

HEP NTUA Weekly Report

30/9/2020

G. Bakas



Summary

- Analysis:
 - Switched to Tune CP5 nominal MC's for 2016
 - Start investigating ttbar Systematic Uncertainties
- Investigate on how to combine the measurements between the three years
 - Combine them in the fiducial level and extract the cross section?
 - Extract the cross sections individually and combine the measurements in the unfolded level?
- Top Angular Distributions: χ , $|\cos\theta^*|$ leading and subleading
 - Responses, Signal Extraction → Unfolding
 - Results!
- AN 2020/156

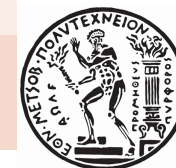


Signal Selection

Variables	Selected Cut
pT (both leading jets)	> 400 GeV
Njets	> 1
N leptons	= 0
eta (both leading jets)	< 2.4
mJJ	> 1000 GeV
jetMassSoftDrop (only for fit)	(50,300) GeV
Top Tagger	> 0.2, 0, 0.1
B tagging (2 btagged jets)	> Medium WP
Signal Trigger	

Control Region Selection

Variables	Selected Cut
pT (both leading jets)	> 400 GeV
Njets	> 1
N leptons	= 0
eta (both leading jets)	< 2.4
mJJ	> 1000 GeV
jetMassSoftDrop (only for fit)	(50,300) GeV
Top Tagger	> 0.2, 0, 0.1
B tagging (0 btagged jets)	< Medium WP
Control Trigger	

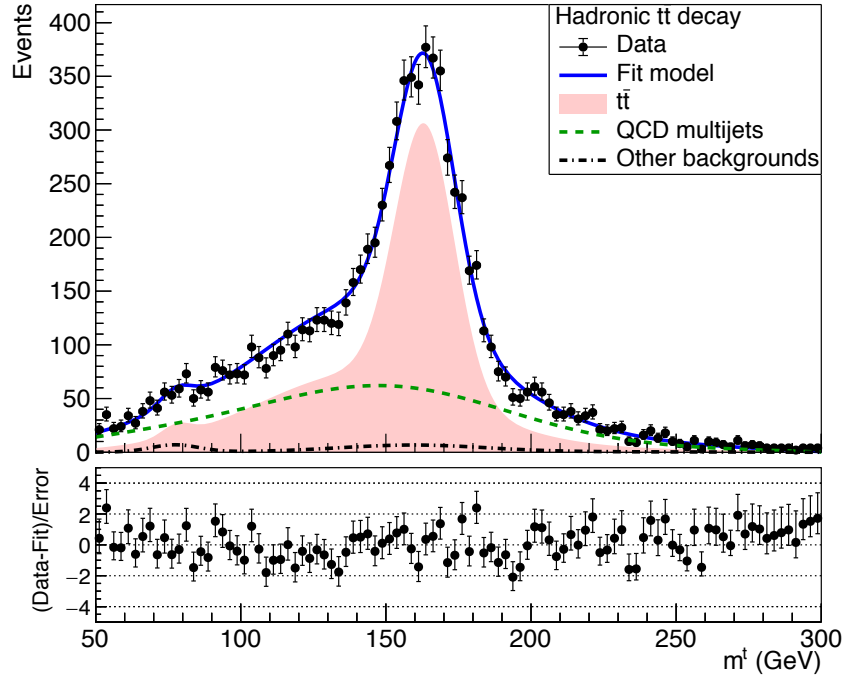


Mass Fit in Extended SR (SR_A)

- Both SR and Control Region use the Medium btag WP.
- Intuition is to remove the $t\bar{t}$ and subdominant bkg contribution from the data Control Region

$$QCD_0(m^t) = D_0(m^t) - T_0(m^t) - Sub_0(m^t)$$

A RooPlot of "mTop"

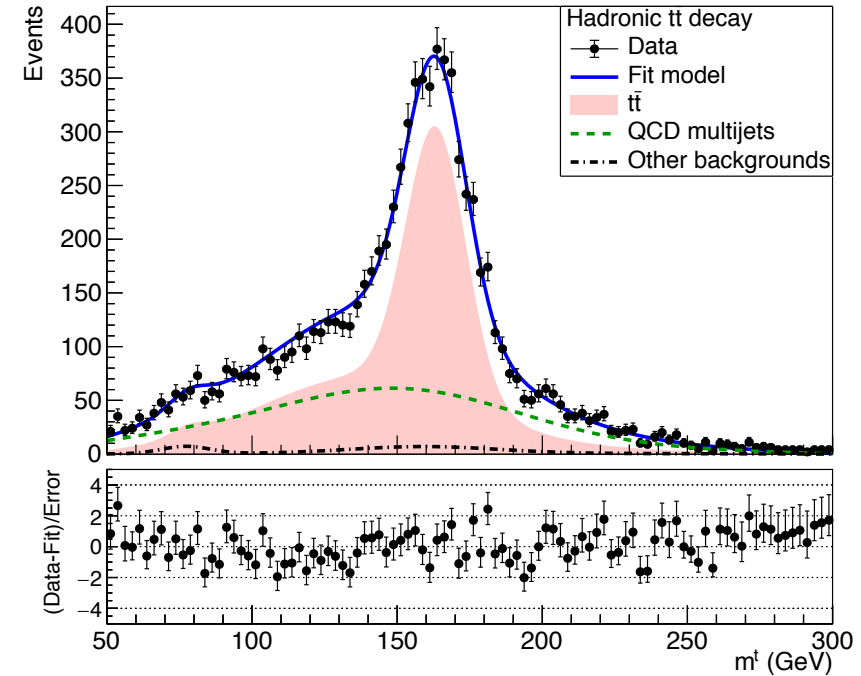


(2016)
Tune CUE

Floating Parameter	FinalValue +/-	Error
kMassResol	9.2251e-01 +/-	2.73e-02
kMassScale	9.9891e-01 +/-	2.01e-03
kQCD_2b	6.9753e-02 +/-	5.26e-02
nFitBkg_2b	2.4472e+02 +/-	1.47e+02
nFitQCD_2b	2.9890e+03 +/-	1.74e+02
nFitSig2b	5.2763e+03 +/-	1.67e+02

$t\bar{t}$ Signal strength: $r = 0.686668 \pm 0.0263103$

A RooPlot of "mTop"



(2016)
Tune CP5

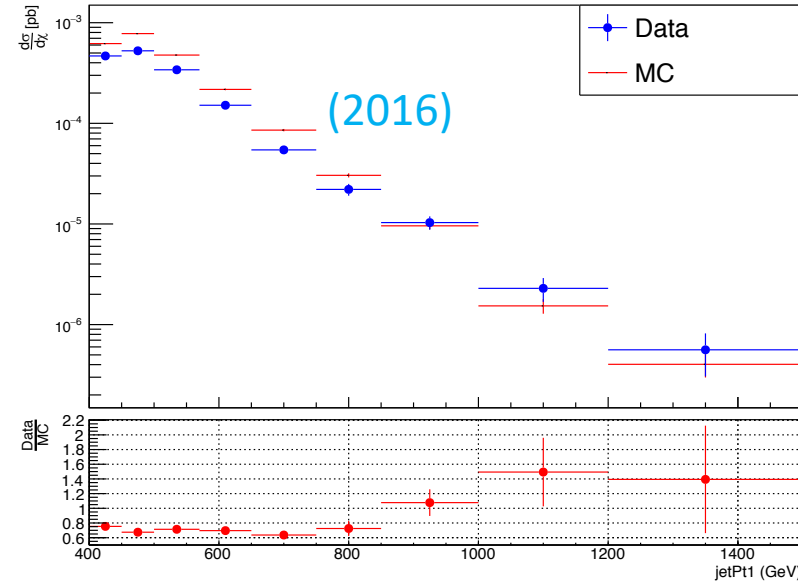
Floating Parameter	FinalValue +/-	Error
kMassResol	9.3798e-01 +/-	2.75e-02
kMassScale	9.9751e-01 +/-	1.99e-03
kQCD_2b	2.3668e-01 +/-	5.05e-01
nFitBkg_2b	2.5758e+02 +/-	1.47e+02
nFitQCD_2b	2.9242e+03 +/-	1.76e+02
nFitSig2b	5.3282e+03 +/-	1.67e+02

$t\bar{t}$ Signal strength: $r = 0.641241 \pm 0.0238714$ (new)

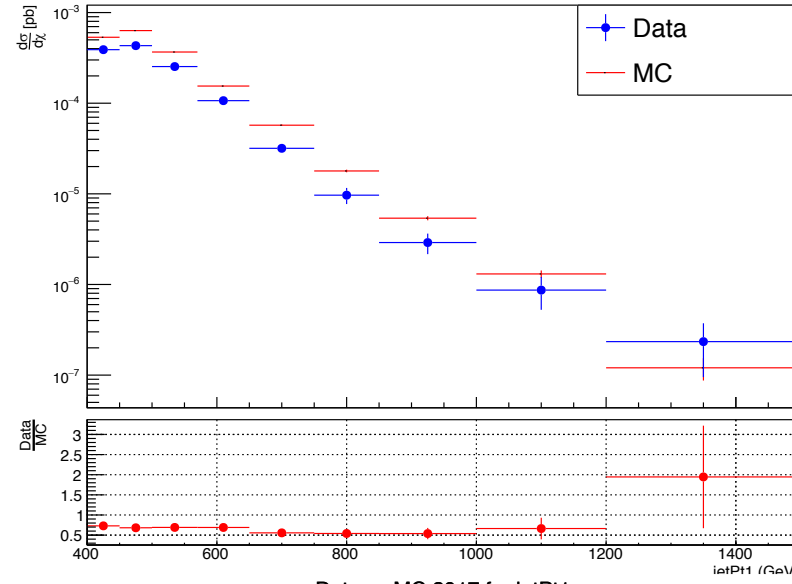


Fiducial Differential Cross Section

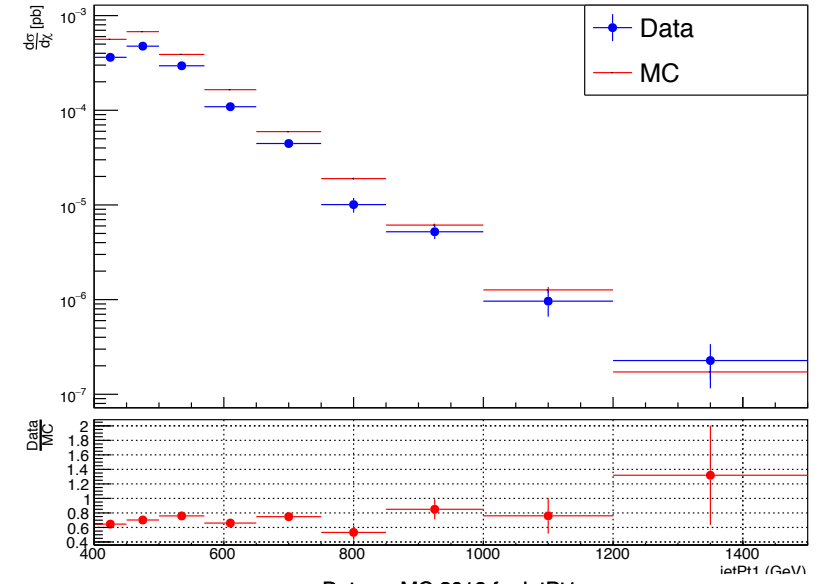
Data vs MC 2016 for jetPt1



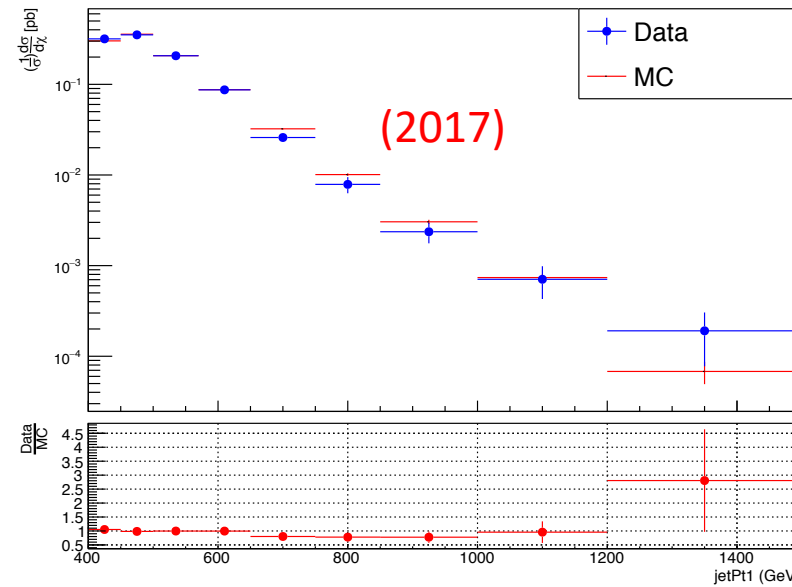
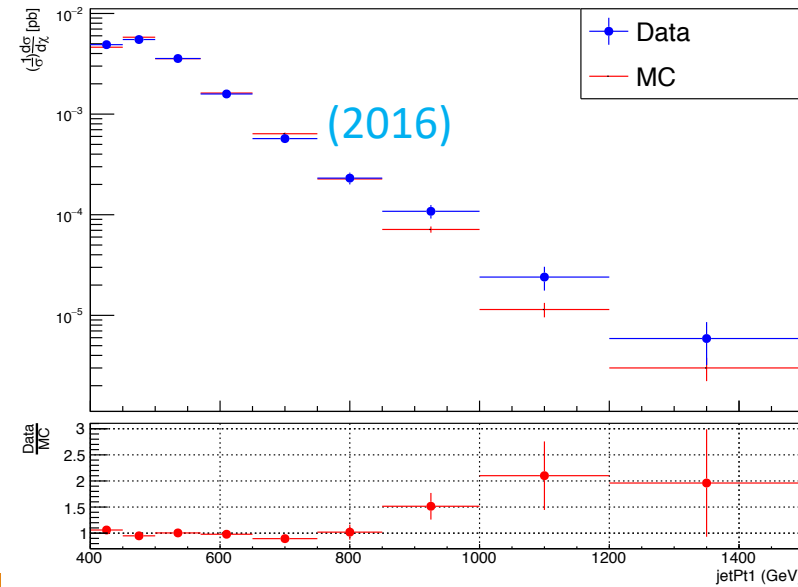
Data vs MC 2017 for jetPt1



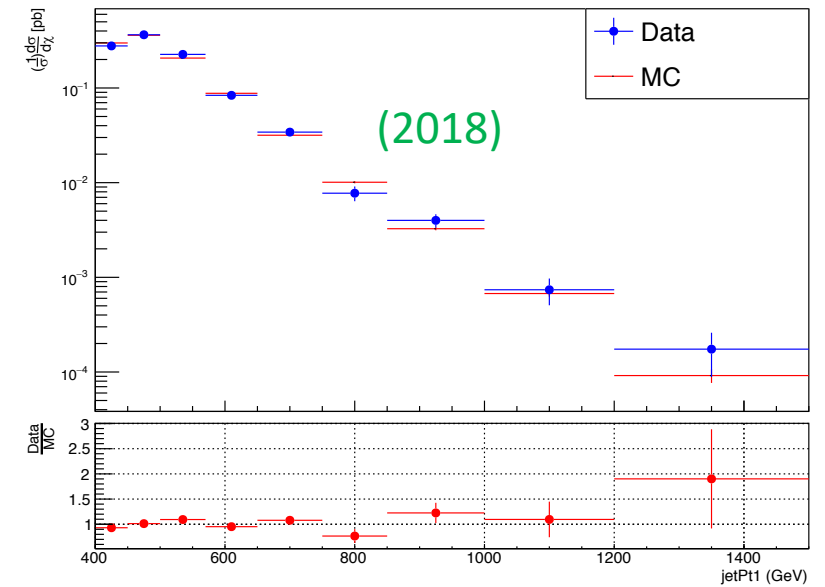
Data vs MC 2018 for jetPt1



Data vs MC 2017 for jetPt1

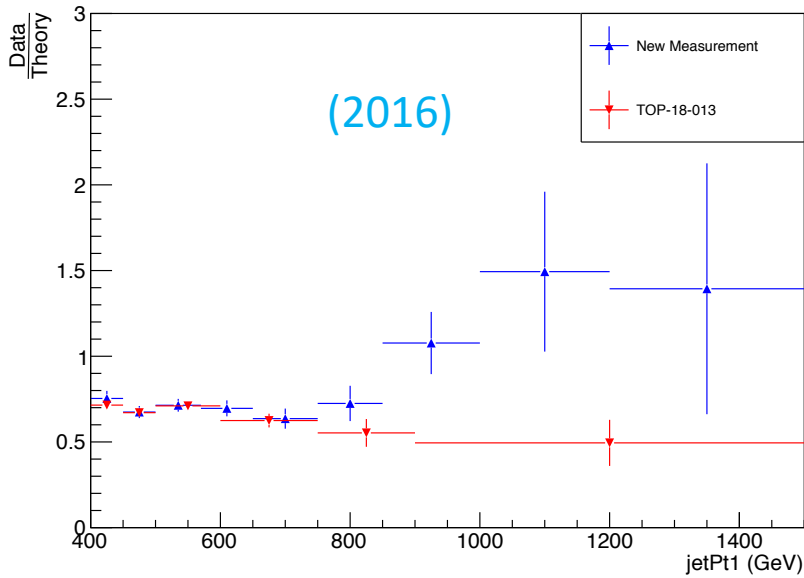


Data vs MC 2018 for jetPt1

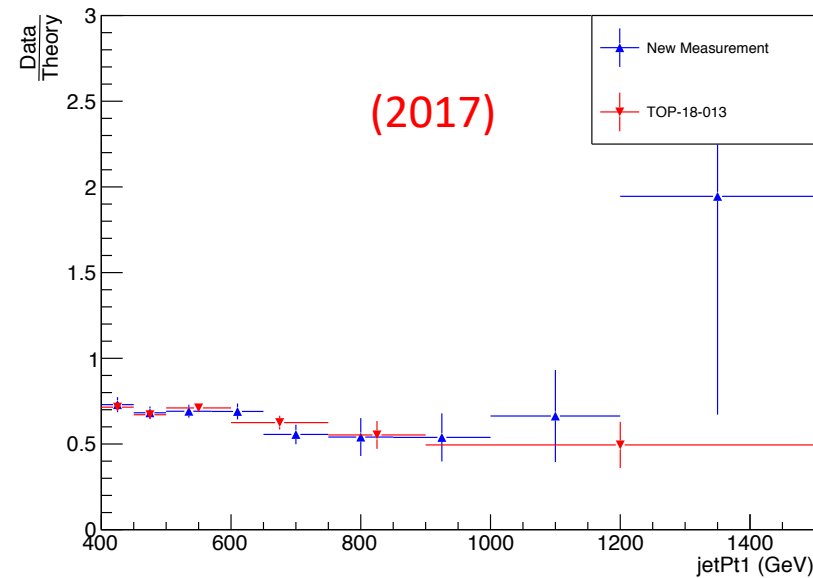


Fiducial Differential Cross Section Comparison

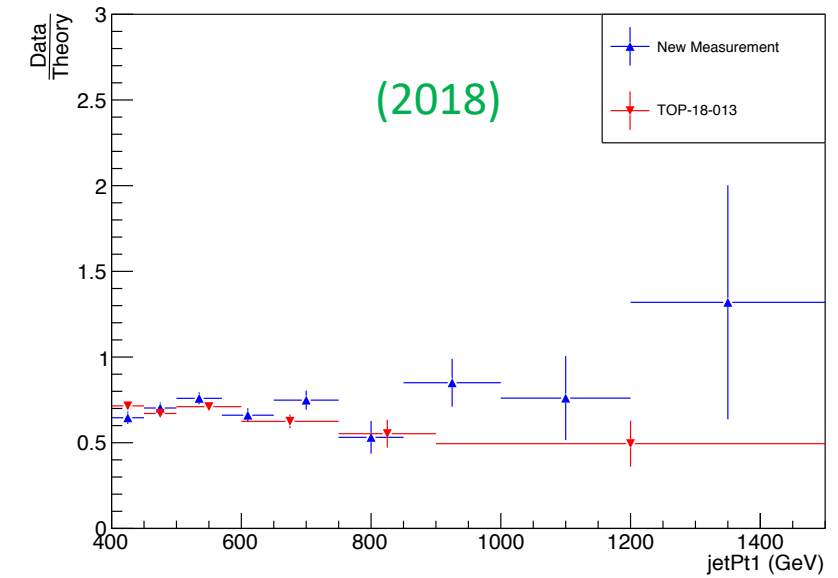
Fiducial DataOverMC ratio (2016, TOP18013)



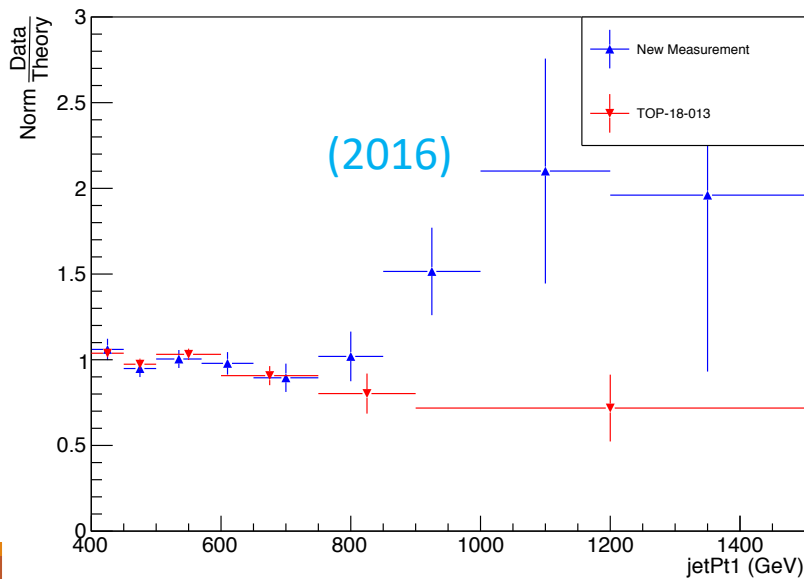
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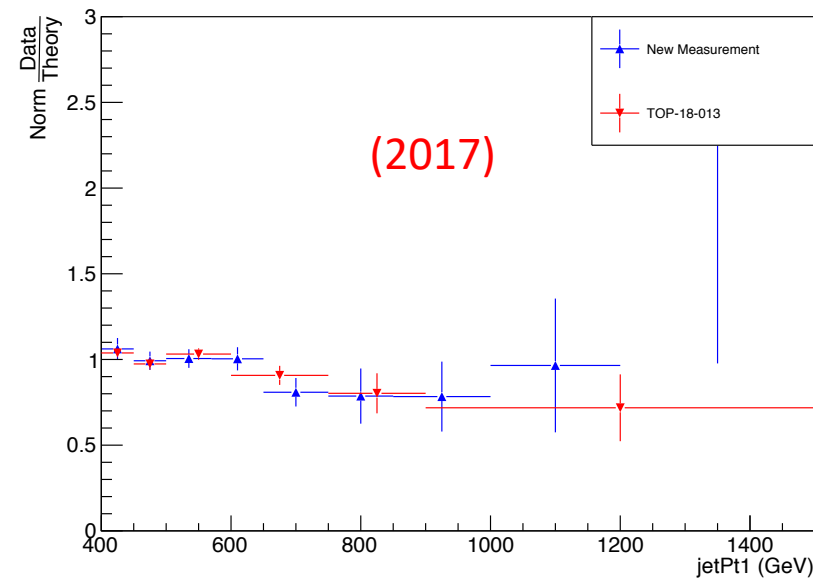
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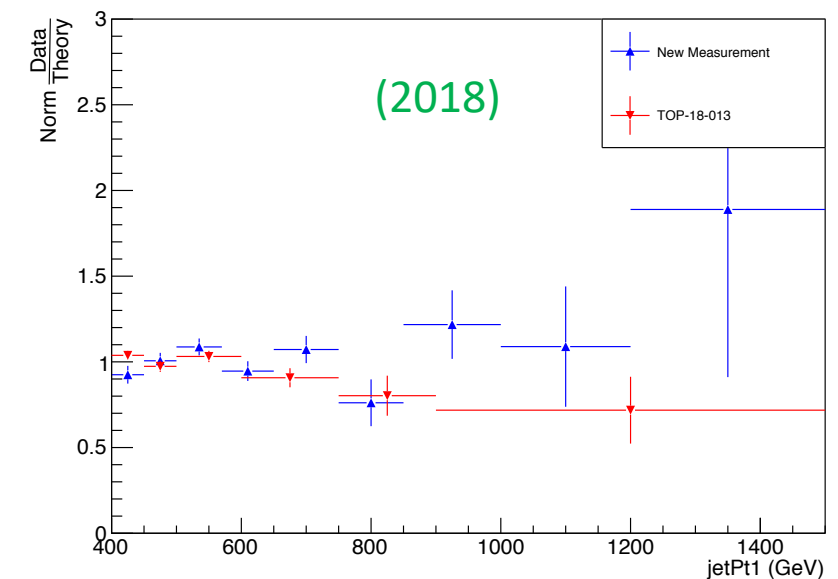
Normalised Fiducial DataOverMC ratio (2016, TOP18013)



Normalised Fiducial DataOverMC ratio (2017, TOP18013)

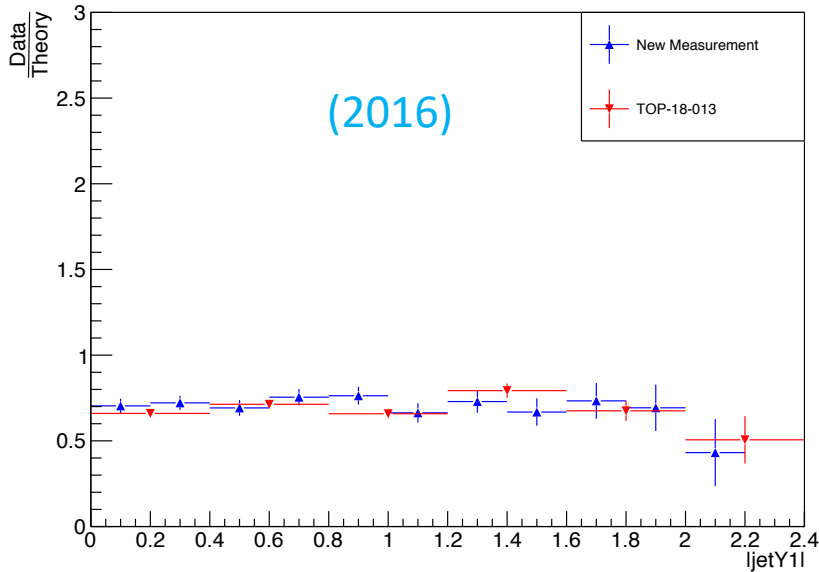


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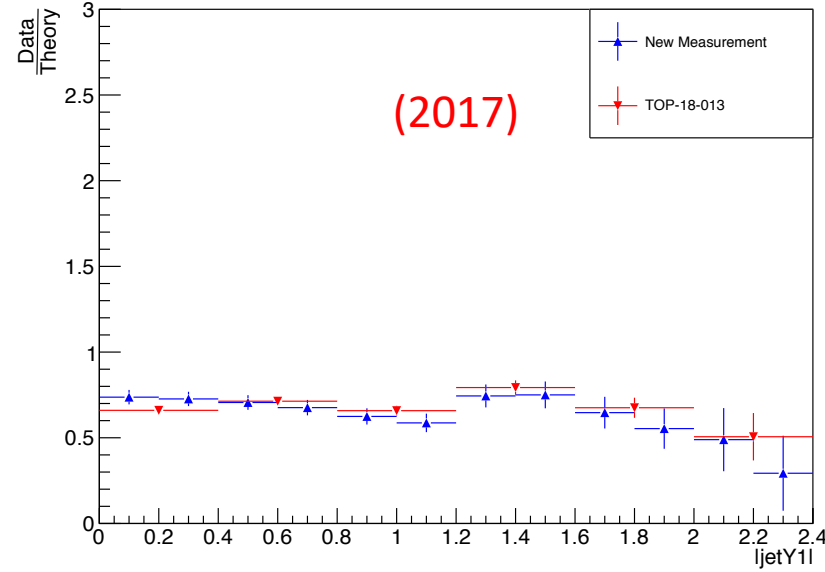


Fiducial Differential Cross Section Comparison

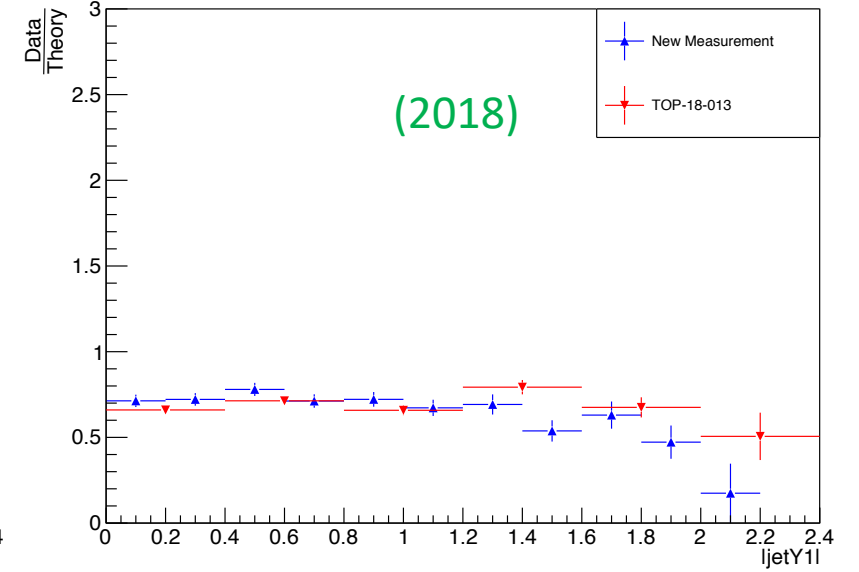
Fiducial DataOverMC ratio (2016, TOP18013)



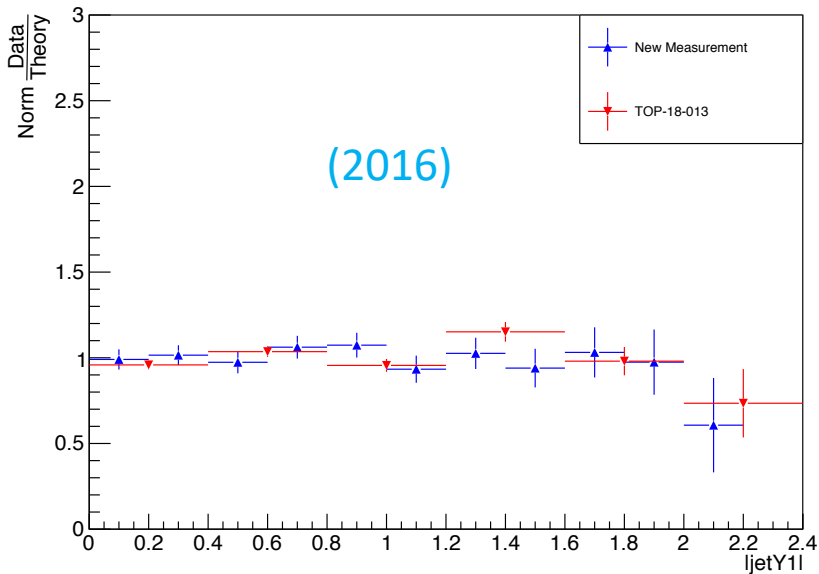
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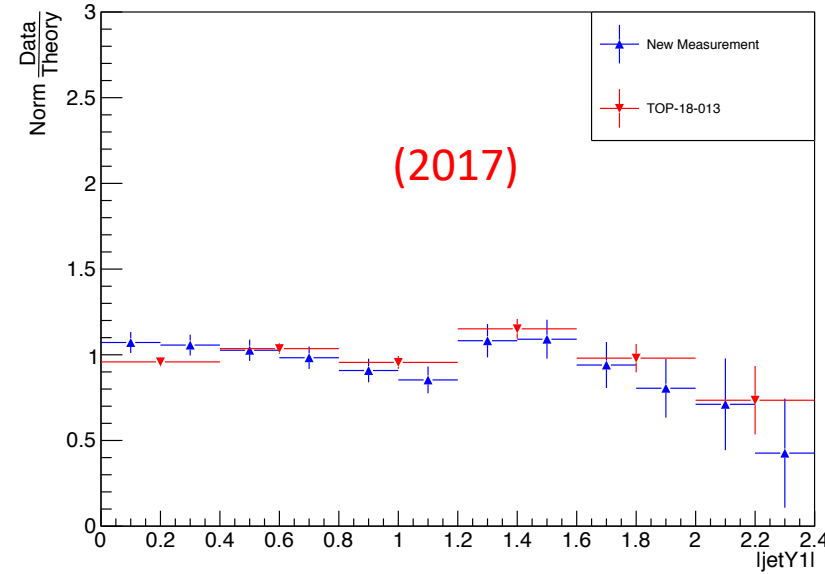
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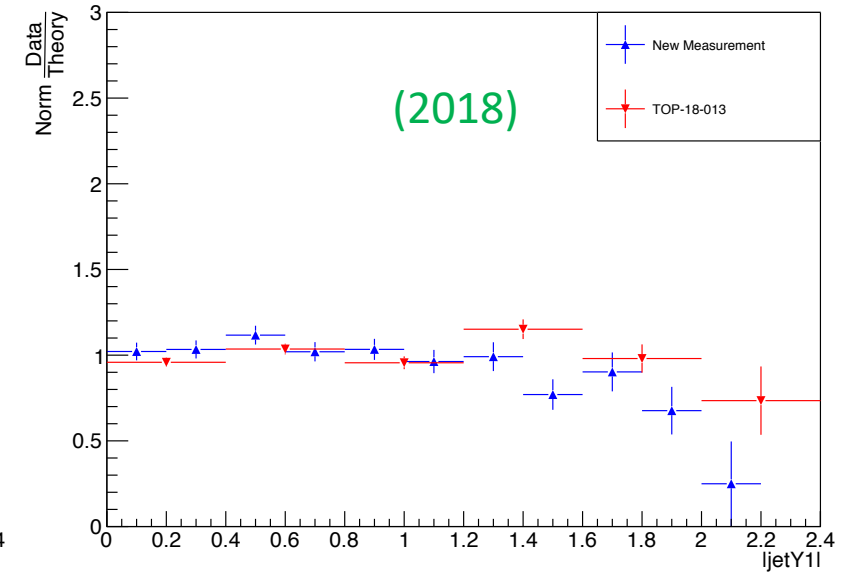
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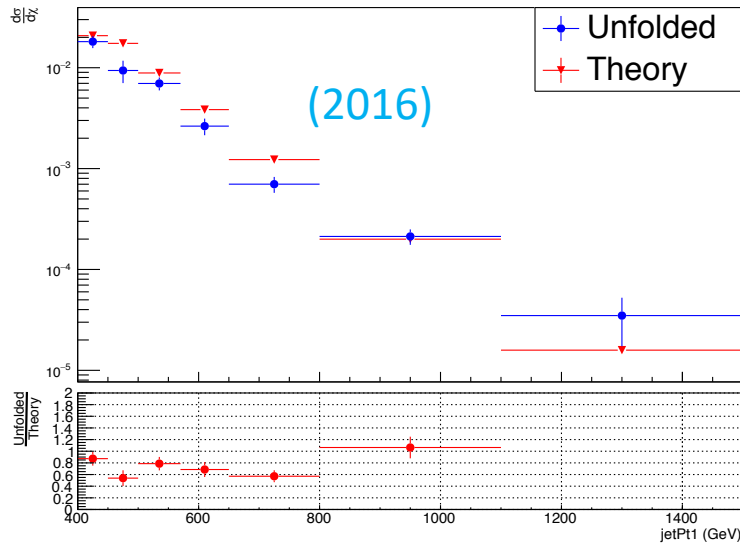


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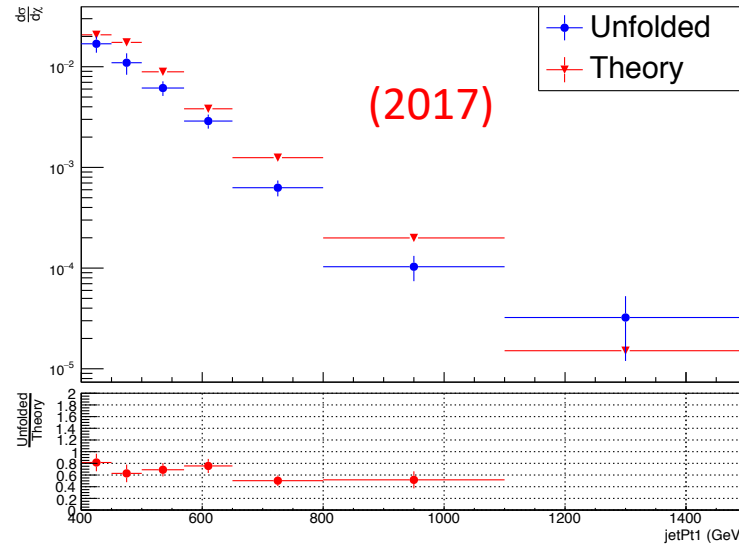


Parton Differential Cross Section

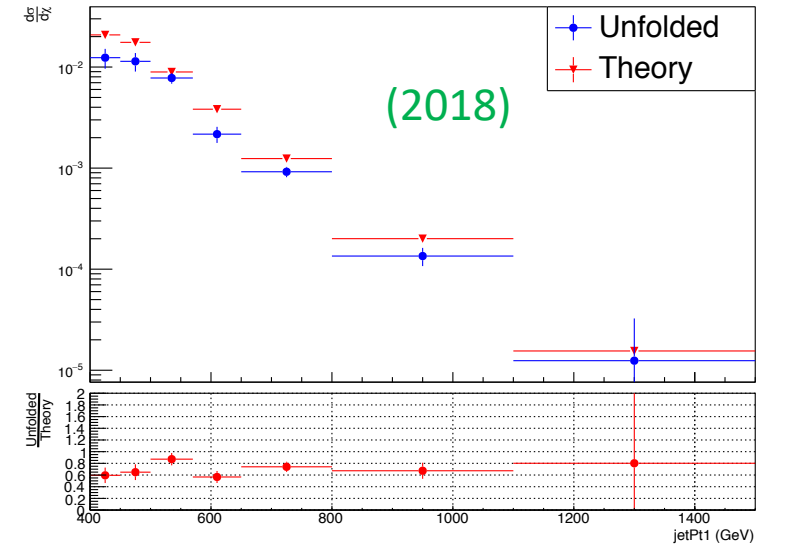
Parton Unfolded vs Theory jetPt1 2016



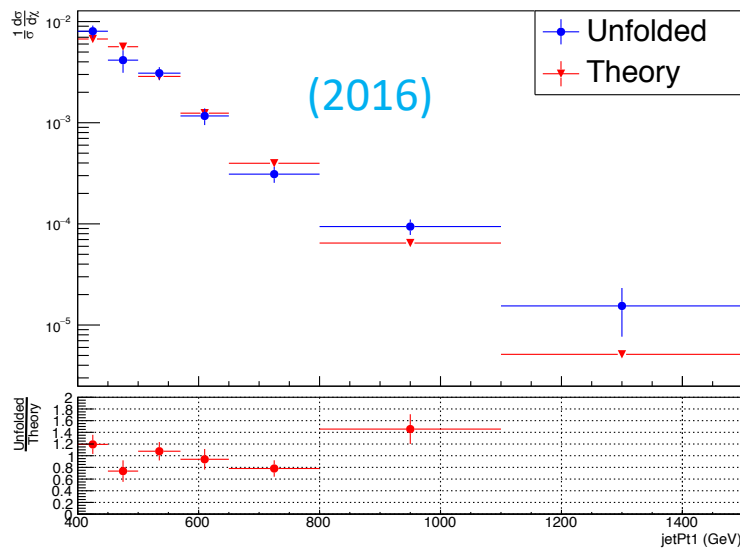
Parton Unfolded vs Theory jetPt1 2017



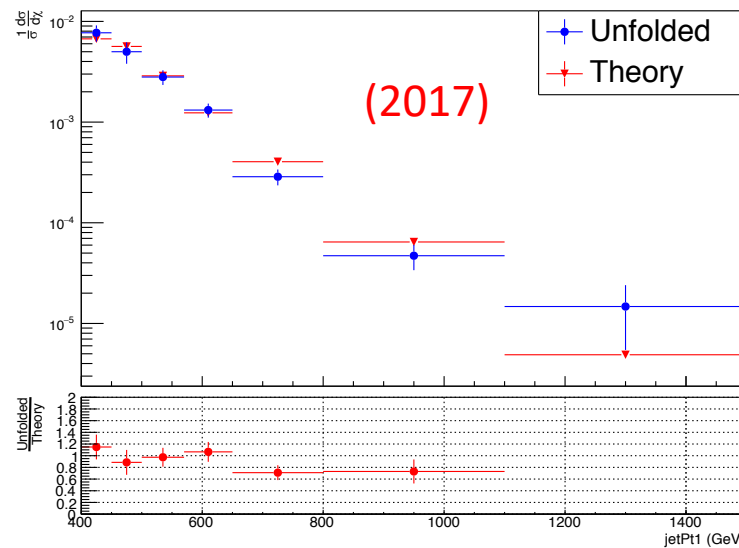
Parton Unfolded vs Theory jetPt1 2018



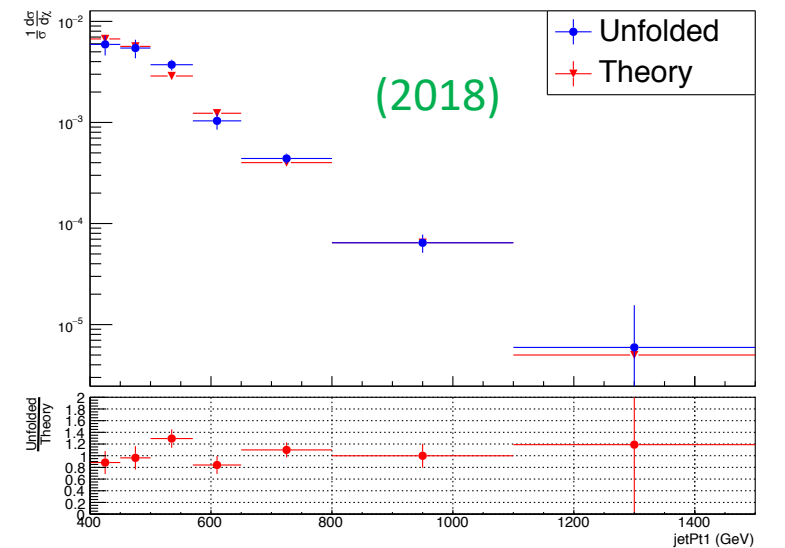
Parton Unfolded vs Theory jetPt1 2016



Parton Unfolded vs Theory jetPt1 2017

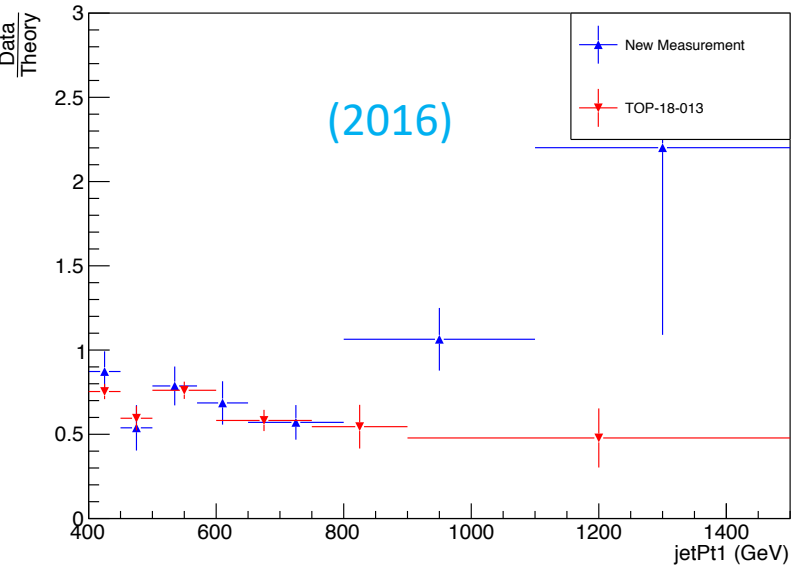


Parton Unfolded vs Theory jetPt1 2018

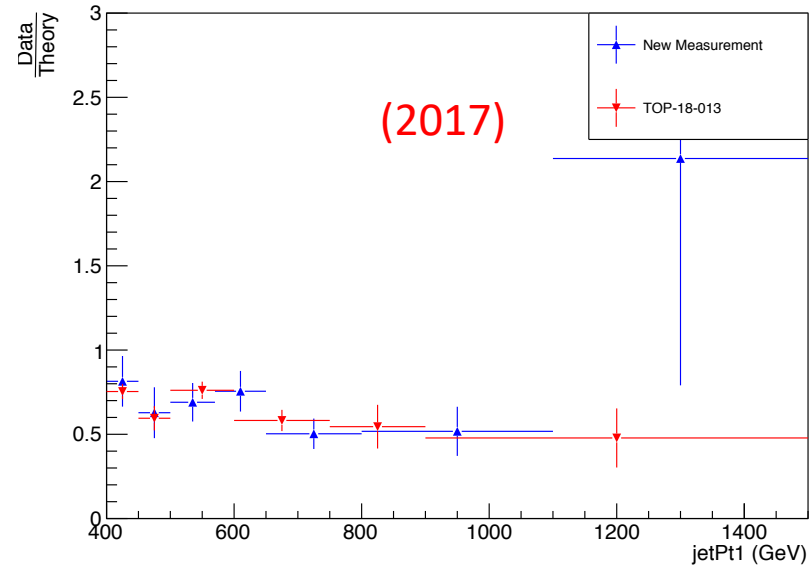


Parton Differential Cross Section Comparison

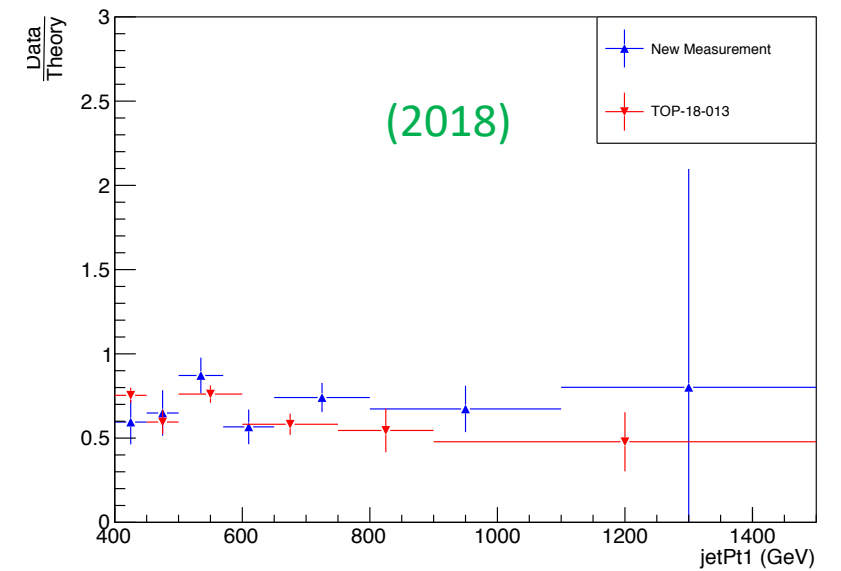
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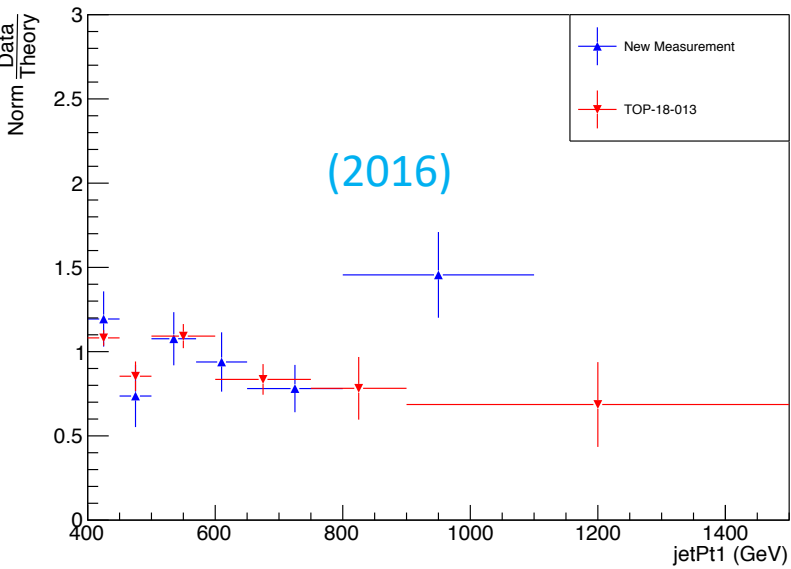
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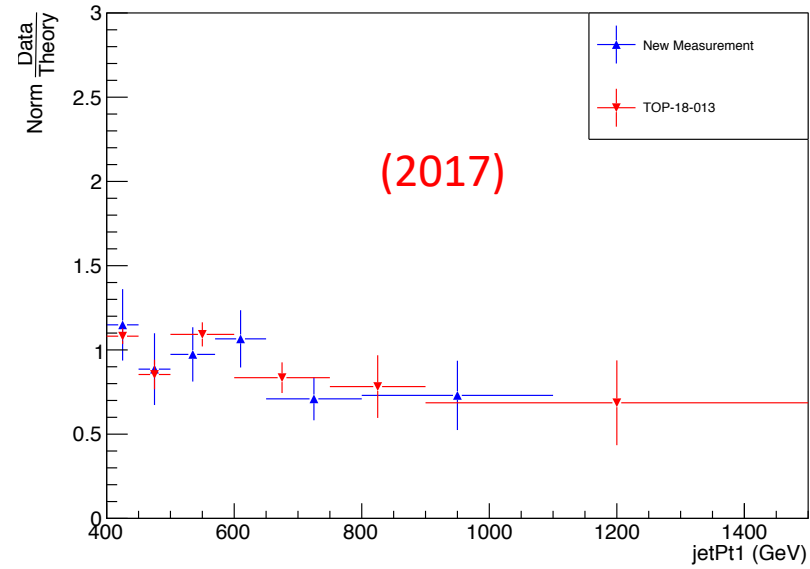
Parton DataOverMC ratio (2018, TOP18013)



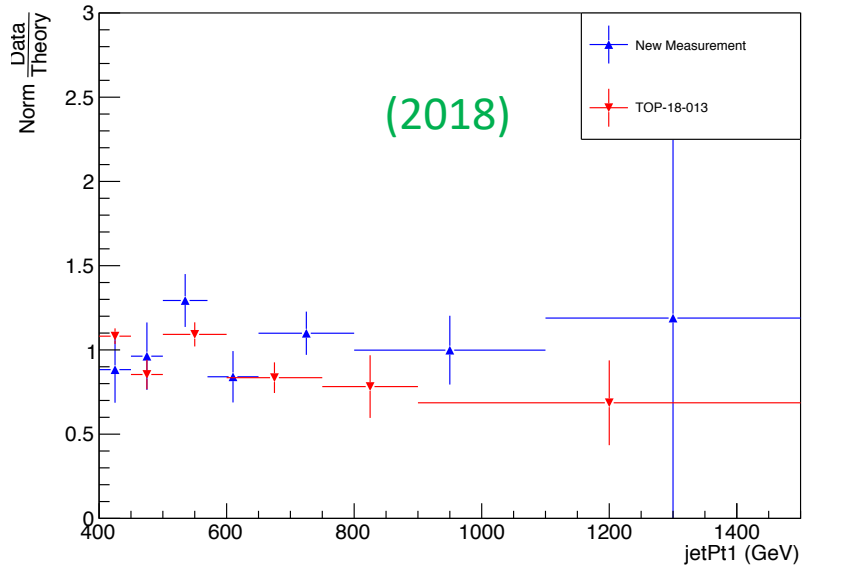
Normalised Parton DataOverMC ratio (2016, TOP18013)



Normalised Parton DataOverMC ratio (2017, TOP18013)

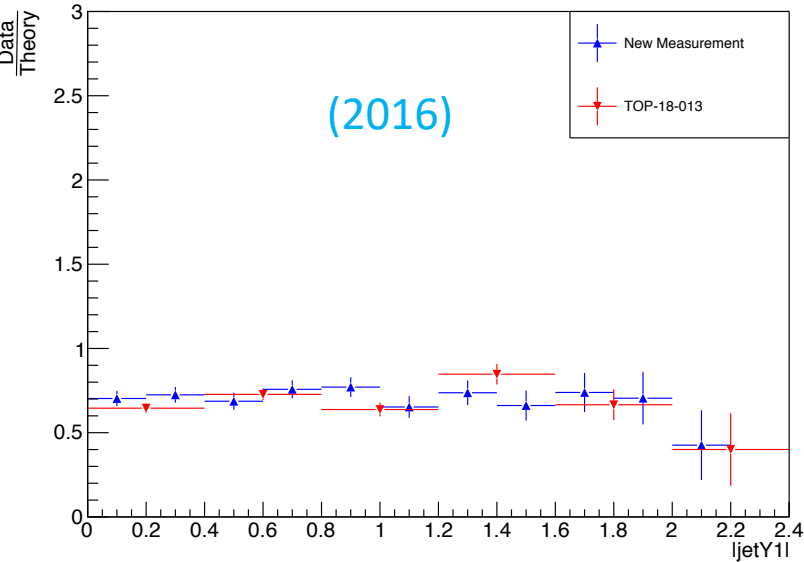


Normalised Parton DataOverMC ratio (2018, TOP18013)

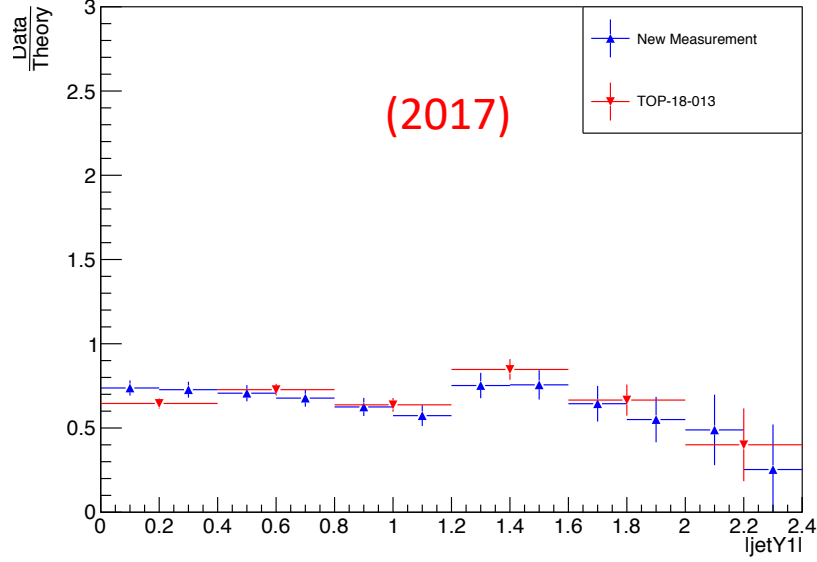


Parton Differential Cross Section Comparison

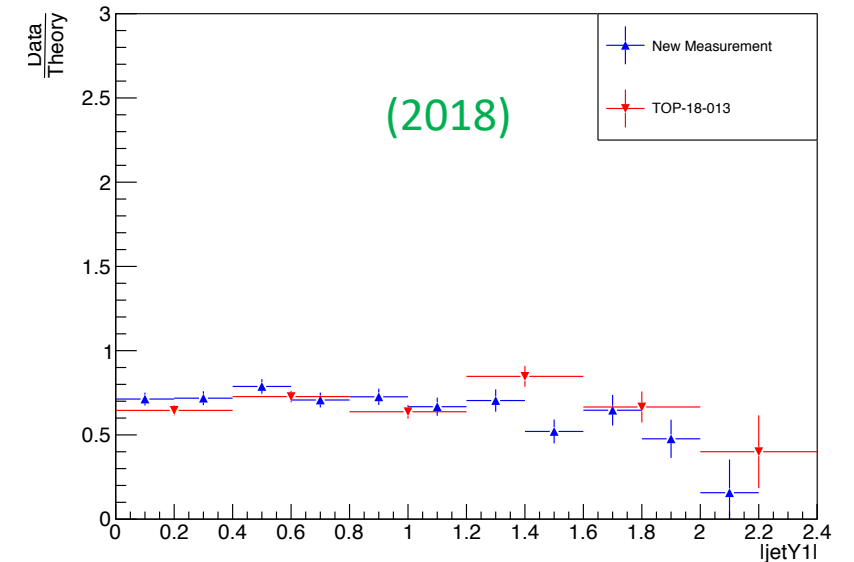
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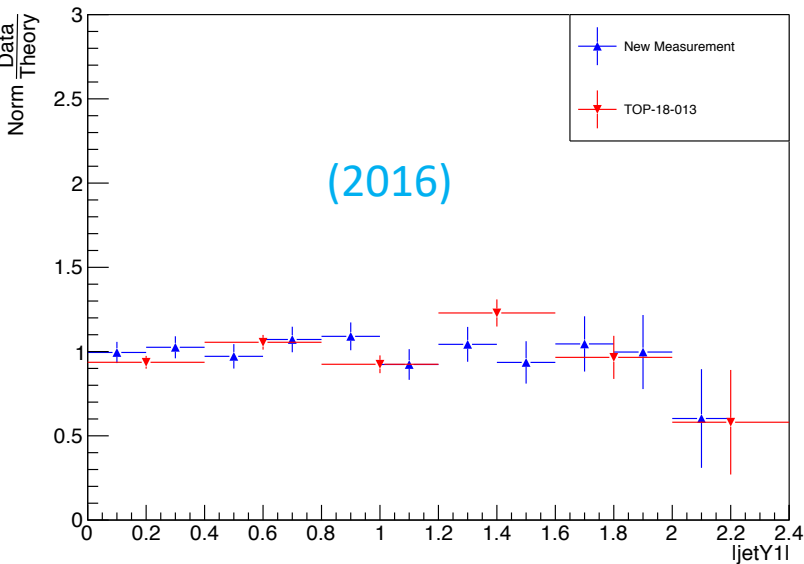
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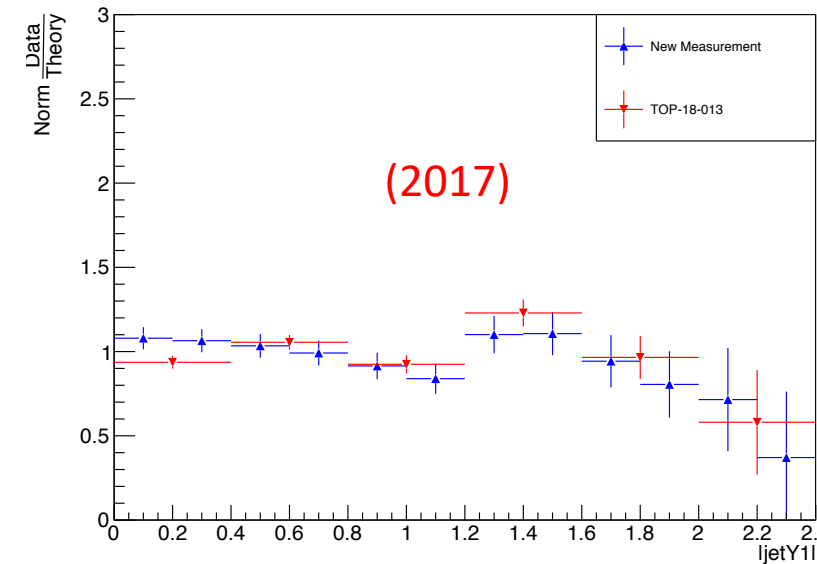
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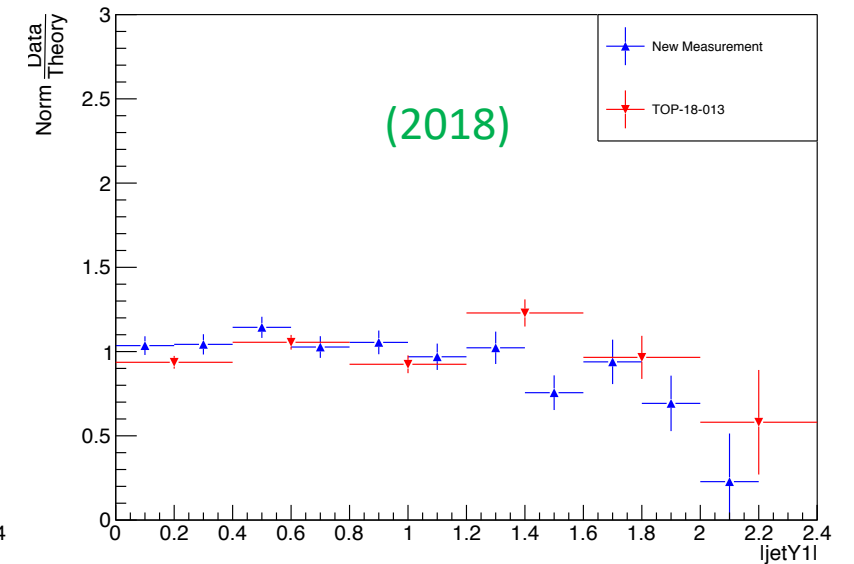
Normalised Parton DataOverMC ratio (2016, TOP18013)



Normalised Parton DataOverMC ratio (2017, TOP18013)



Normalised Parton DataOverMC ratio (2018, TOP18013)



Tag And Probe Calculations

(2016) Tune CUE

eff data: 0.781 ± 0.038

eff ttbar: 0.772 ± 0.014

Efficiency per Pt region

eff data pT[400-600]: 0.761 ± 0.042

eff ttbar pT[400-600]: 0.778 ± 0.016

eff data pT[600-800]: 0.851 ± 0.100

eff ttbar pT[600-800]: 0.748 ± 0.031

eff data pT[800-Inf]: 0.886 ± 0.160

eff ttbar pT[800-Inf]: 0.775 ± 0.063

(2016) Tune CP5

eff data: 0.781 ± 0.038

eff ttbar: 0.813 ± 0.01

Efficiency per Pt region

eff data pT[400-600]: 0.761 ± 0.042

eff ttbar pT[400-600]: 0.812 ± 0.011

eff data pT[600-800]: 0.851 ± 0.100

eff ttbar pT[600-800]: 0.820 ± 0.023

eff data pT[800-Inf]: 0.886 ± 0.160

eff ttbar pT[800-Inf]: 0.806 ± 0.053



Top Angular Distributions

- We employ the dijet angular variable χ from the rapidities of the two leading jets
- Why χ ?
 - The distributions associated with the final states produced via QCD interactions are relatively flat in comparison with the distributions of the BSM models or new particles, which typically peak at low values of χ
- We can measure the variable χ in two ways

1. By measuring the difference of the rapidities of the two leading jets such as the corresponding rapidity in the ZMF is:

$$y^* = \frac{1}{2}(y_1 - y_2)$$

χ is defined as $\chi = e^{|y^*|} = e^{|y_1 - y_2|}$ (1) and can be measured by creating the TLorentzVector, boost it to the ZMF and find the rapidity difference of the two leading jets

2. By measuring the scattering angle θ^* (angle between top quark and z-axis in the Zero Momentum Frame)

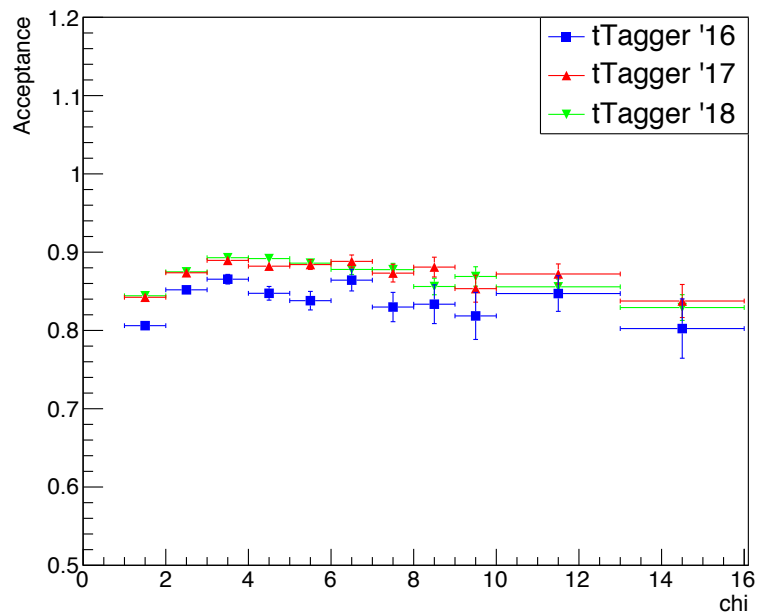
We define as $y^* = \frac{1}{2} \ln\left(\frac{1+|\cos\theta^*|}{1-|\cos\theta^*|}\right)$ and from (1) we can find that:

$$\chi = \frac{1 + |\cos\theta^*|}{1 - |\cos\theta^*|}$$

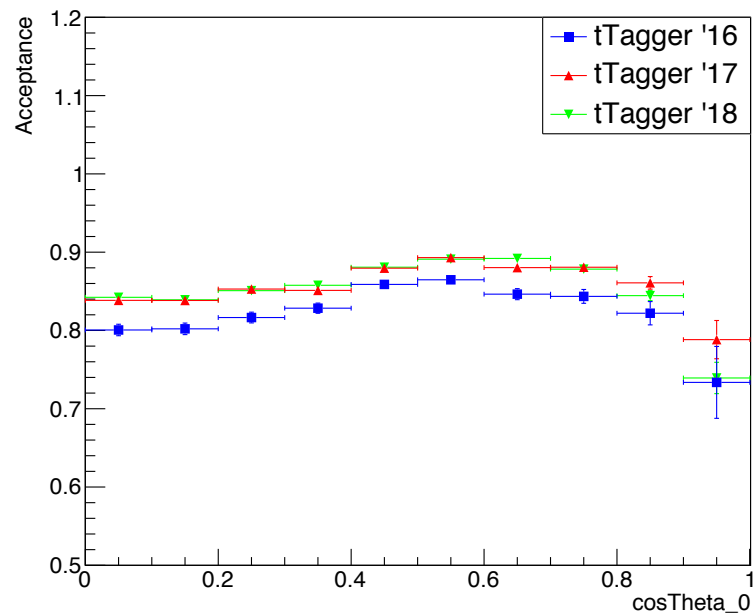


Efficiency, Acceptance

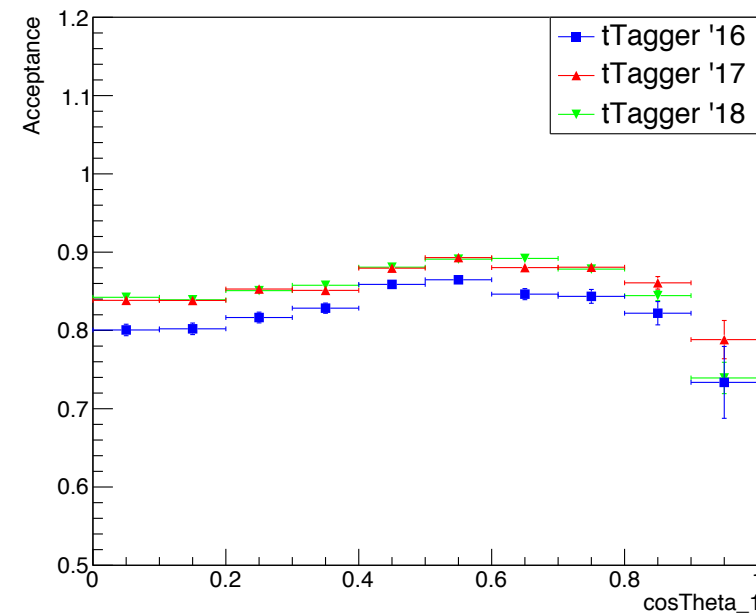
Parton Acceptance '16,'17,'18 NominalMC



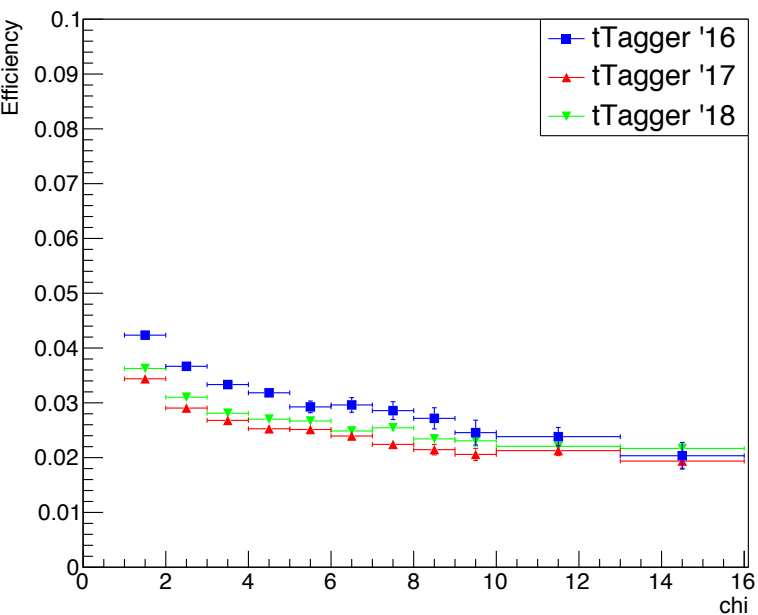
Parton Acceptance '16,'17,'18 NominalMC



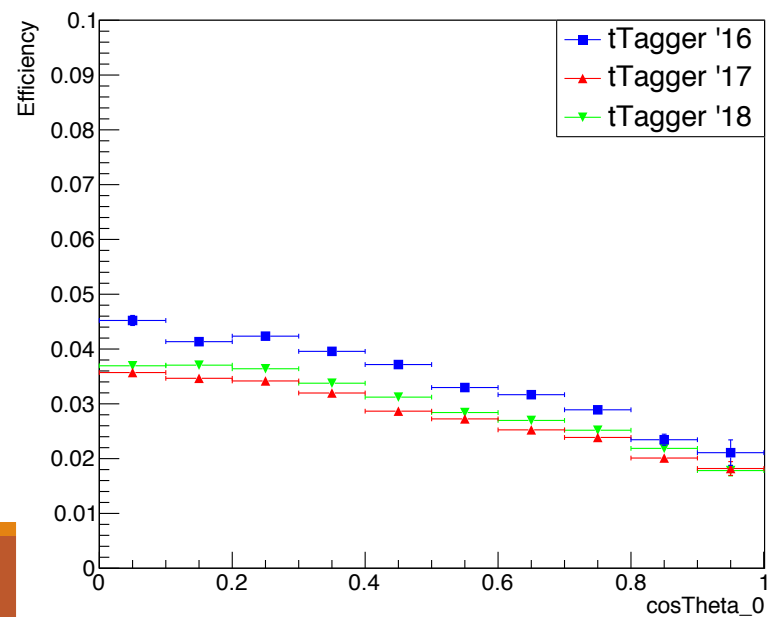
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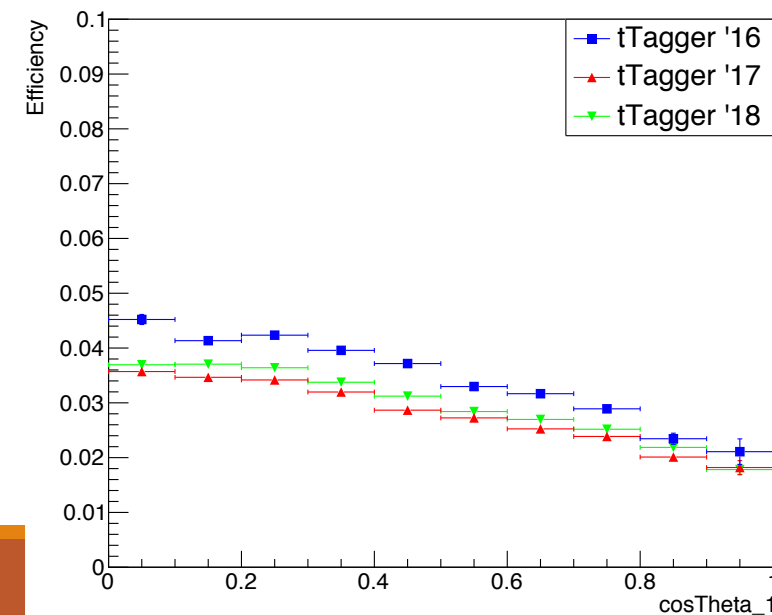
Parton Efficiency '16,'17,'18 NominalMC



Parton Efficiency '16,'17,'18 NominalMC

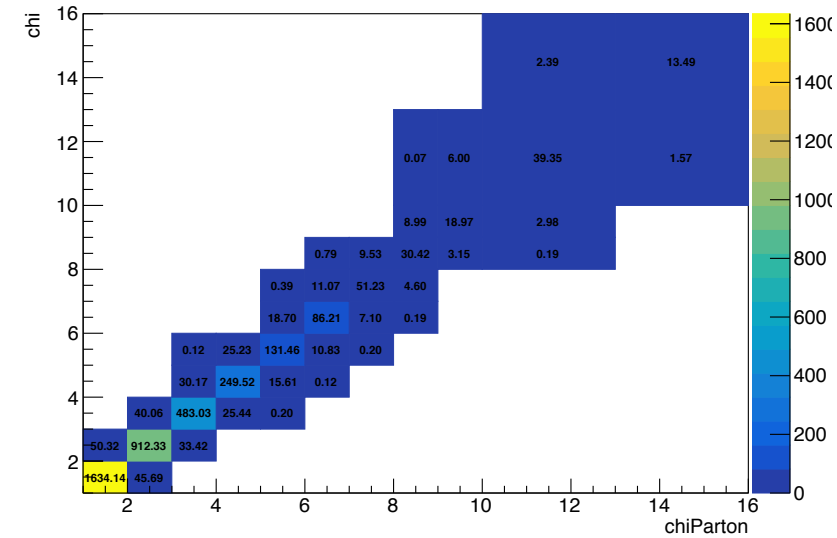


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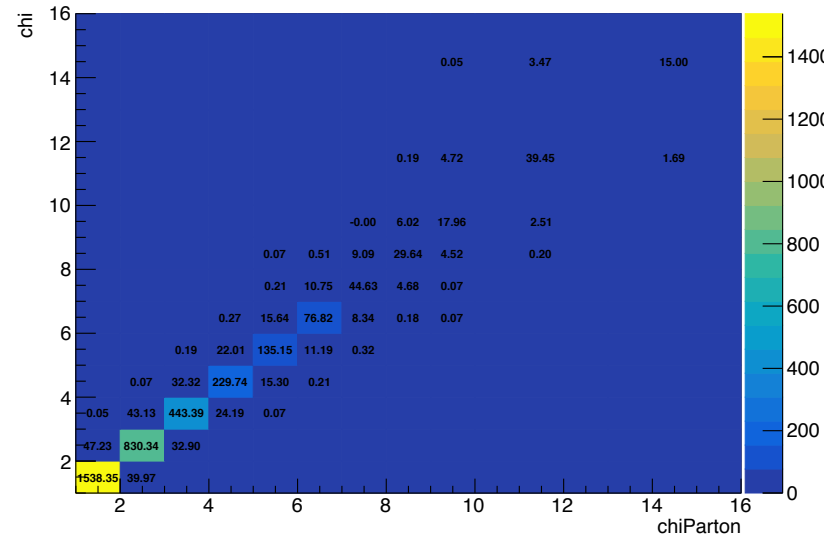


Response Matrices

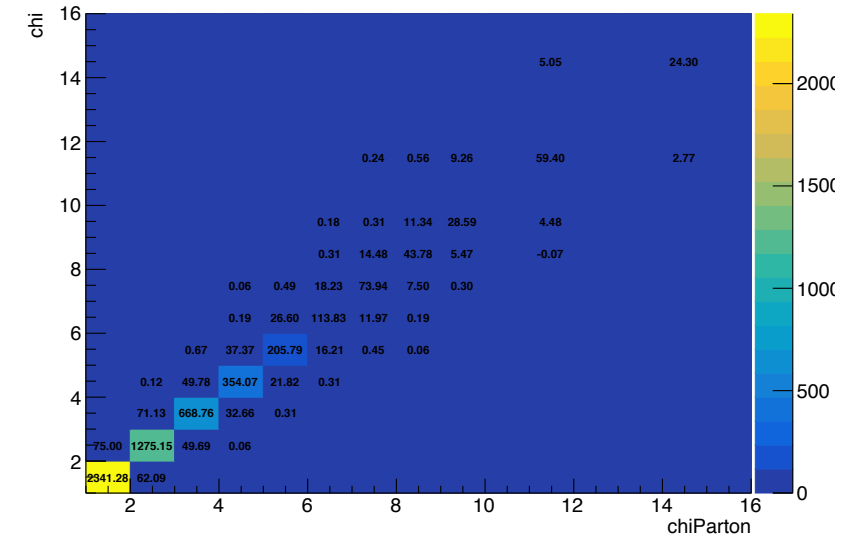
Response Reco-Parton chi 2016 NominalMC



Response Reco-Parton chi 2017 NominalMC

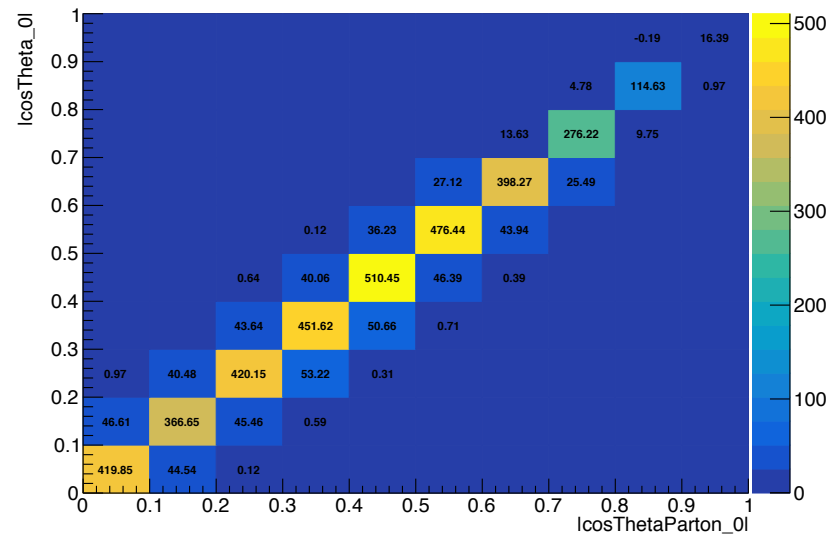


Response Reco-Parton chi 2018 NominalMC

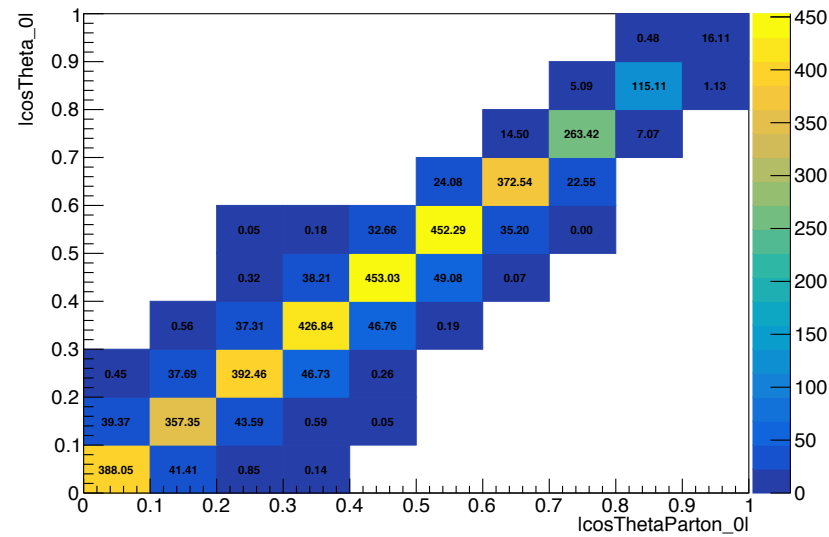


Response Matrices

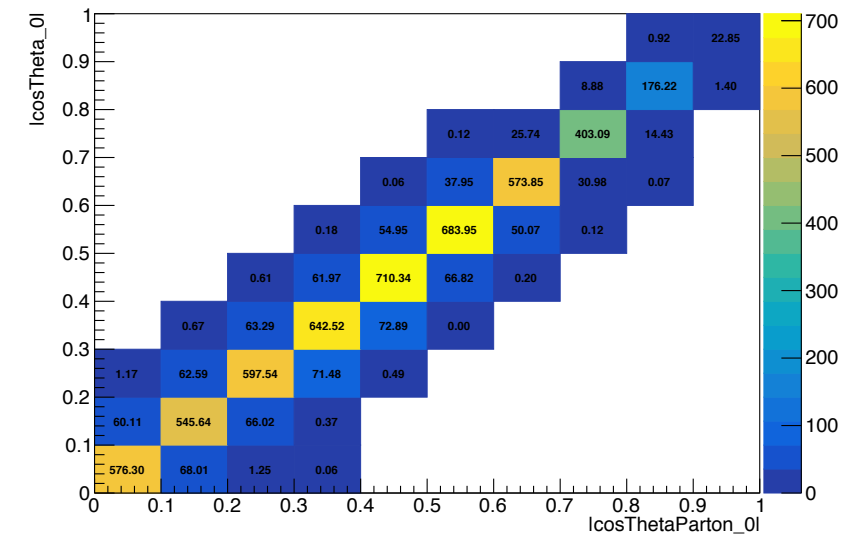
Response Reco-Parton cosTheta_0 2016 NominalMC



Response Reco-Parton cosTheta_0 2017 NominalMC



Response Reco-Parton cosTheta_0 2018 NominalMC



Signal Extraction

$$S(x_{reco}) = D(x_{reco}) - C_{bkg}^{yield} N_{QCD}^{fit} C_{QCD}^{shape}(x_{reco}) Q(x_{reco}) - B(x_{reco})$$

Diagram illustrating the components of the signal extraction equation:

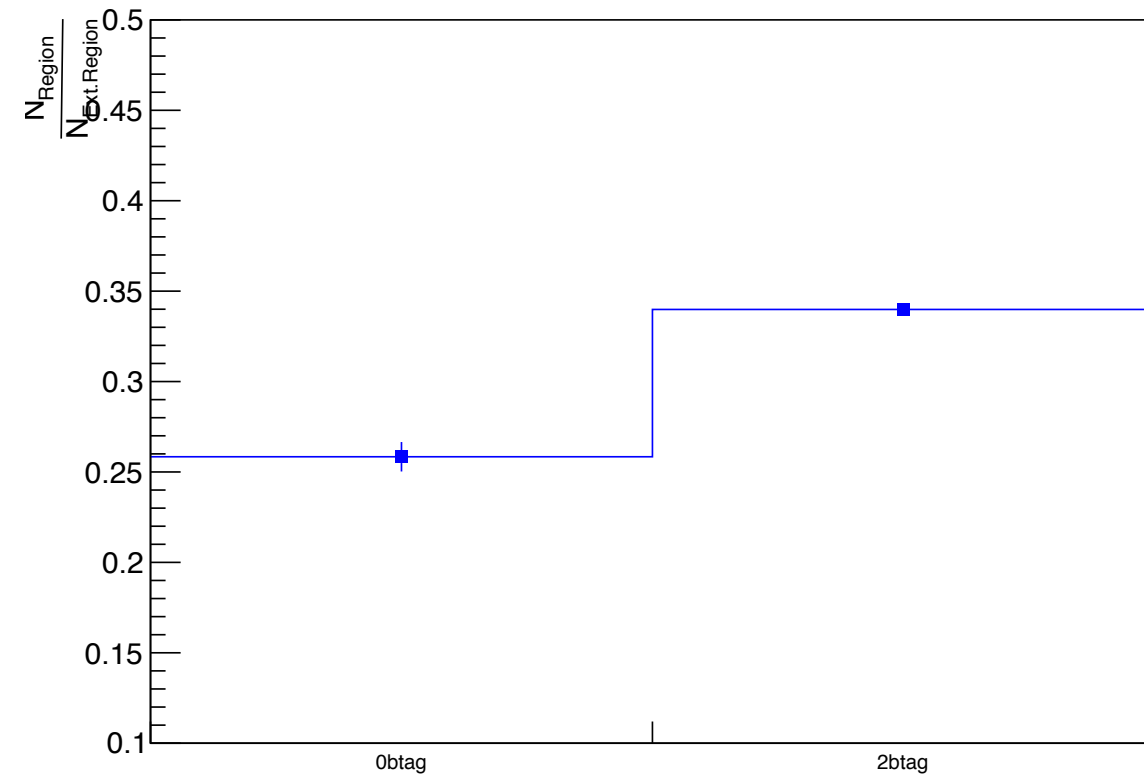
- Fiducial Yield**: Points to $S(x_{reco})$
- Measured dist from data**: Points to $D(x_{reco})$
- Fitted number of QCD events in SR_A** : Points to N_{QCD}^{fit}
- QCD shape taken from Data (CR)**: Points to $C_{QCD}^{shape}(x_{reco})$
- Transfer factor from SR_A to SR**: Points to C_{bkg}^{yield}
- QCD shape correction factor**: Points to $C_{QCD}^{shape}(x_{reco})$
- Subdominant bkg shape and contribution (MC)**: Points to $B(x_{reco})$

- Where x_{reco} is the respected variable of interest (ttbar mass, pt, rapidity, leading and subleading jetPt and |jetY|)
 - We deploy a fit in the Signal Region (2btag) to extract the N_{QCD}^{fit}
- $$D(m^t)^{(i)} = N_{tt}^{(i)} T^{(i)}(m^t, k_{MassScale}, k_{MassResolution}) + N_{bkg}^{(i)} B(m^t)(1 + k_1 x) + N_{sub}^{(i)} O^{(i)}(m^t)$$
- Our data CR is contaminated from ttbar and subdominant bkg which has to be dealt with.

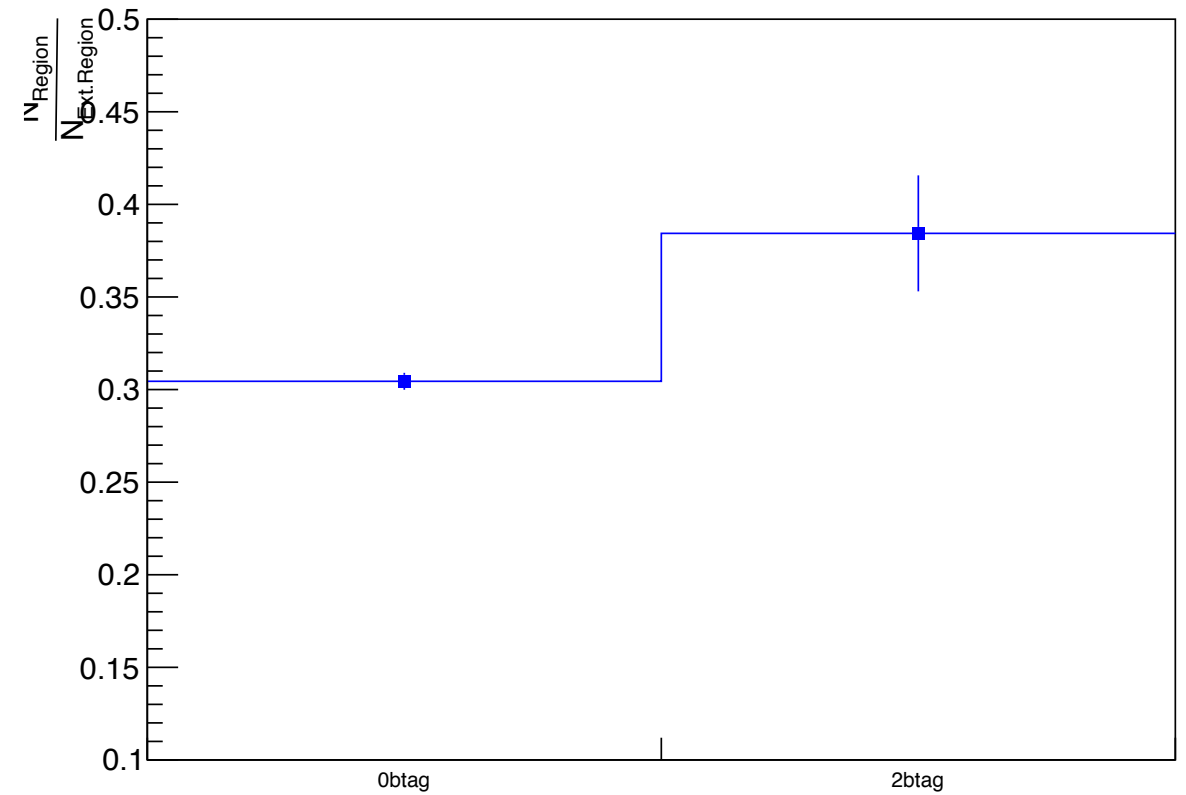


R yield

R_{yield} transfer factor 2016 cosTheta_0

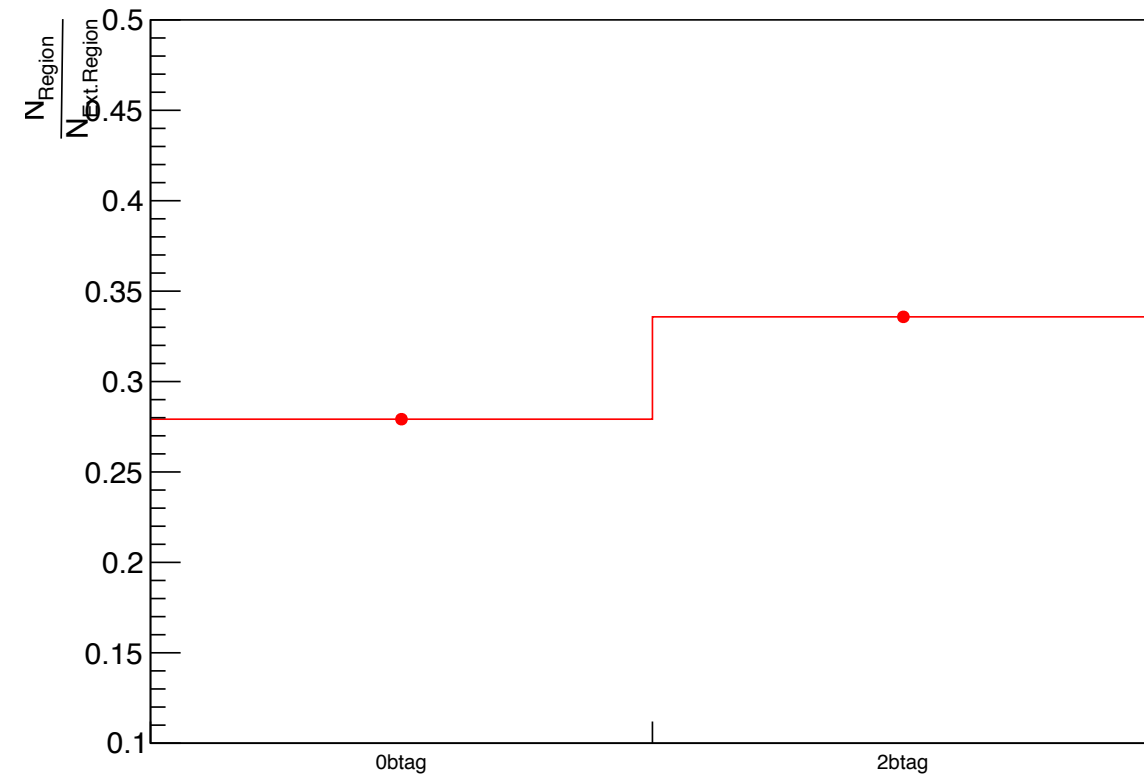


R_{yield} transfer factor 2016 cosTheta_0(Closure Test)

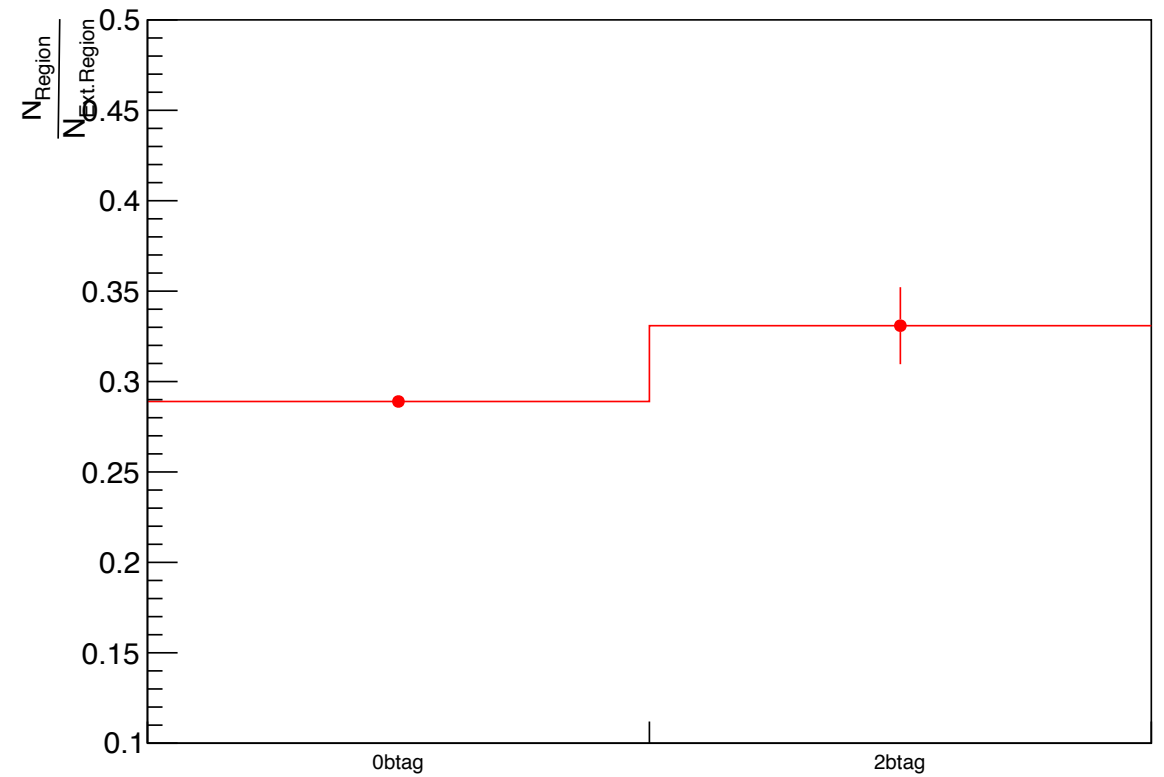


R yield

R_{yield} transfer factor 2017 cosTheta_0

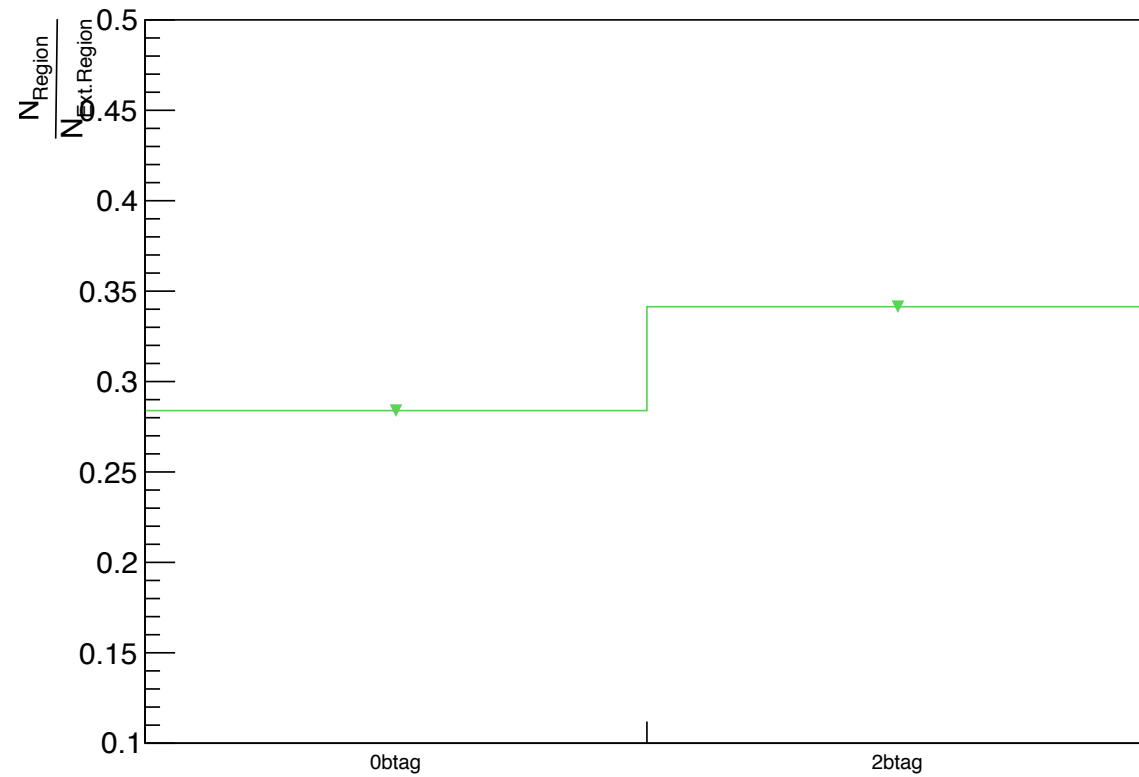


R_{yield} transfer factor 2017 cosTheta_0(Closure Test)

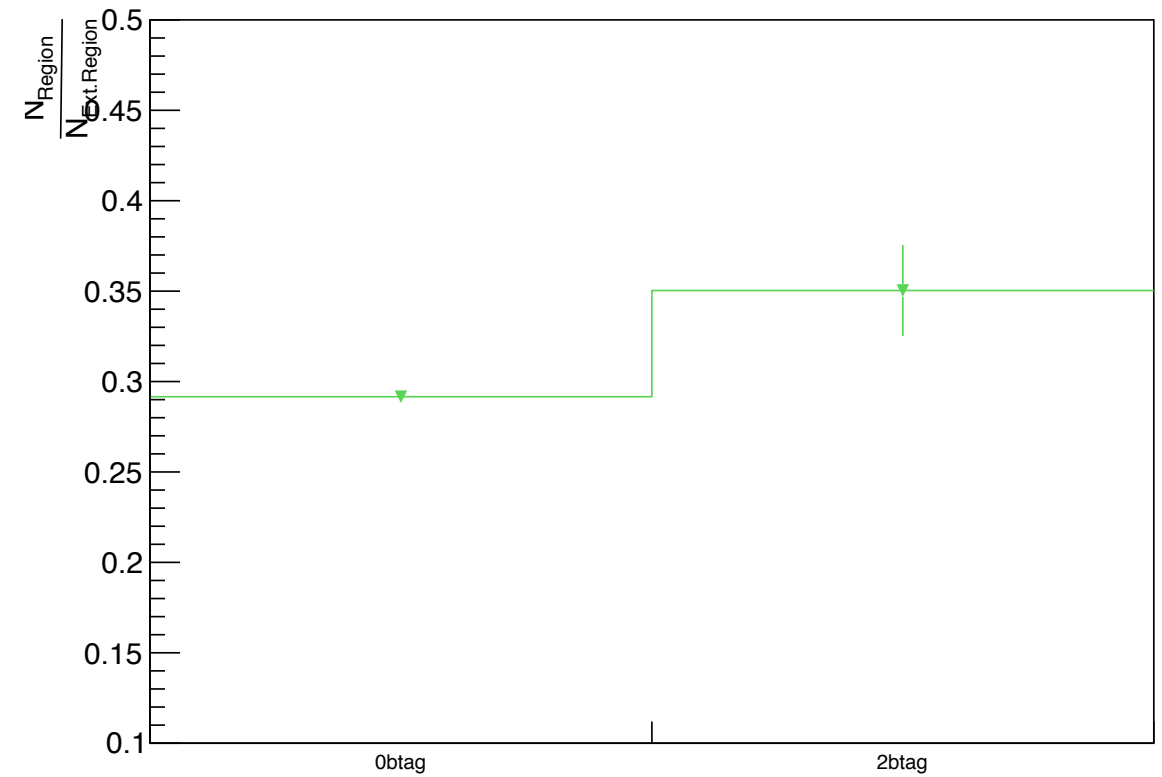


R yield

R_{yield} transfer factor 2018 cosTheta_0

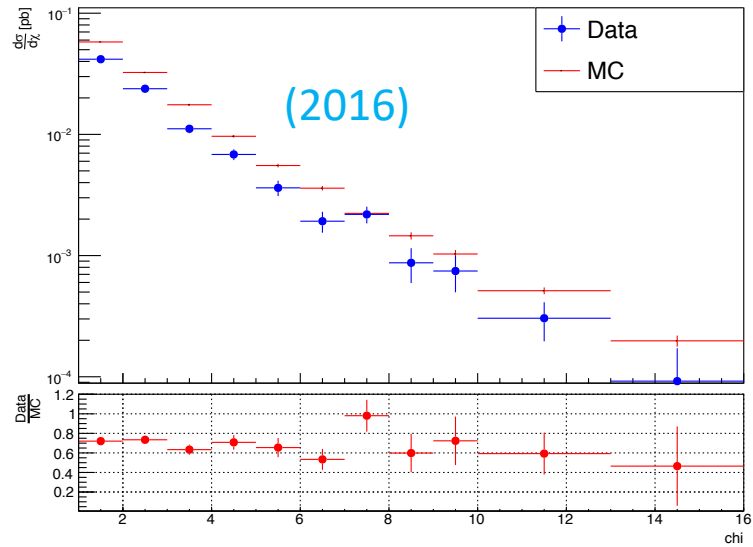


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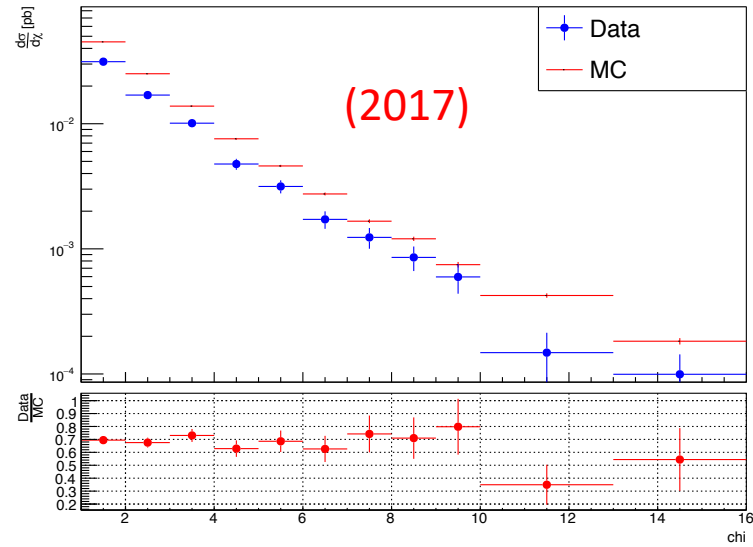


Fiducial Measurement

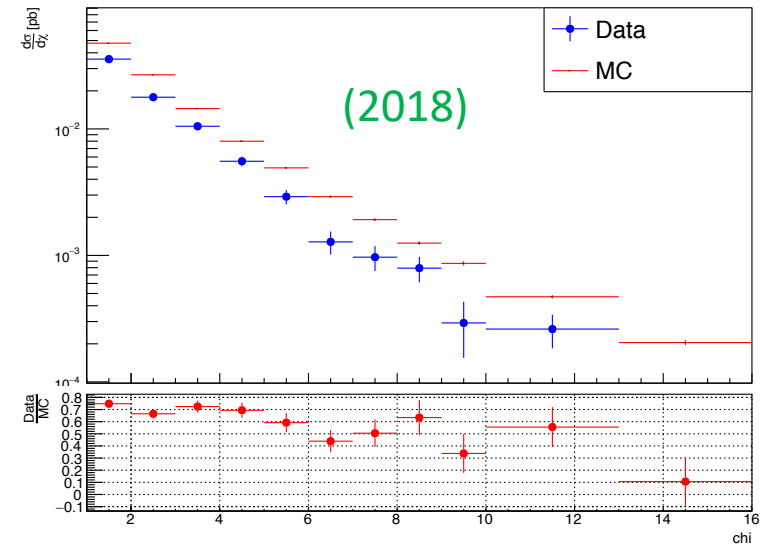
Data vs MC 2016 for chi



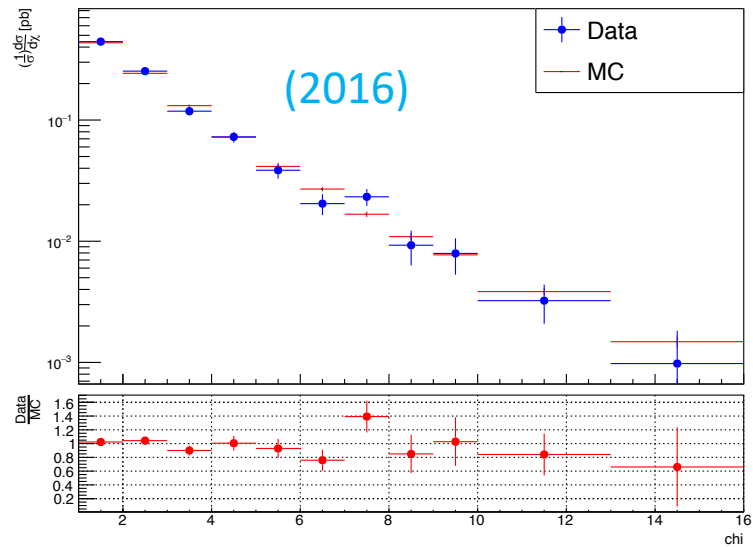
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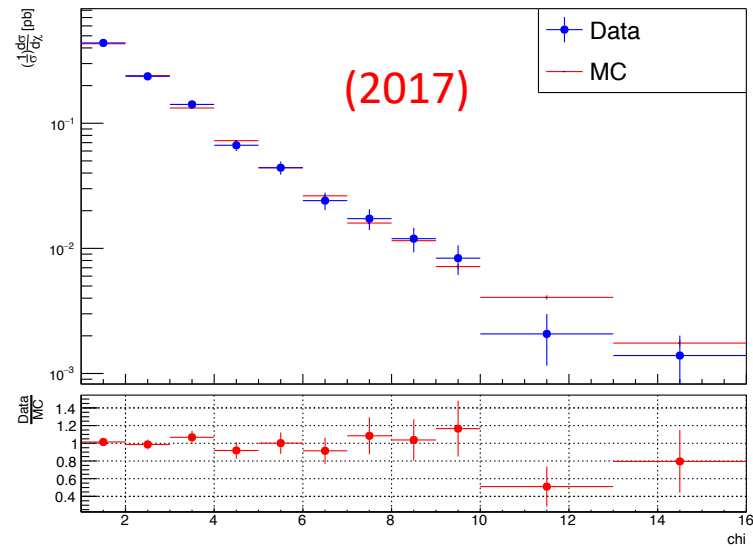
Data vs MC 2018 for chi



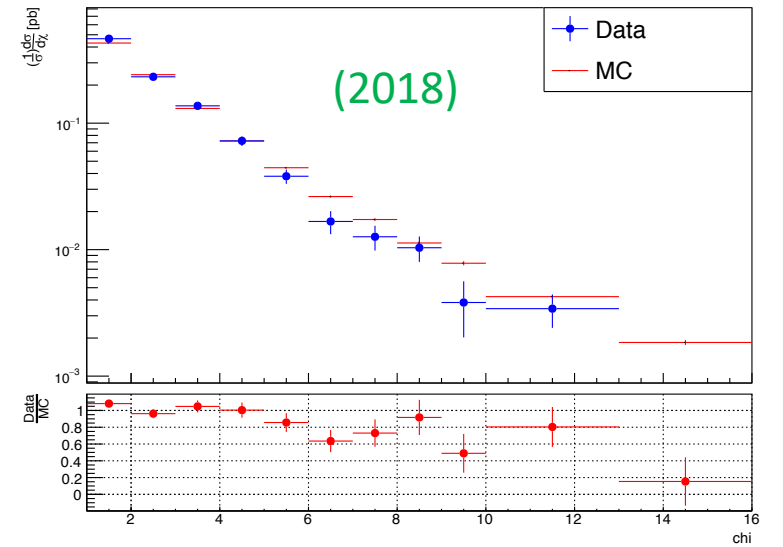
Data vs MC 2016 for chi



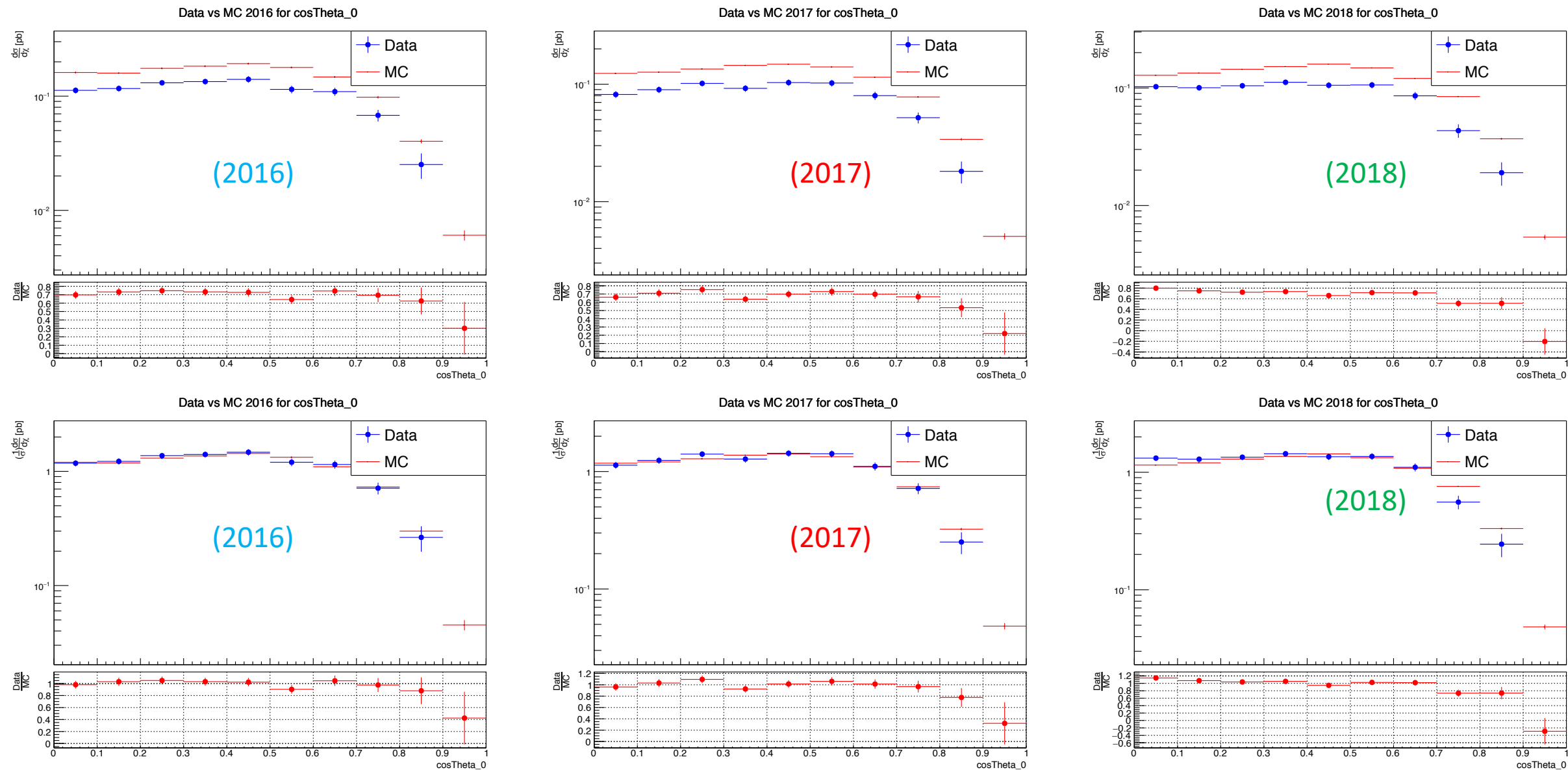
Data vs MC 2017 for chi



Data vs MC 2018 for chi

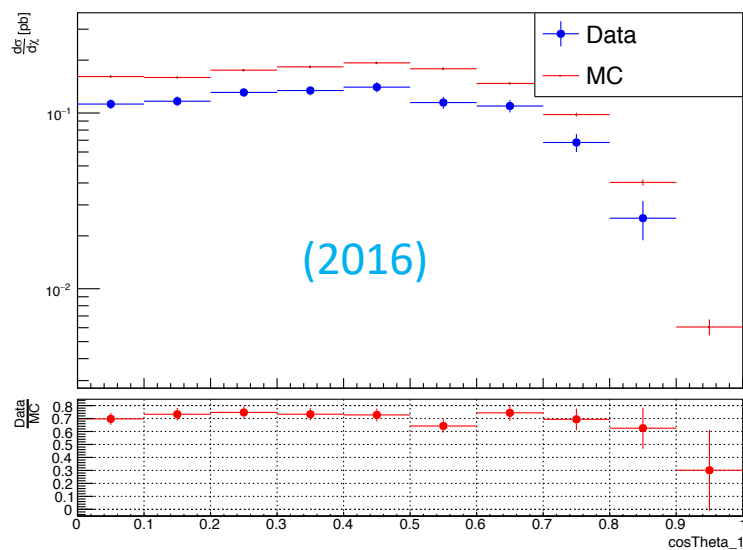


Fiducial Measurement

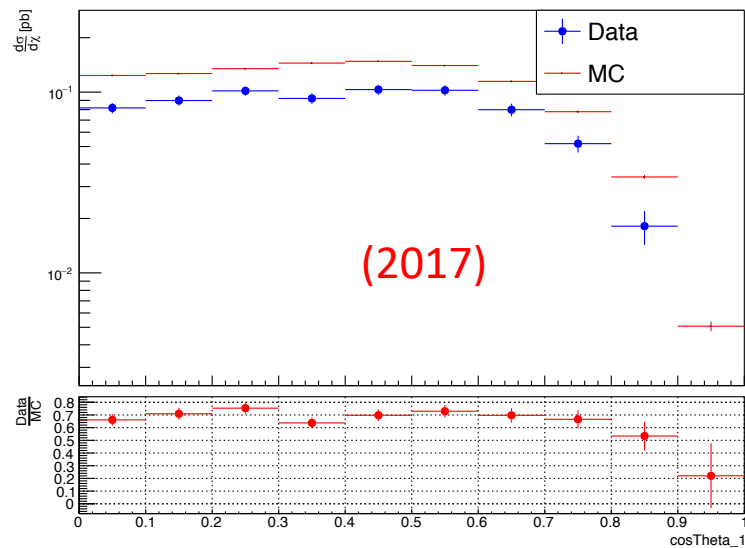


Fiducial Measurement

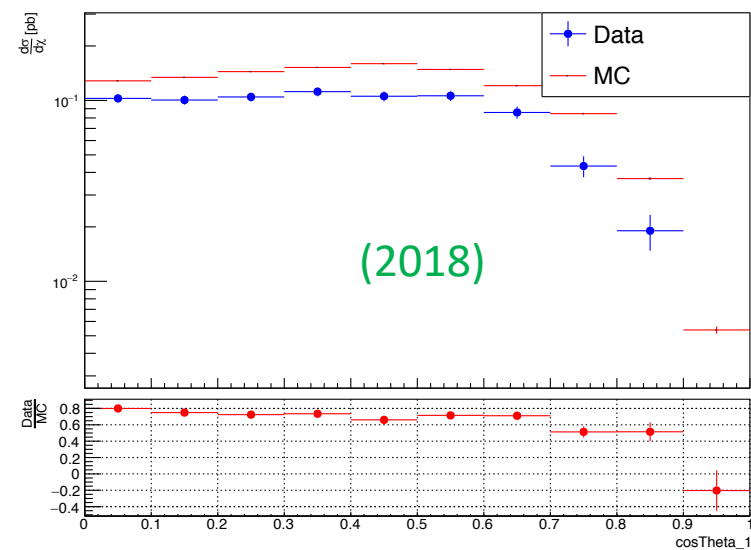
Data vs MC 2016 for $\cos\Theta_{11}$



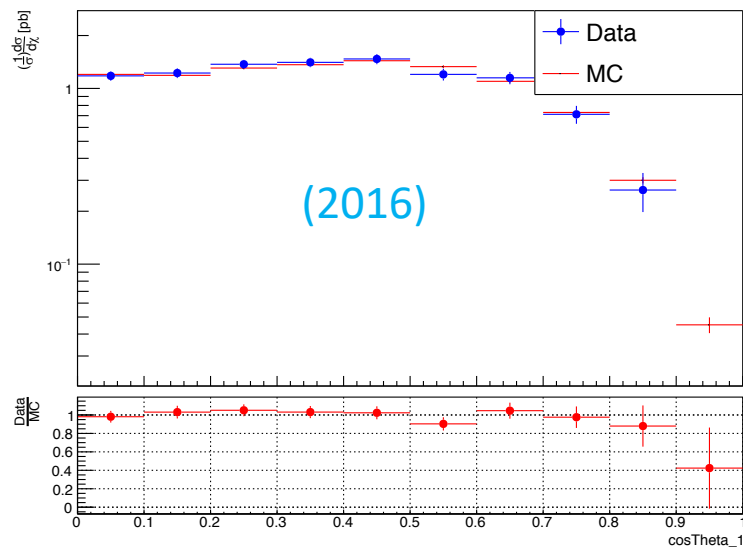
Data vs MC 2017 for $\cos\Theta_{11}$



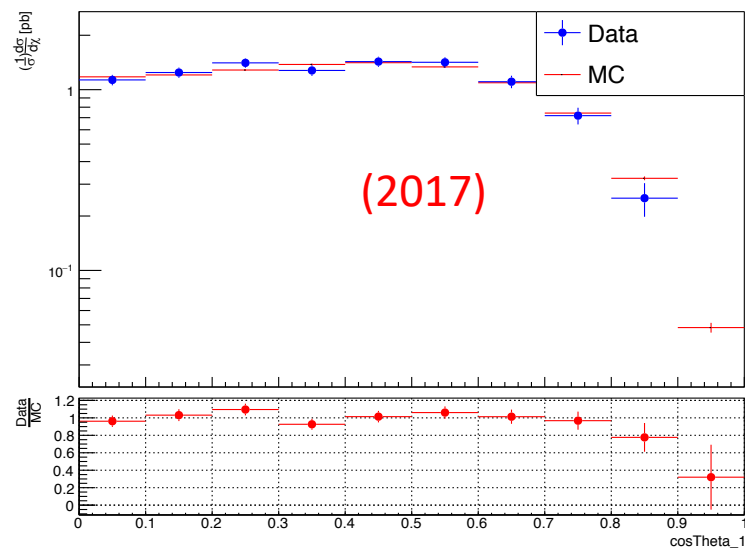
Data vs MC 2018 for $\cos\Theta_{11}$



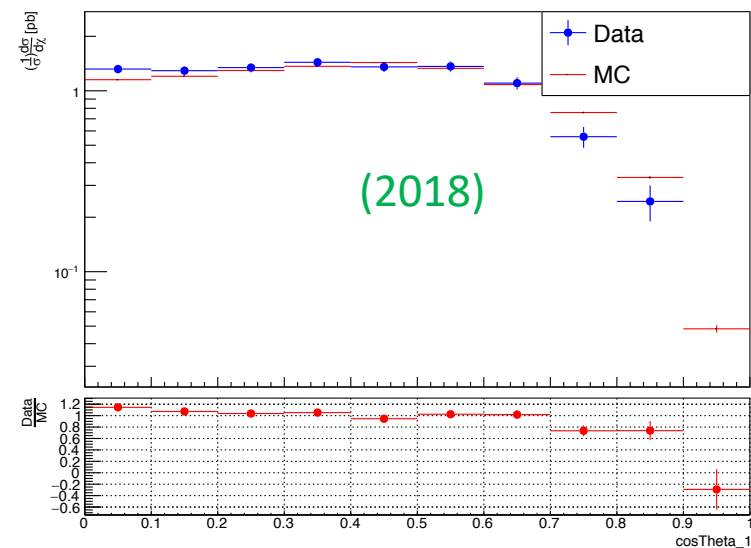
Data vs MC 2016 for $\cos\Theta_{11}$



Data vs MC 2017 for $\cos\Theta_{11}$

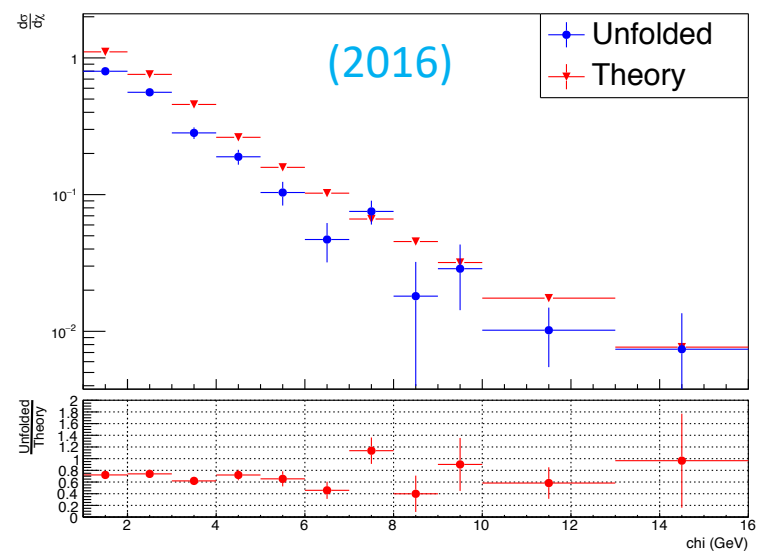


Data vs MC 2018 for $\cos\Theta_{11}$

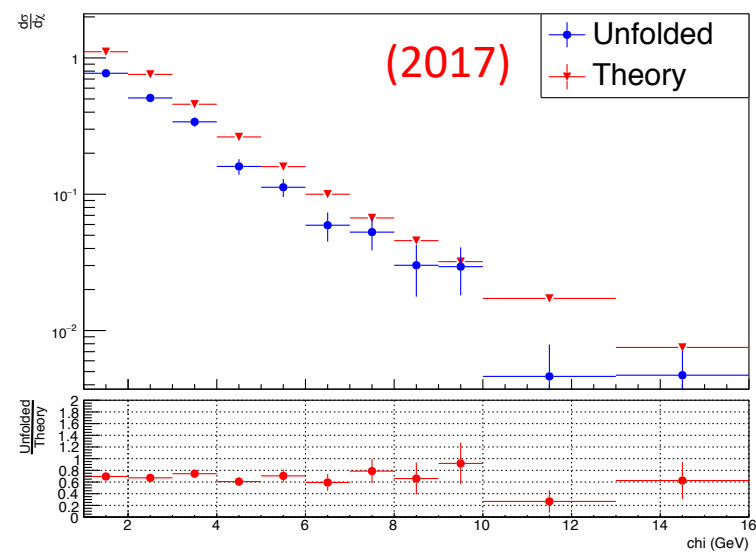


Parton Measurement

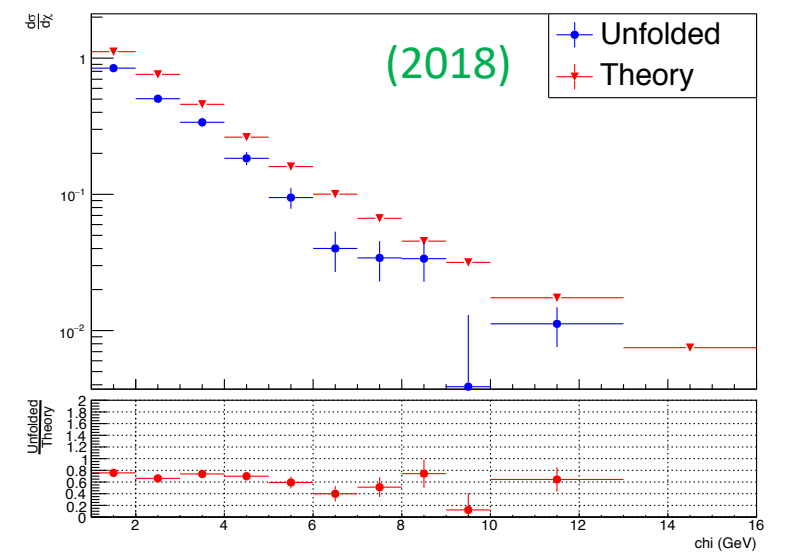
Parton Unfolded vs Theory chi 2016



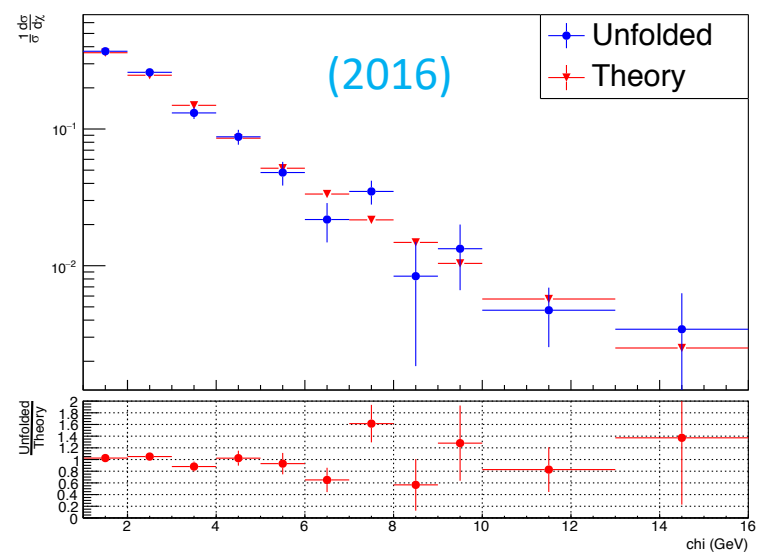
Parton Unfolded vs Theory chi 2017



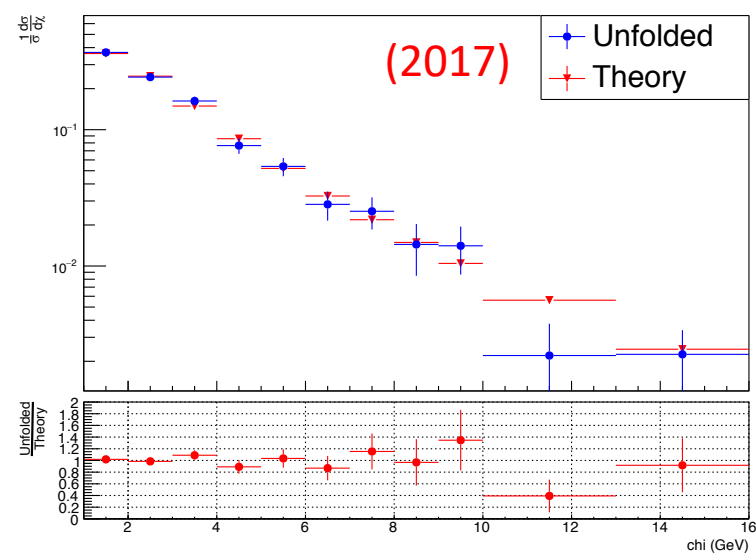
Parton Unfolded vs Theory chi 2018



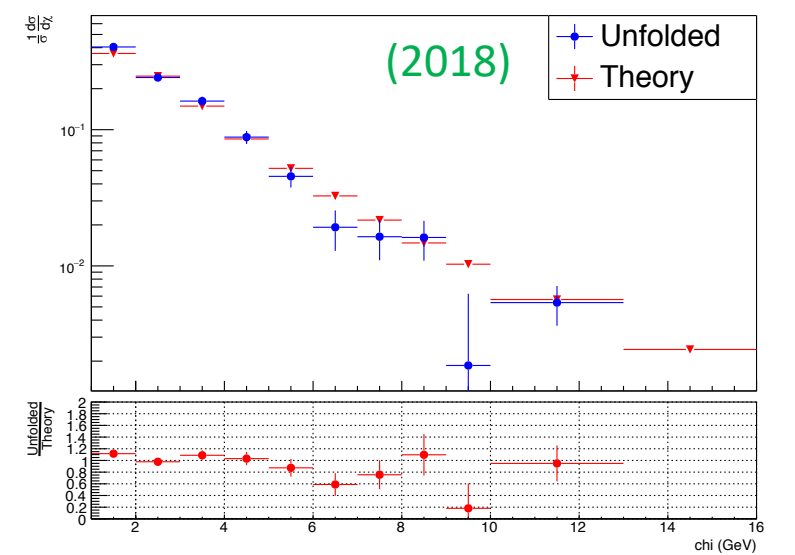
Parton Unfolded vs Theory chi 2016



Parton Unfolded vs Theory chi 2017

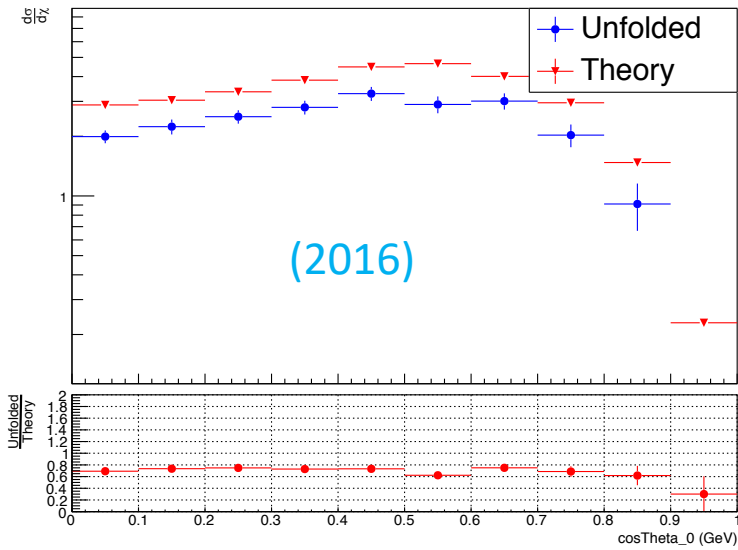


Parton Unfolded vs Theory chi 2018

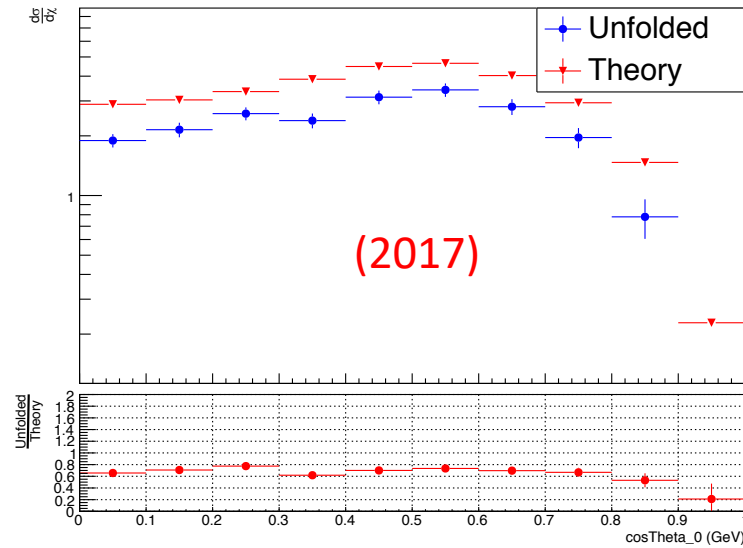


Parton Measurement

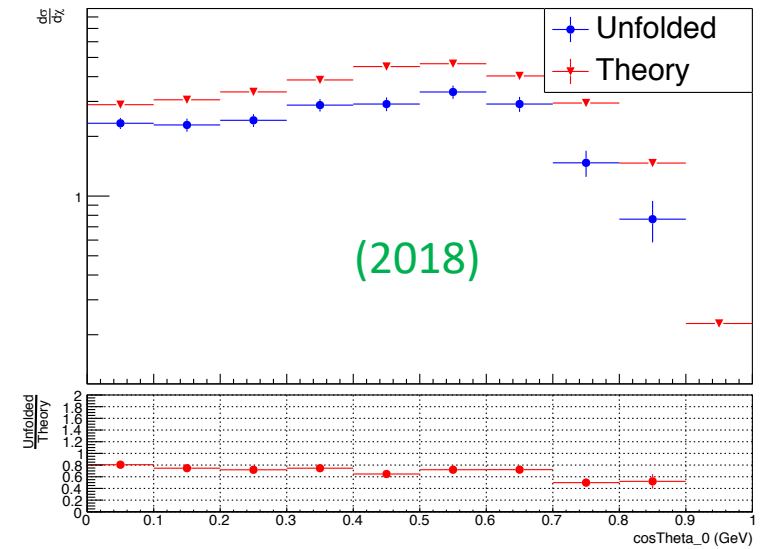
Parton Unfolded vs Theory cosTheta_0 2016



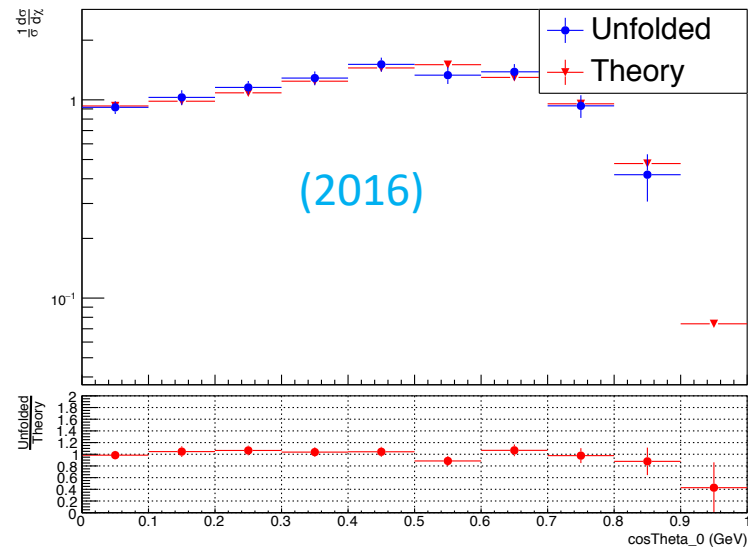
Parton Unfolded vs Theory cosTheta_0 2017



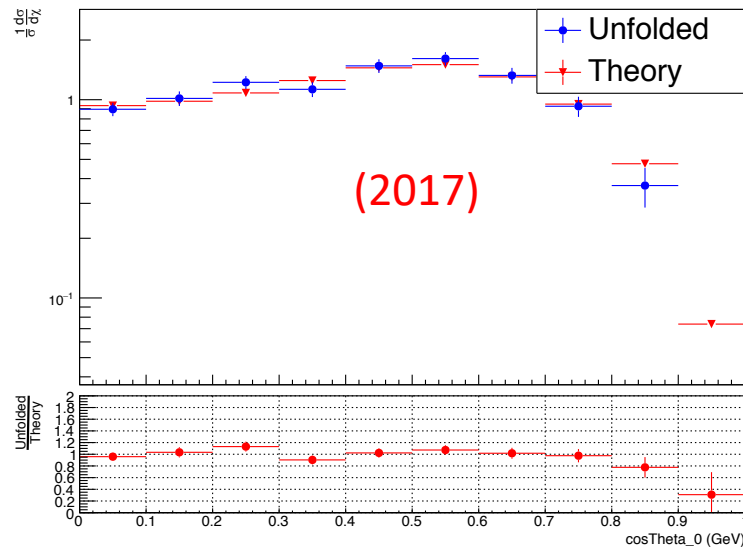
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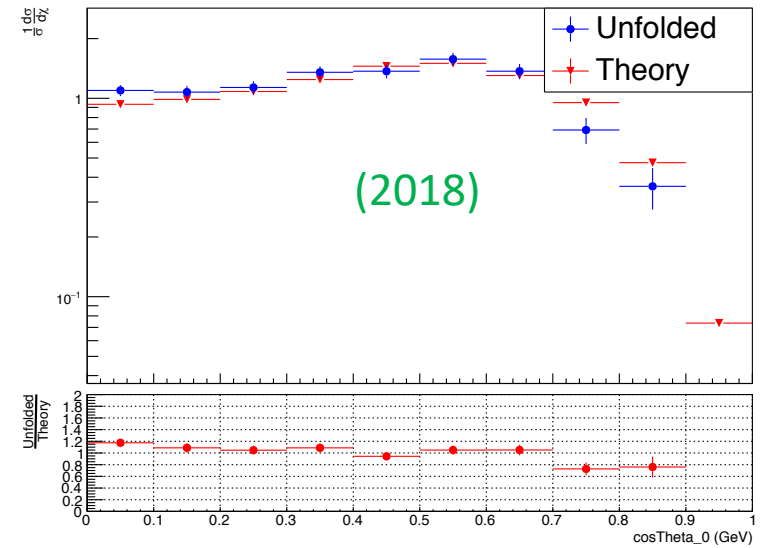
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Parton Unfolded vs Theory cosTheta_0 2017

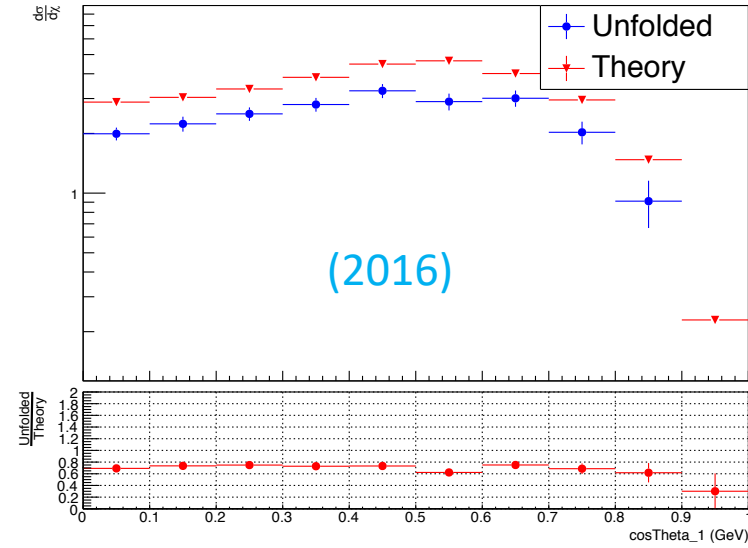


Parton Unfolded vs Theory cosTheta_0 2018

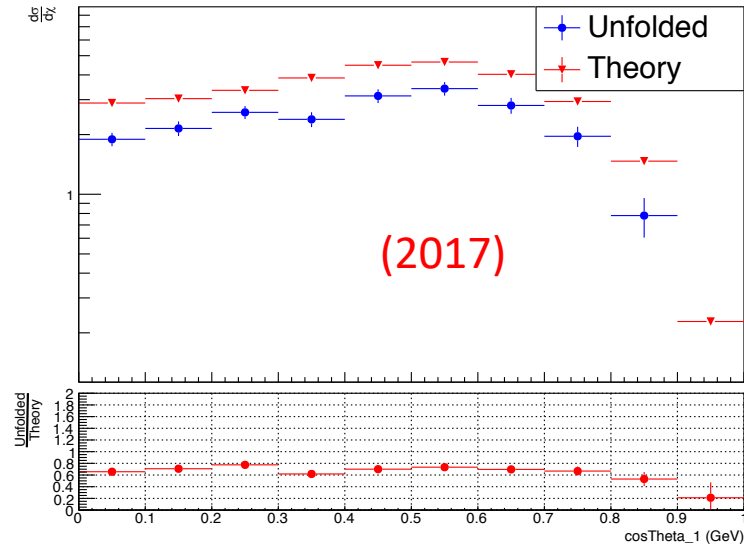


Parton Measurement

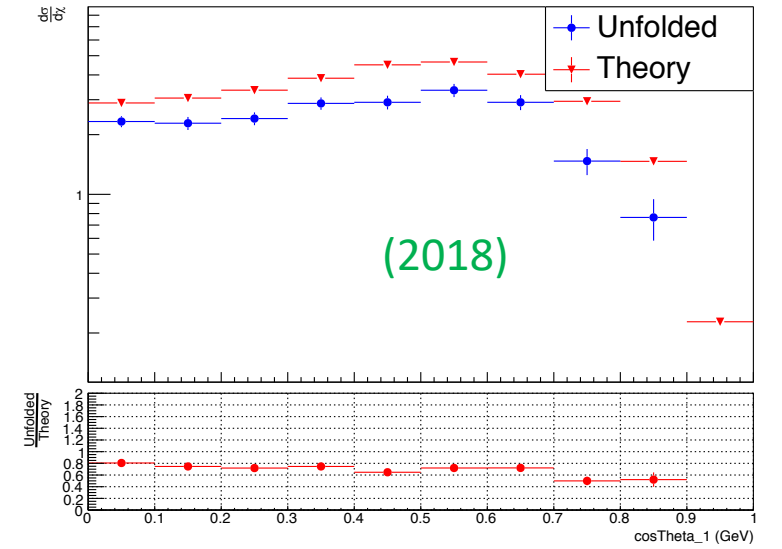
Parton Unfolded vs Theory cosTheta_1 2016



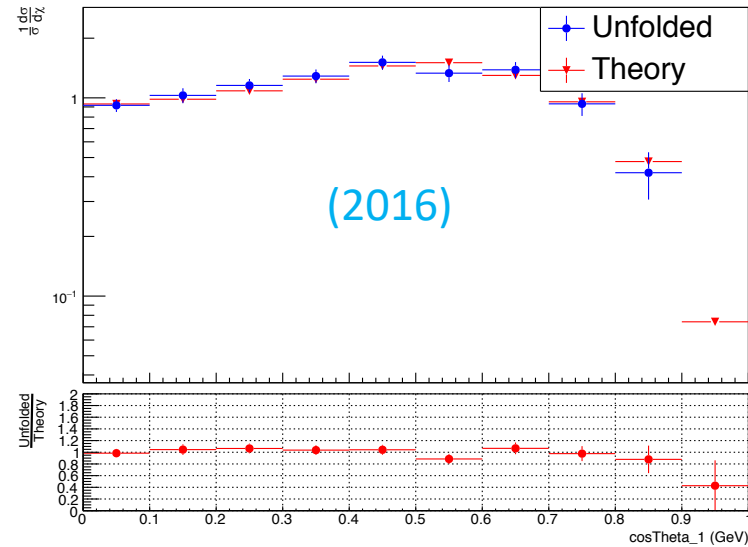
Parton Unfolded vs Theory cosTheta_1 2017



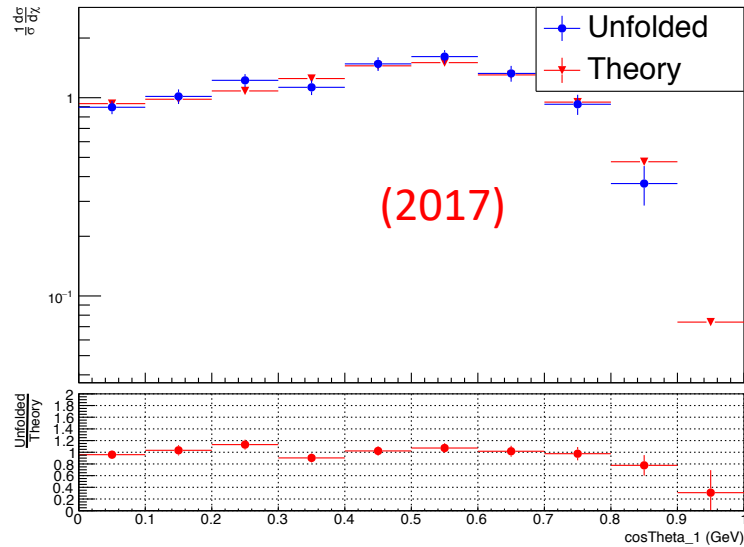
Parton Unfolded vs Theory cosTheta_1 2018



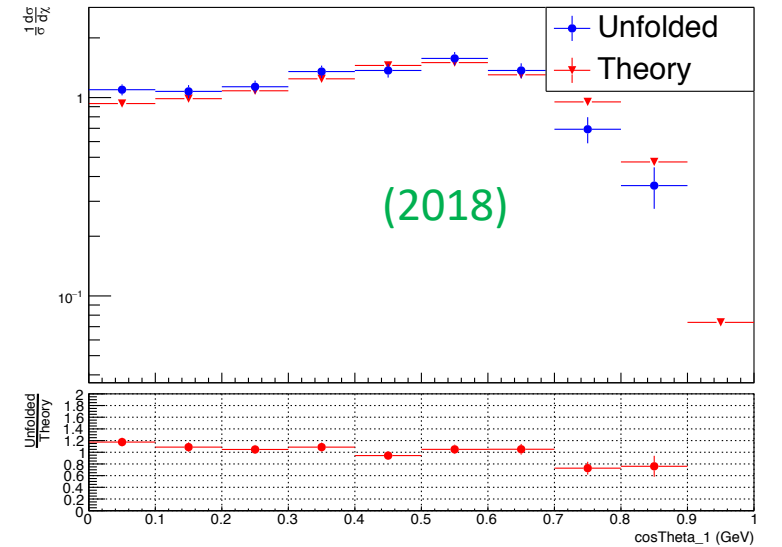
Parton Unfolded vs Theory cosTheta_1 2016



Parton Unfolded vs Theory cosTheta_1 2017



Parton Unfolded vs Theory cosTheta_1 2018



BACKUP



Signal Extraction

$$S(x_{reco}) = D(x_{reco}) - C_{bkg}^{yield} N_{QCD}^{fit} C_{QCD}^{shape}(x_{reco}) Q(x_{reco}) - B(x_{reco})$$

Diagram illustrating the components of the signal extraction equation:

- Fiducial Yield**: Points to $S(x_{reco})$
- Measured dist from data**: Points to $D(x_{reco})$
- Fitted number of QCD events in SR_A** : Points to N_{QCD}^{fit}
- QCD shape taken from Data (CR)**: Points to $C_{QCD}^{shape}(x_{reco})$
- Transfer factor from SR_A to SR**: Points to C_{bkg}^{yield}
- QCD shape correction factor**: Points to $C_{QCD}^{shape}(x_{reco})$
- Subdominant bkg shape and contribution (MC)**: Points to $B(x_{reco})$

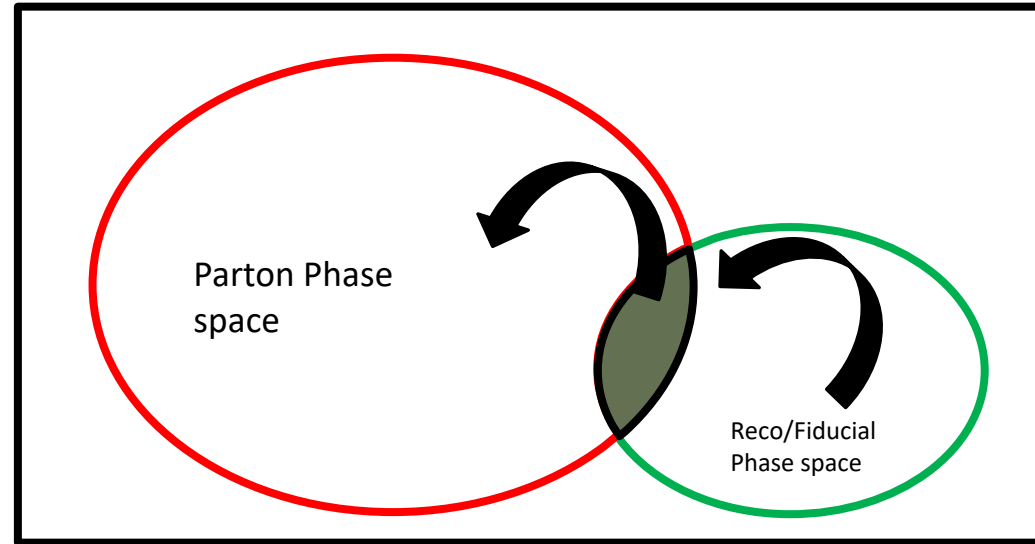
- Where x_{reco} is the respected variable of interest (ttbar mass, pt, rapidity, leading and subleading jetPt and |jetY|)
 - We deploy a fit in the Signal Region (2btag) to extract the N_{QCD}^{fit}
- $$D(m^t)^{(i)} = N_{tt}^{(i)} T^{(i)}(m^t, k_{MassScale}, k_{MassResolution}) + N_{bkg}^{(i)} B(m^t)(1 + k_1 x) + N_{sub}^{(i)} O^{(i)}(m^t)$$
- Our data CR is contaminated from ttbar and subdominant bkg which has to be dealt with.



Parton & Particle levels

Parton

Observable	Requirement
$p_T^{t,\bar{t}}$	$> 400 \text{ GeV}$
$ \eta^{t,\bar{t}} $	< 2.4
$m_{t\bar{t}}$	$> 1000 \text{ GeV}$



Particle level Top Candidates

Observable	Requirement
N_{jets}	> 1
$p_T^{jet1,2}$	$> 400 \text{ GeV}$
$ \eta^{jet1,2} $	< 2.4
$m_{SD}^{jet1,2}$	$(120, 220) \text{ GeV}$
m_{jj}	$> 1000 \text{ GeV}$

$$\frac{d\sigma_i^{\text{unf}}}{dx} = \frac{1}{\mathcal{L} \cdot \Delta x_i} \cdot \frac{1}{f_{2,i}} \cdot \sum_j \left(R_{ij}^{-1} \cdot f_{1,j} \cdot S_j \right)$$

efficiency of the reco+true selection

migration matrix

reco efficiency of the reco+true selection

Unfolding: simple response matrix inversion w/o regularisation



Tag And Probe

- Top Tagger Scale Factors

- Validation method to ensure that no SF's are needed
- From data we subtract QCD and Subdominant bkg (MC) so that the data sample is pure

$$efficiency = \frac{\# (1 \text{ jet pass baseline} + \textit{Tight TopTagger Cut AND 1 jet pass SR})}{\# (1 \text{ jet pass baseline} + \textit{Tight TopTagger Cut AND 1 jet pass only baseline})}$$

- Randomization: Randomly select leading/subleading jet to use as tag or probe to avoid pT bias
- Divide the phase space into pT regions based on the topTagger categories: [400-600] GeV, [600-800] GeV, [800-Inf] GeV



Year	Type of File	DAS
2016	TT Mtt 700-1000	/TT_Mtt-700to1000_TuneCUETP8M2T4_13TeV-powheg-pythia8/RunIISummer16MiniAODv3-PUMoriond17_94X_mcRun2_asymptotic_v3-v2/MINIAODSIM
	TT Mtt 1000-Inf	/TT_Mtt-1000toInf_TuneCUETP8M2T4_13TeV-powheg-pythia8/RunIISummer16MiniAODv3-PUMoriond17_94X_mcRun2_asymptotic_v3-v2/MINIAODSIM
	TT Nominal	/TT_TuneCUETP8M2T4_13TeV-powheg-pythia8/RunIISummer16MiniAODv3-PUMoriond17_94X_mcRun2_asymptotic_v3-v1/MINIAODSIM
2017	TT Mtt 700-1000	--
	TT Mtt 1000-Inf	--
	TT Nominal Hadronic	/TTToHadronic_TuneCP5_13TeV-powheg-pythia8/RunIIFall17MiniAODv2-PU2017_12Apr2018_94X_mc2017_realistic_v14-v1/MINIAODSIM
	TT Nominal Semilepton	/TTToSemiLeptonic_TuneCP5_13TeV-powheg-pythia8/RunIIFall17MiniAODv2-PU2017_12Apr2018_94X_mc2017_realistic_v14-v2/MINIAODSIM
	TT Nominal Dilepton	/TTTo2L2Nu_TuneCP5_13TeV-powheg-pythia8/RunIIFall17MiniAODv2-PU2017_12Apr2018_94X_mc2017_realistic_v14-v2/MINIAODSIM
2018	TT Mtt 700-1000	--
	TT Mtt 1000-Inf	--
	TT Nominal Hadronic	/TTToHadronic_TuneCP5_13TeV-powheg-pythia8/RunIIAutumn18MiniAOD-102X_upgrade2018_realistic_v15-v1/MINIAODSIM
	TT Nominal Semilepton	/TTToSemiLeptonic_TuneCP5_13TeV-powheg-pythia8/RunIIAutumn18MiniAOD-102X_upgrade2018_realistic_v15-v1/MINIAODSIM
	TT Nominal Dilepton	/TTTo2L2Nu_TuneCP5_13TeV-powheg-pythia8/RunIIAutumn18MiniAOD-102X_upgrade2018_realistic_v15-v1/MINIAODSIM

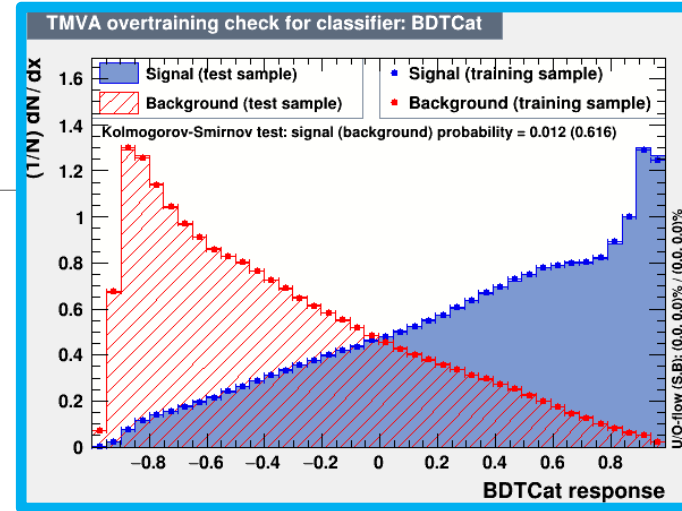


Overview: Discriminator, Efficiency and Acceptance

The discriminator is a BDT trained individually for 2016, 2017 and 2018

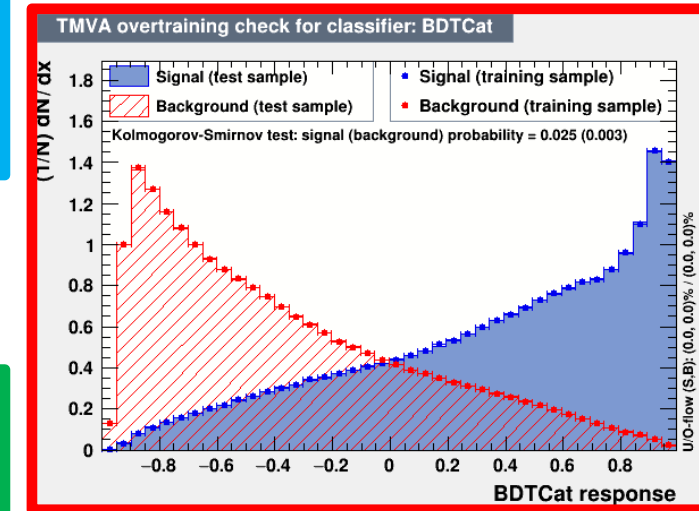
Category training: split the sample in categories based on Pt

- Bins:
 - [400, 600] GeV
 - [600, 800] GeV
 - [800, 1200] GeV
 - [1200, inf] GeV
- BDT, used variables:
 - Leading and Sub-leading subjet mass
 - N-Subjetiness variables (tau1, tau2, tau3)
 - fraction of the jetPt over the total pt sum of the event.
 - Energy correlation functions (ecfB1N2,ecfB1N3,ecfB2N2, ecfB2N3)
- BDT Output consistency for the 3 years
- Calculation of Efficiency and acceptance for each year
 - We choose the WP's for each year so that the leading jet p_T efficiency is similar for all years

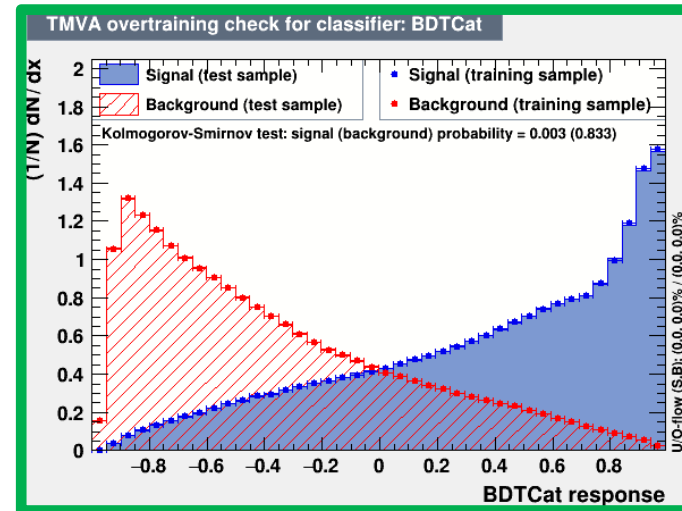


2016

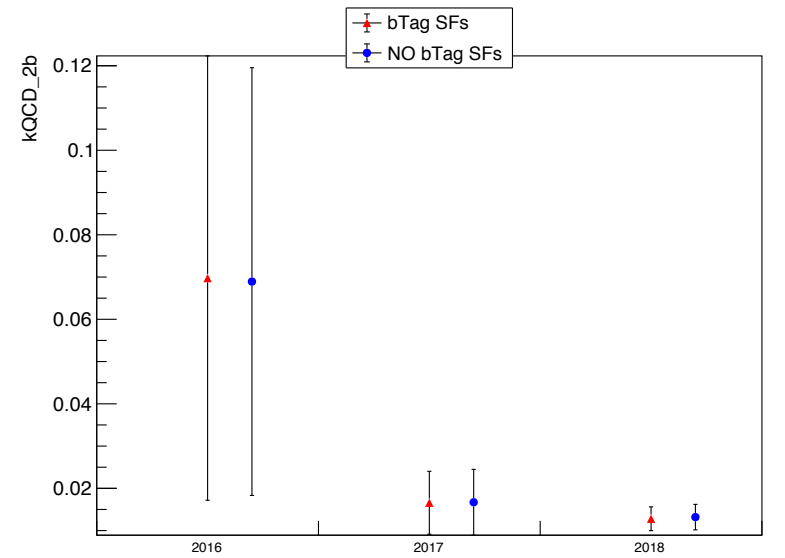
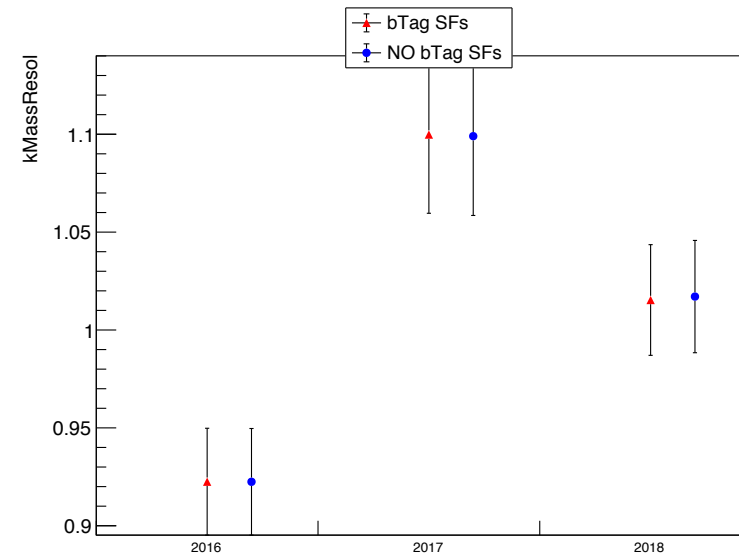
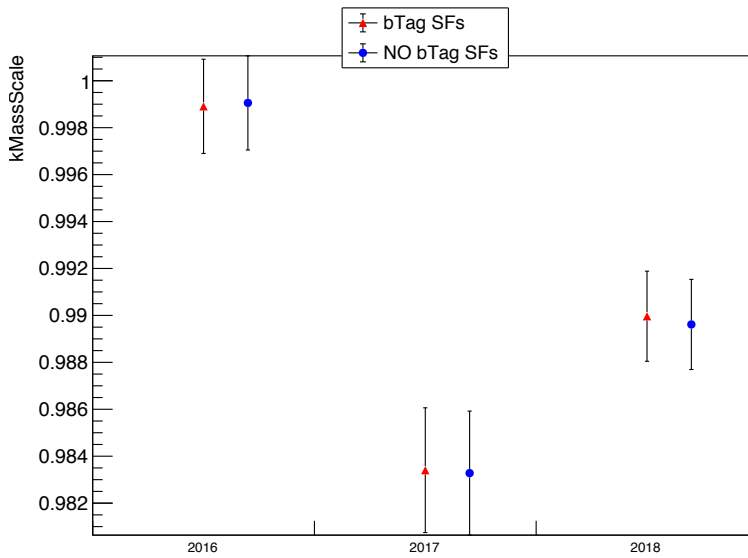
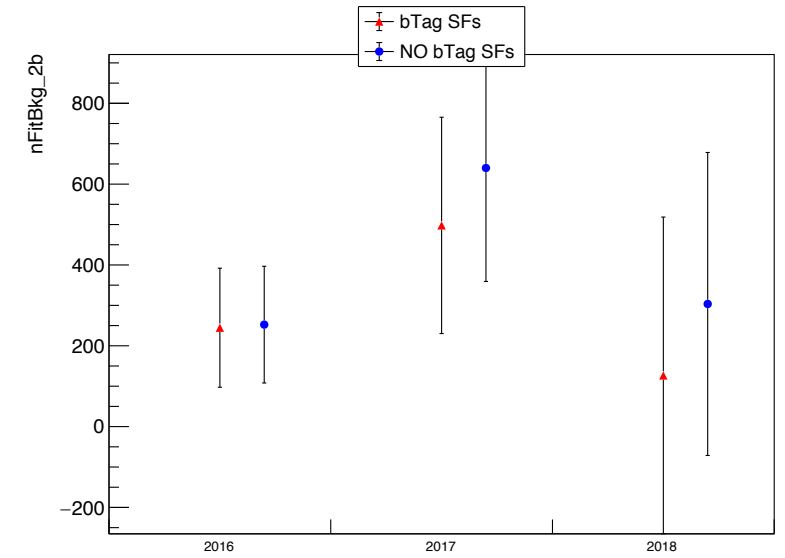
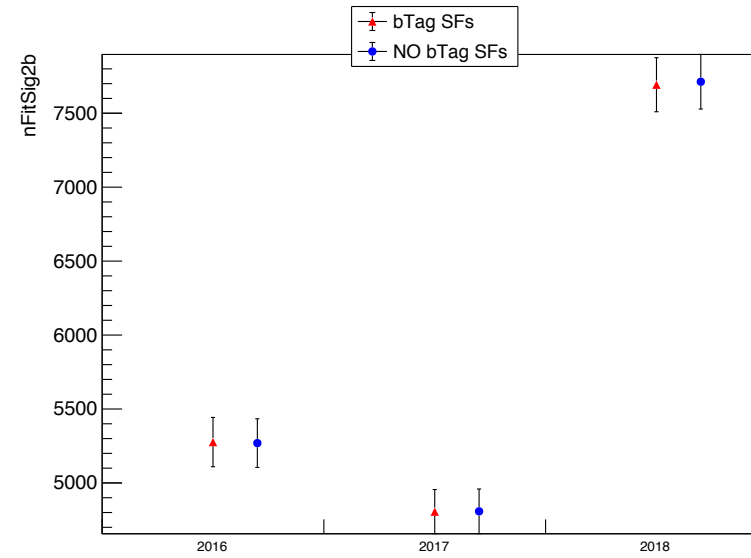
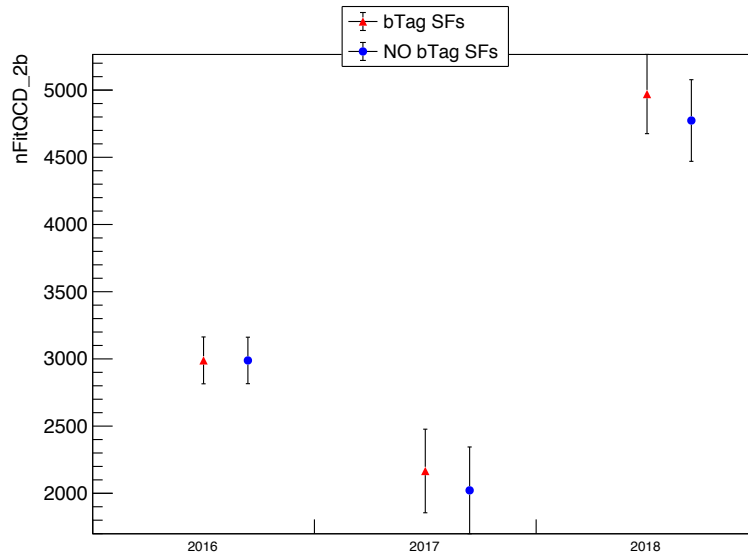
2017



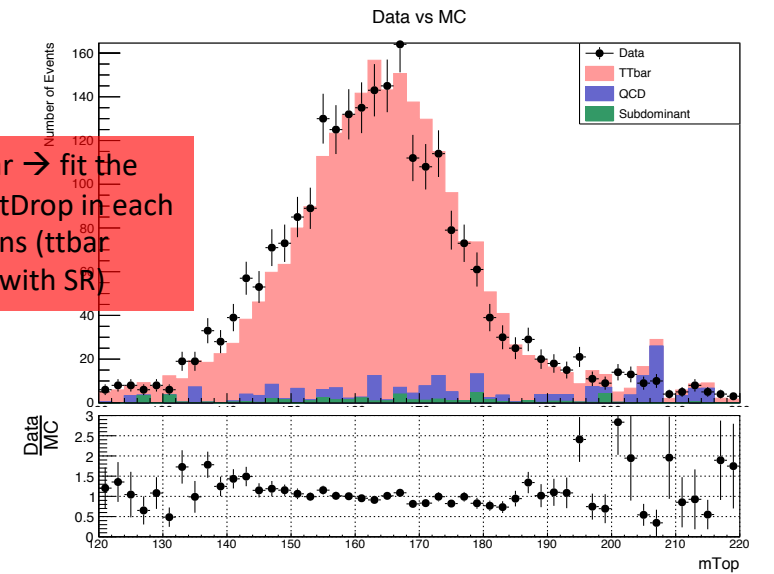
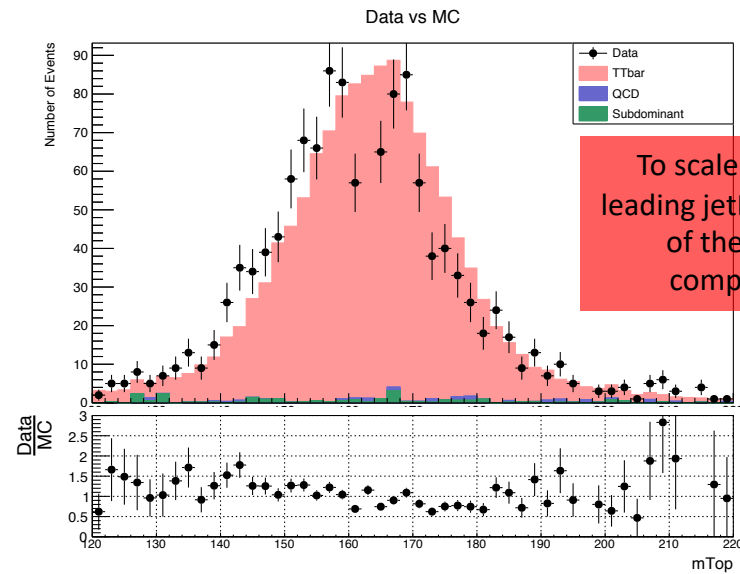
2018



Fit Params Results Comparison



TagAndProbe Efficiency Plots



Tight TopTagger + Probe

