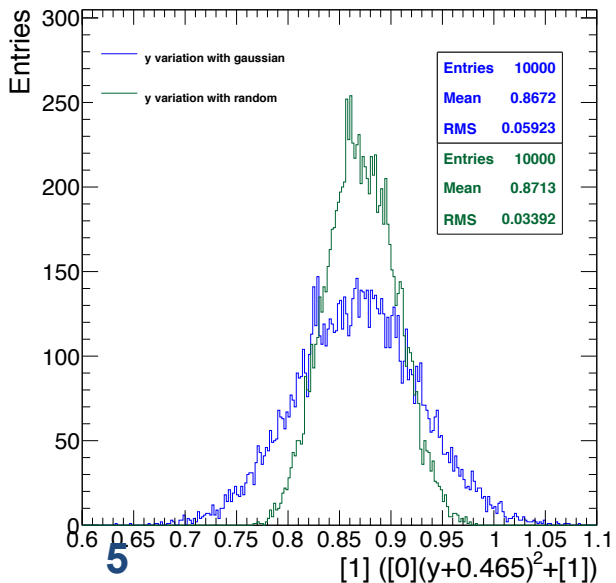
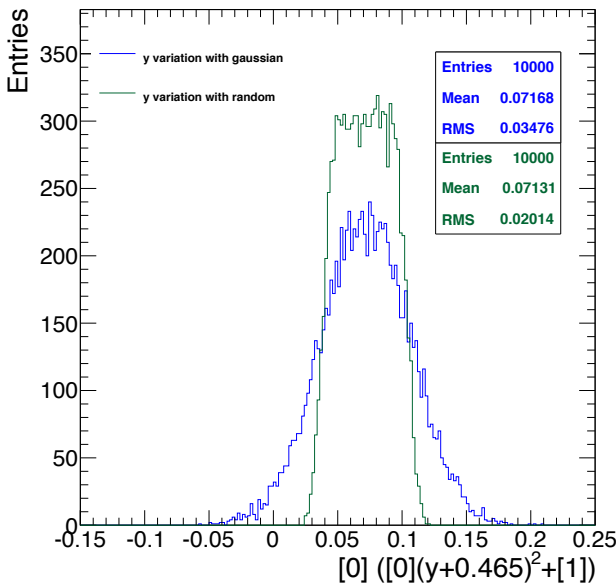
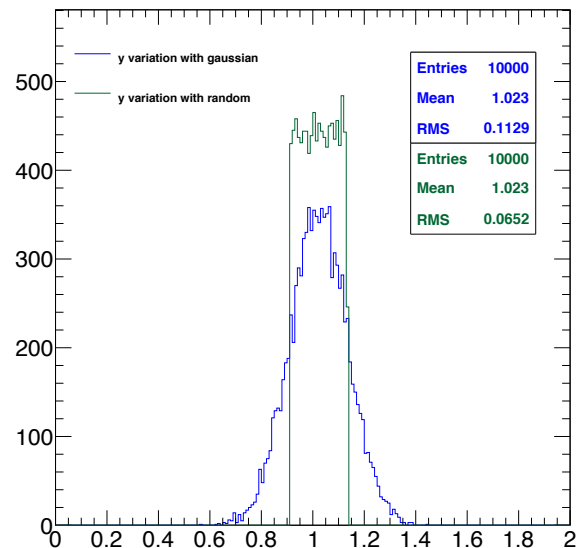
- Run random generator and vary the Data/MC ratio in each bin
 - Mean = Data/MC ratio, Sigma = error of the Data/MC
 - Run the random generator 10k times
 - Get parameters of 1st order polynomial
 - Test with flat and gaussian distribution

Ratio variation

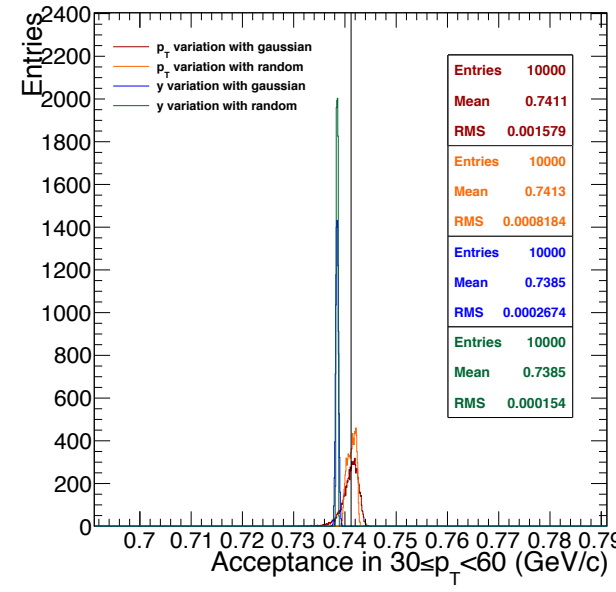
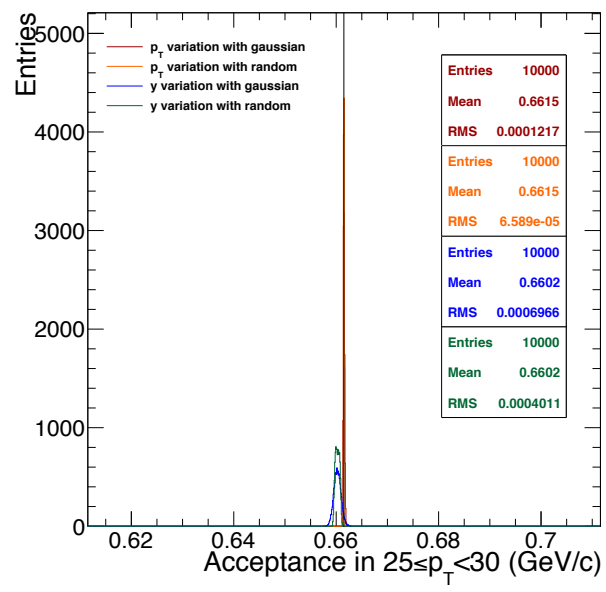
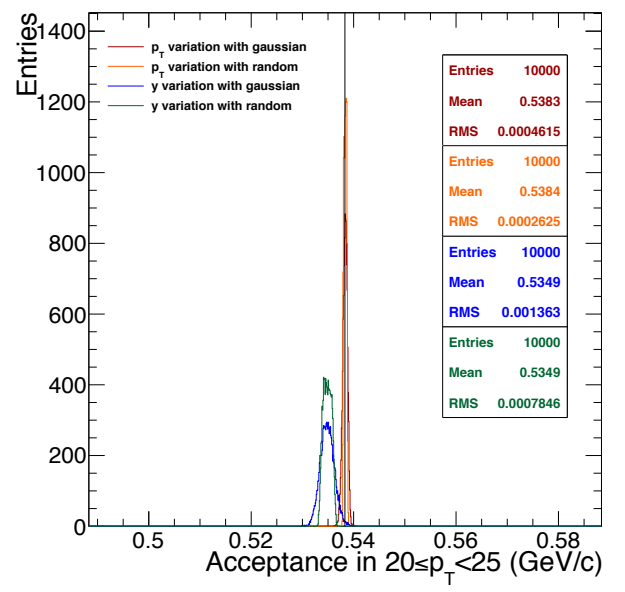
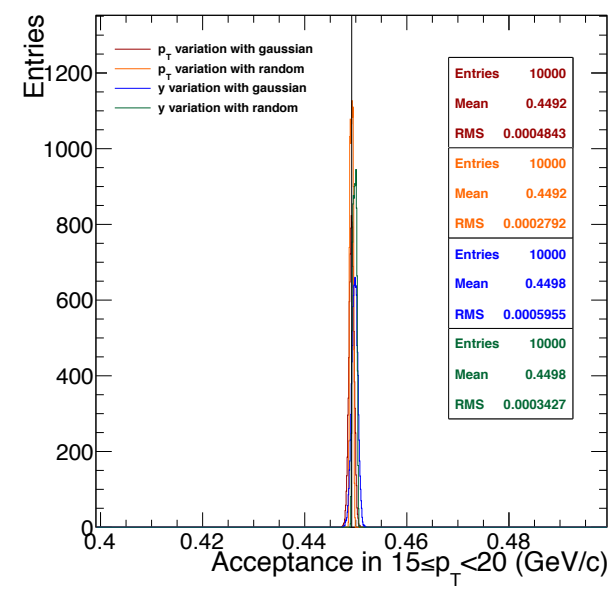
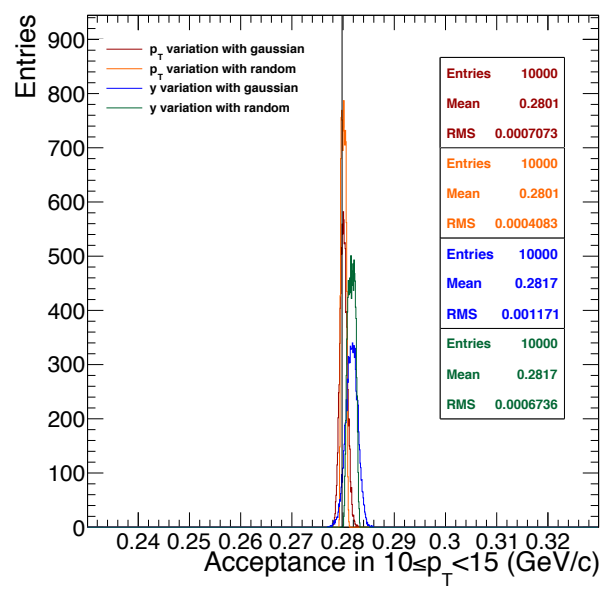
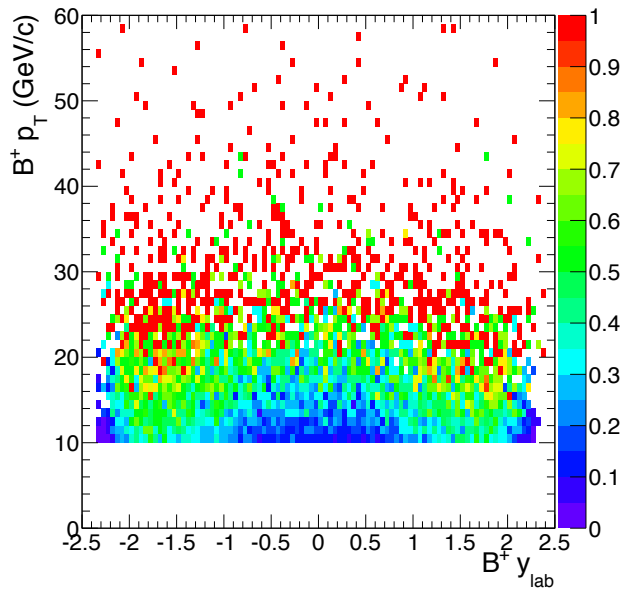
- pT dependence



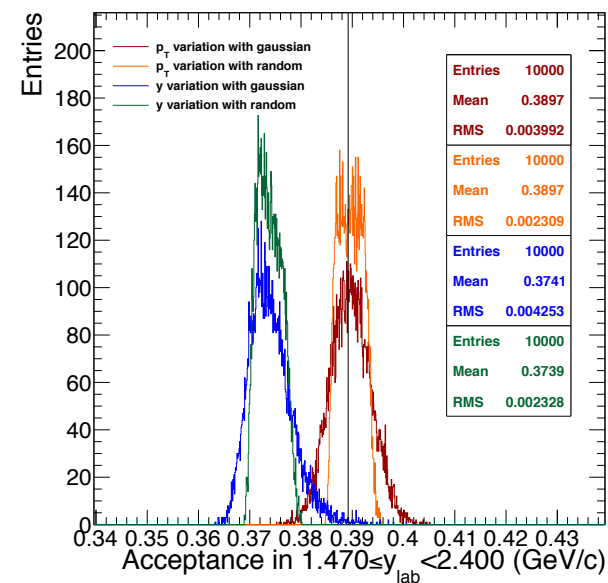
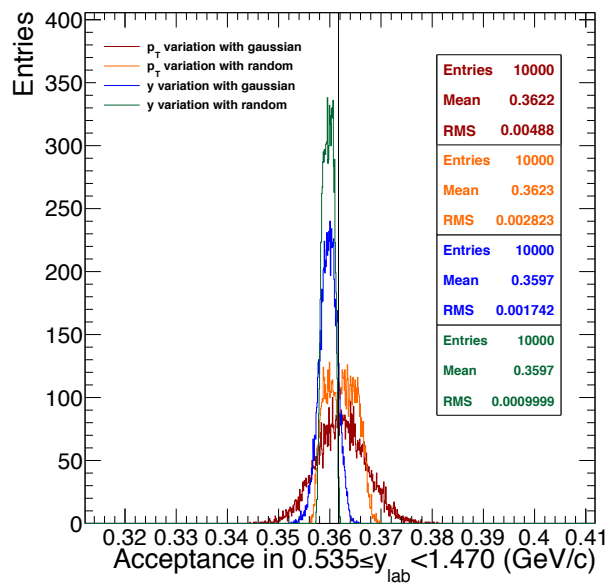
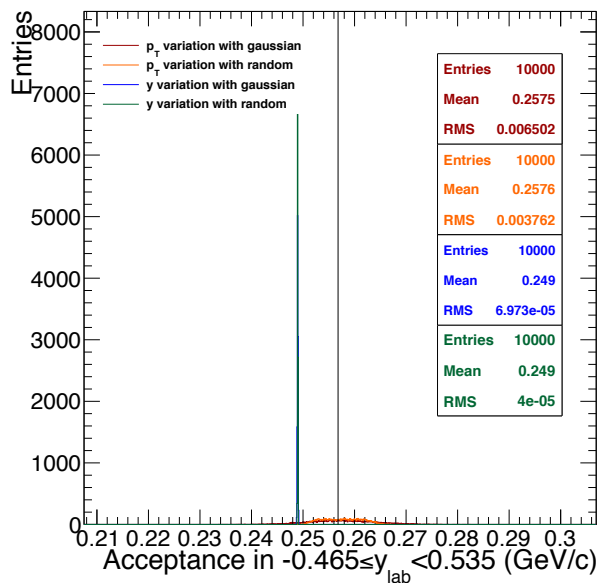
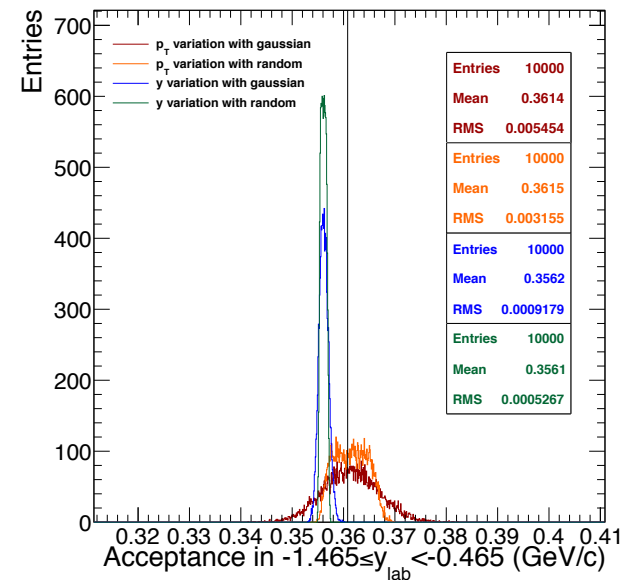
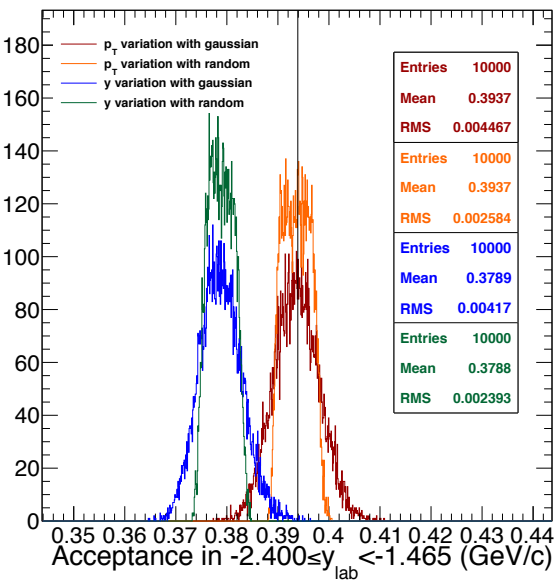
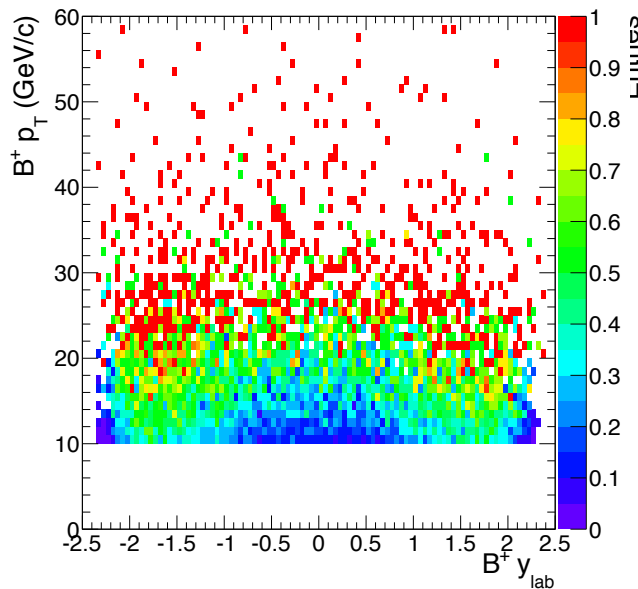
- y dependence (in lab frame)



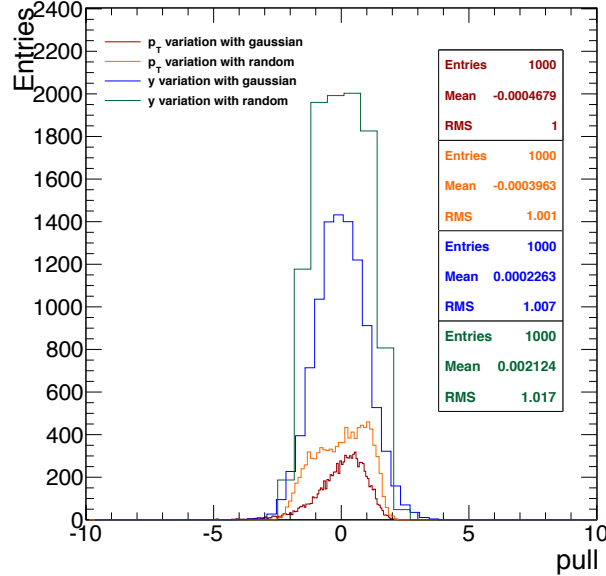
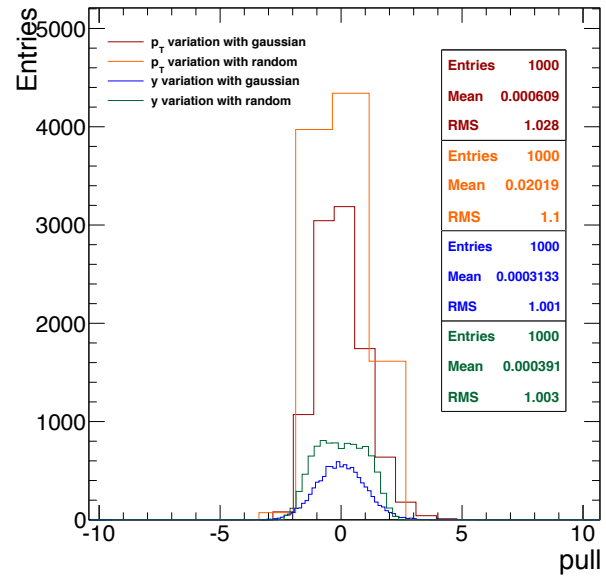
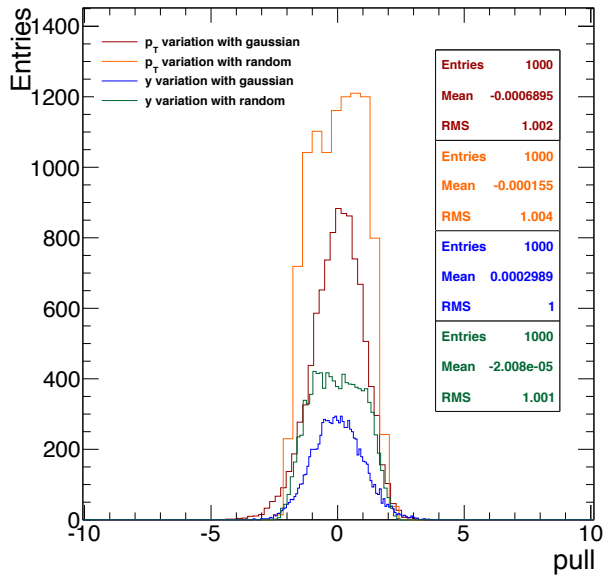
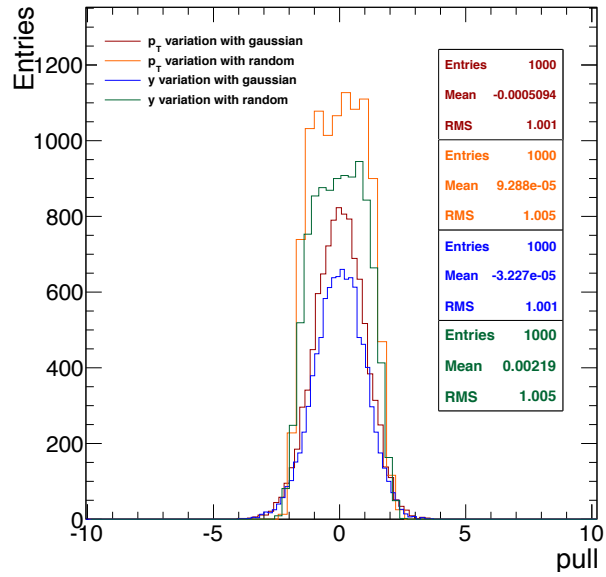
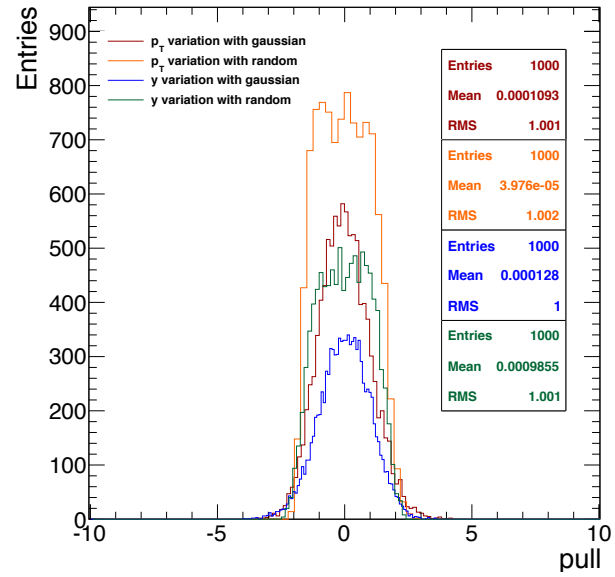
p_T dependence



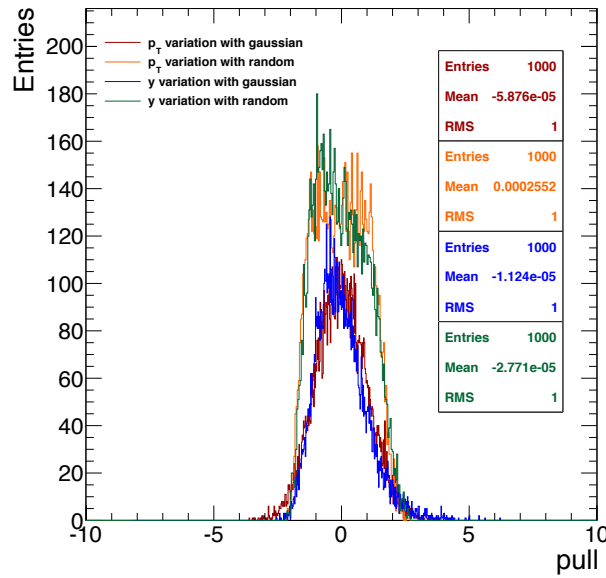
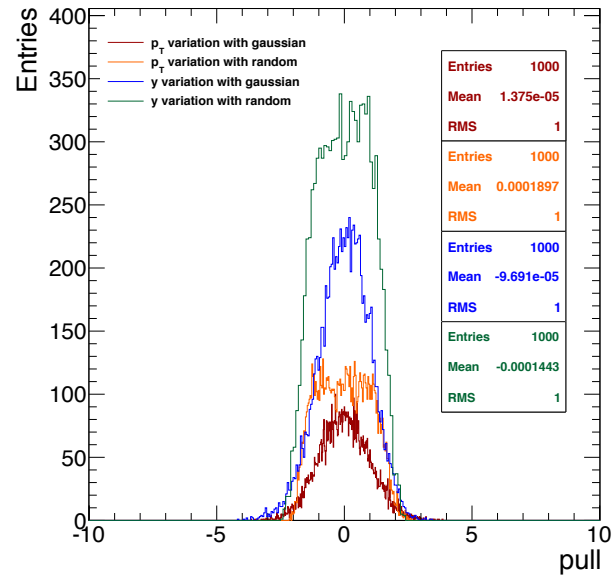
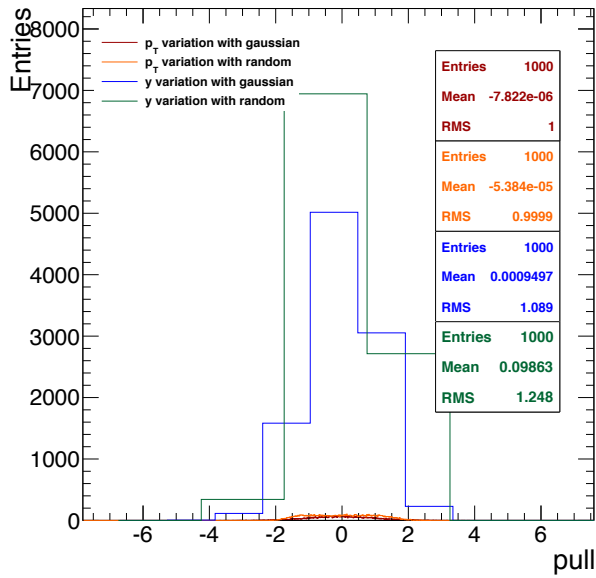
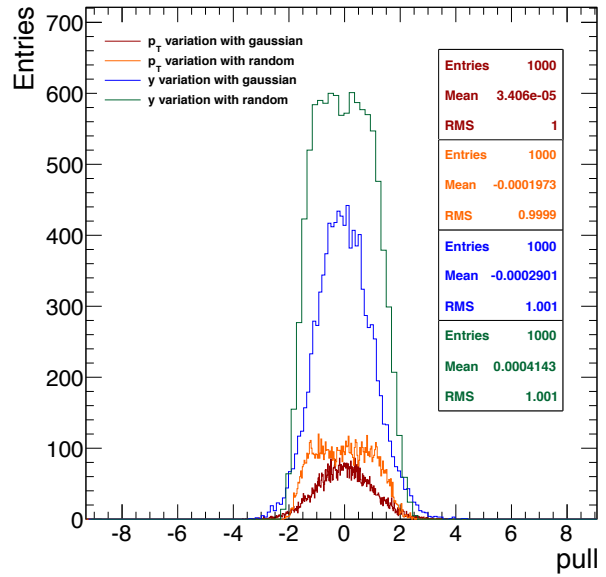
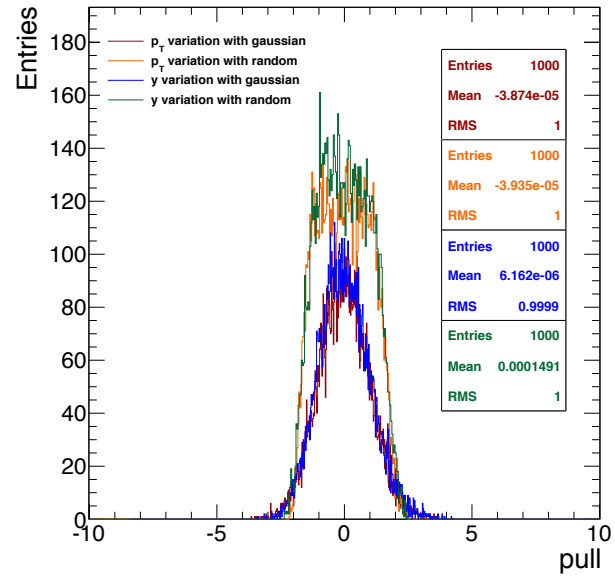
y dependence



Pull distribution, p_T dependence



Pull distribution, y dependence



Discussion

- Acceptance by 10k variation with gaussian or flat distribution
- Variation is affected by acceptance trends versus p_T or y , so varied acceptance value have complicated trends
- How we should choose the difference from that?
 - Mean value from toy study – Acceptance value without variation
 - Difference between max and min from toy study
 - Max or min from toy study – Acceptance value without variation
 - shown table in next slide

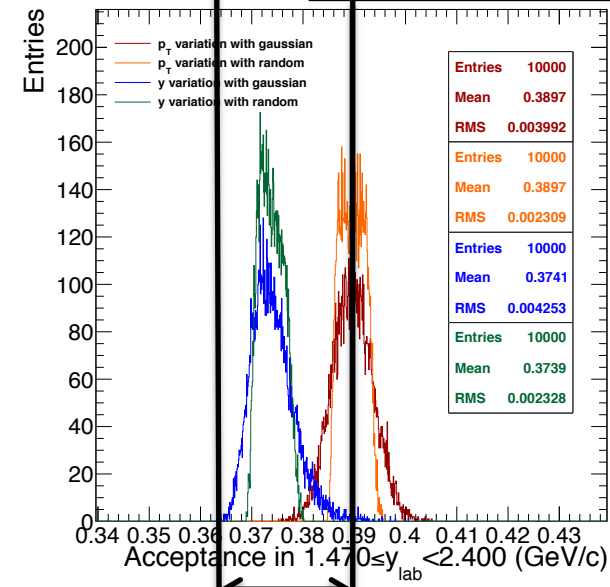
Acceptance variation

p_T bin	Acceptance	p_T variation Delta Acc.	y variation Delta Acc.
1 (10,15)	0.2798	0.33%	0.61%
2 (15,20)	0.4492	0.20%	0.28%
3 (20,25)	0.5383	0.22%	0.84%
4 (25,30)	0.6615	0.08%	0.39%
5 (30,60)	0.7412	1.34%	0.38%

y bin (lab)	Acceptance	p_T variation Delta Acc.	y variation Delta Acc.
1 (-2.4,-1.465)	0.3939	1.70%	2.93%
2 (-1.465,-0.465)	0.3608	2.15%	0.78%
3 (-0.465,0.535)	0.2568	2.56%	0.93%
4 (0.535, 1.470)	0.3617	1.94%	0.93%
5 (1.470, 2.4)	0.3892	1.60%	2.60%

Minimum acceptance
w/ y variation

Acceptance
w/o variation



- **Red** : Select maximum in two cases, use as systematics?

Backup



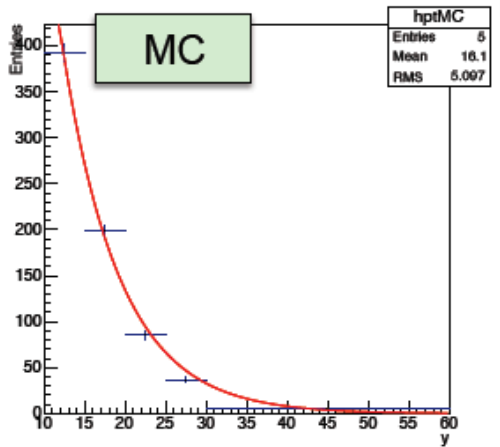
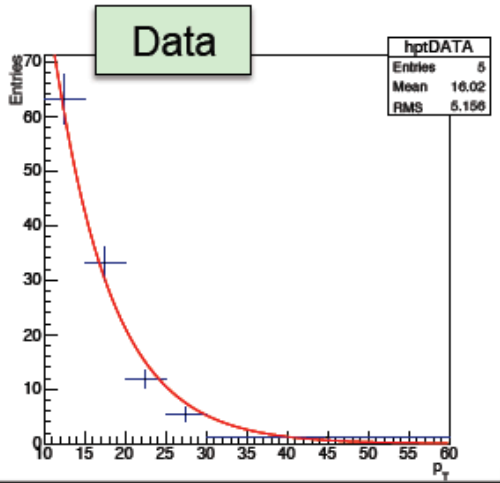
Last decision

- QM2014 methods : rule out with problems
- Fit and divide methods : proper for p_T dependence
- Divide and fit methods : rule out because of failing fit

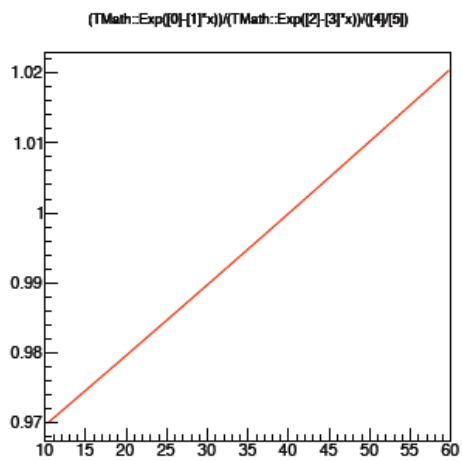
New proposal after collection of y

- QM2014 methods : rule out with problems
- Fit and divide methods : proper for p_T dependence
- **Divide and fit methods : proper for y dependence – find the candidate fitting function**

Method 2 : Fit and divide methods

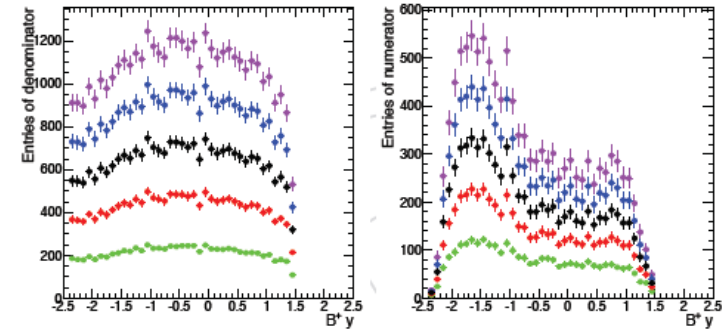


$$\frac{e^{p^0 - xp^1} |_{data}}{Total\ B\ cand. |_{data}} \div \frac{e^{p^0 - xp^1} |_{MC}}{Total\ B\ cand. |_{MC}}$$

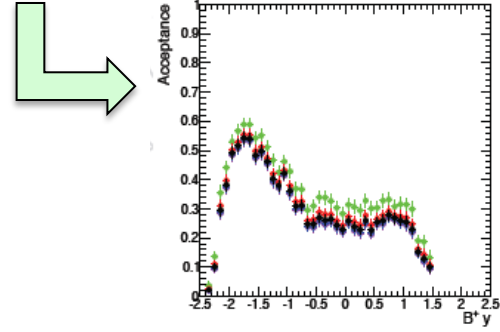


With the raw yields from MC and data, fit by exponential function

Divide fitting functions from MC and data, then get the weighting function W
Consider the errors from each parameters

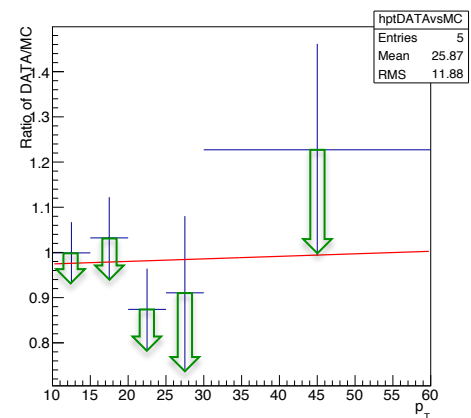
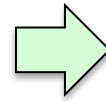
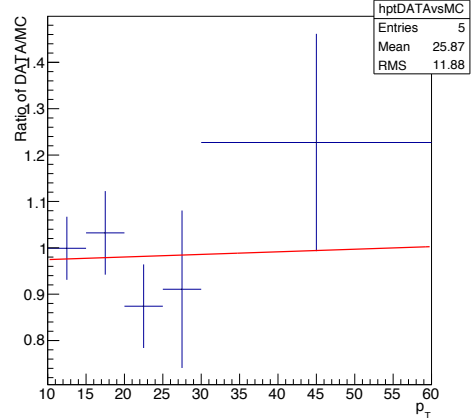
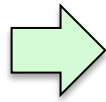
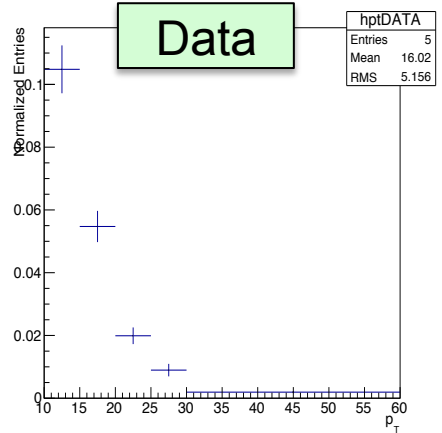


Weight the denominator and numerator : Fill(y,W)



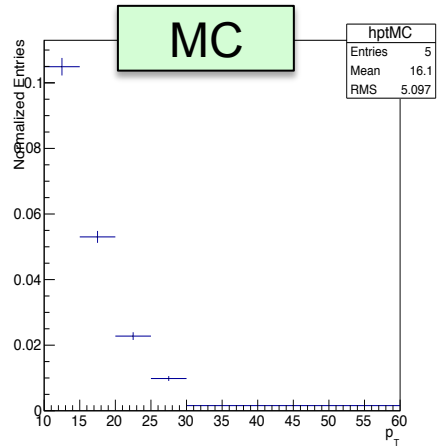
Calculate the acceptance from dividing the numerator by the denominator, and redo with the varying the parameters consideration with errors, then compare the difference and set the systematics

Method 3 : Divide and fit methods



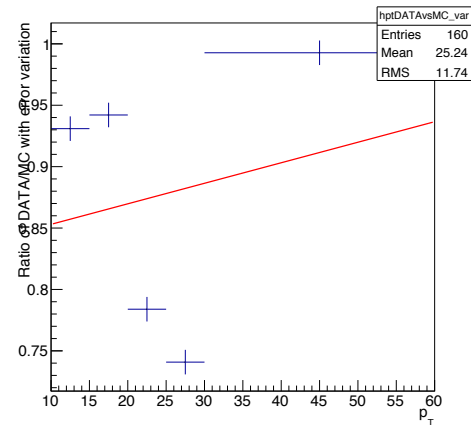
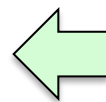
Divide the normalized raw yields then fit points by 1st order polynomial function (like method 1), that is weighting function W

With up or down variation of error in each points, get the fitting function again (ex. down, down, down, down)



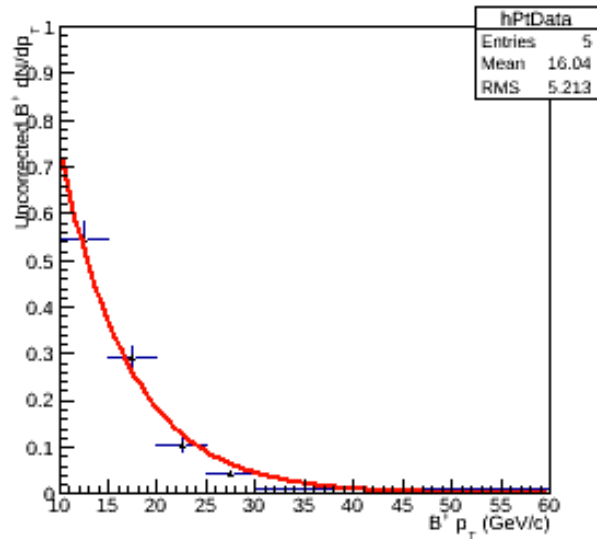
Get the normalized raw yields from MC and data

Get the acceptance with various weighting function and check the difference between maximum and minimum, set as the systematics

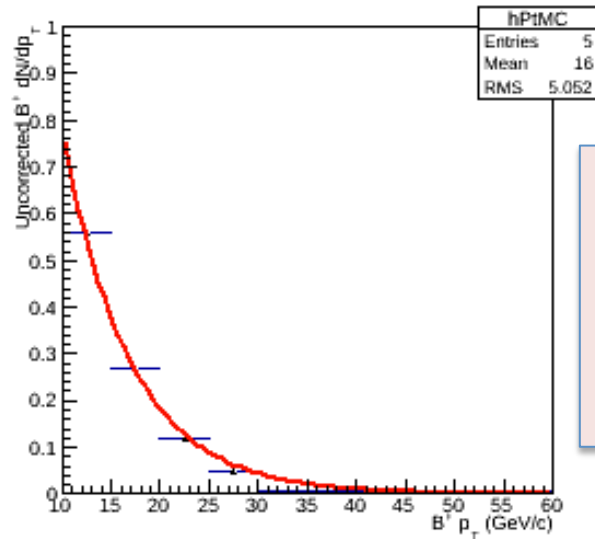


p_T dependence

Reco. B+ in Data

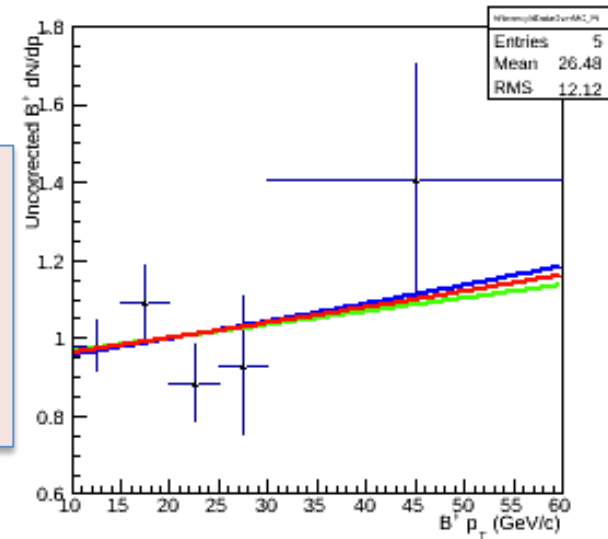


Reco. B+ in MC



exp/exp without fitting
 exp/exp with fitting
 1st order polynomial with fitting

Data/MC

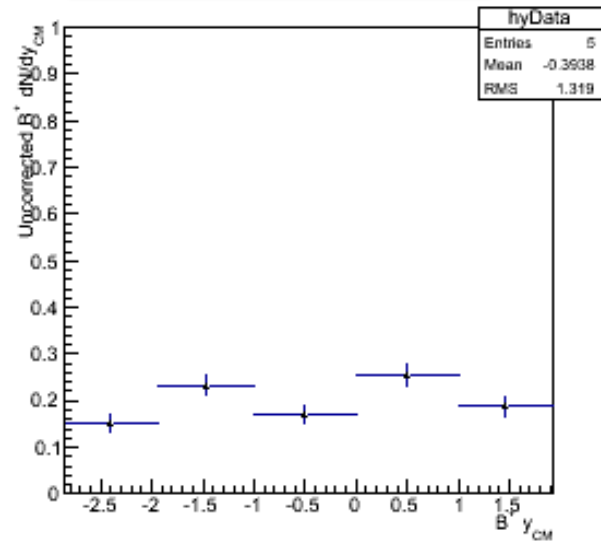


Divide with “B” option

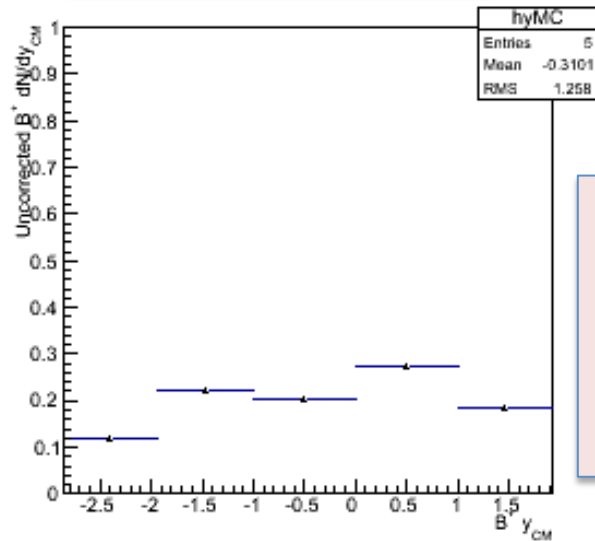
- Apply method 2 (fit and divide)

y dependence

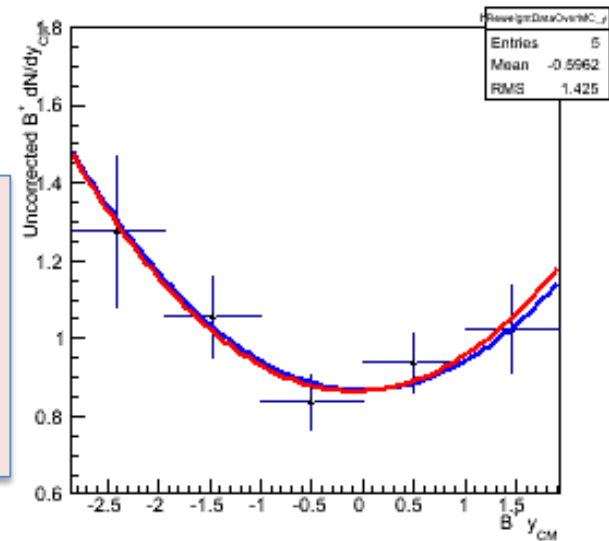
Reco. B+ in Data



Reco. B+ in MC



Data/MC



Divide with “B” option

2nd order polynomial with fixed y shift
(minimum is at y=0)
2nd order polynomial without fixed y shift
(minimum y is controlled by fitting)

- Apply method 3 (divide and fit)
- Find proper fitting function, revive method 3

Rapidity conversion in between lab and CM frame

- **General**

- Proton going direction have plus rapidity in CM frame
- Merge bins with same rapidity in CM frame (same color in tables)

- **1st run**

- proton going to minus eta $y_{CM} = -y_{lab} - 0.465$

yLAB	-2.4	-1.465	-0.465	+0.535	+1.470	+2.4
yCM	1.935	1.0	0.0	-1.0	-1.935	-2.865

← proton going direction

- **2nd run**

- proton going to plus eta $y_{CM} = y_{lab} - 0.465$

yLAB	-2.4	-1.470	-0.535	+0.465	+1.465	+2.4
yCM	-2.865	-1.935	-1.0	0.0	1.0	1.935

→ proton going direction