

$B^0 \rightarrow \pi^0 \pi^0$ analysis

Sebastiano

Overview

BF and A_{CP} of $B^0 \rightarrow \pi^0\pi^0$ decays: fundamental measurements at Belle II.

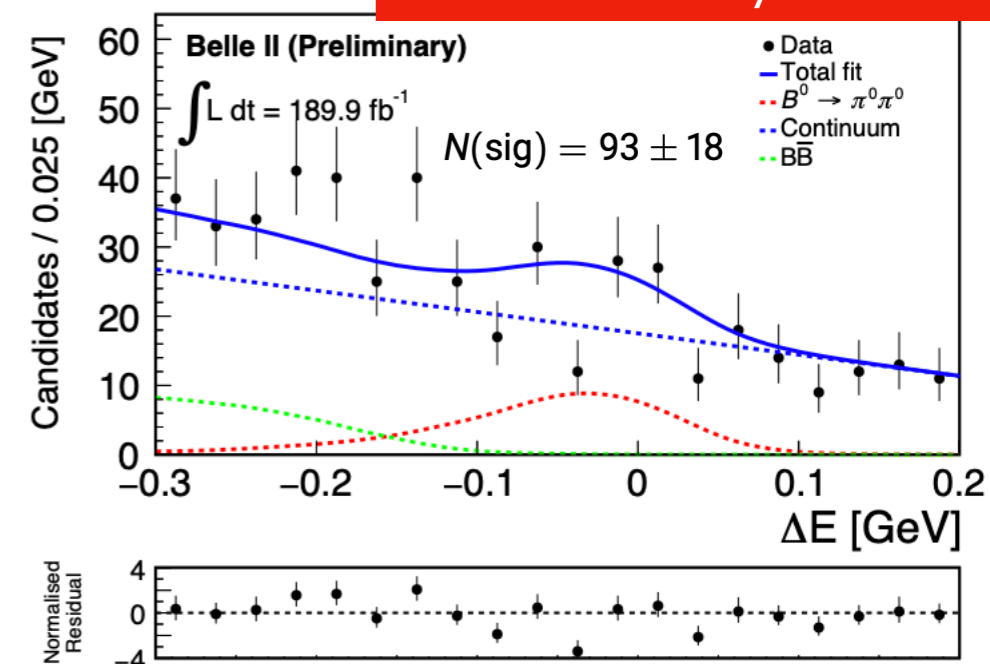
Results (@189.9fb⁻¹) by Francis shown at ICHEP2022.

Now: prepare new analysis for pre-LS1 dataset.

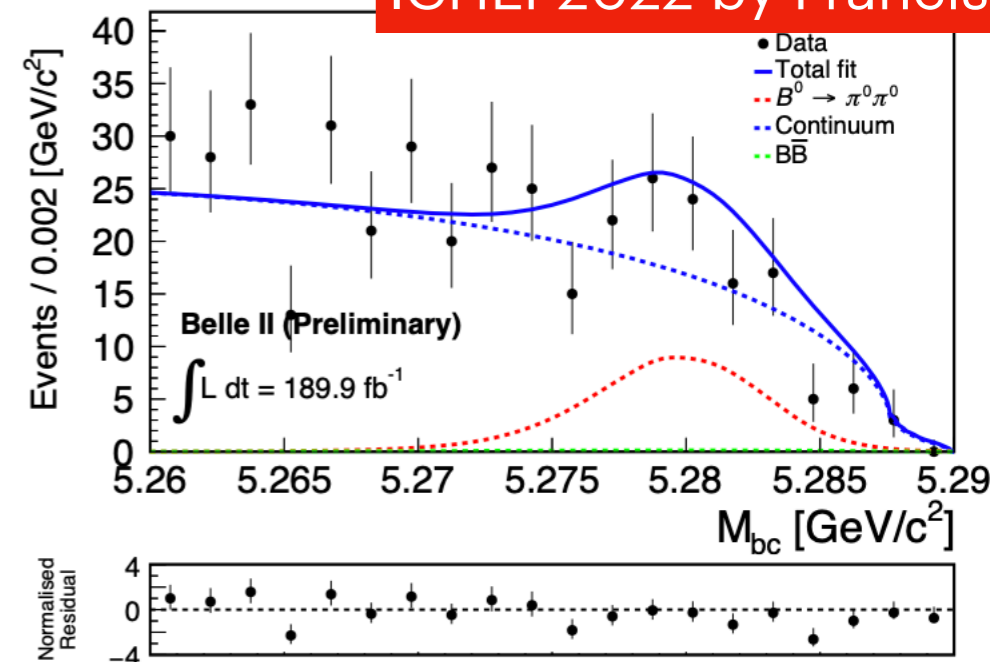
Plan:

- revisit photonMVA looking at variables with good data/MC agreement
- revisit CSBDT adding BTag variables to suppress even more $e^+e^- \rightarrow q\bar{q}$
- Introduce specific BDT trained against continuum ρ 's

ICHEP2022 by Francis



ICHEP2022 by Francis



$$A^{CP} = -0.14 \pm 0.46 \text{ (stat.)} \pm 0.07 \text{ (syst.)}$$

$$B = (1.27 \pm 0.25 \text{ (stat.)} \pm 0.17 \text{ (syst.)}) \cdot 10^{-6}$$

Photon MVA

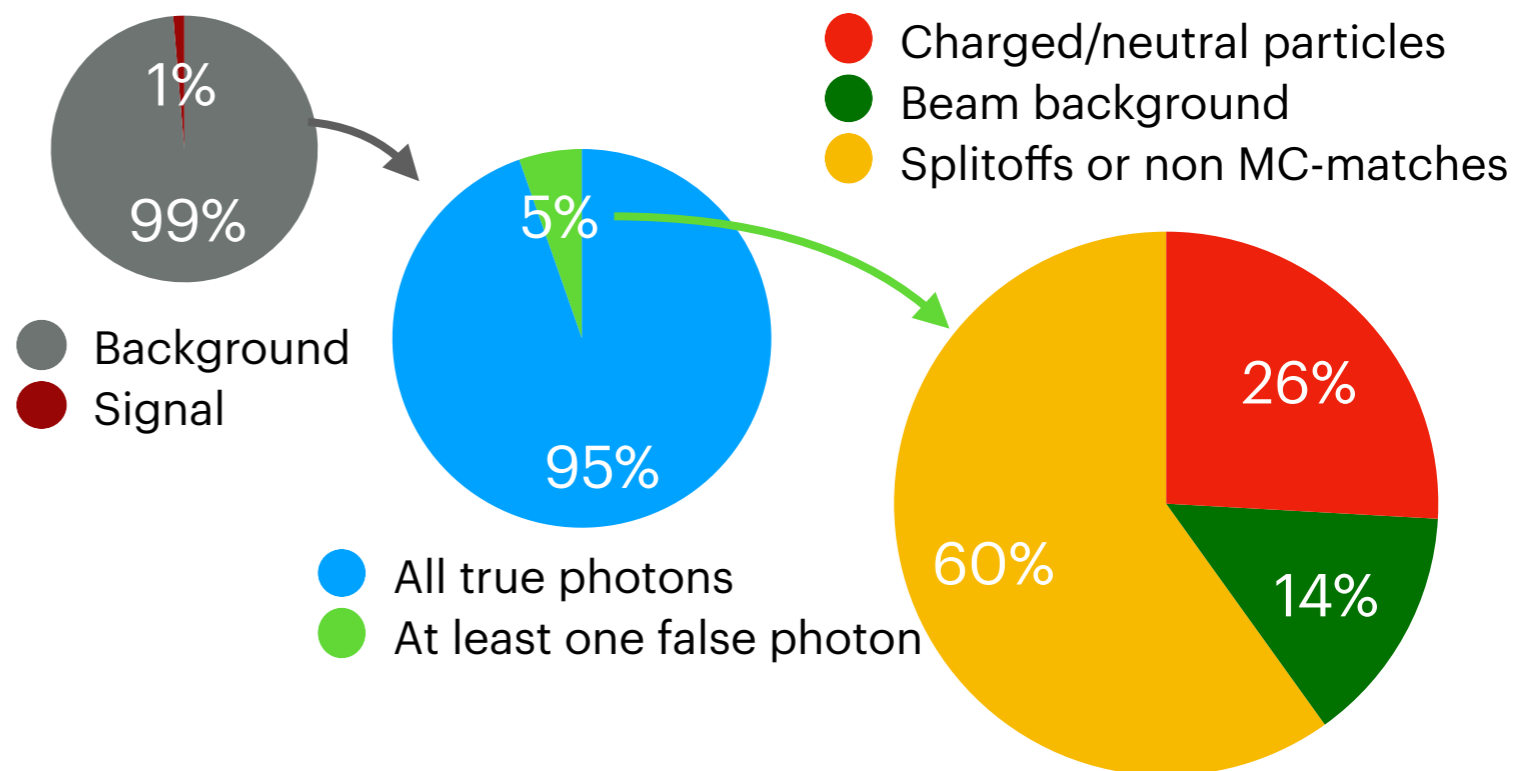
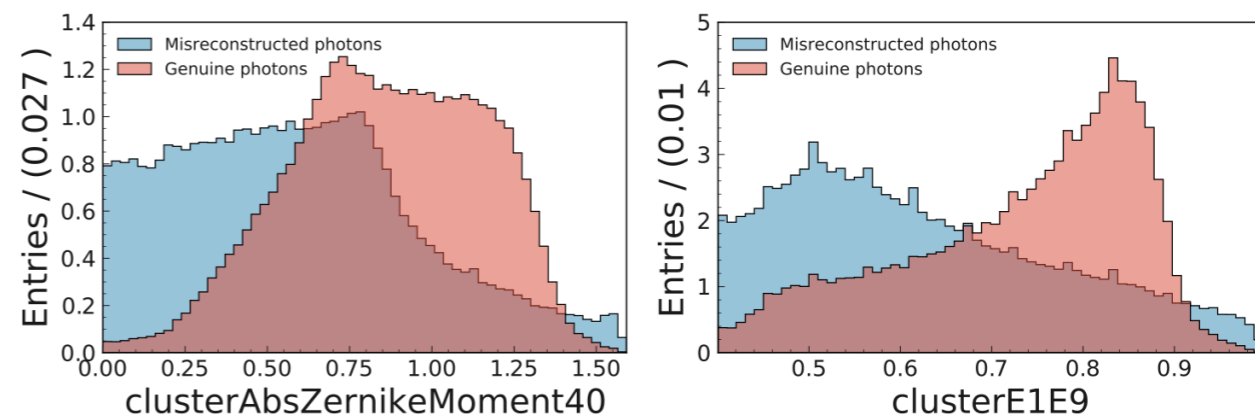
Photon MVA

Distinguish between real photons and “false” photons: beam backgrounds, other particles, energy releases from other particles (split-offs)....

Combine highly-discriminant cluster- and photon-variables in a MVA.

False photons have usually low energies, while $B^0 \rightarrow \pi^0\pi^0$ photons high-energy.

After the default selection on photons and π^0 's, the residual bkg is mainly composed by true combinatorial π^0 's.



Photon MVA: inputs validation

Ideally we need a sample of true photons and a sample of false photons (difficult to obtain).

Use **inclusive sample of photons from $D^* \rightarrow D^0(K\pi\pi^0)\pi$ decays**: apply same π^0 selections of my analysis \rightarrow same π^0 kinematic distributions.

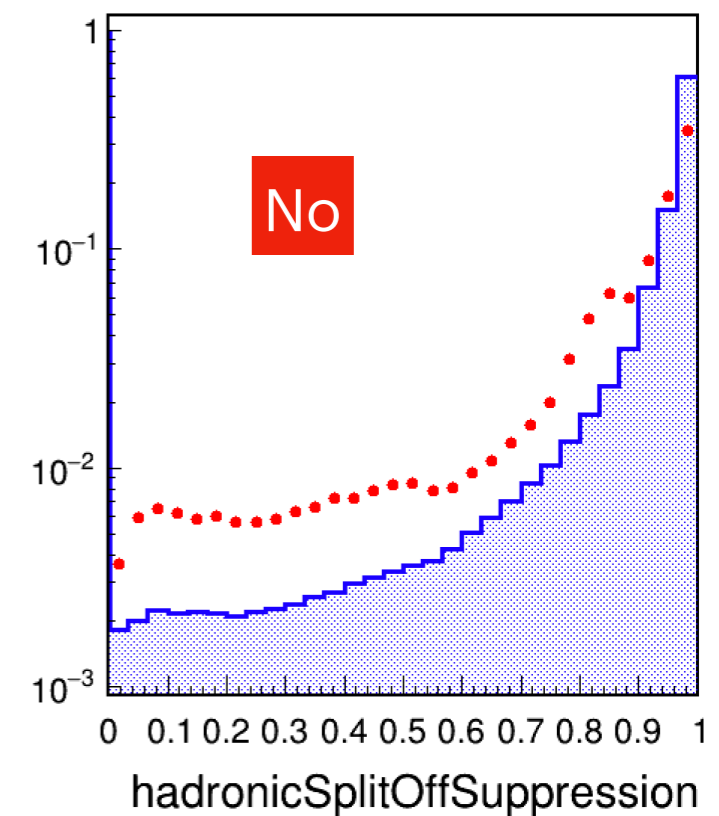
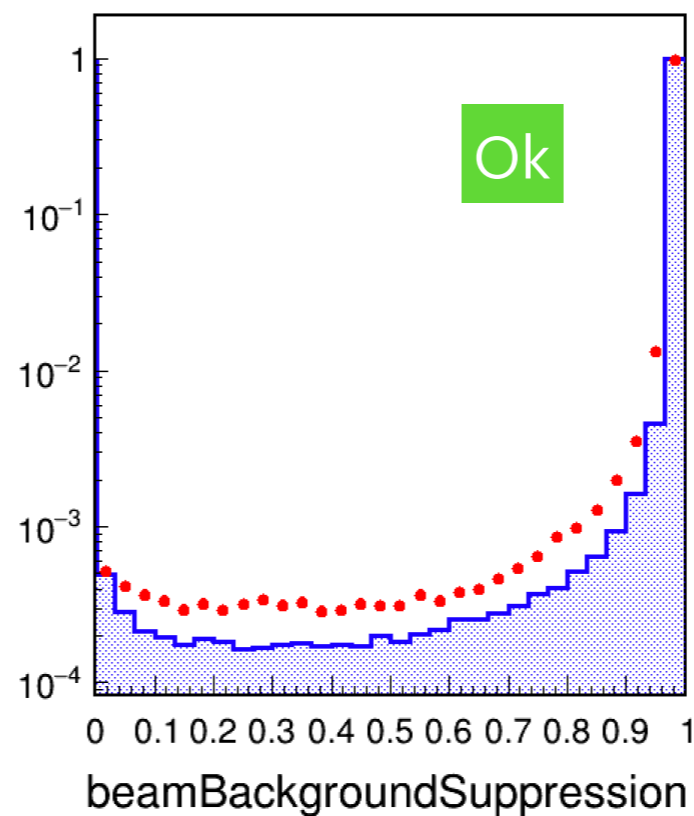
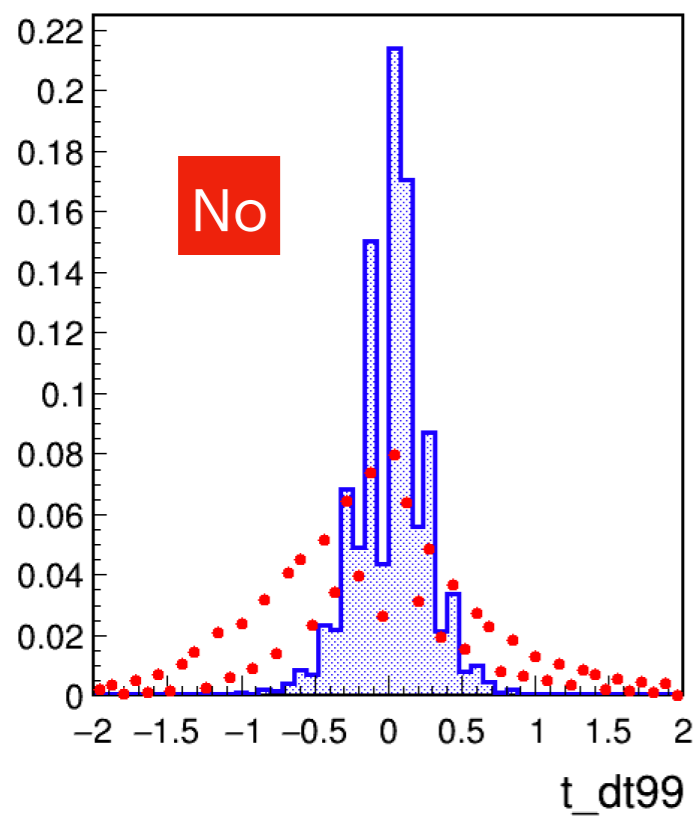
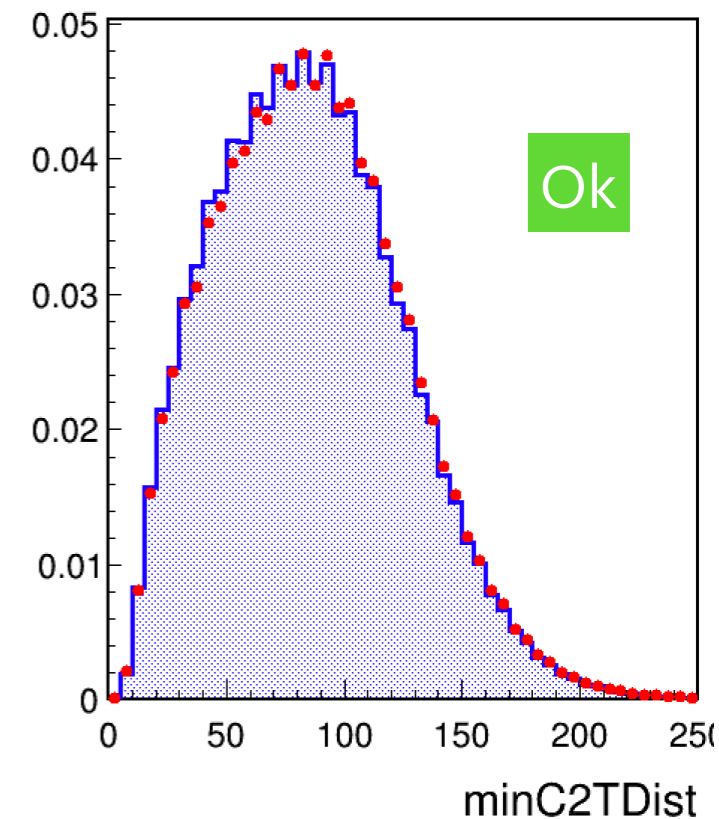
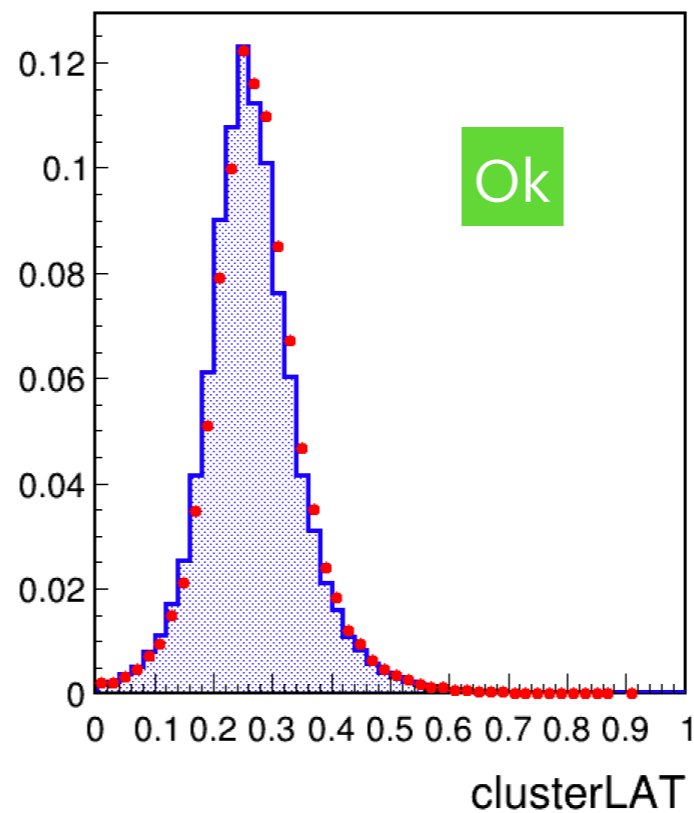
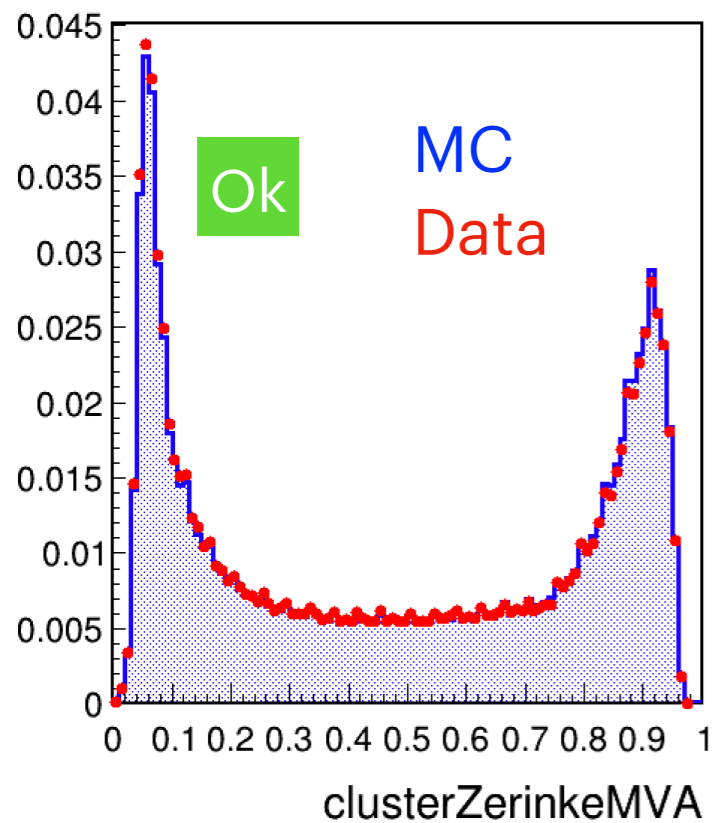
Sample is signal dominated \rightarrow ~all true photons (as in $B^0 \rightarrow \pi^0\pi^0$).

Compare input distributions using MC14rd (1 ab⁻¹)/Proc12+AllBuckets(189 fb⁻¹) and MC15ri (200 fb⁻¹)/Proc13c1(8 fb⁻¹).

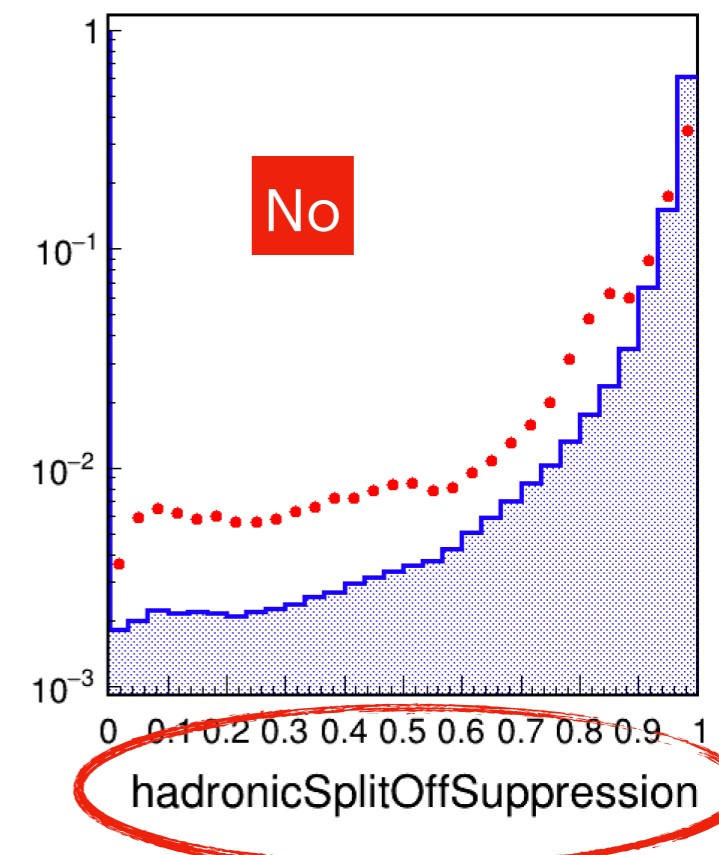
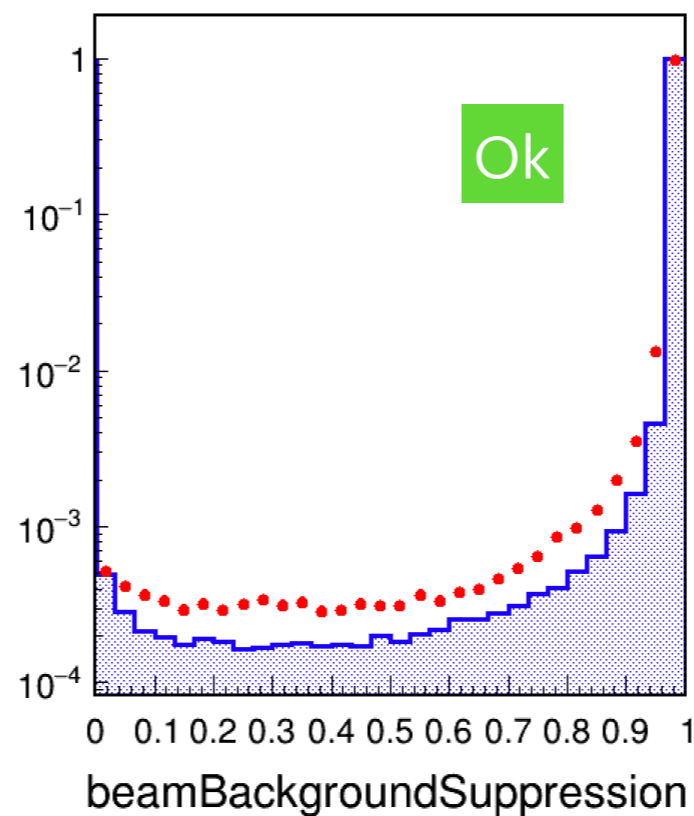
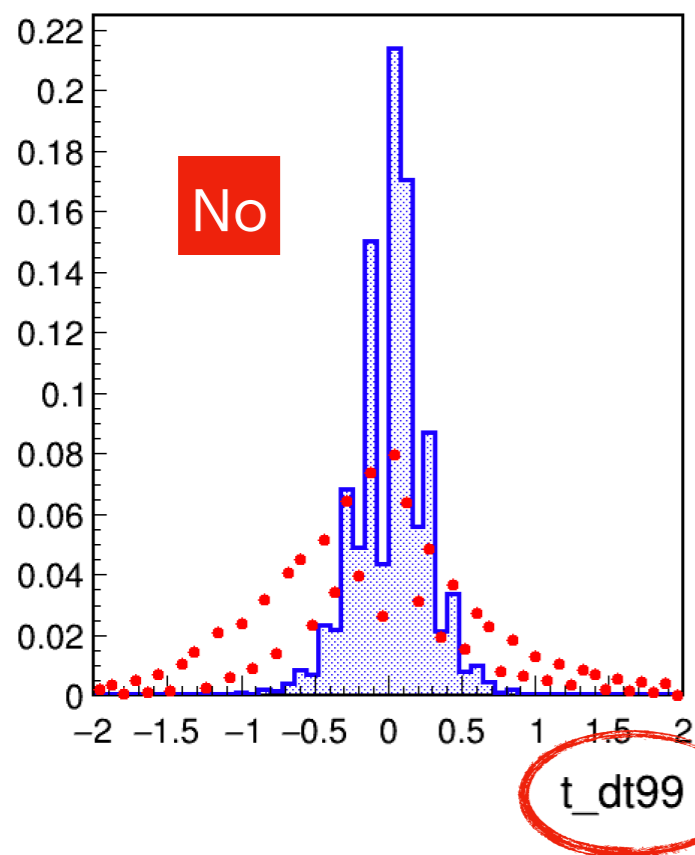
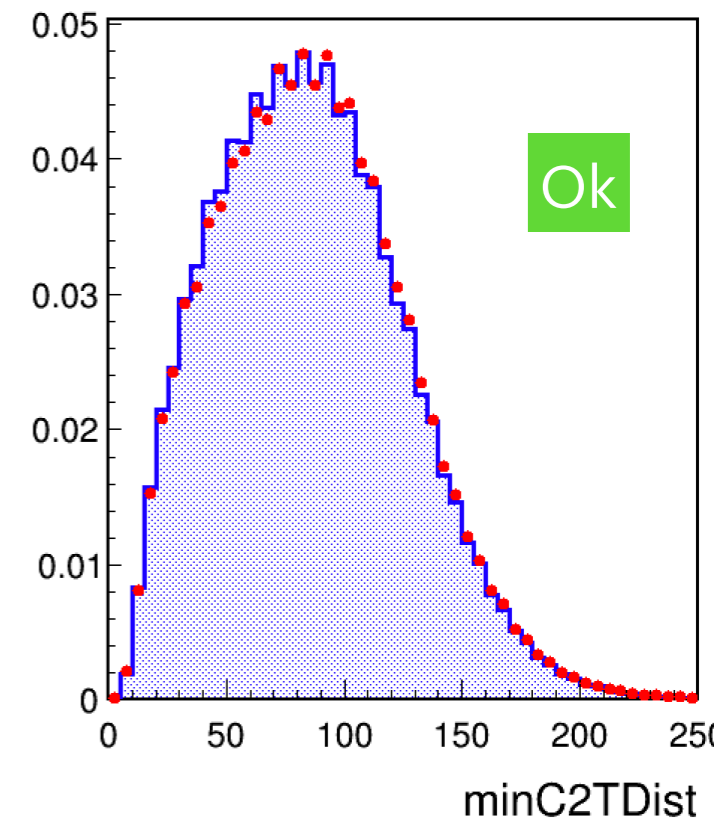
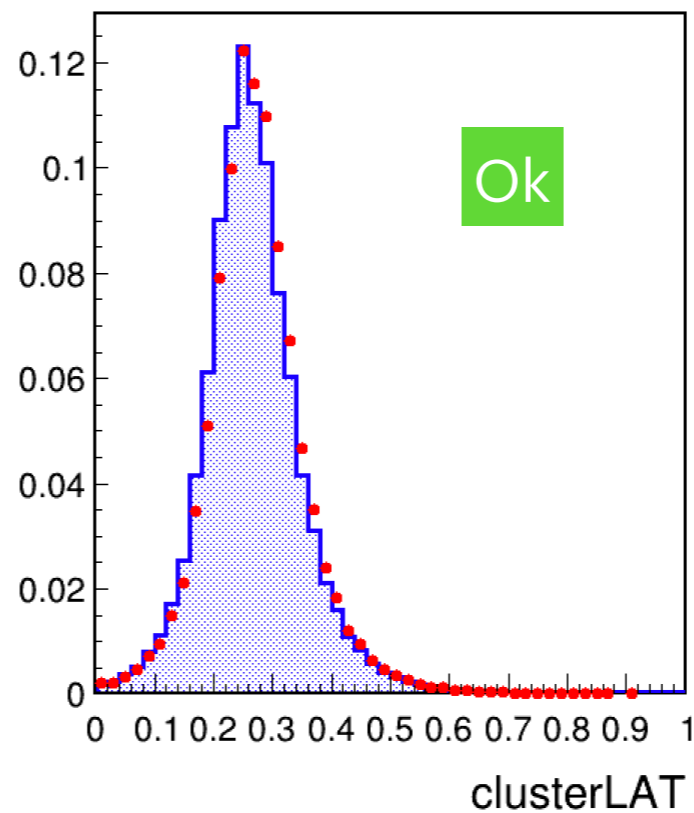
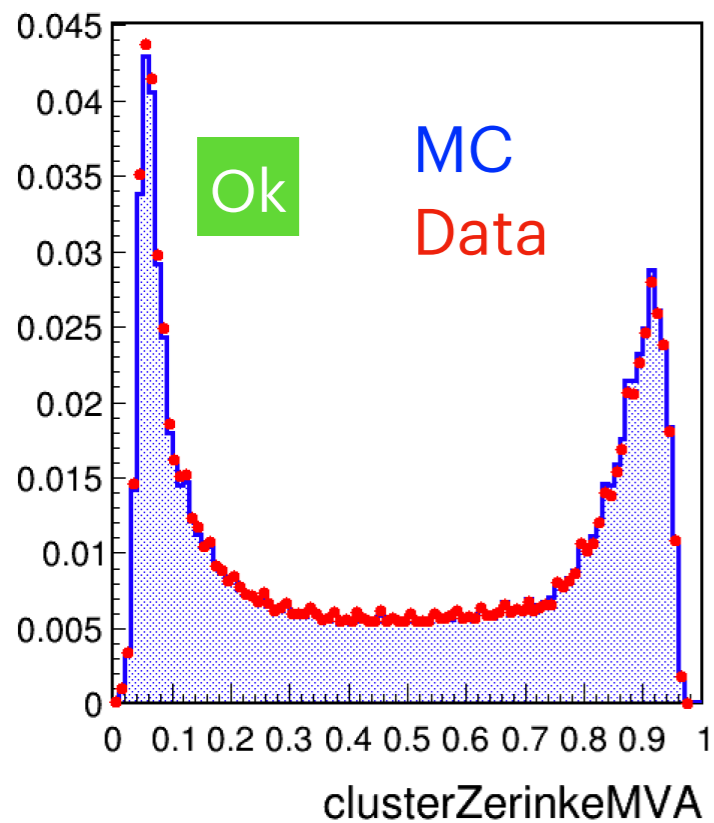
MC14 vs Proc12+AllBuckets

Release-05

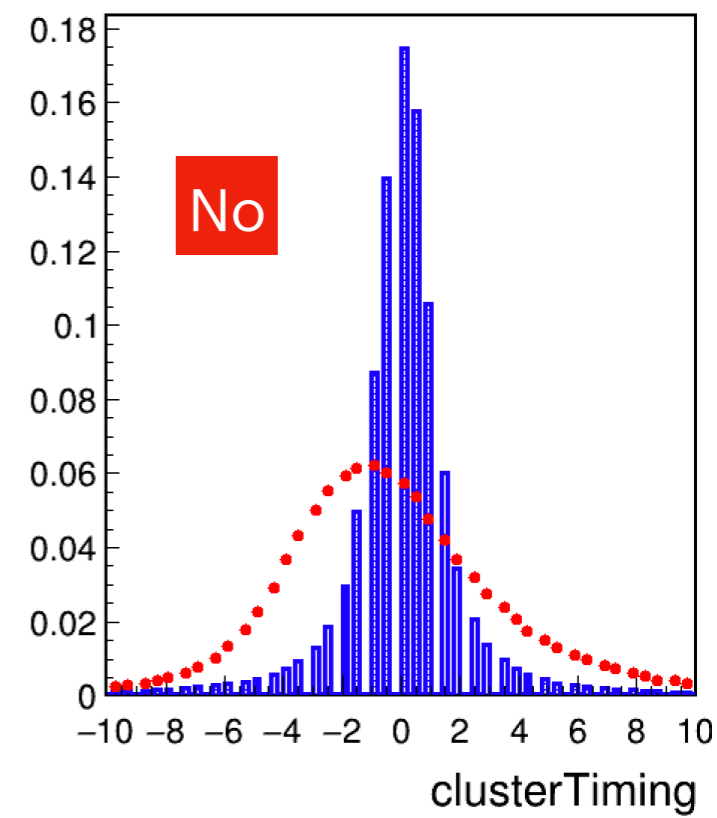
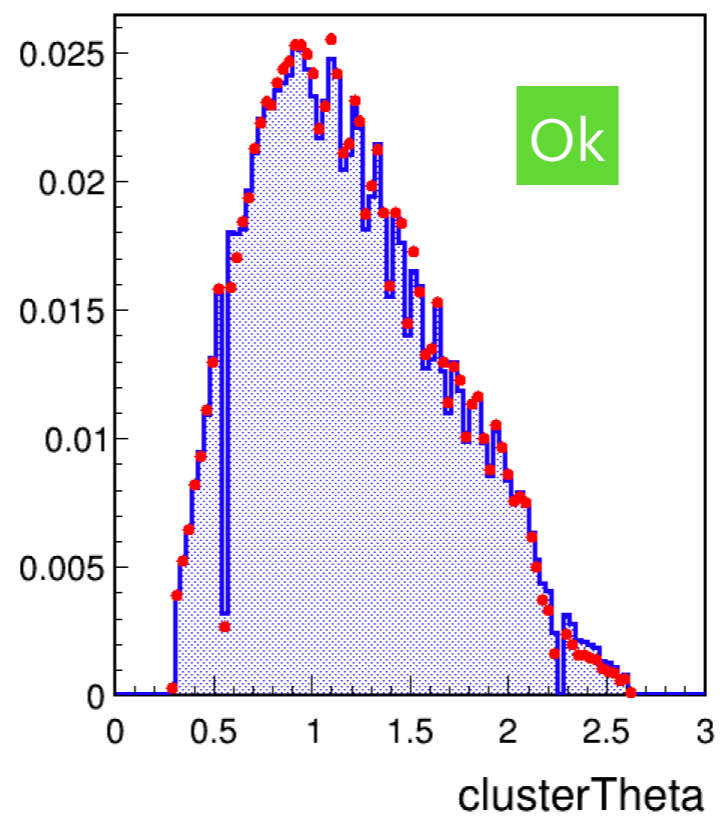
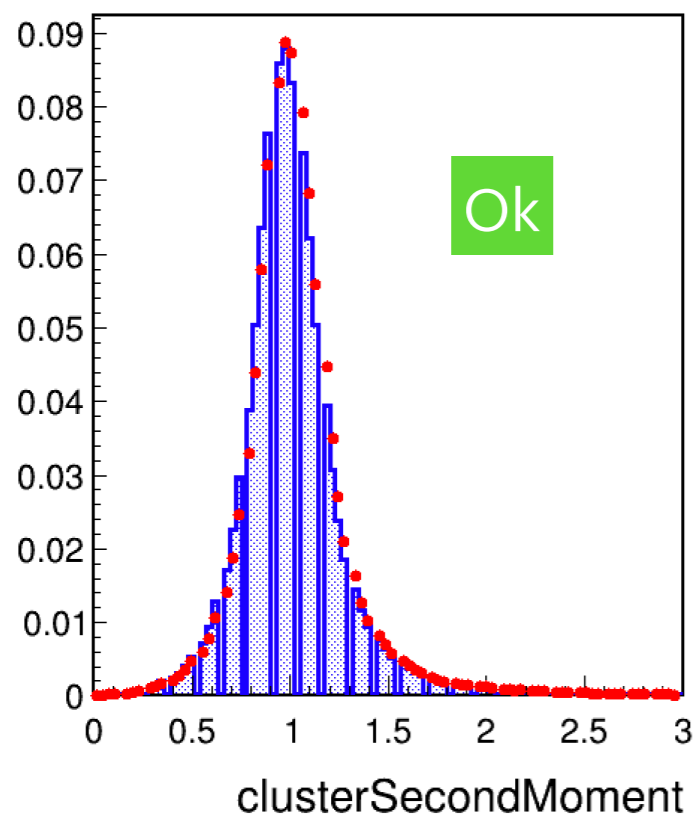
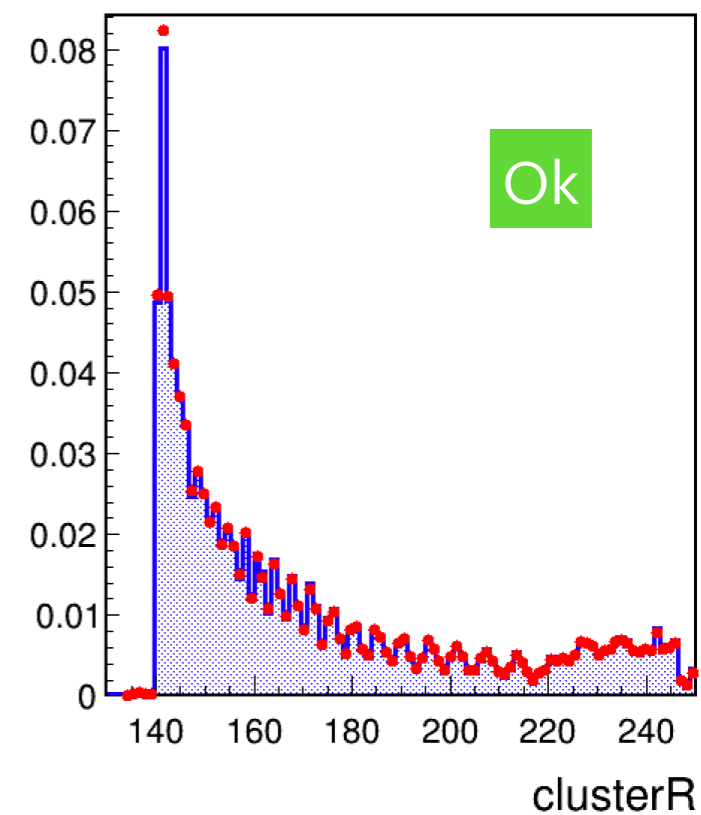
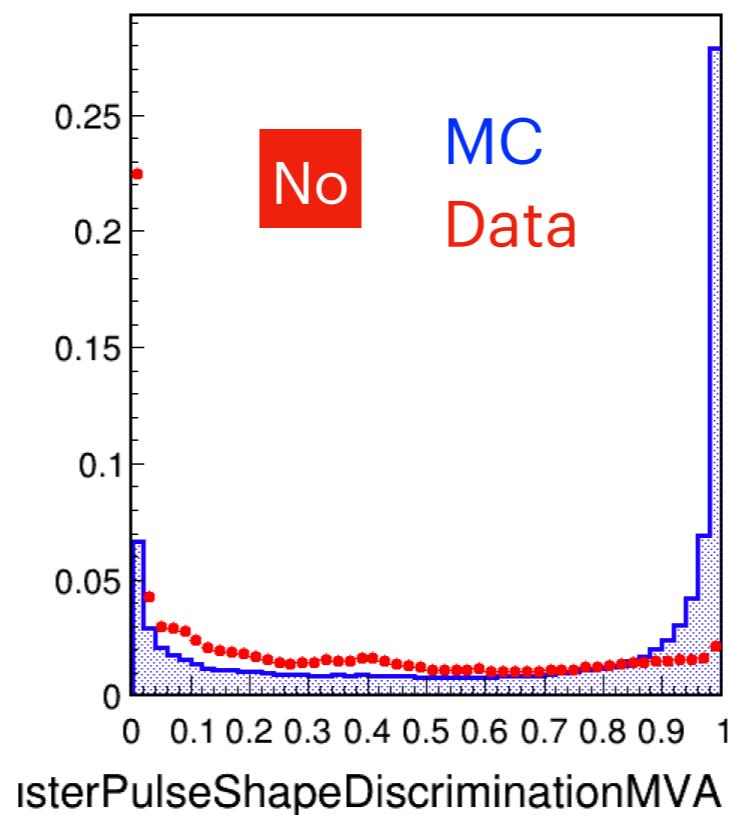
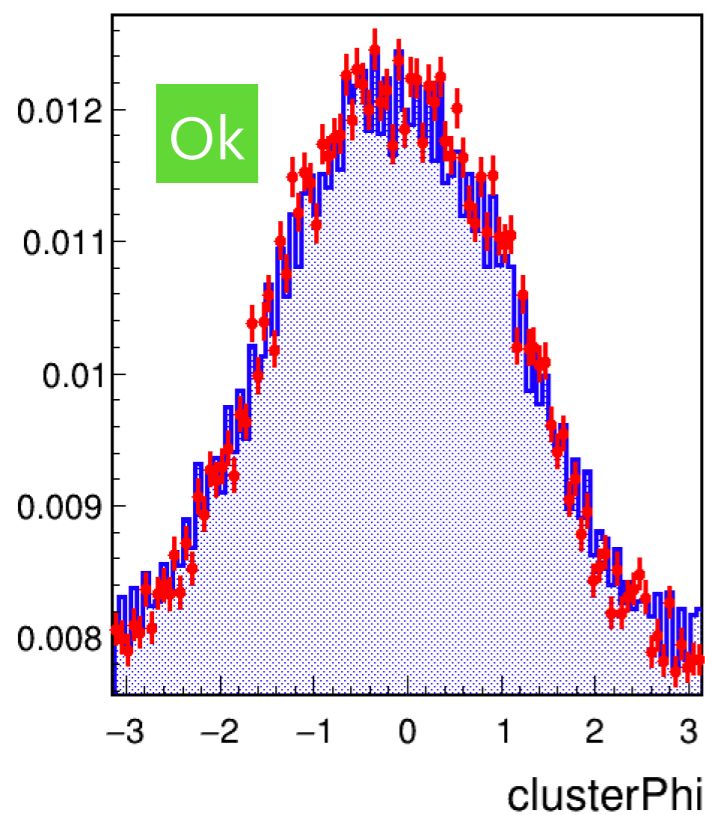
Photon MVA: inputs validation (rel-05)



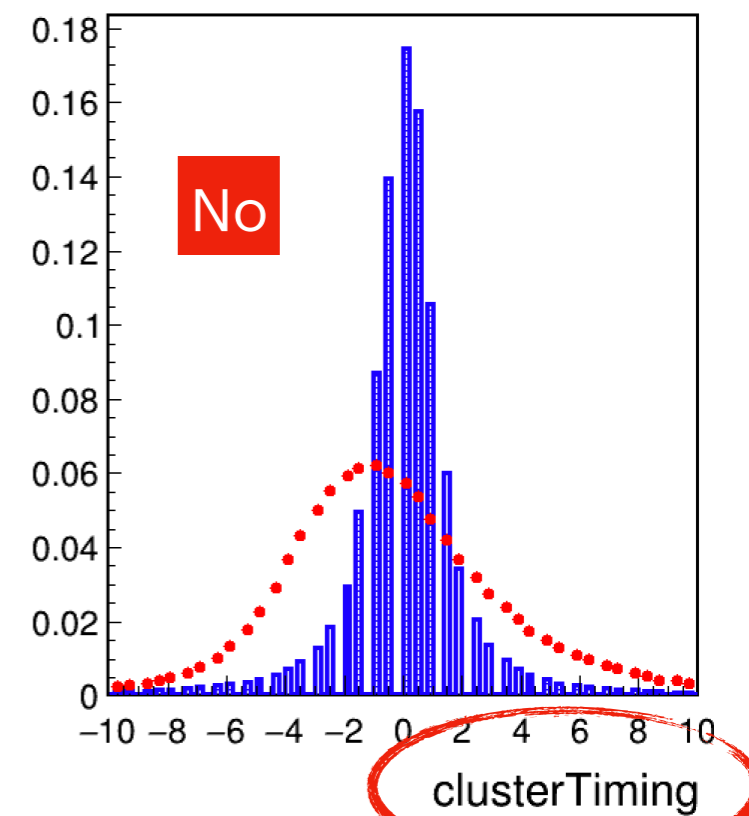
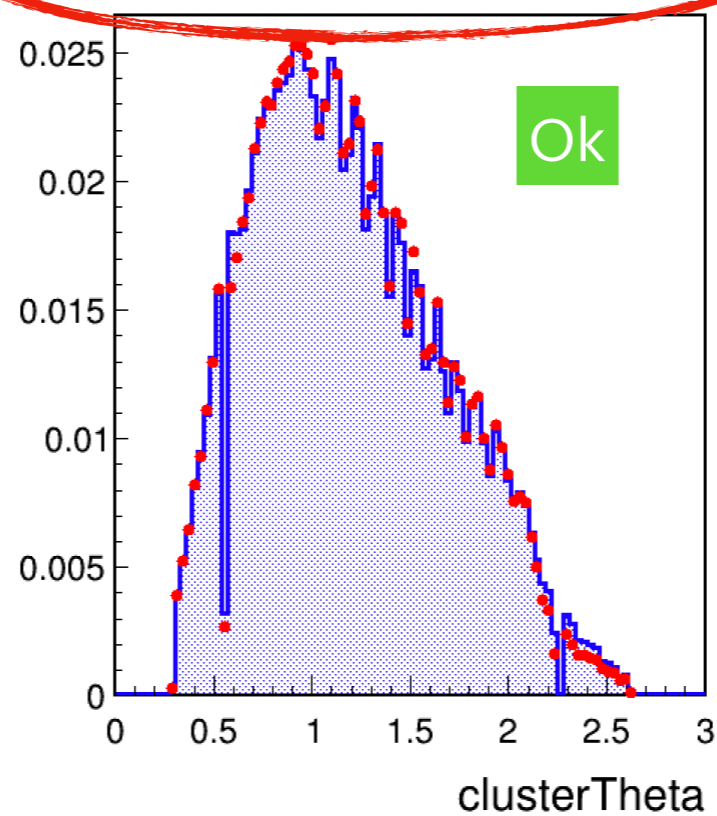
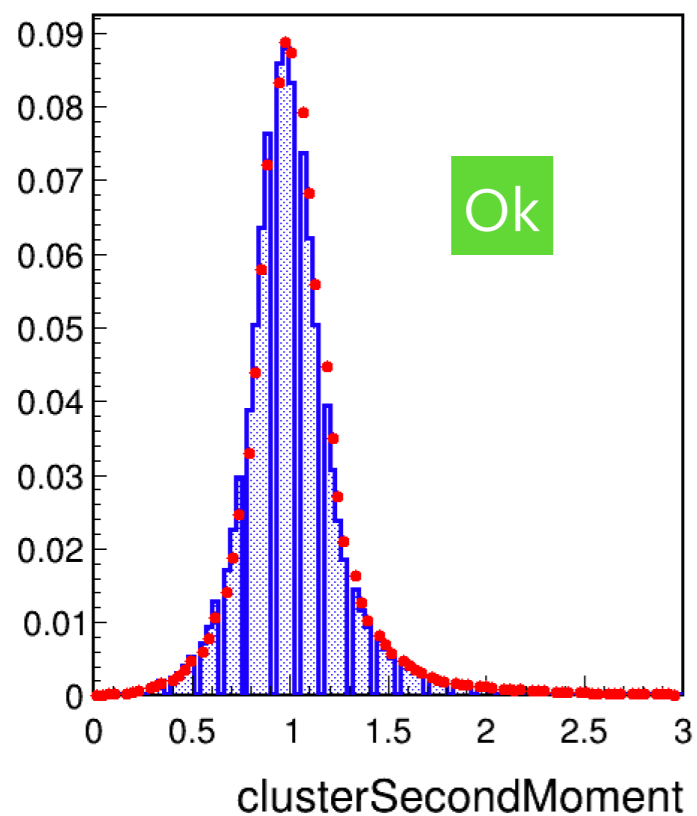
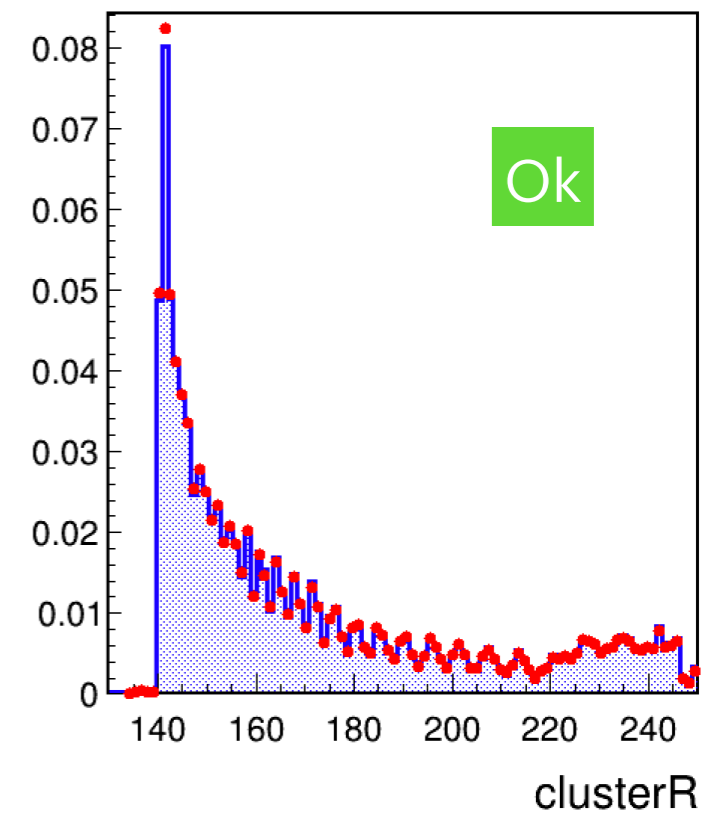
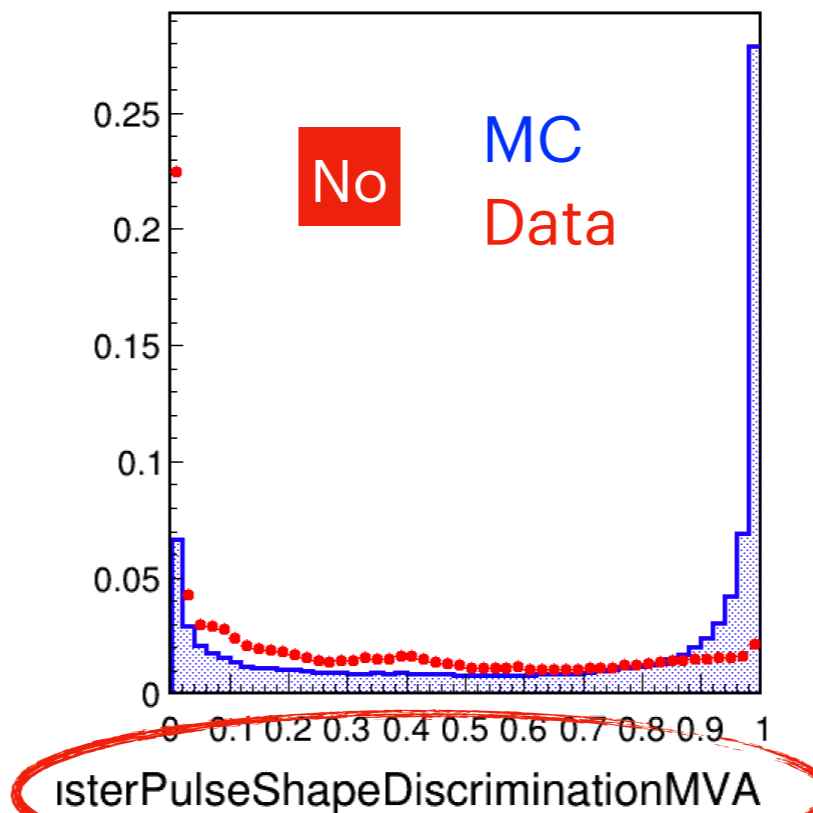
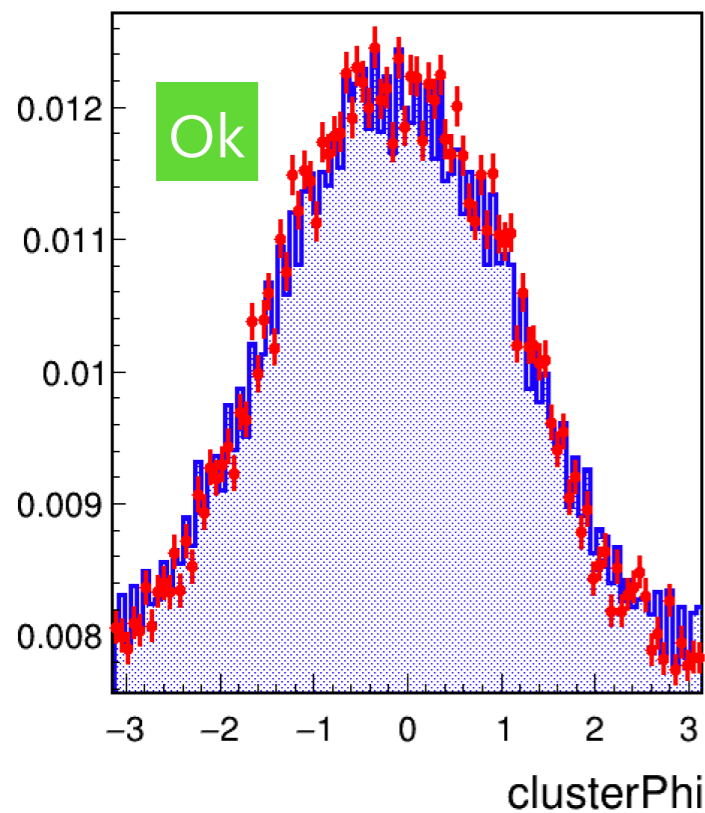
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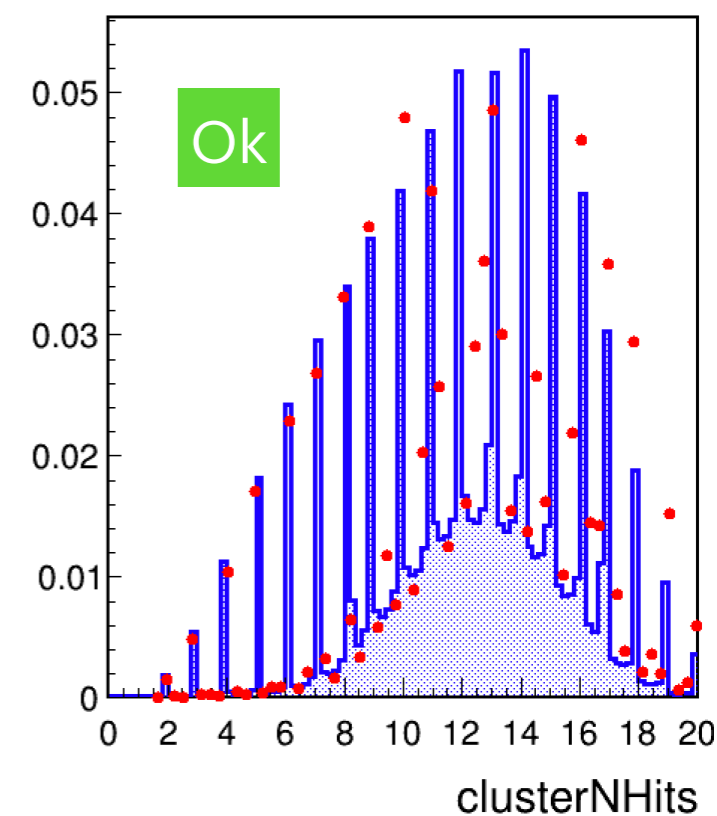
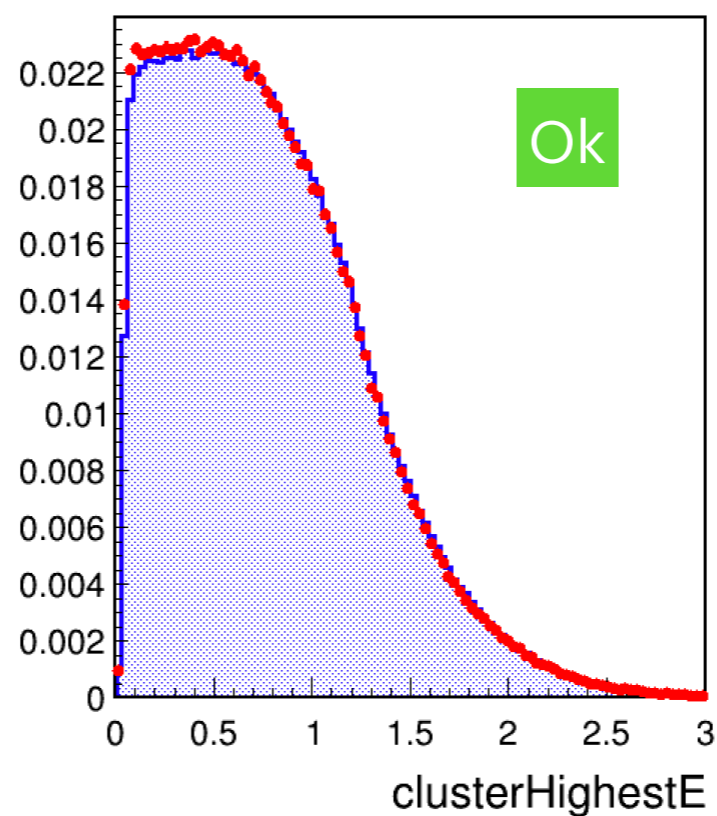
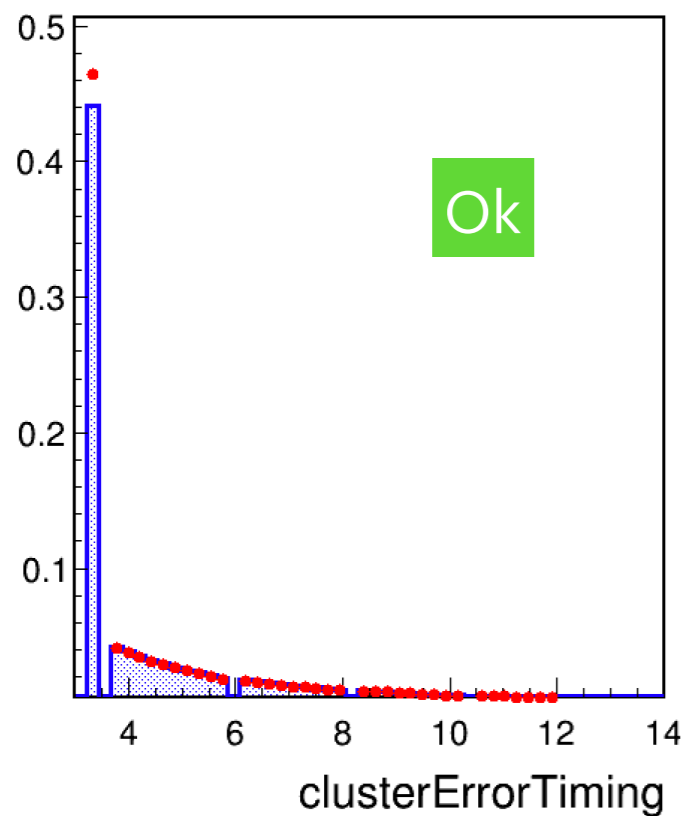
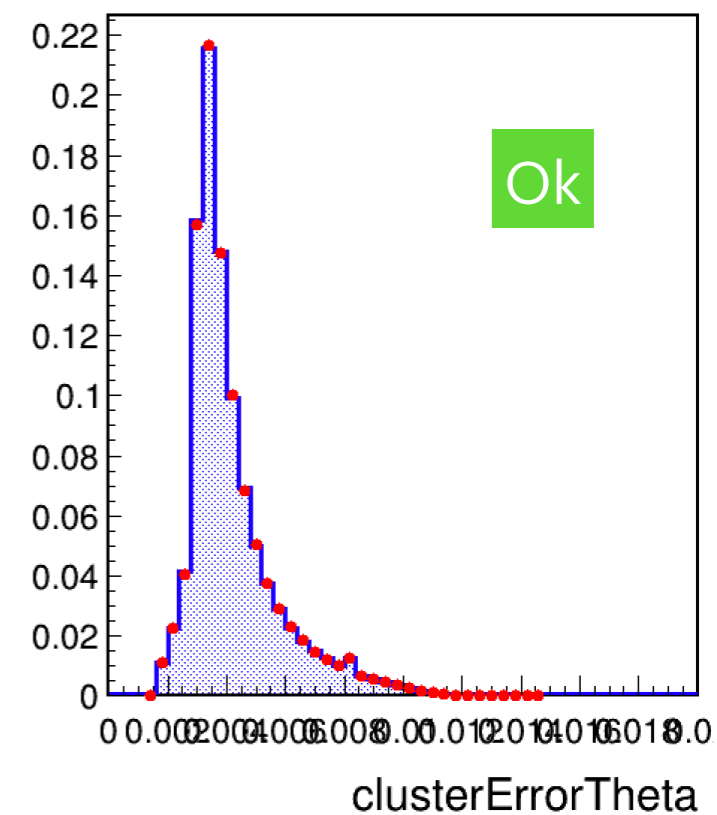
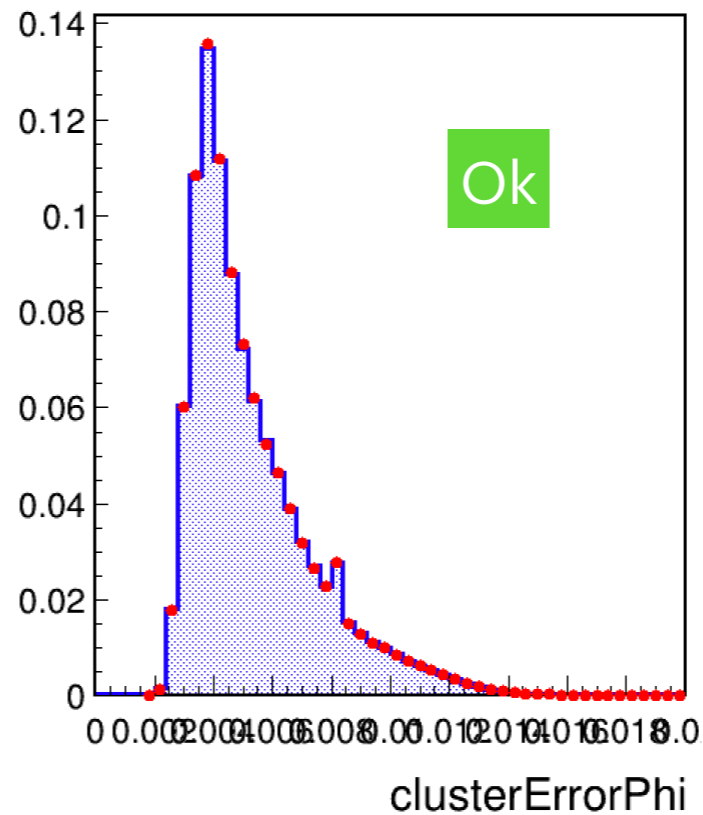
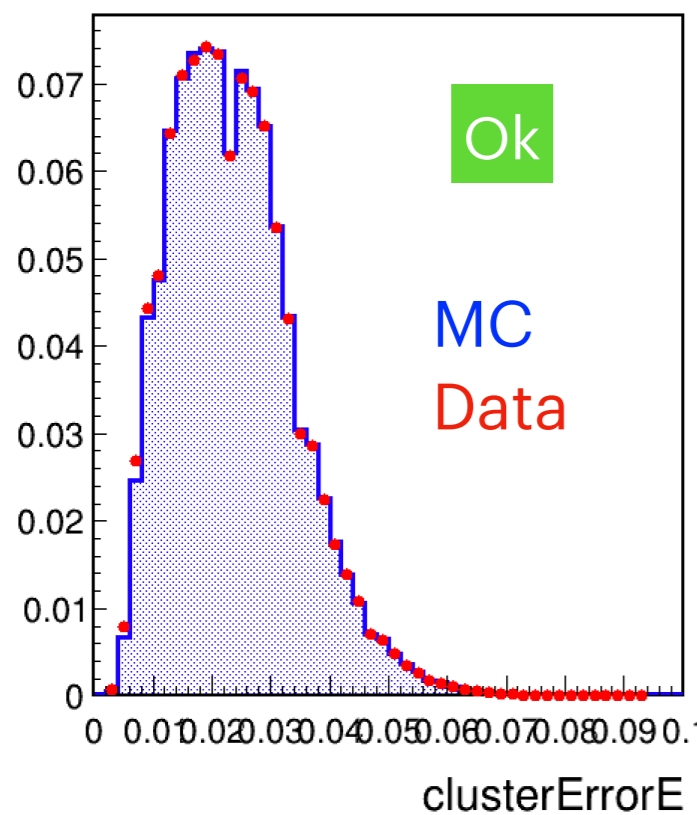
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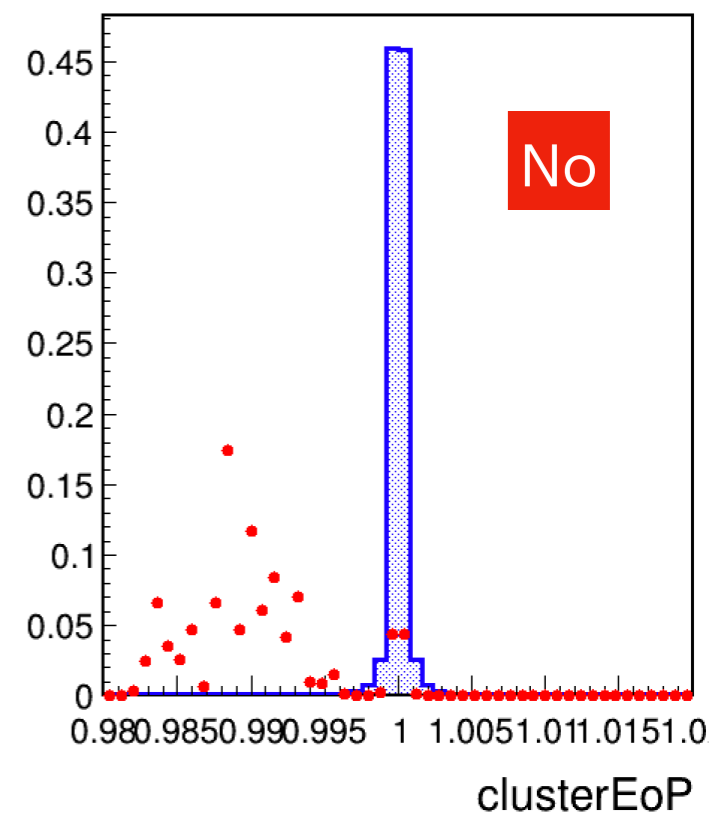
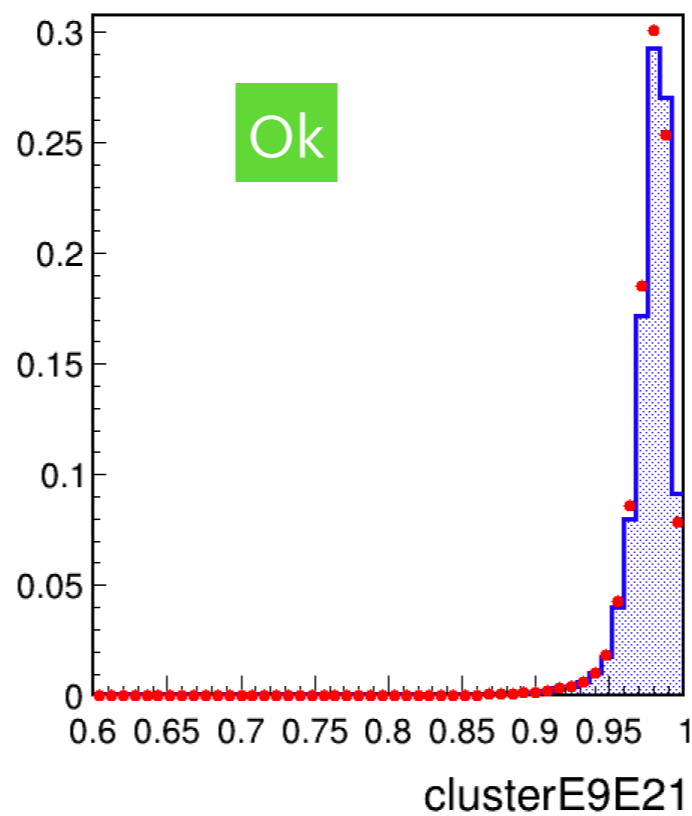
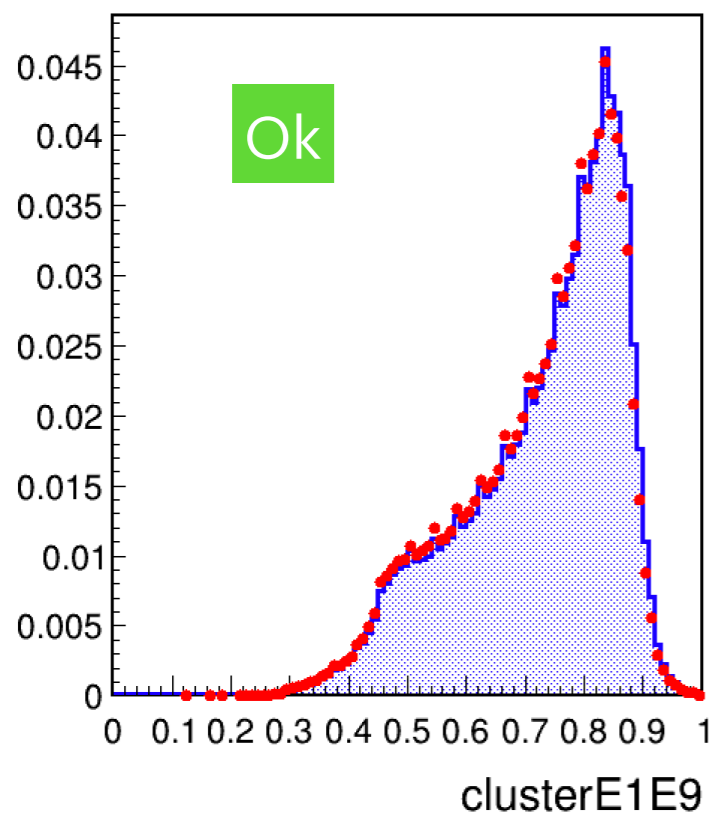
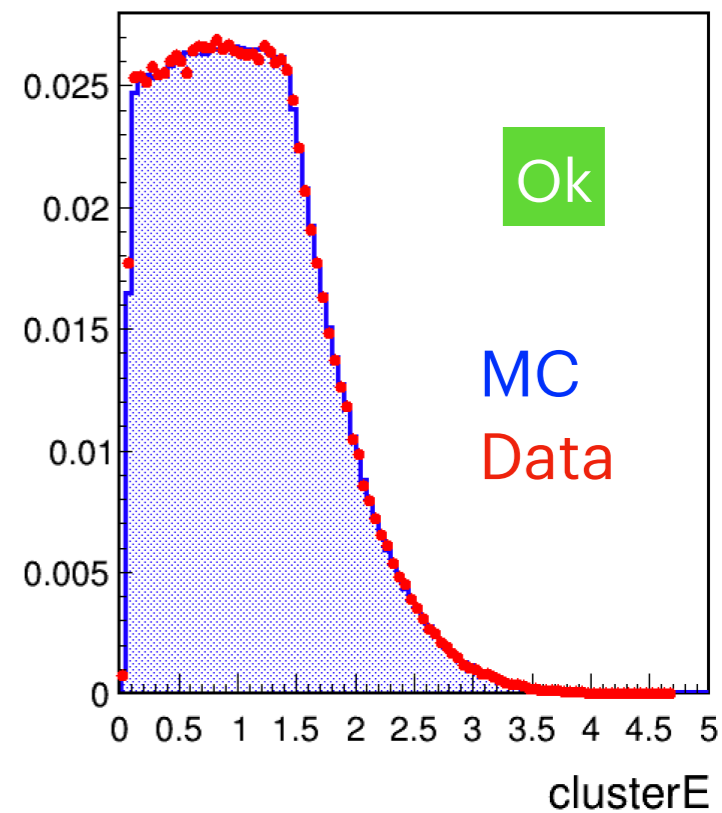
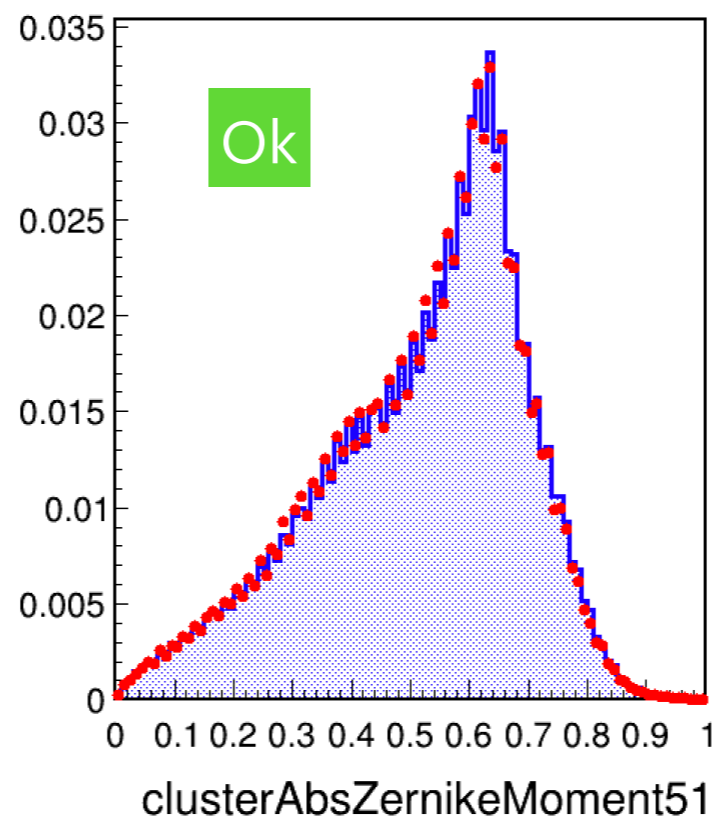
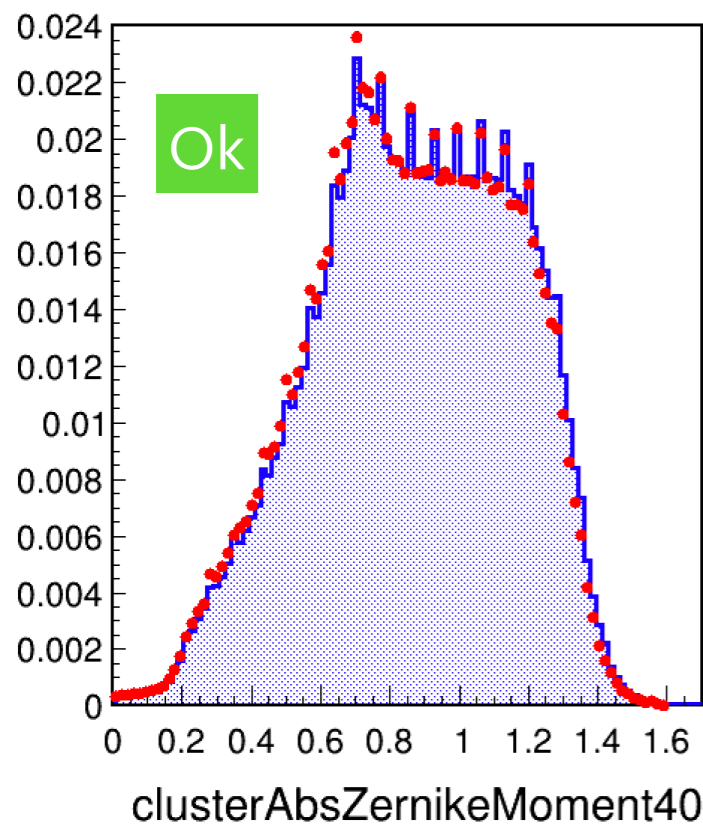
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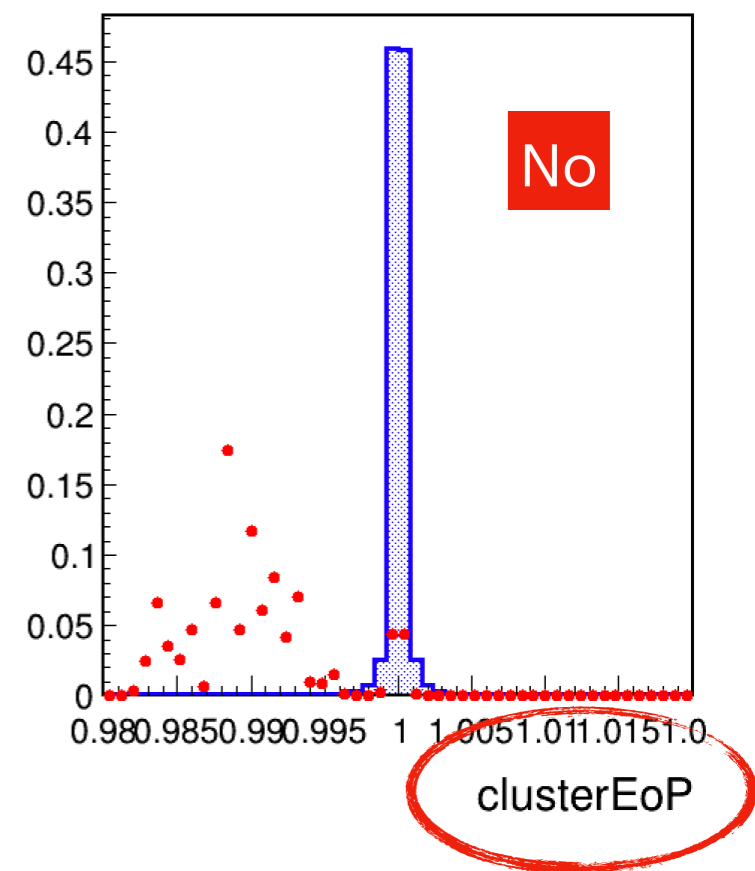
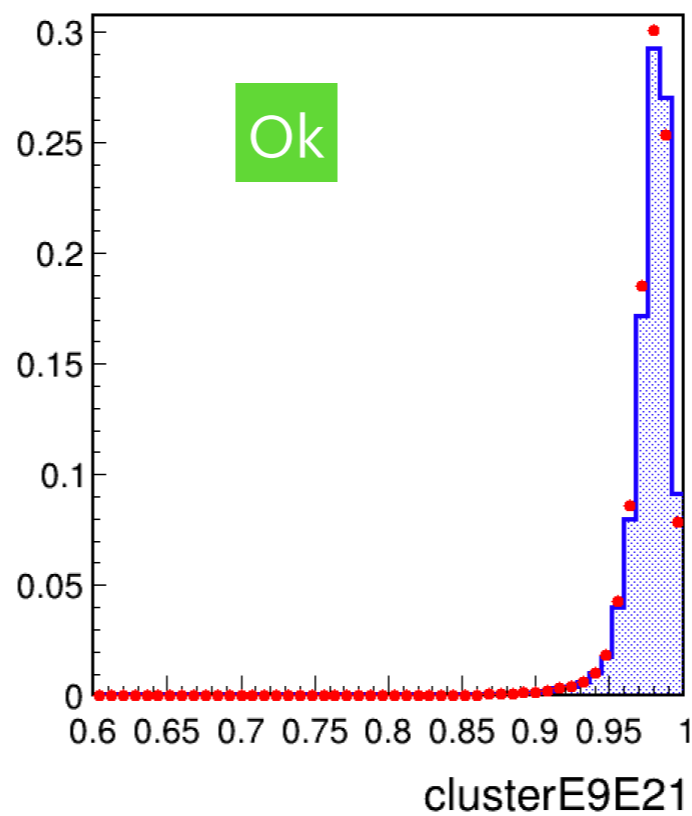
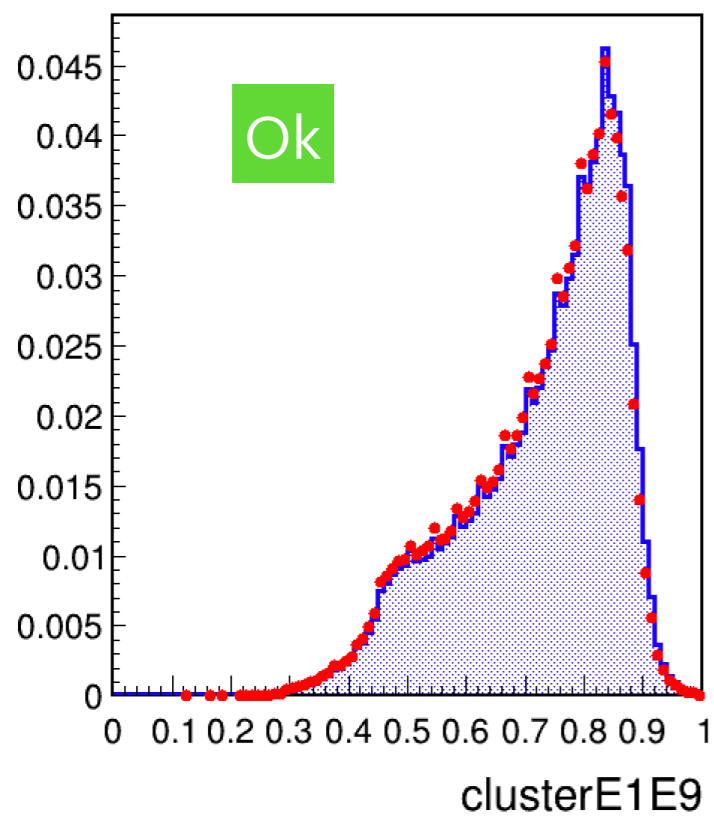
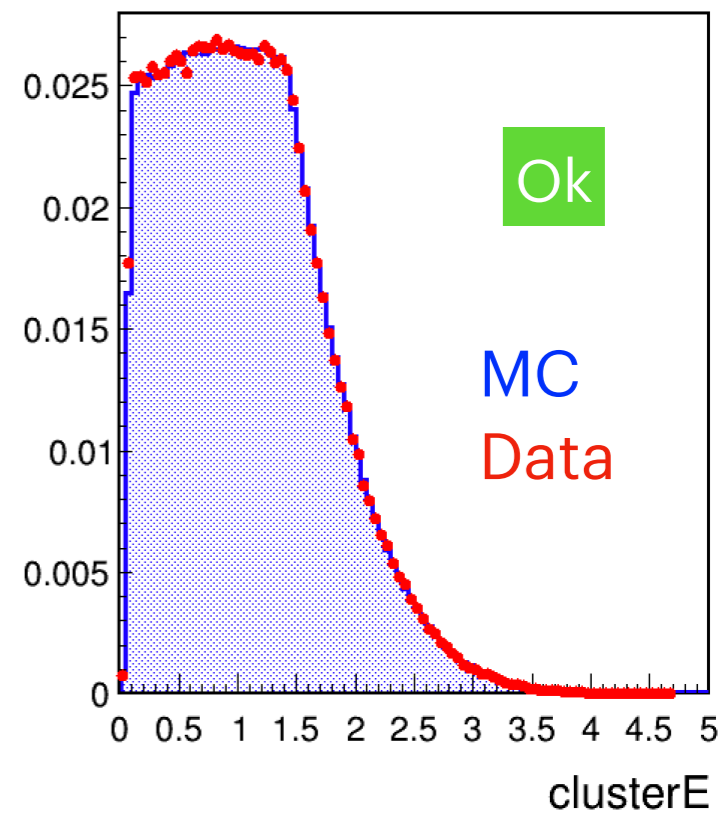
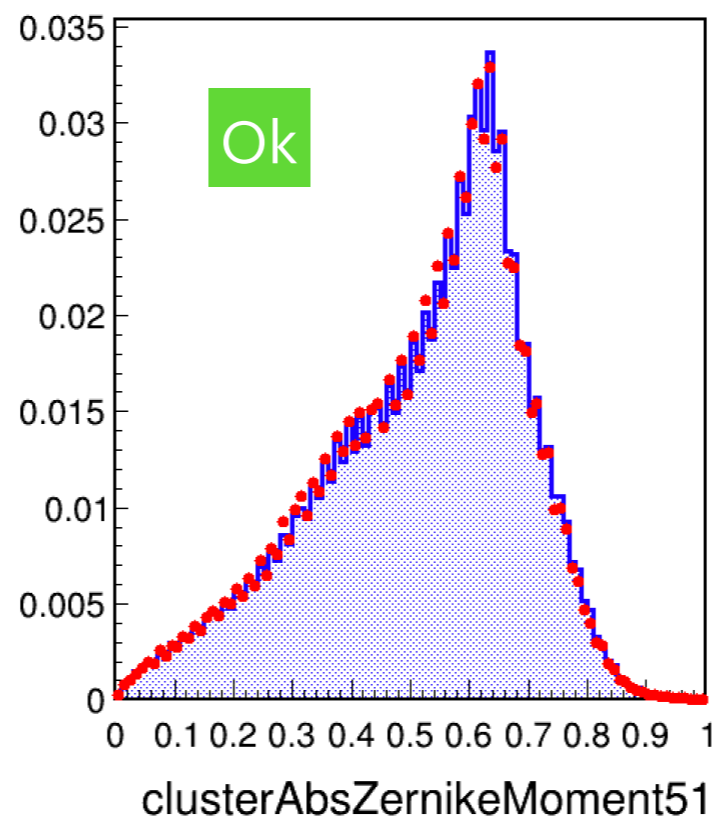
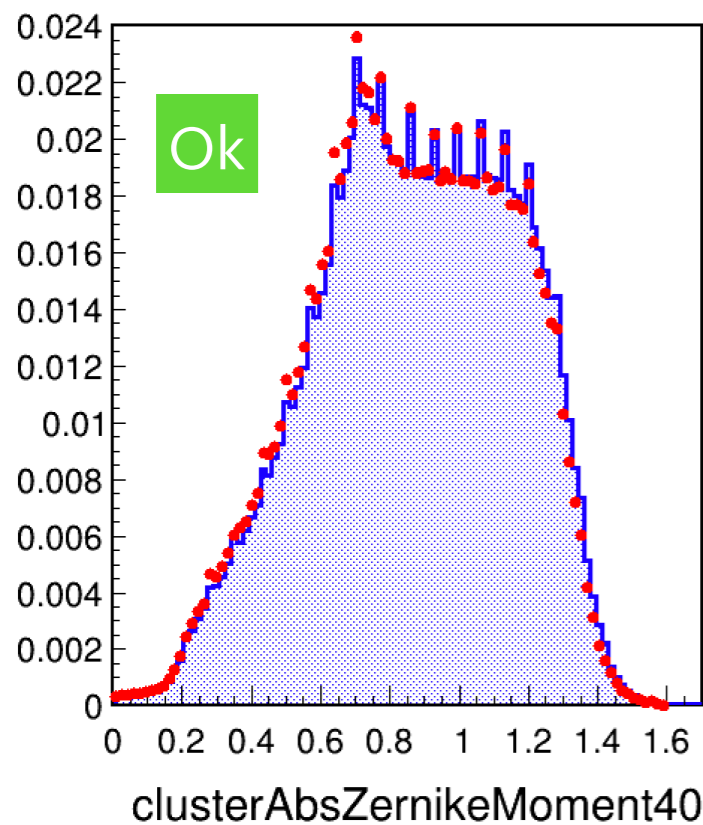
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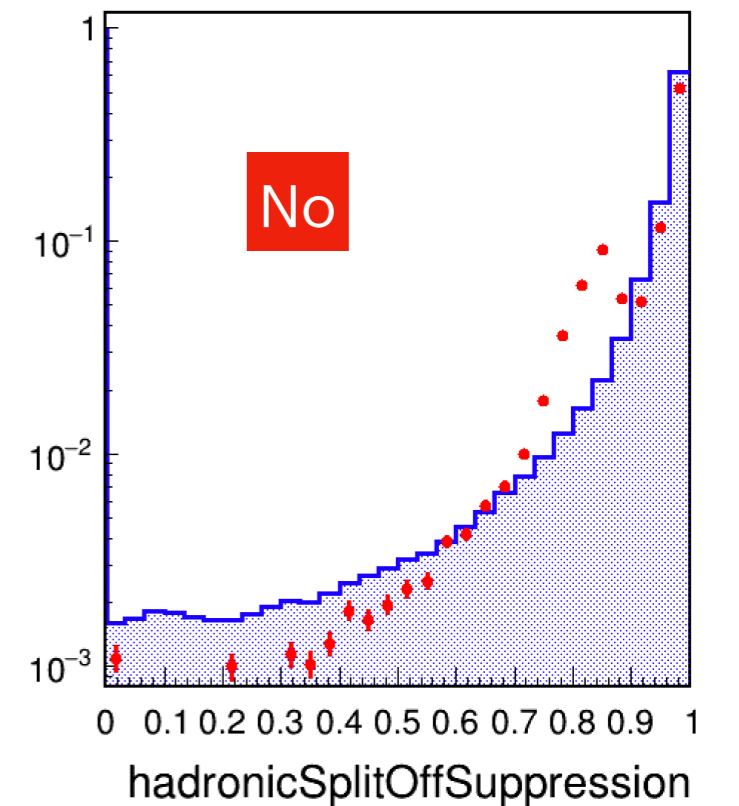
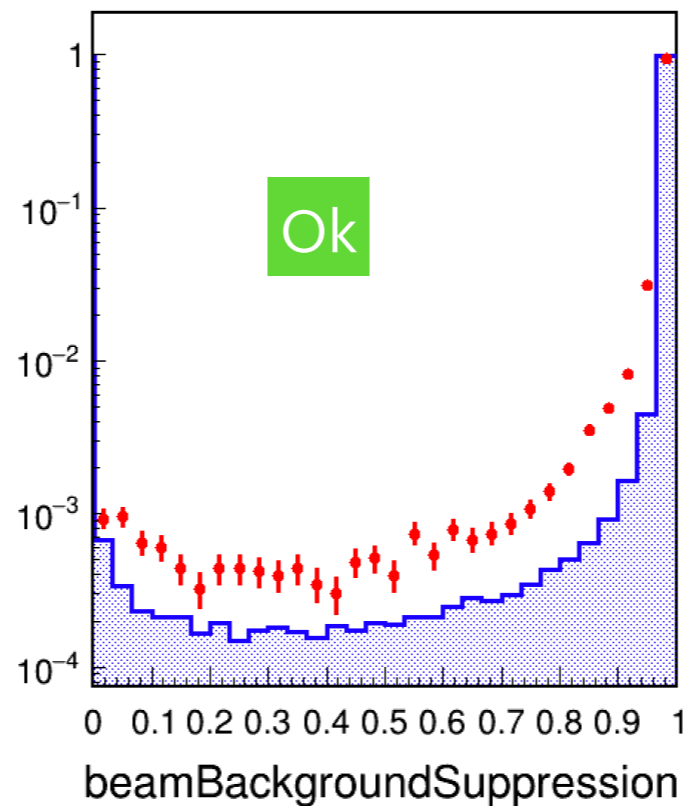
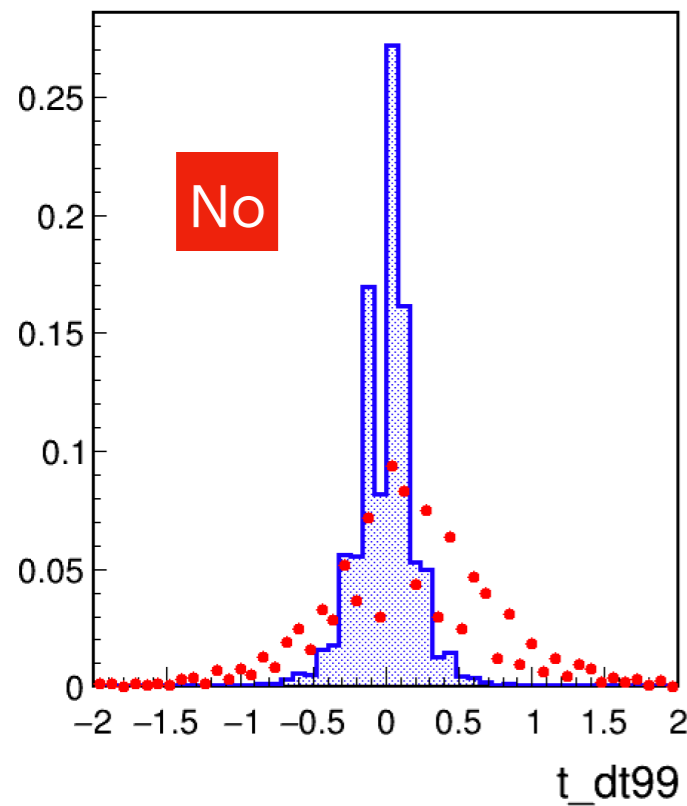
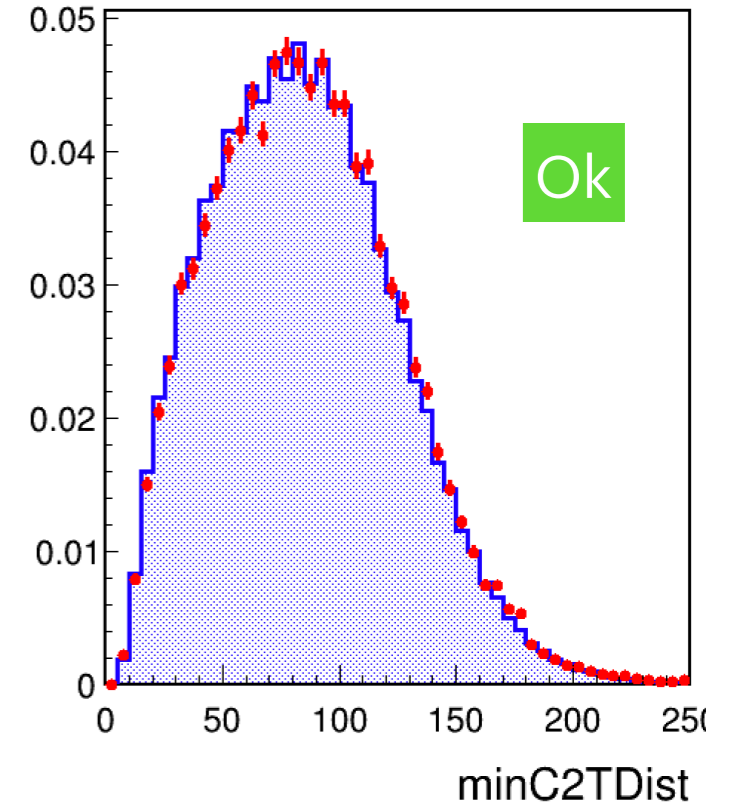
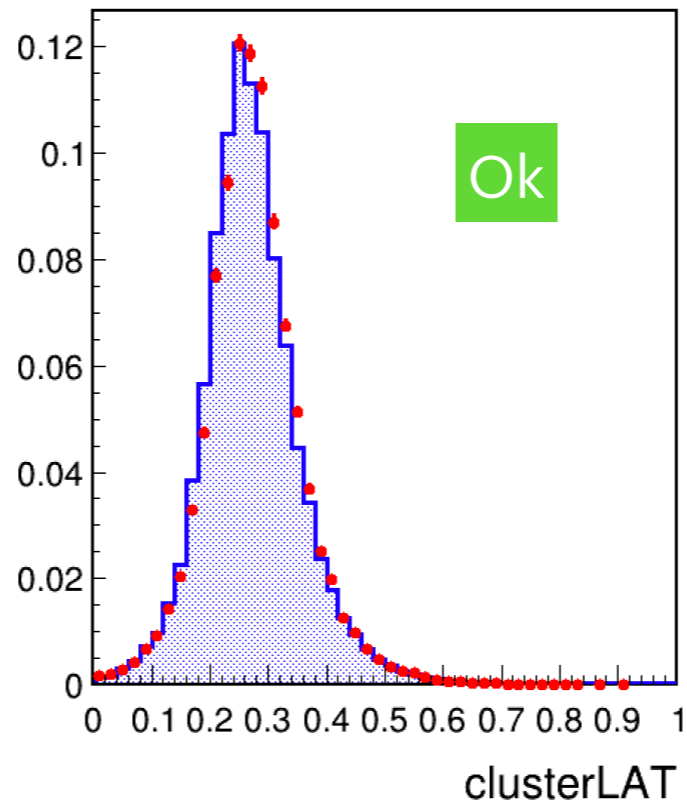
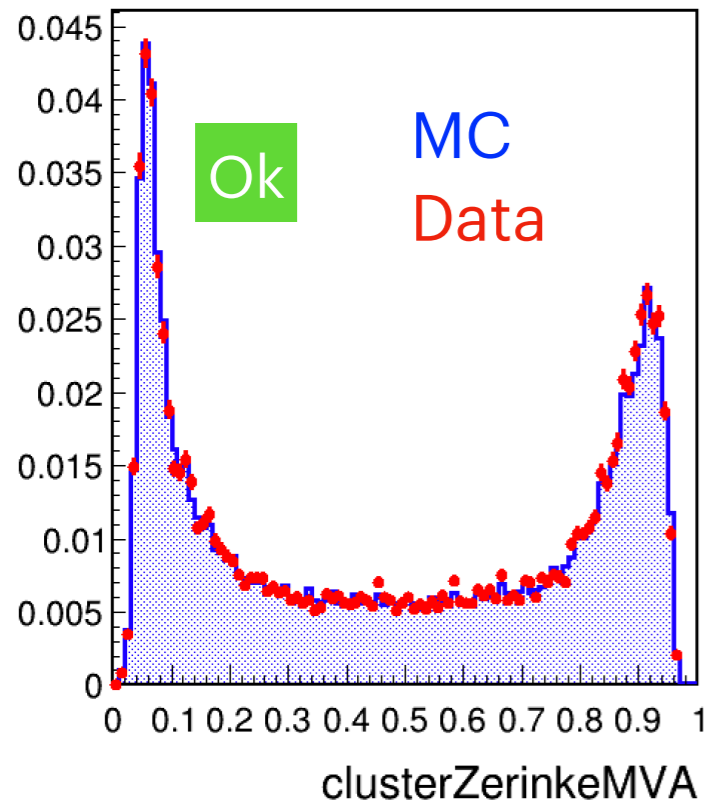
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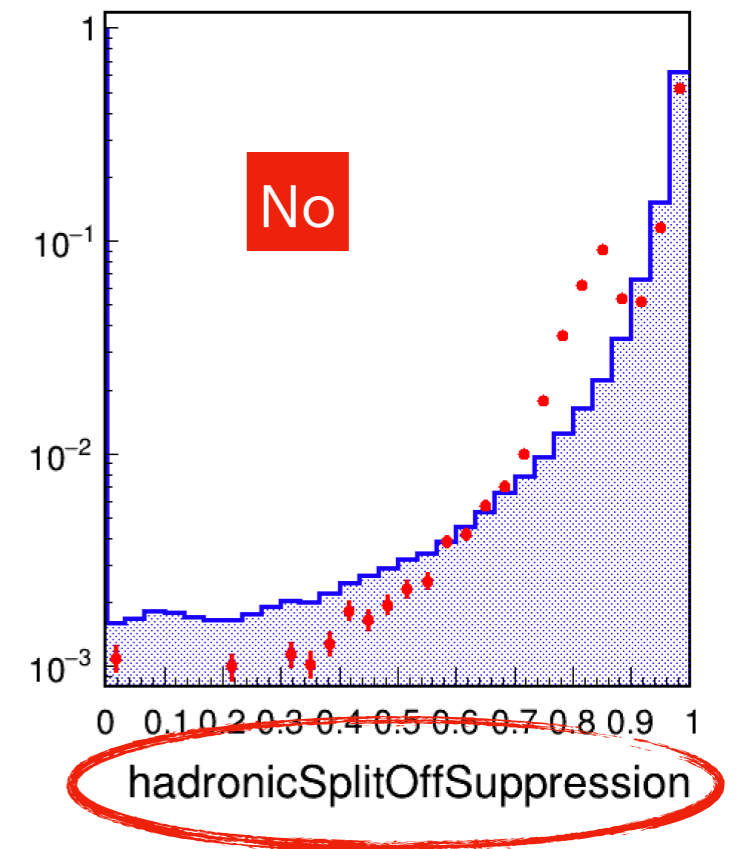
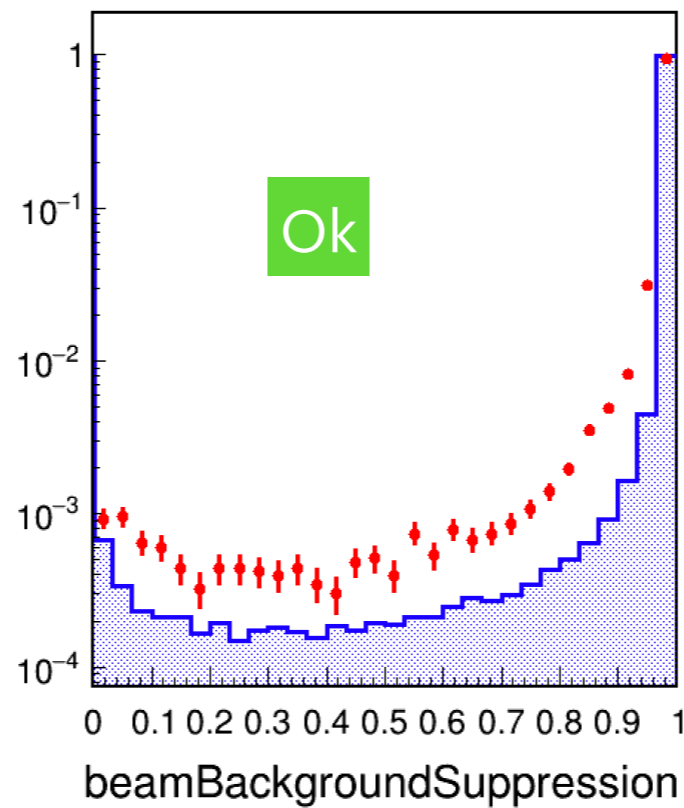
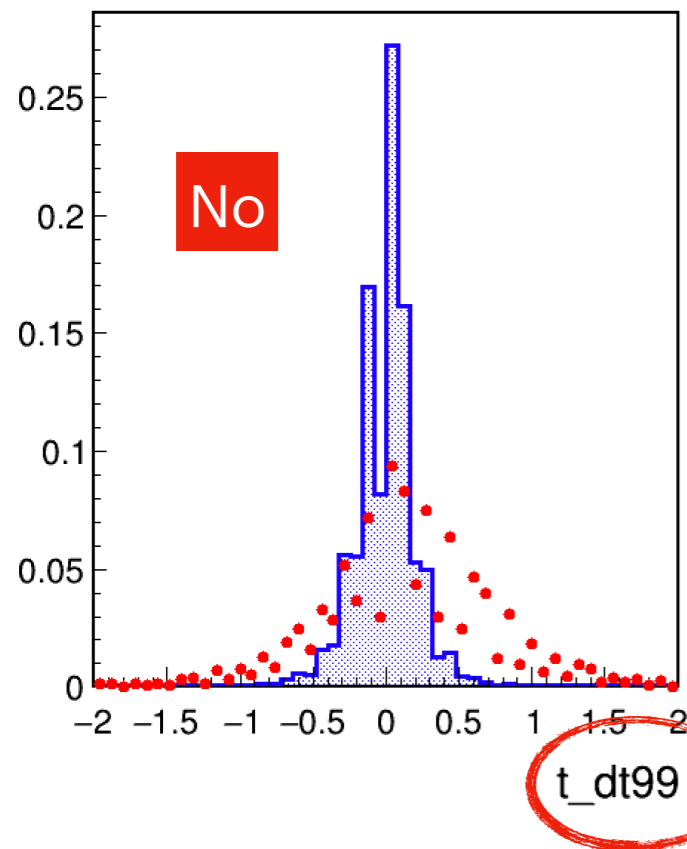
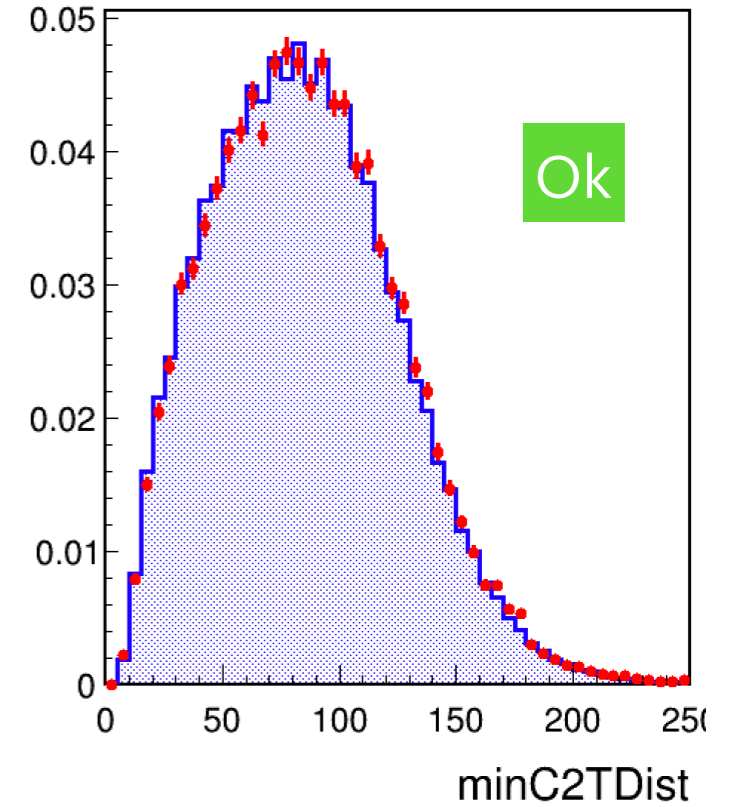
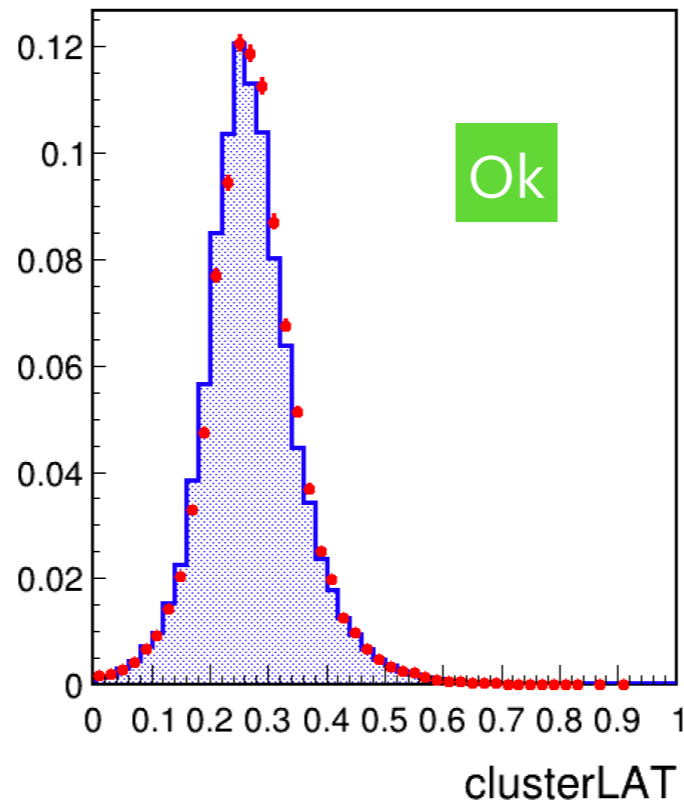
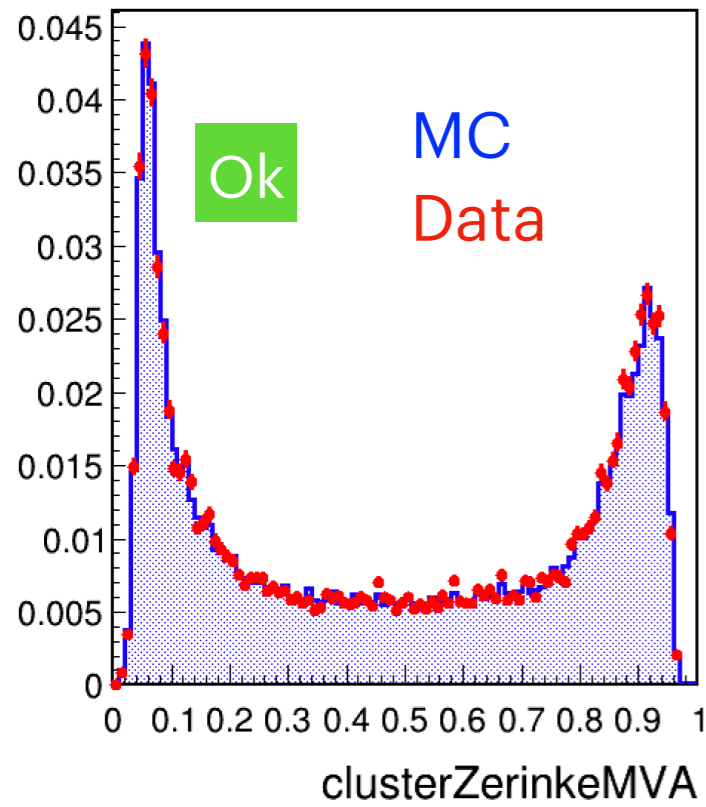
MC15 vs Proc13

Release-06

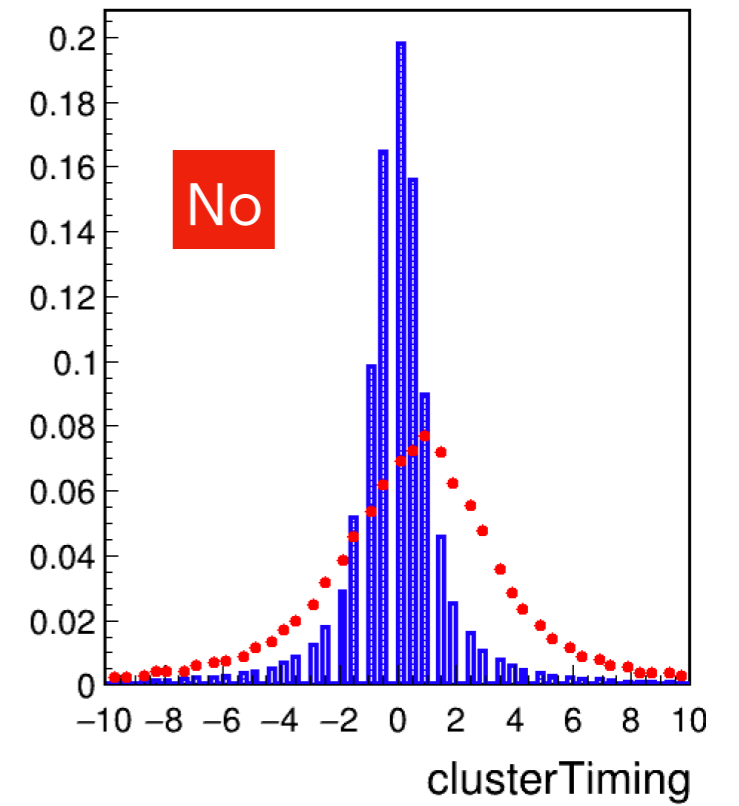
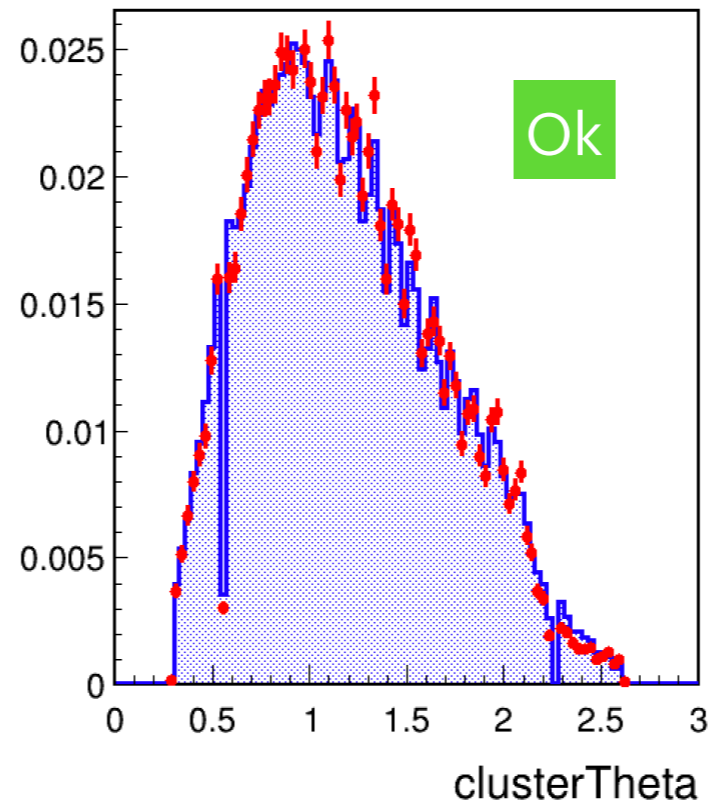
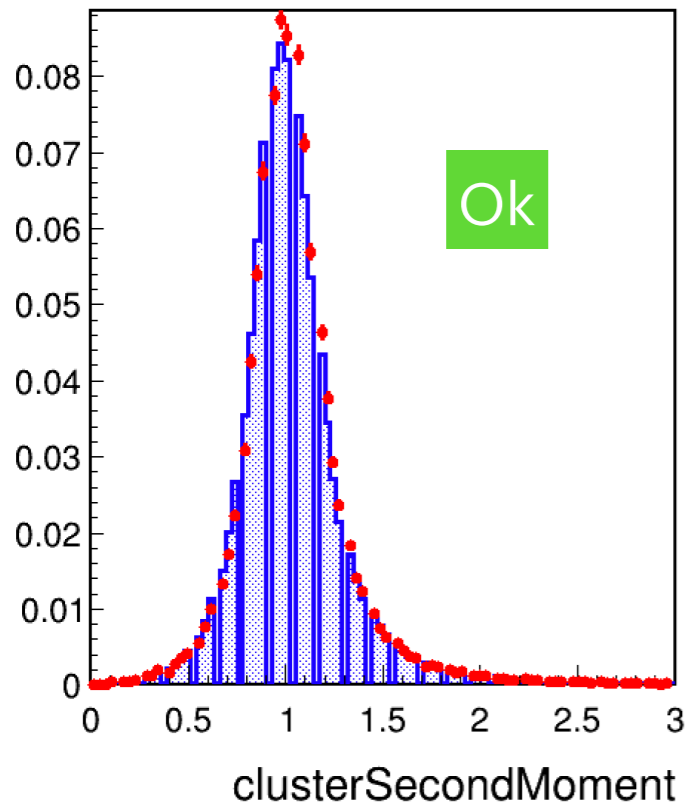
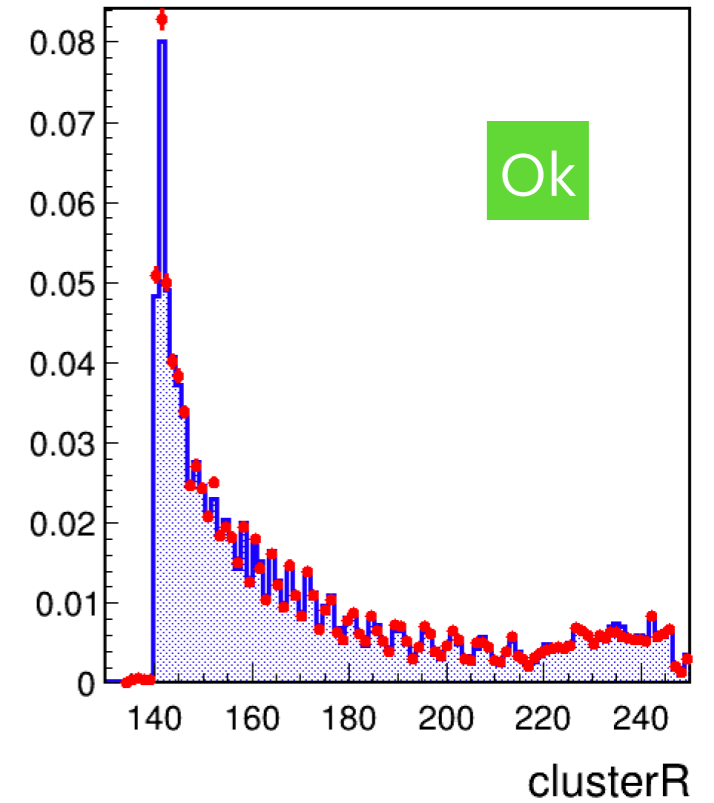
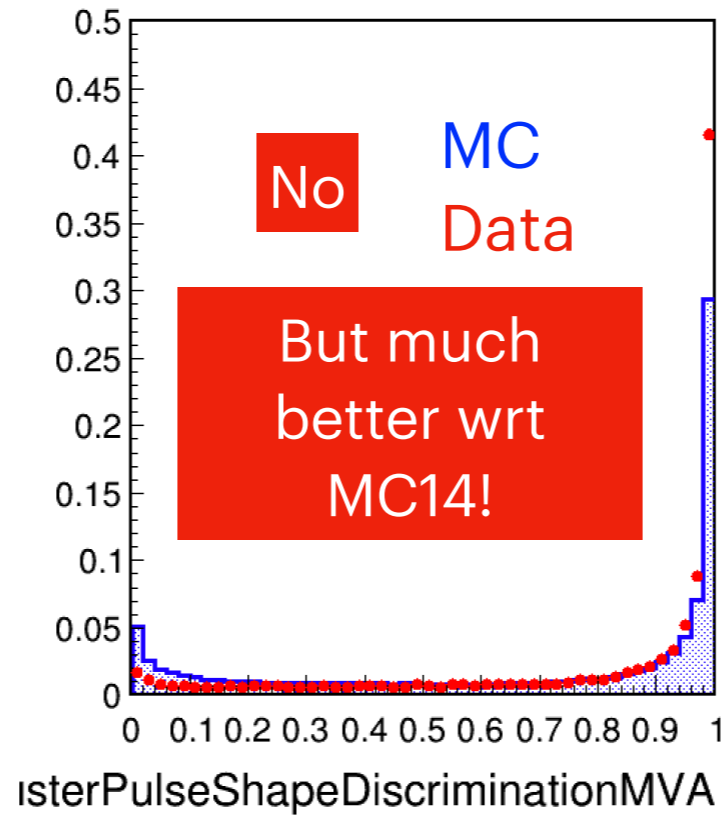
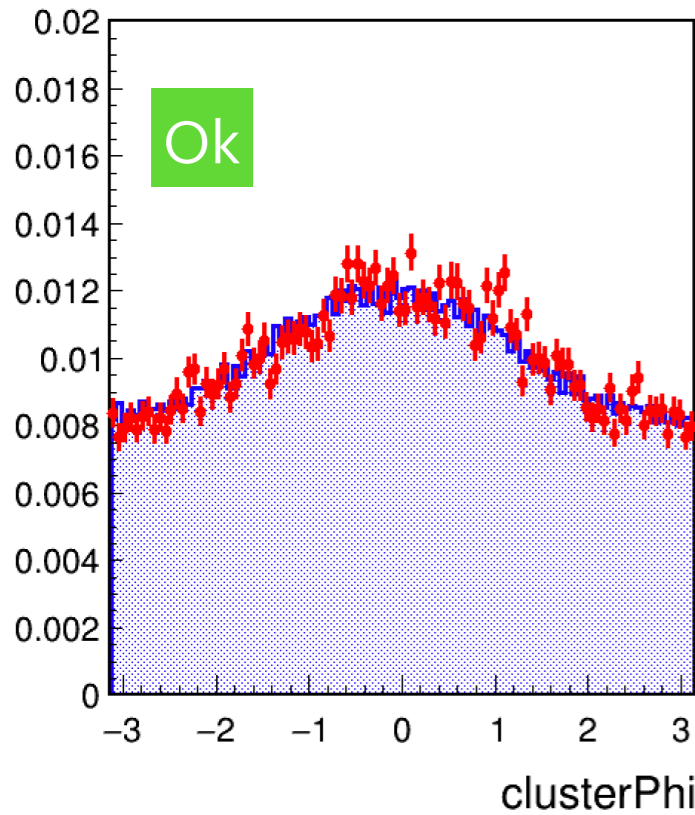
Photon MVA: inputs validation (rel-06)



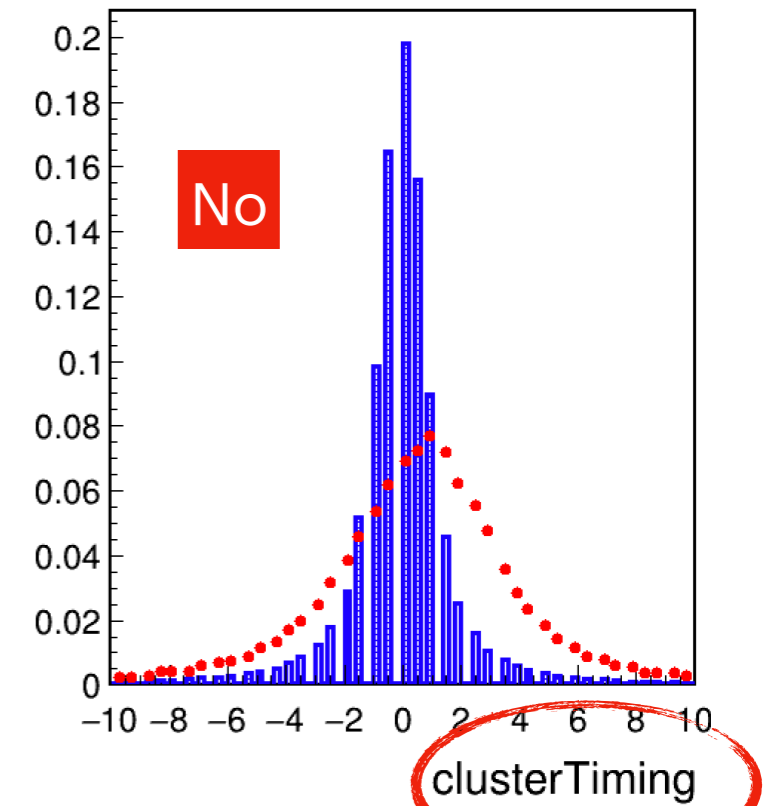
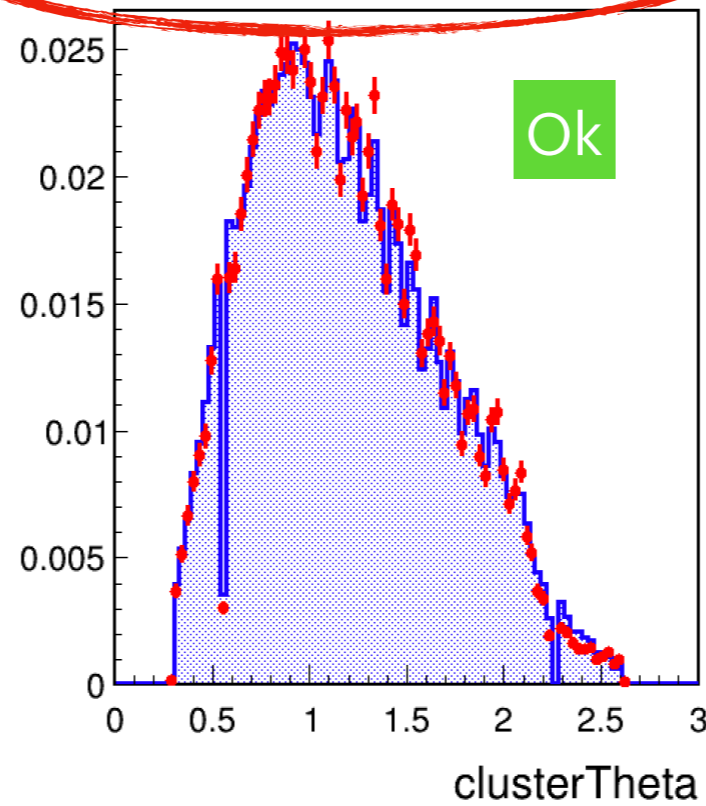
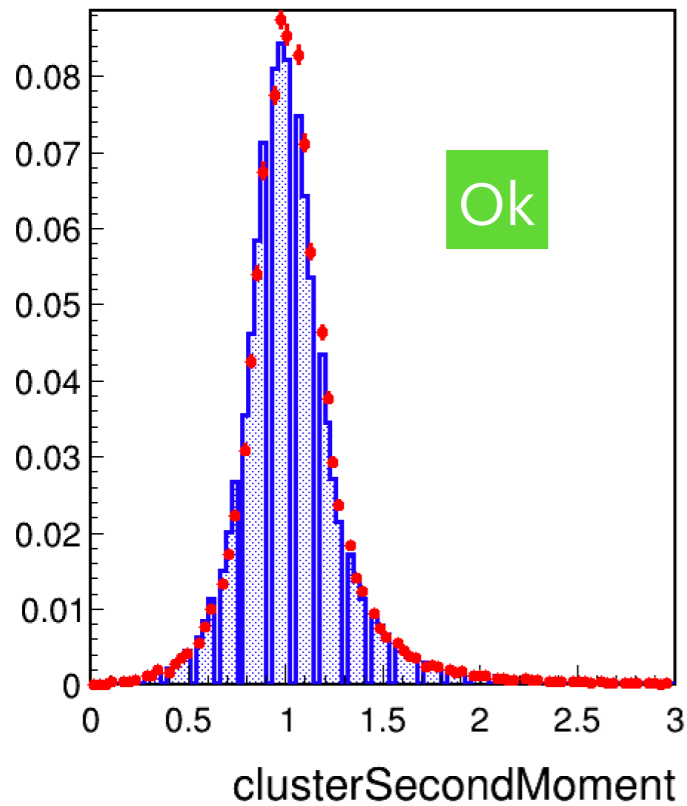
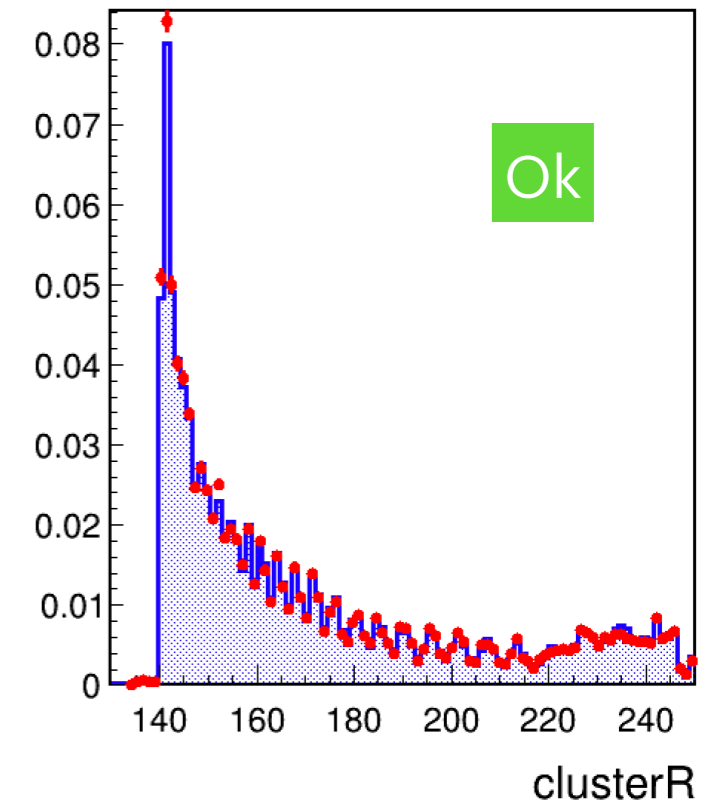
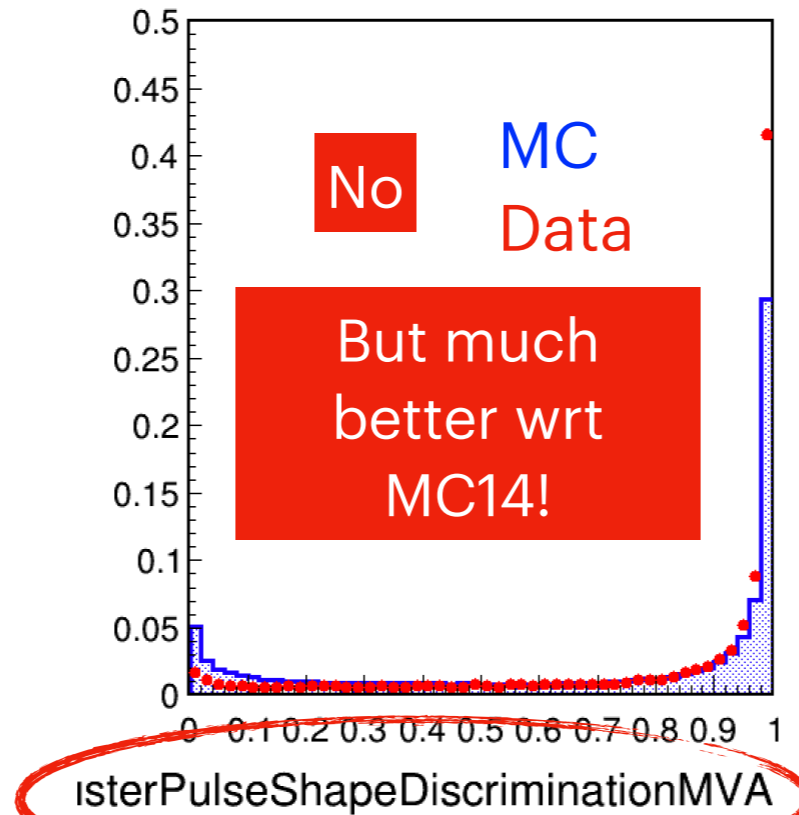
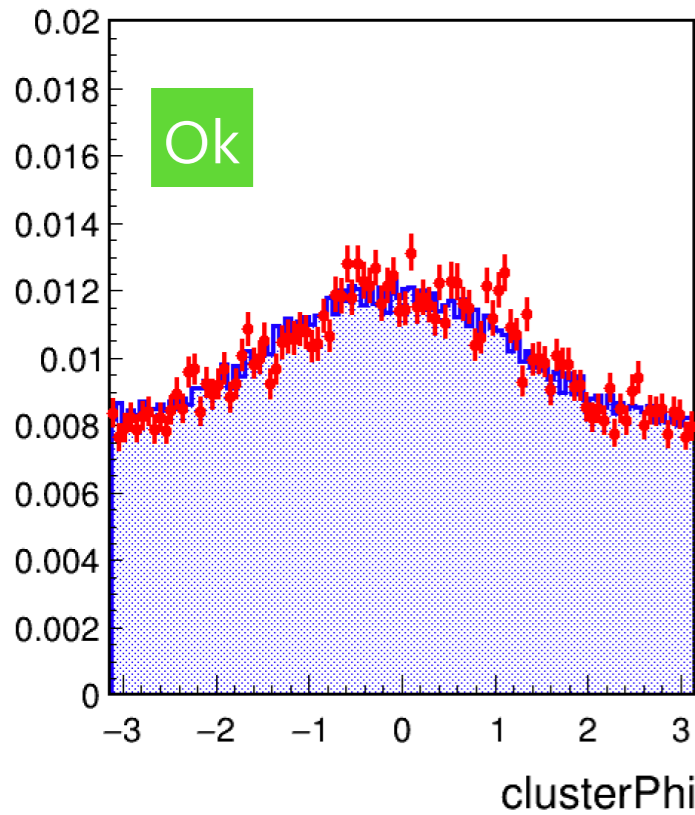
Photon MVA: inputs validation (rel-06)



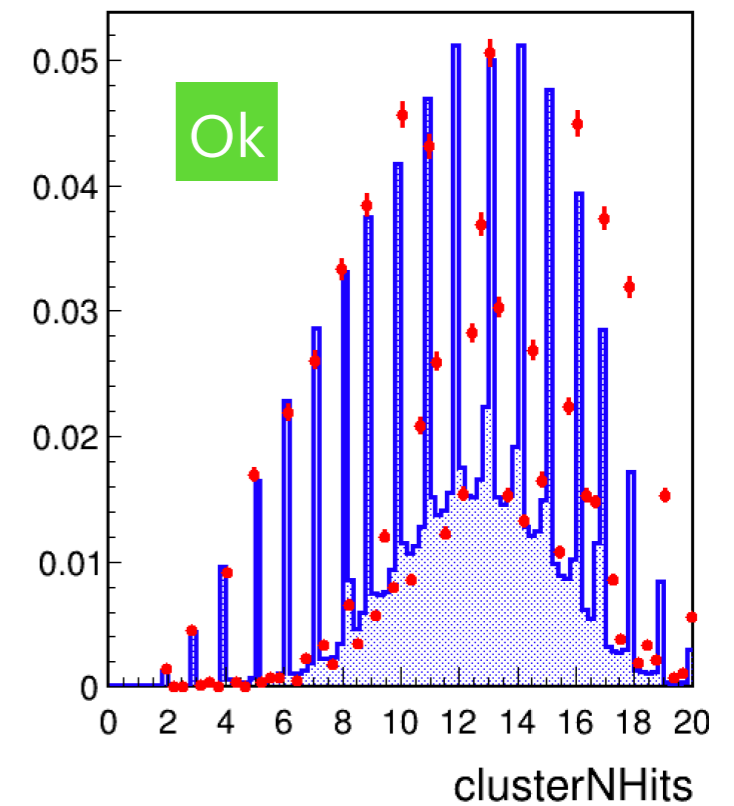
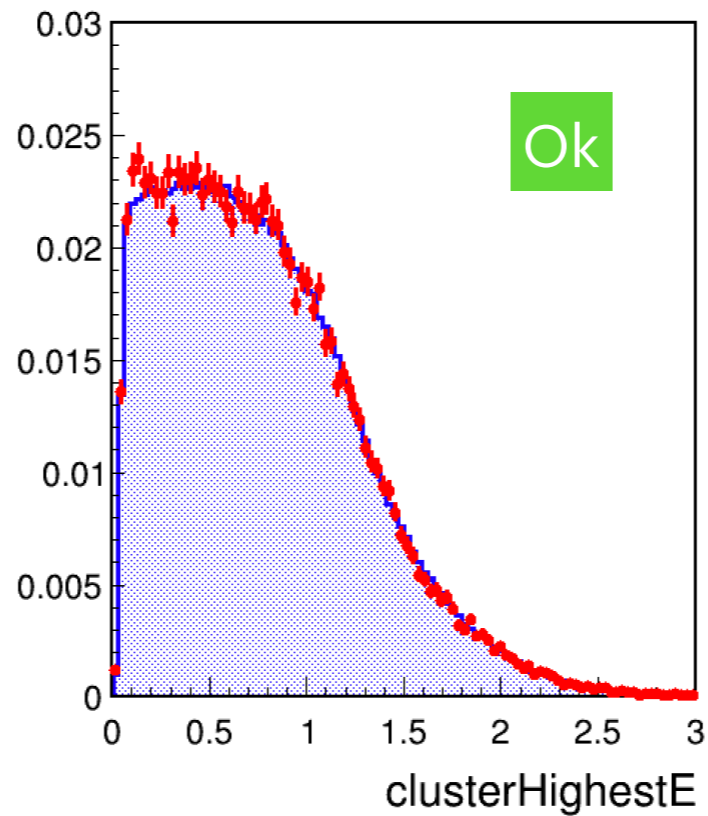
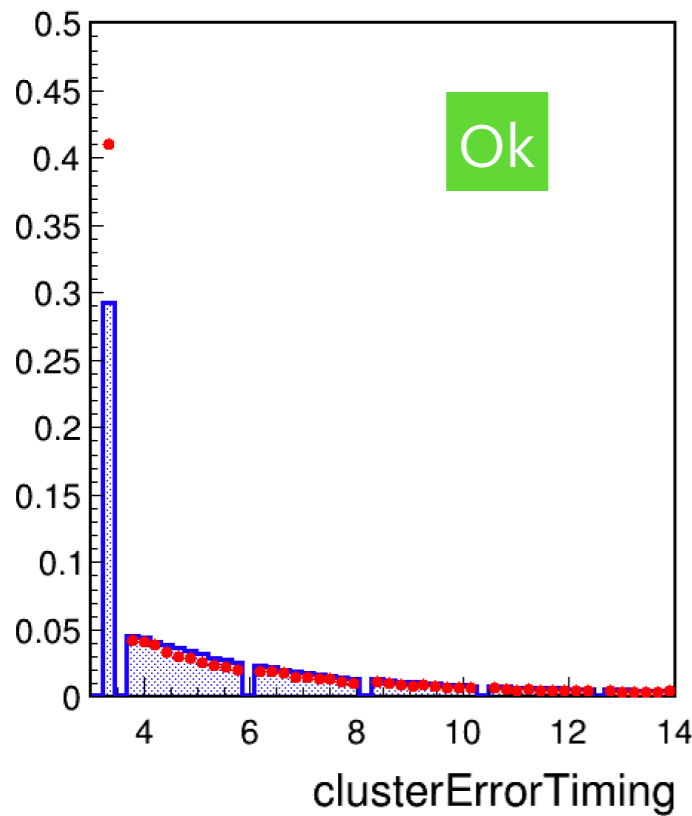
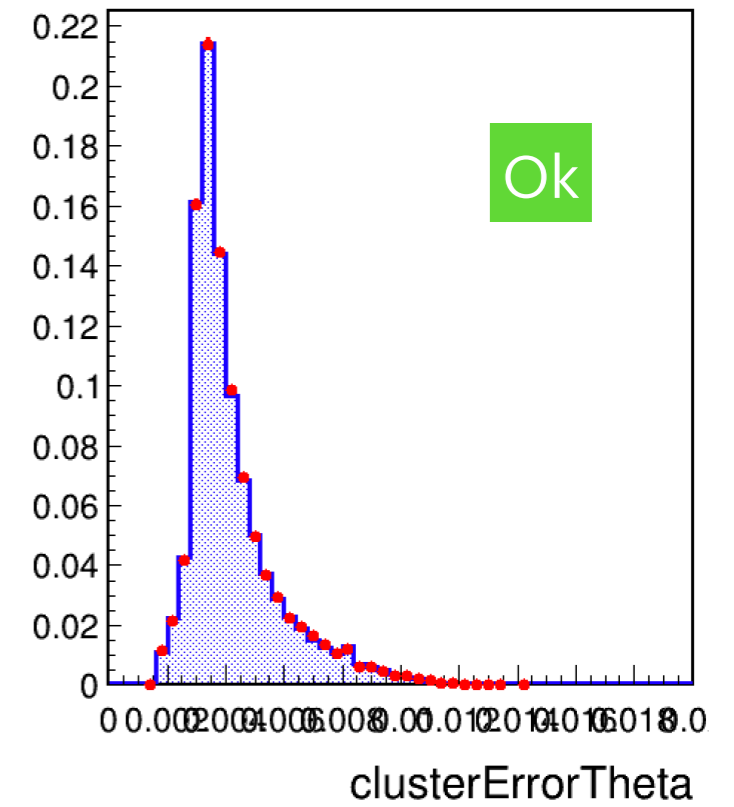
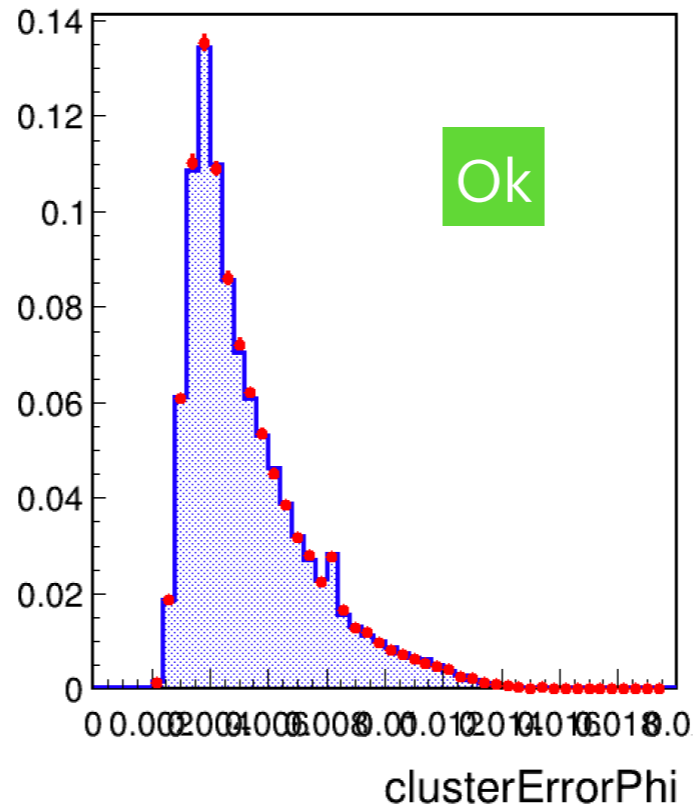
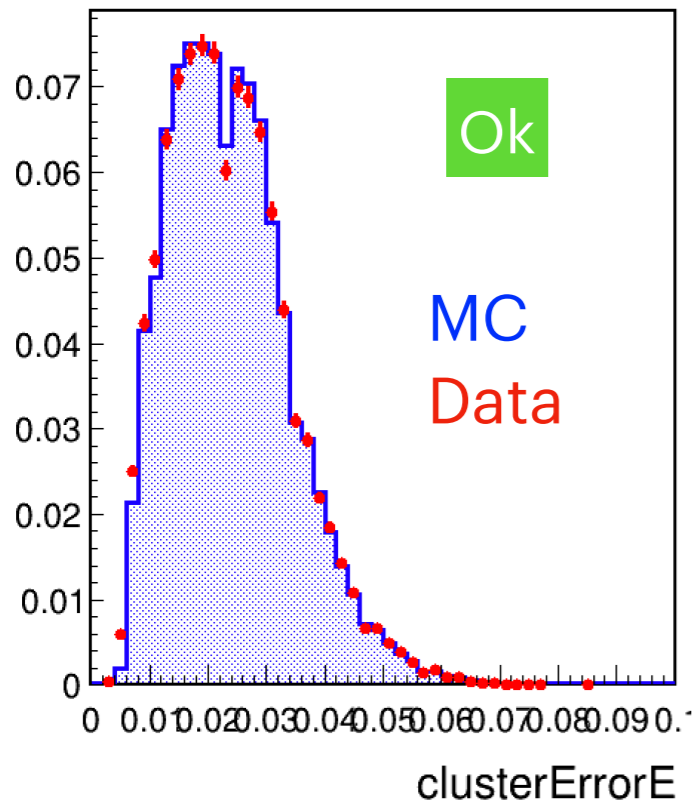
Photon MVA: inputs validation (rel-06)



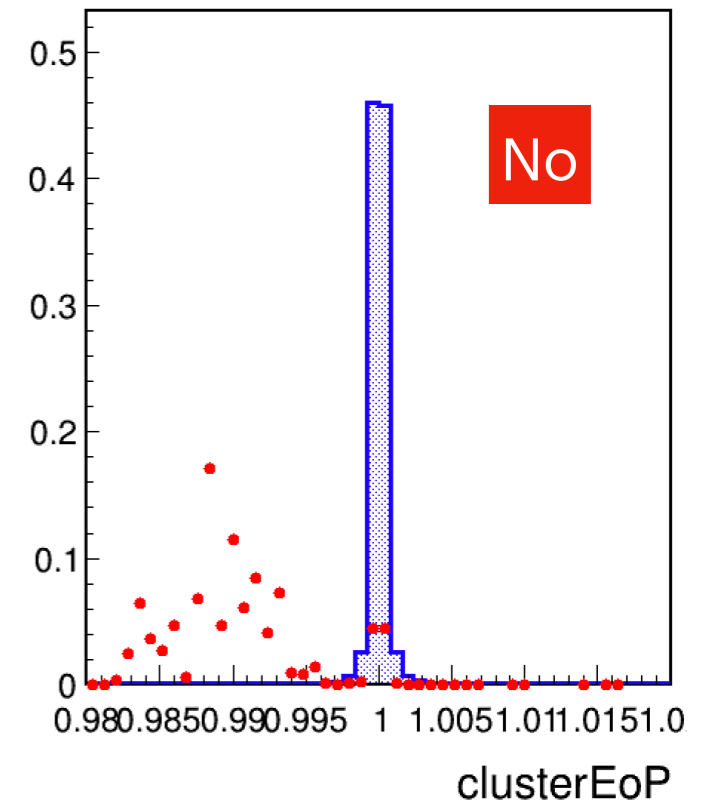
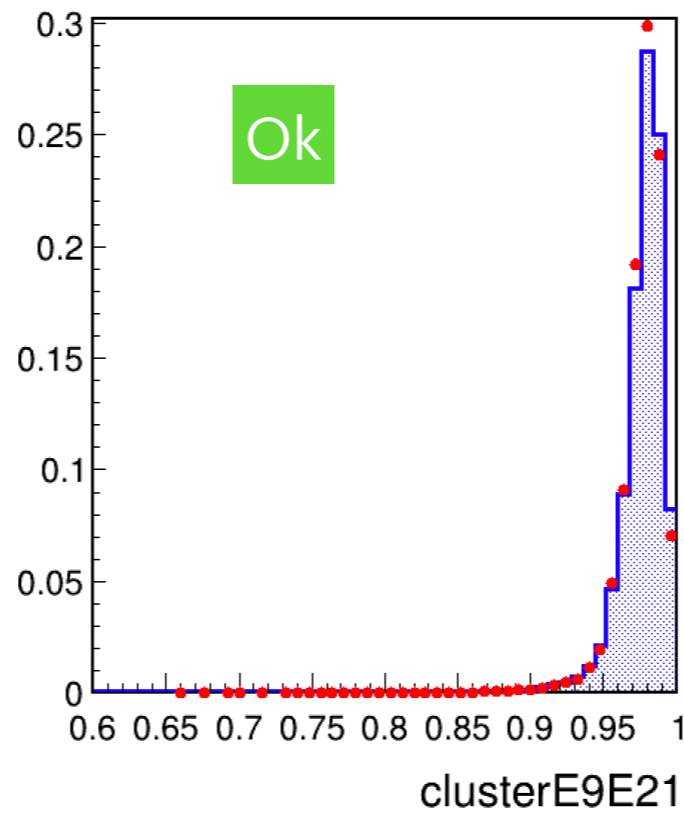
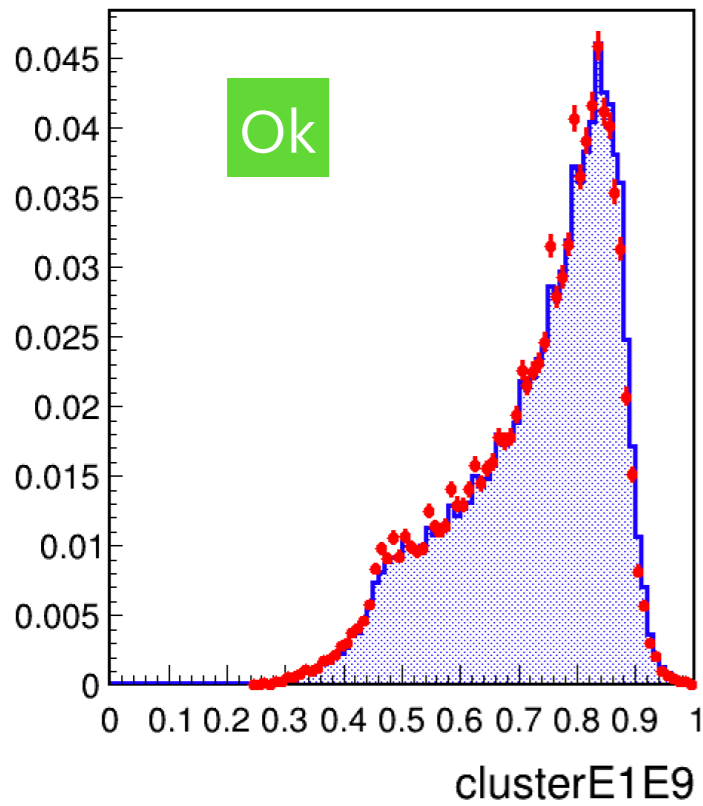
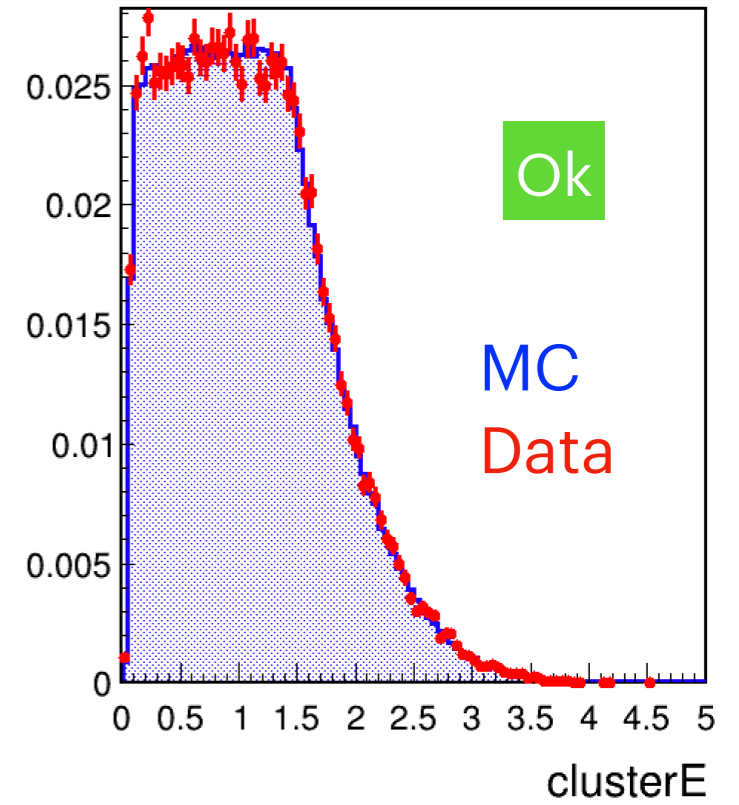
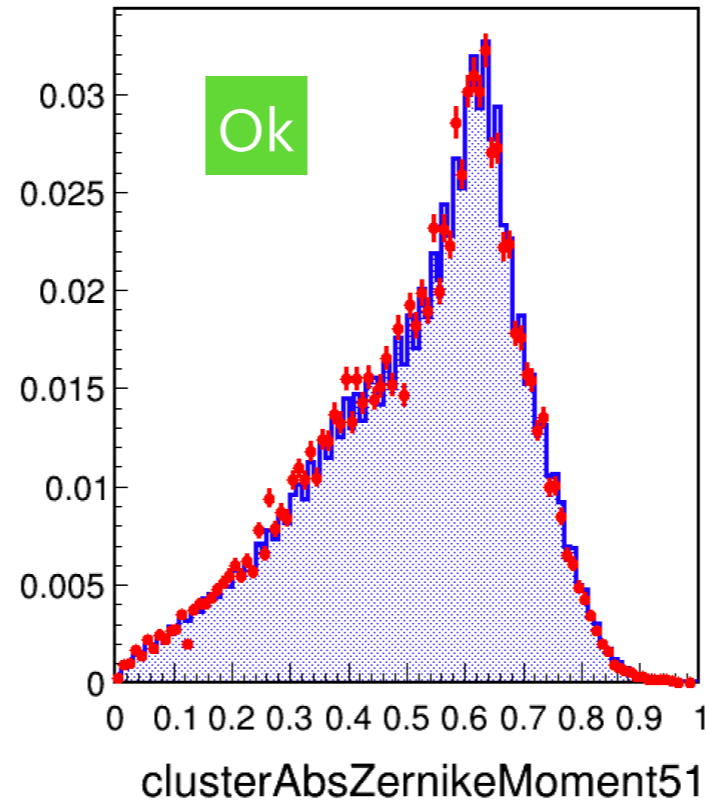
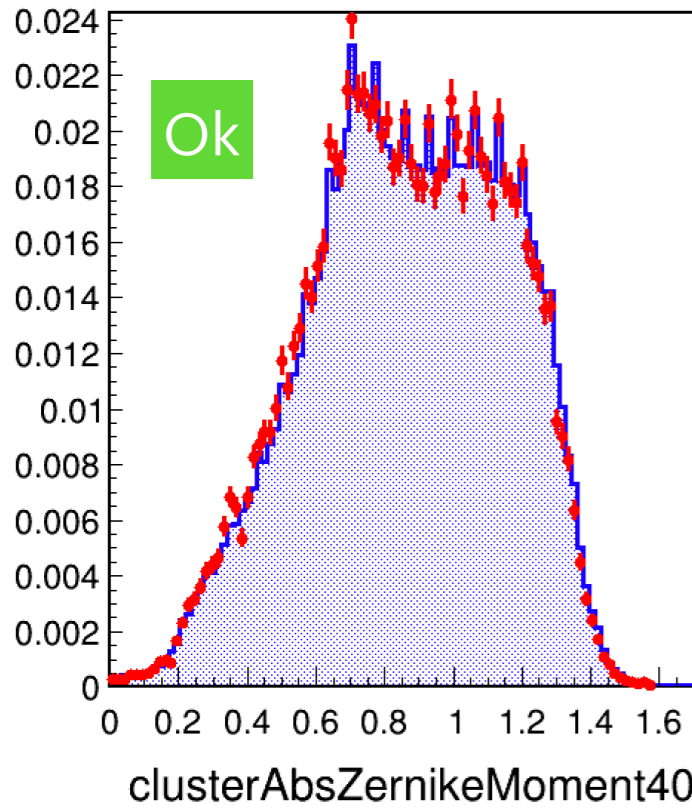
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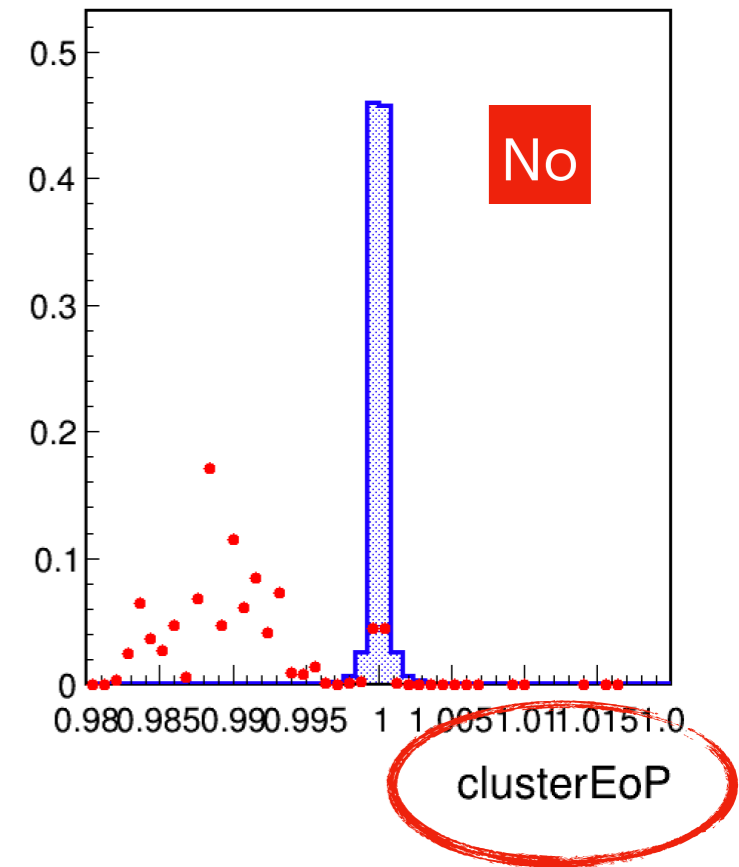
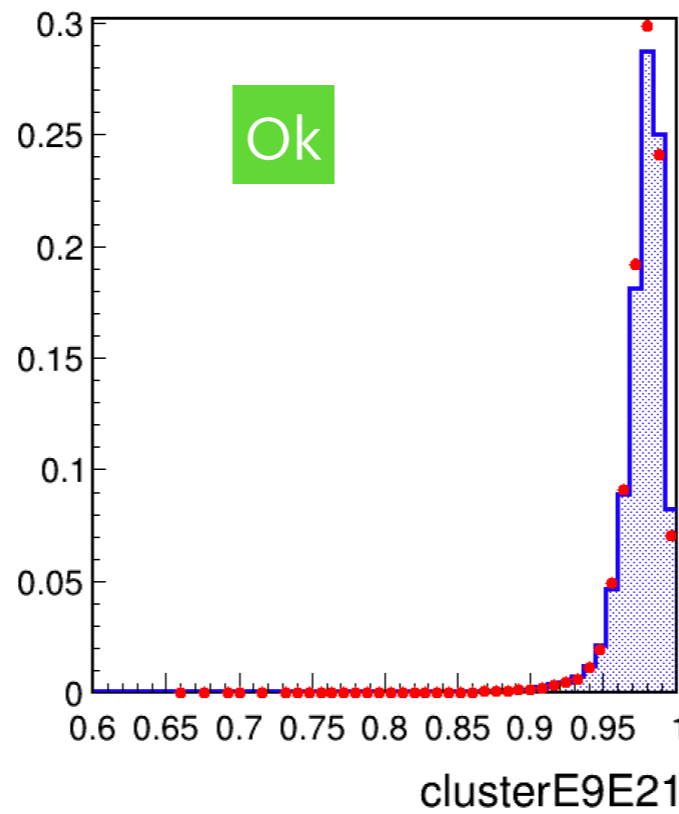
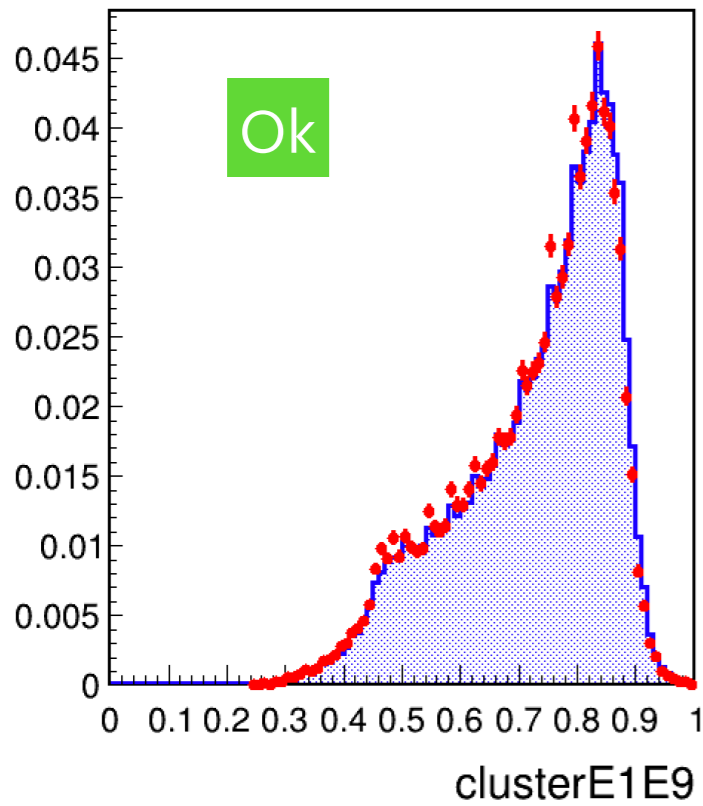
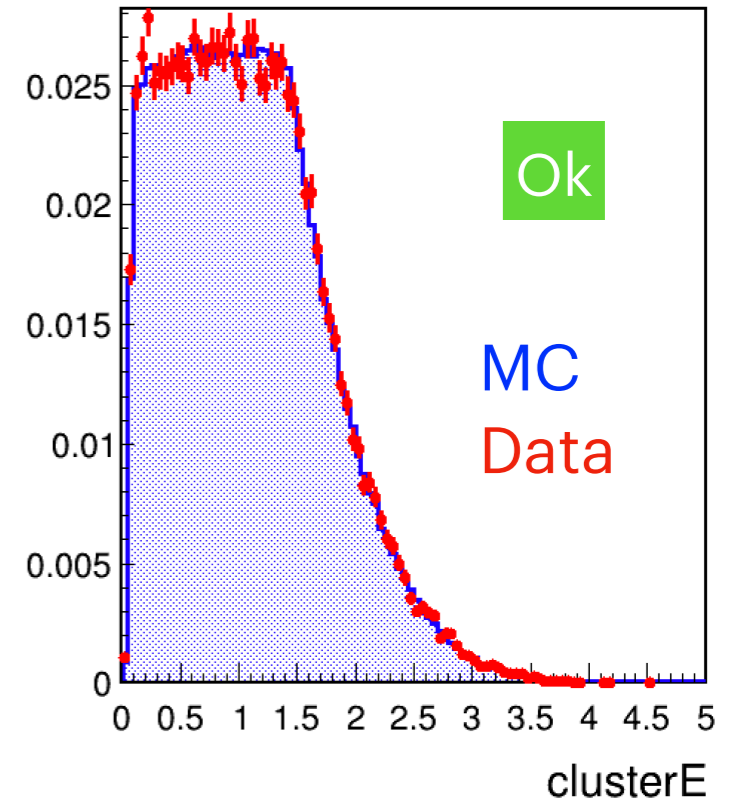
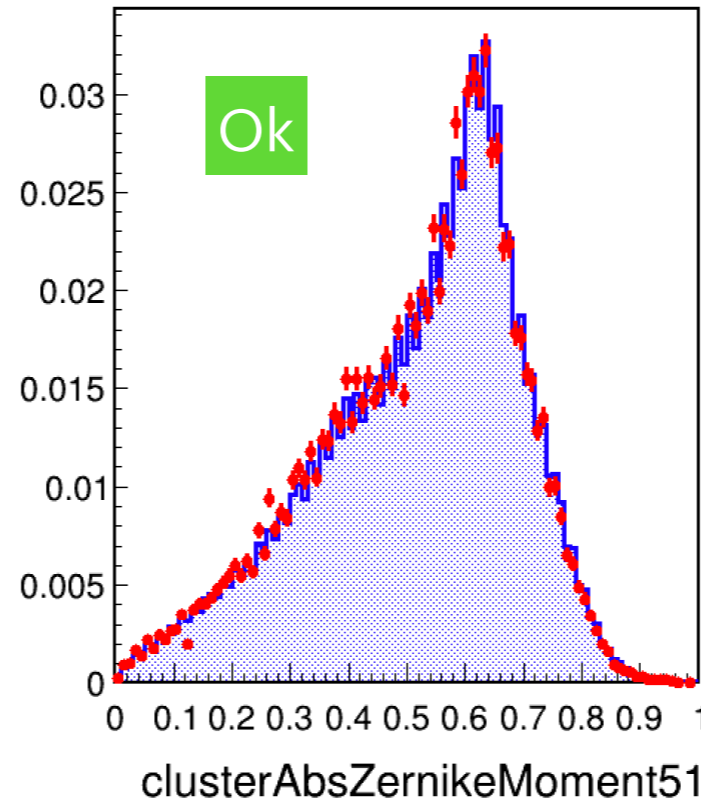
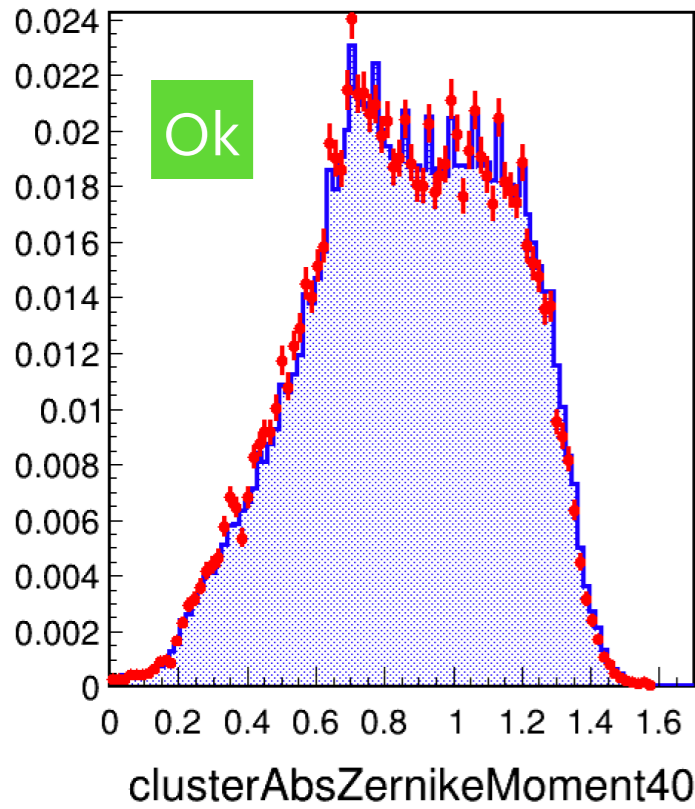
Photon MVA: inputs validation (rel-06)



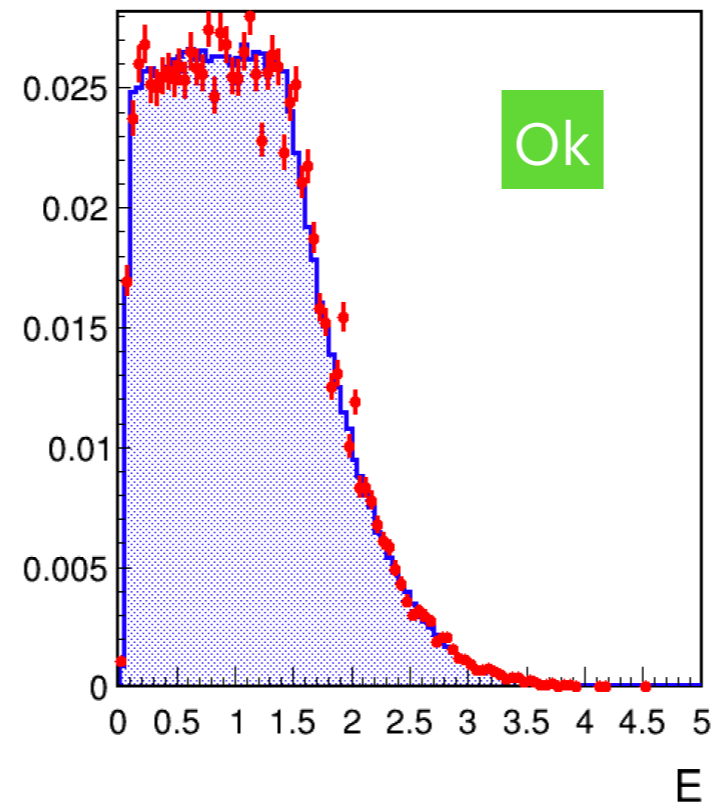
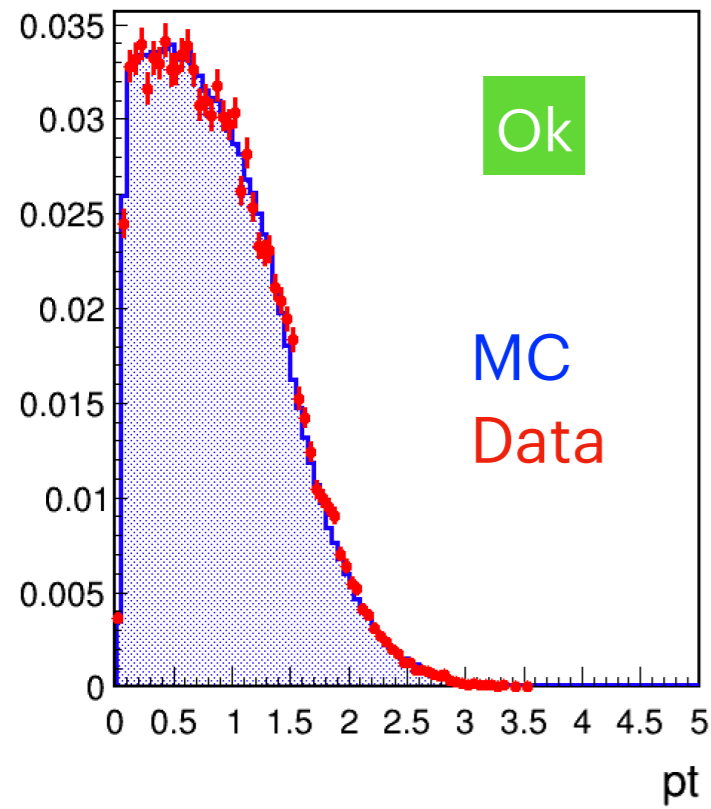
Photon MVA: inputs validation (rel-06)



Photon MVA: inputs validation (rel-06)



Photon MVA: inputs validation (rel-06)



Photon MVA results using release-06

Train on MC sample after applying all π^0 selections.

Inputs (after pruning)

pt

clusterE1E9

clusterErrorPhi

clusterHighestE

clusterSecondMoment

clusterZernikeMVA

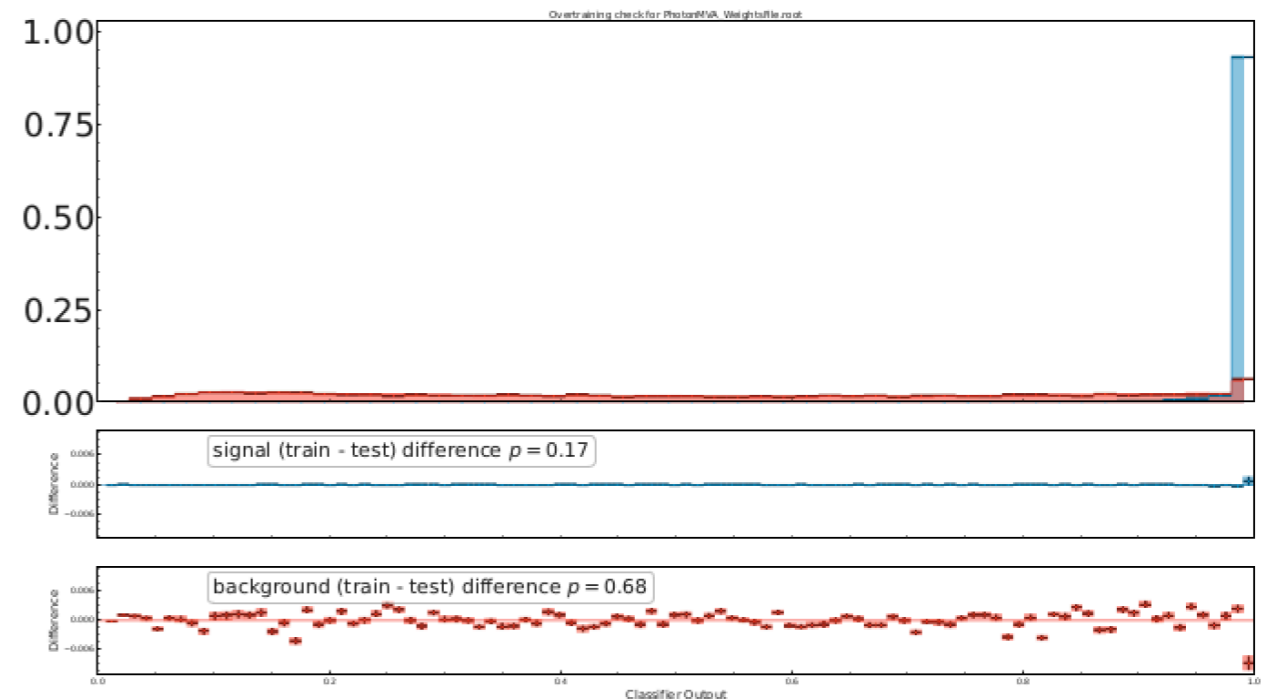
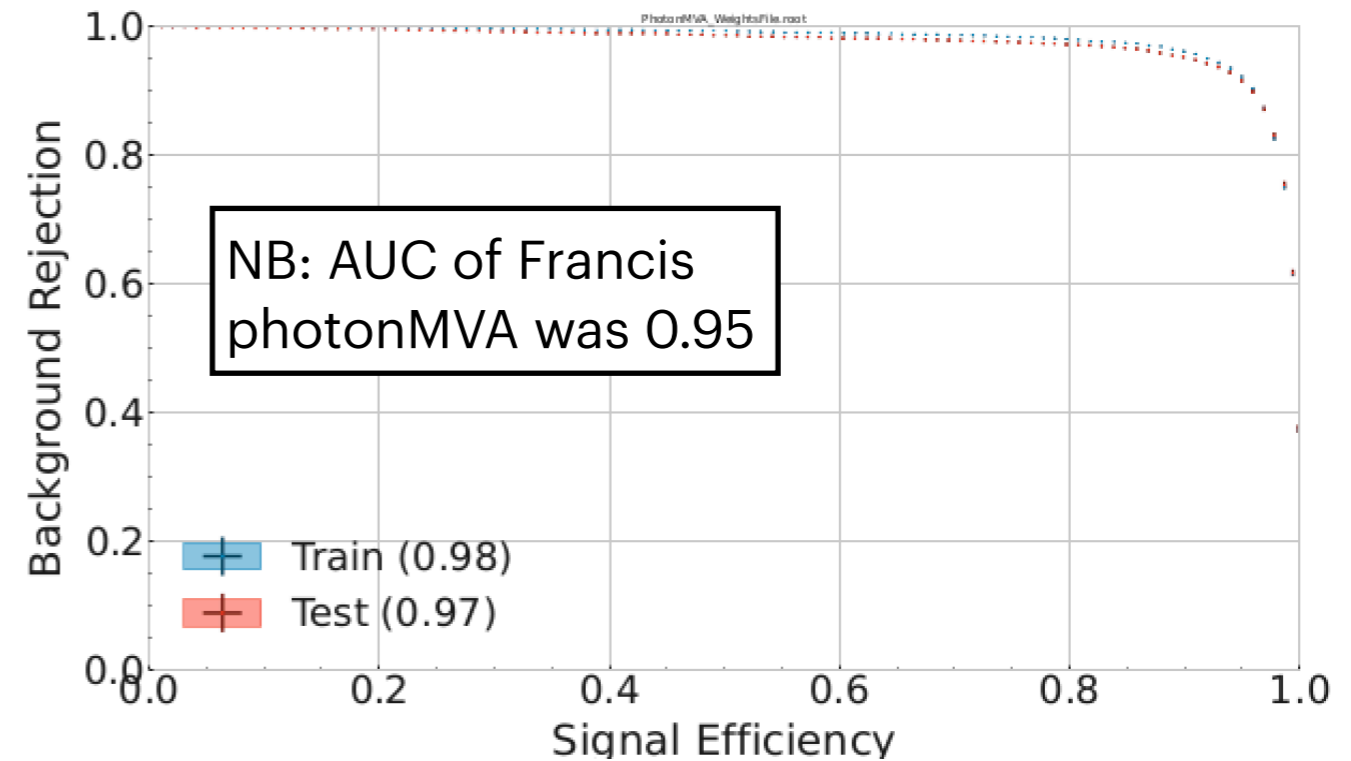
minC2TDist

clusterLAT

clusterNHits

clusterTheta

beamBackgroundSuppression

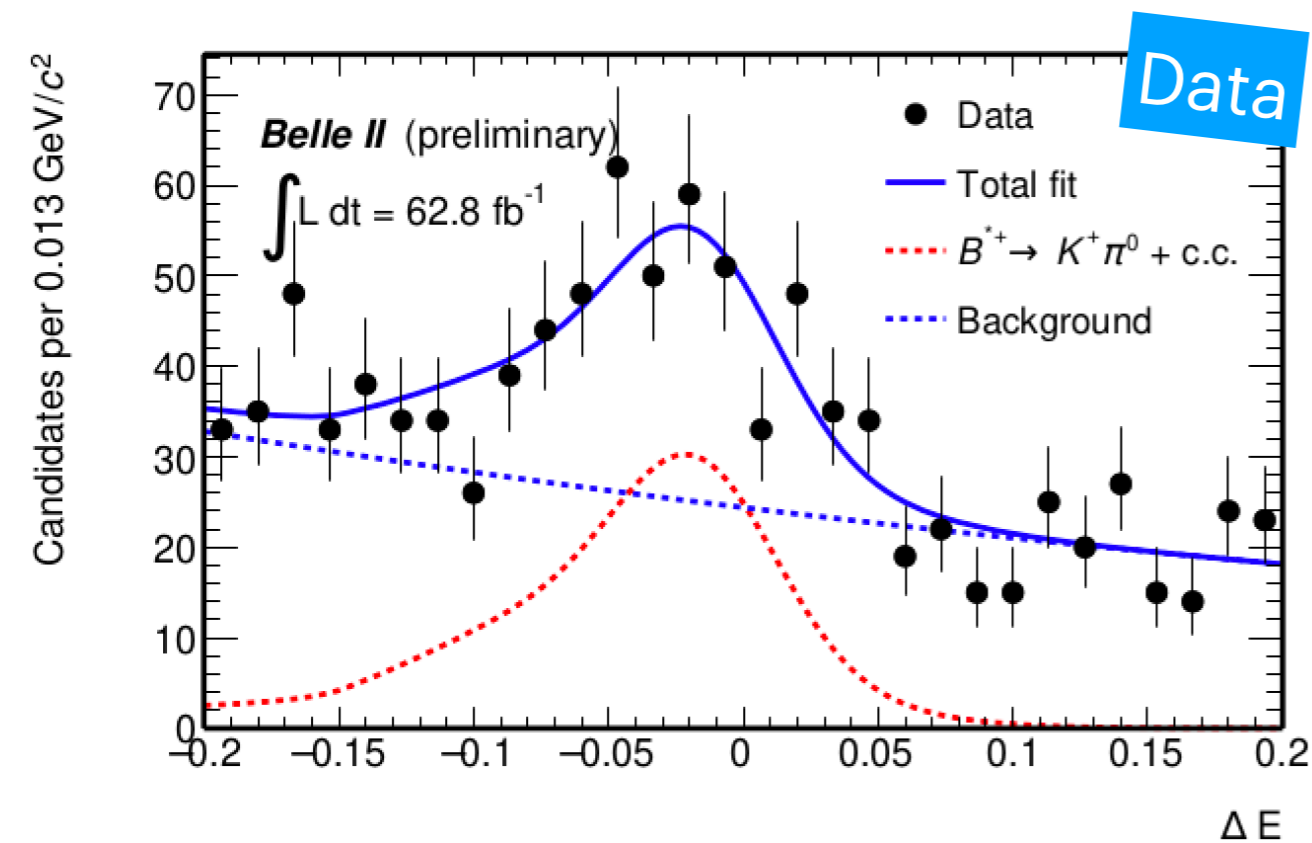


Photon MVA validation

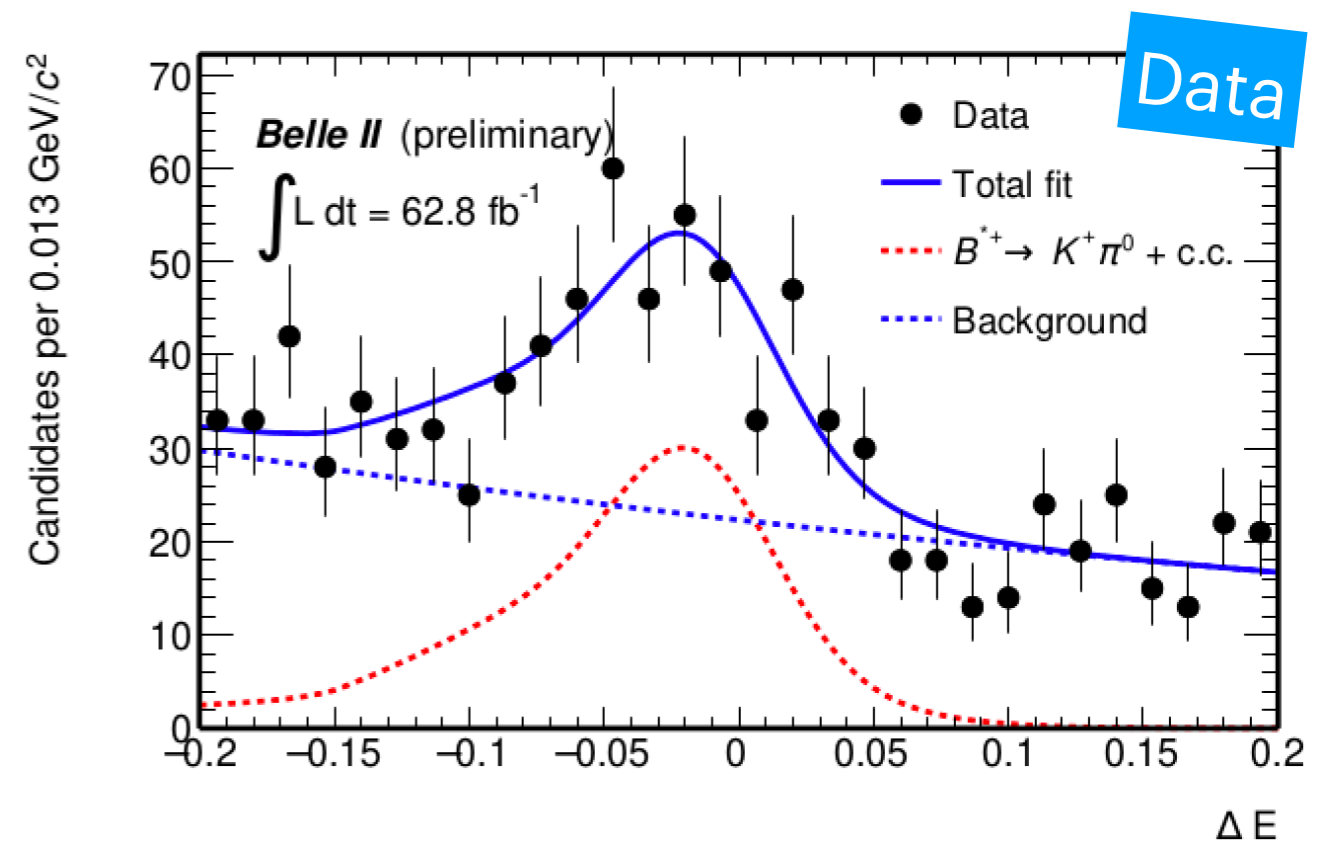
Apply photonMVA to $B^+ \rightarrow K^+ \pi^0$ proc13 sample (chunk1+chunk2 — 62fb⁻¹).

No photonMVA

PhotonMVA > 0.2



Background: 742.64 ± 40.1
Signal: 260.35 ± 33.6



Background: 679.23 ± 38.6 (-8,5%)
Signal: 258.76 ± 32.6 (-0.6%)

PhotonMVA works well.

Still to do: check performance of Francis photonMVA, optimise selection.

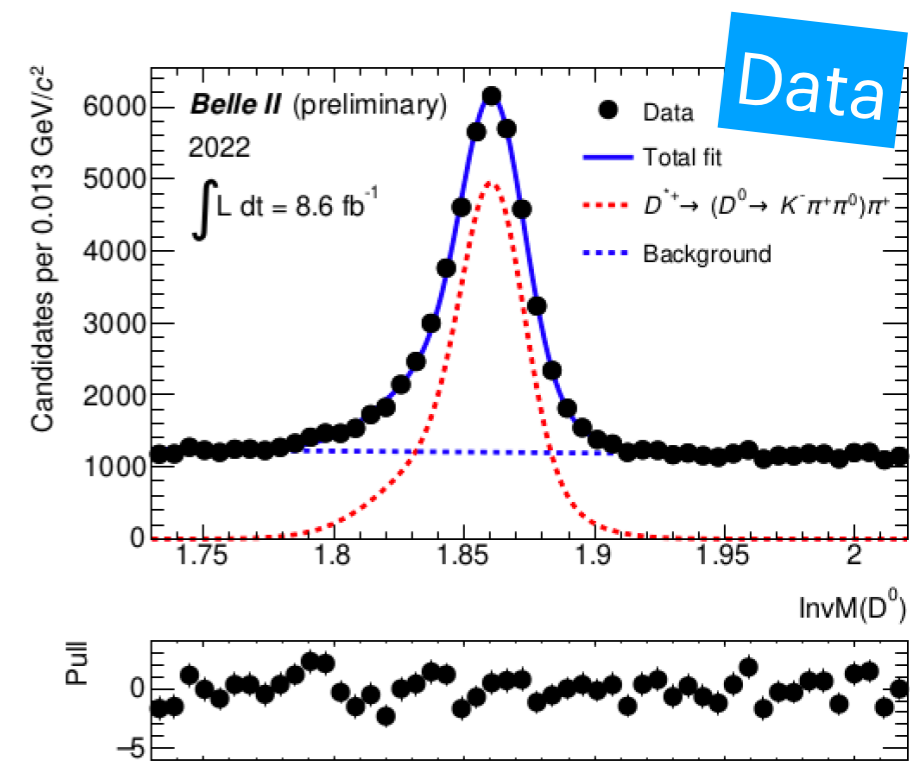
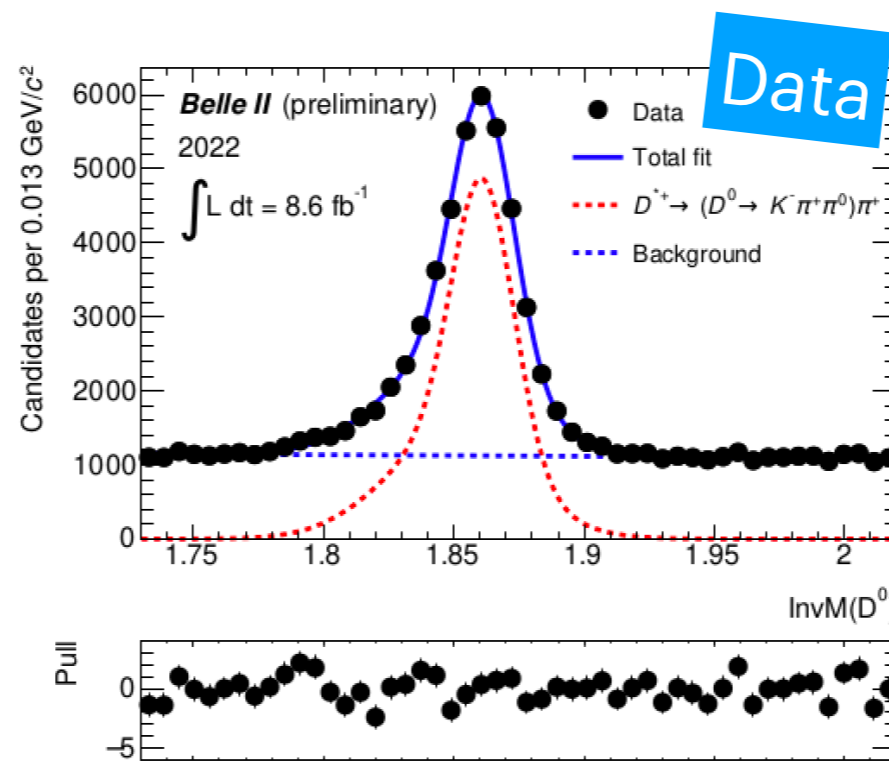
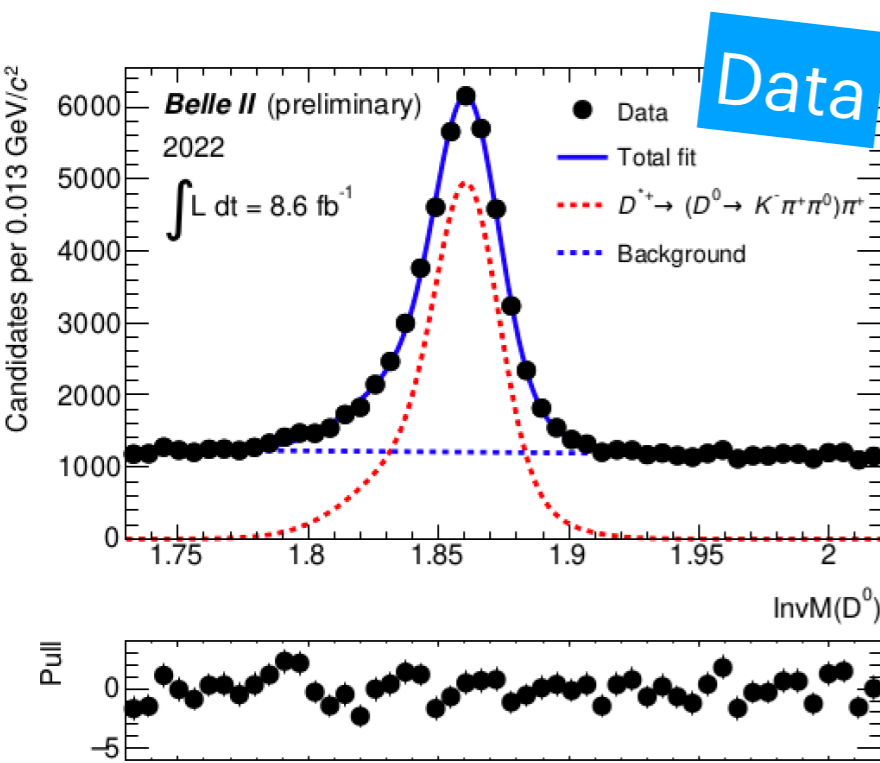
Photon MVA validation

Apply photonMVA to $D^* \rightarrow D^0(K\pi\pi^0)\pi$ proc13 sample (chunk1 — 8.6fb⁻¹).

No photonMVA

My PhotonMVA > 0.05 (optimised)

Francis PhotonMVA > 0.05



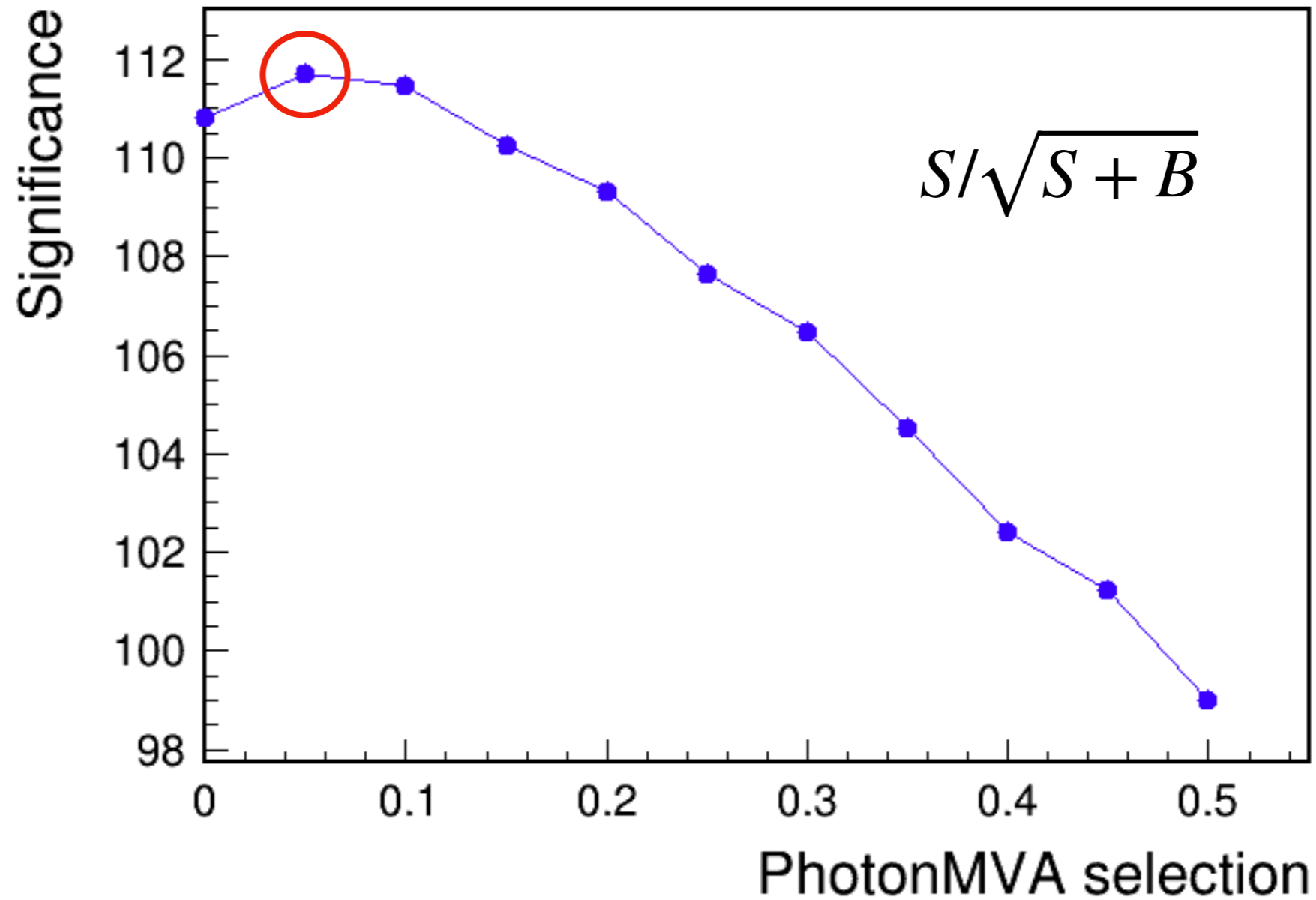
Background: 59902 ± 74
 Signal: 33944 ± 69
 Significance: 110.804

Background: 56066 ± 490 (-6.4%)
 Signal: 33422 ± 528 (-1.5%)
 Significance: 111.724

Background: 57410 ± 70 (-4.1%)
 Signal: 33519 ± 61 (-1.3%)
 Significance: 111.158

PhotonMVA works well (and better wrt Francis).

PhotonMVA selection optimisation



CSBDT

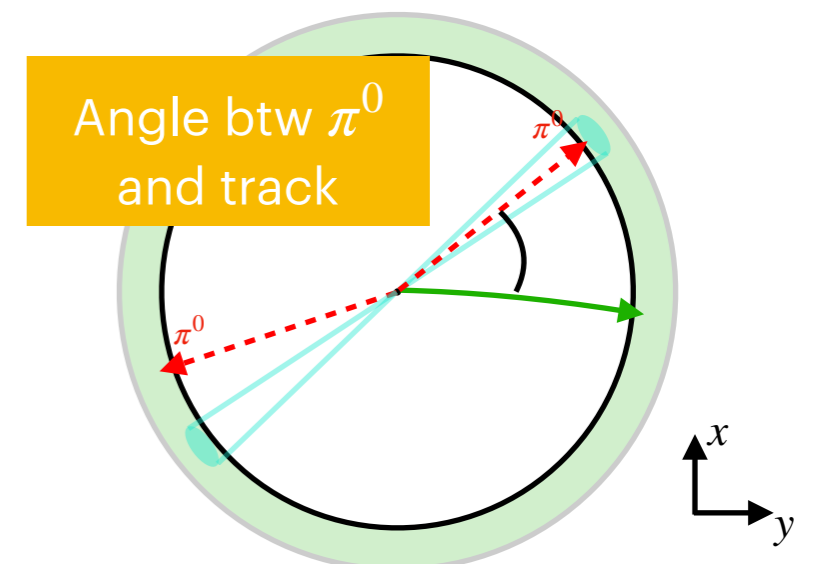
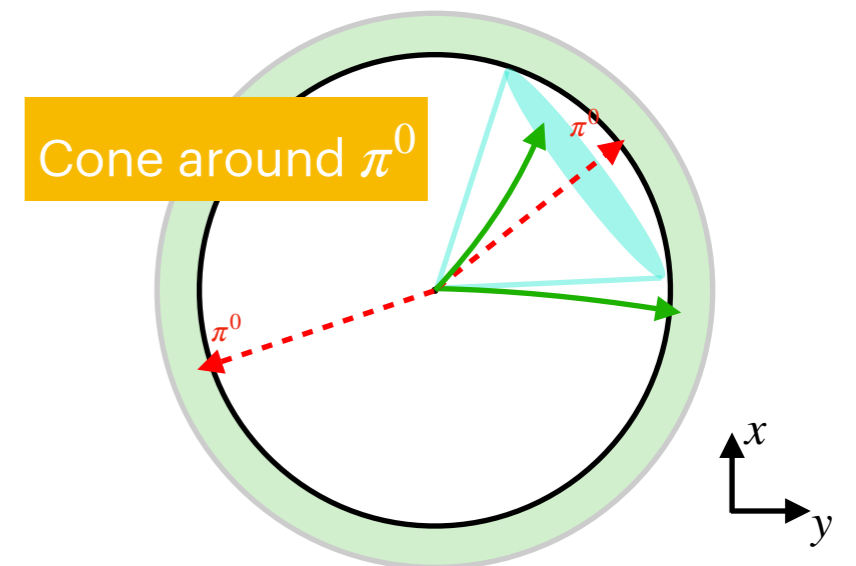
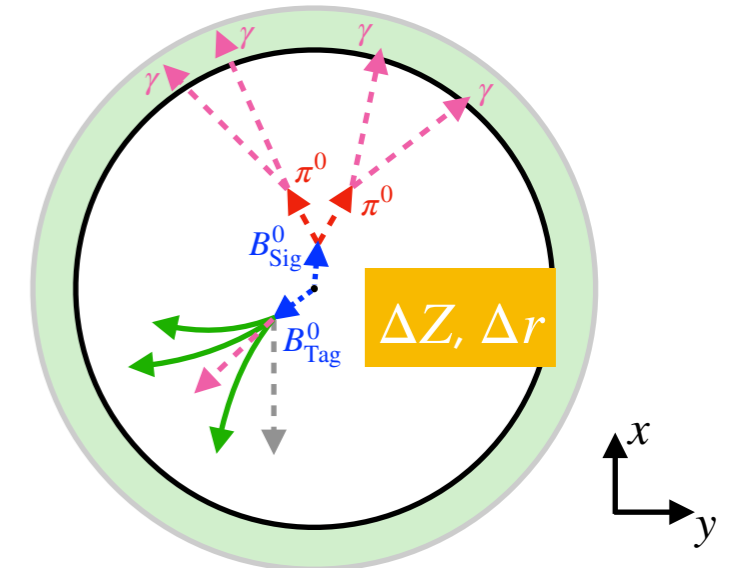
CSBDT summary

New possible inputs:

Create continuum-suppression BDT using event-shape variables and B_{Tag} variables, avoiding large correlations (<10% — was 5% for Francis) and/or sculpting.

Must check if the use of B_{Tag} variables sculpts or introduces large correlations in the flavour tagger variables.

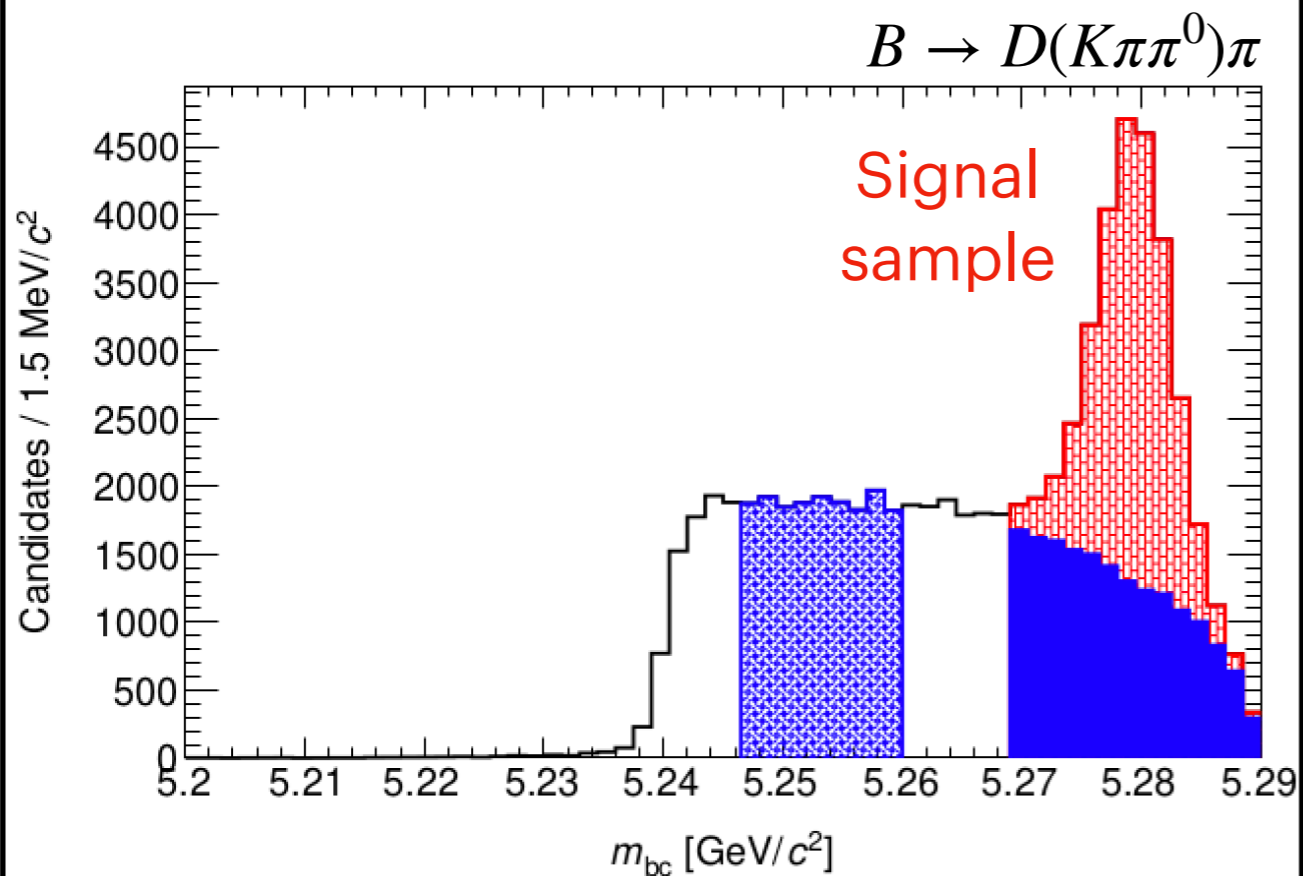
Note: 6.7% of the signal events doesn't have a B_{Tag} vertex \rightarrow remove these events (bkg: -9.4%).



CSBDT: inputs validation

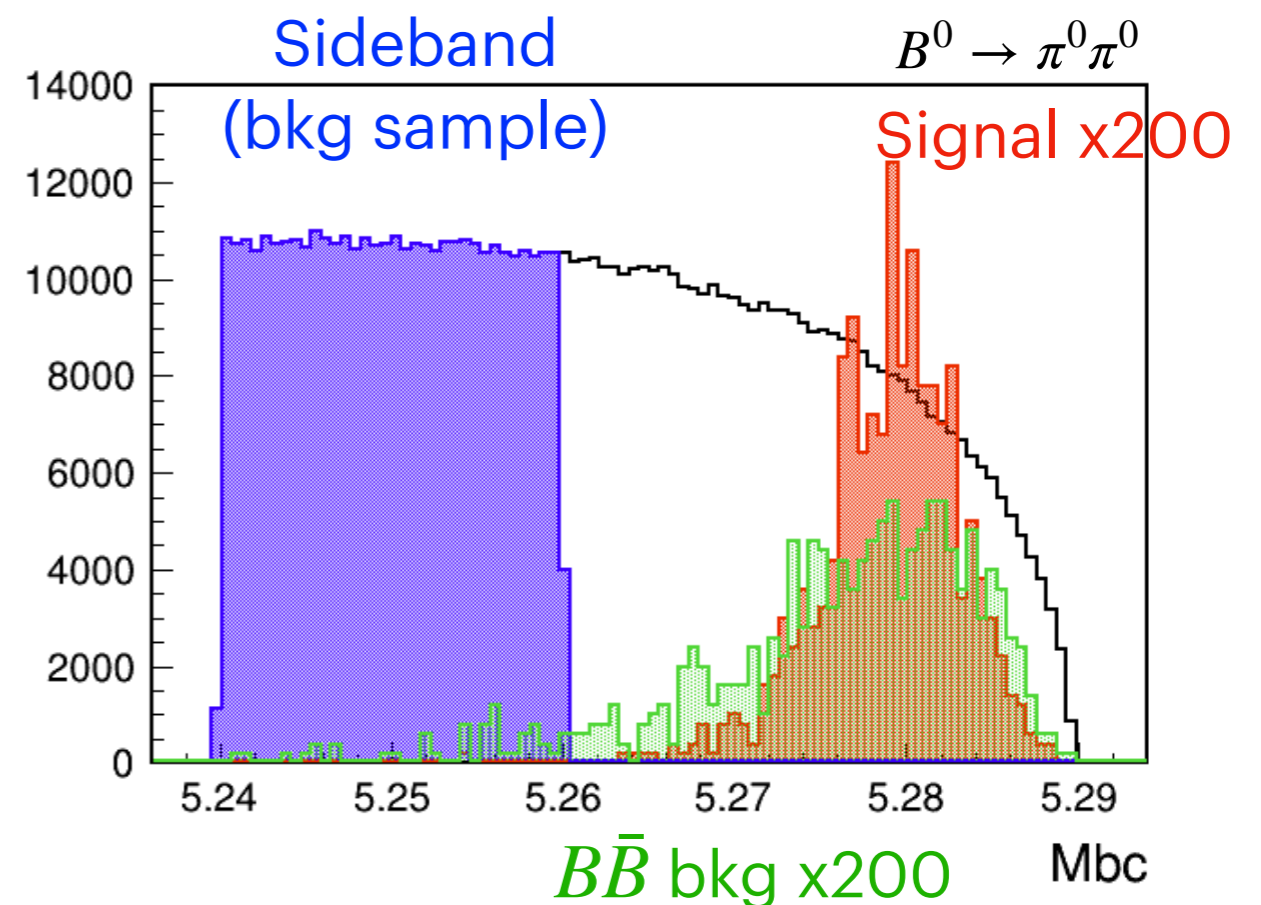
Signal: use $B \rightarrow D(K\pi\pi^0)\pi$ sideband-subtracted data (proc13) and sideband-subtracted $B \rightarrow D(K\pi\pi^0)\pi$ MC15

Do not use $B \rightarrow D(K\pi\pi^0)\pi$ for bkg because of the different compositions



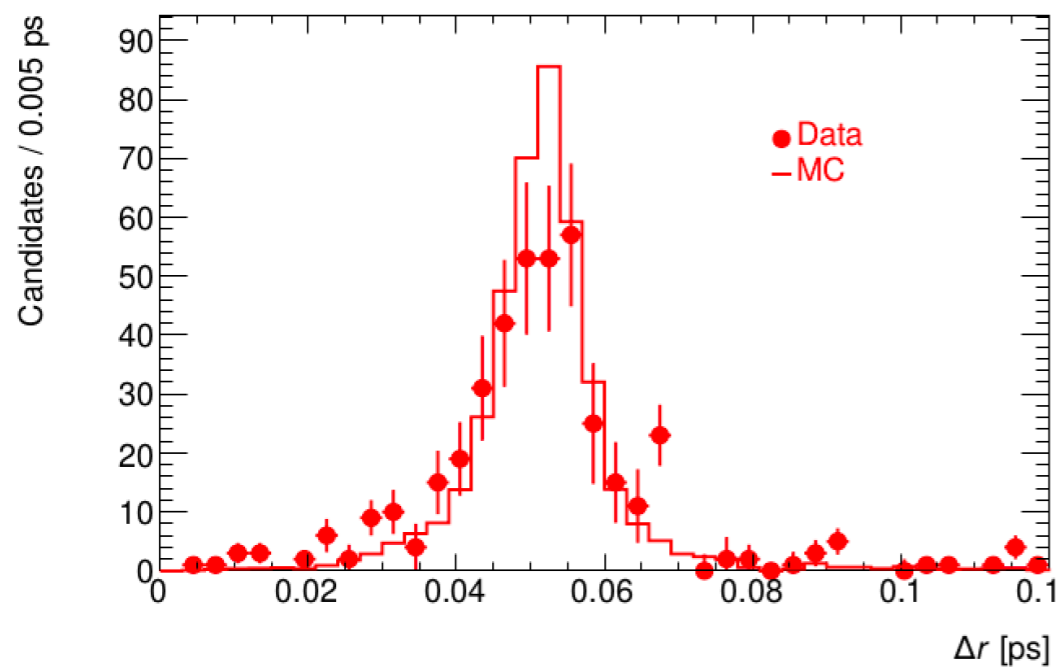
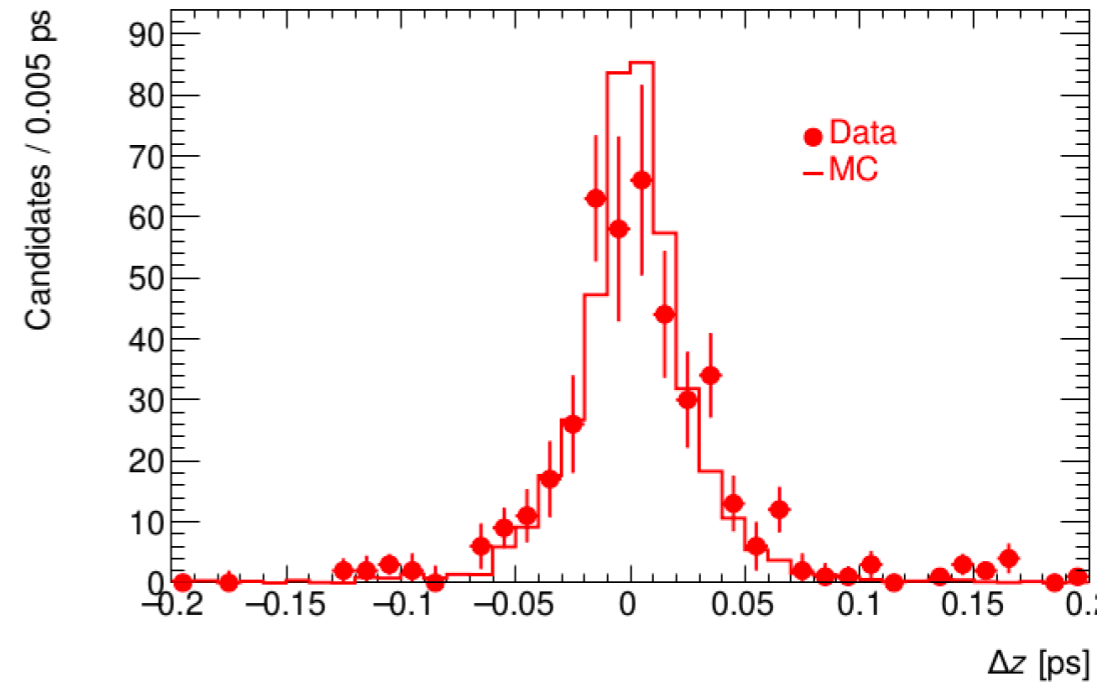
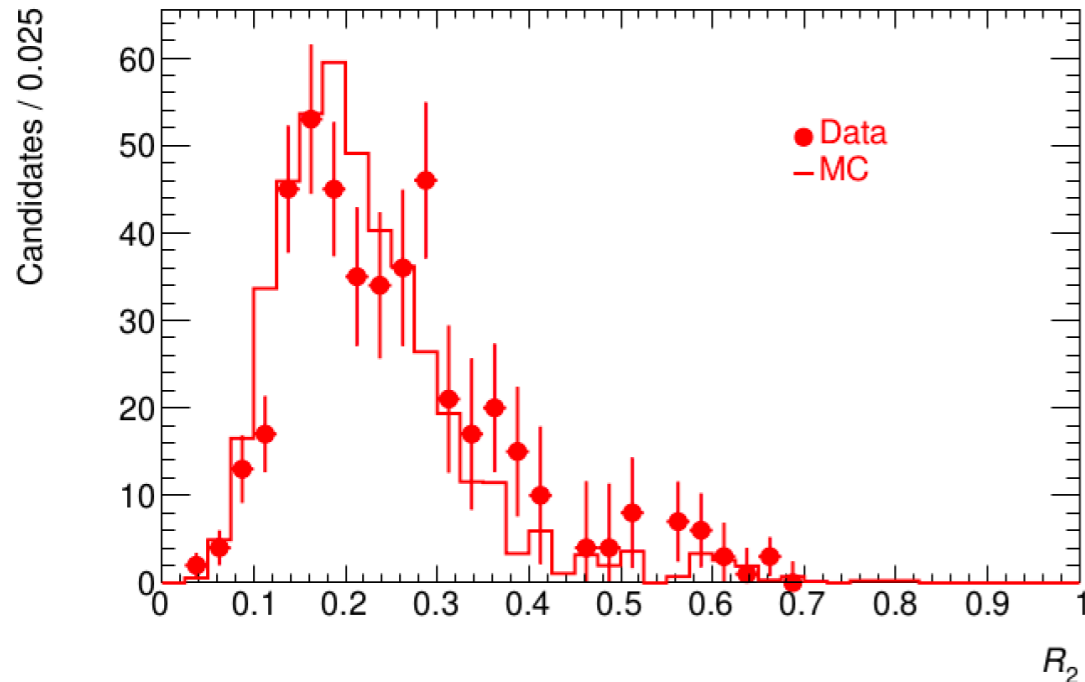
Background: use $B^0 \rightarrow \pi^0\pi^0$ sideband data (proc13) and $B^0 \rightarrow \pi^0\pi^0$ sideband MC15

Need to check if bkg composition is the same in sideband and signal region



Inputs validation — Signal only

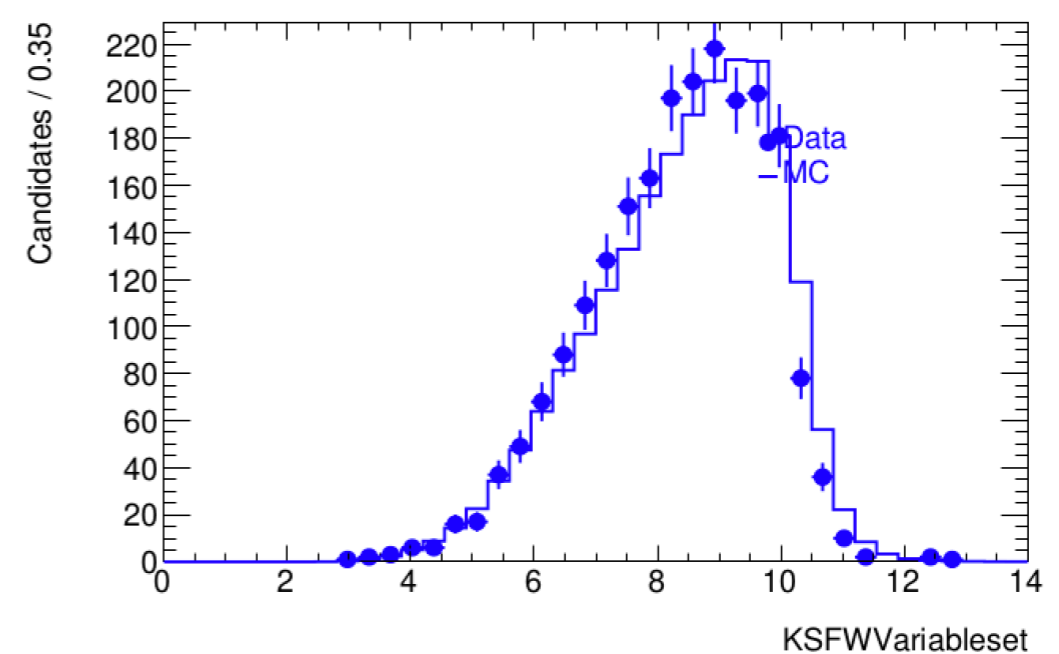
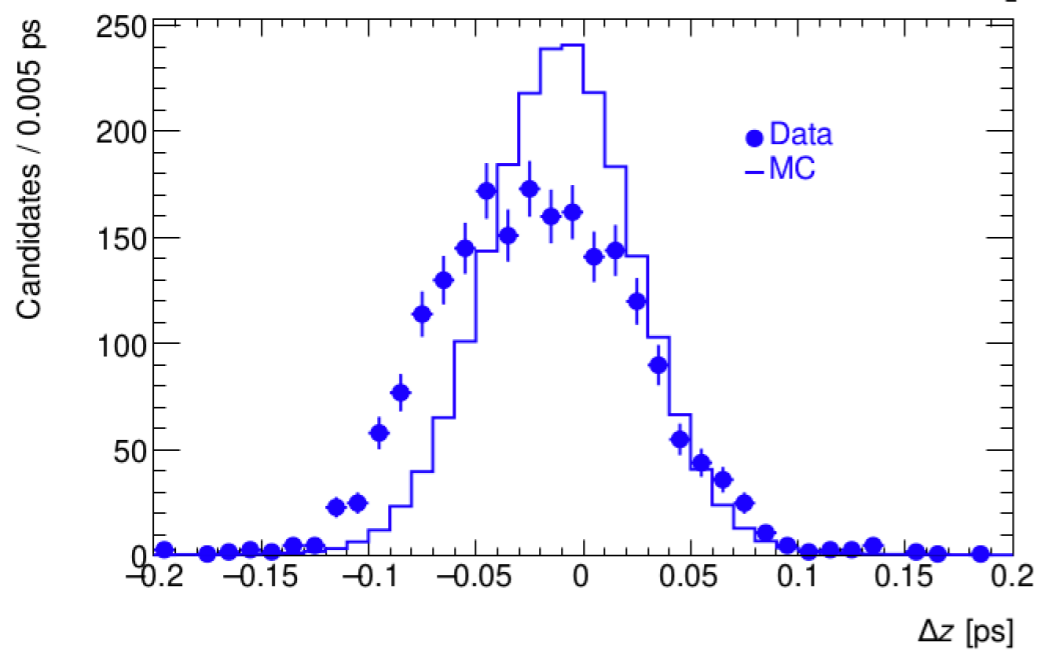
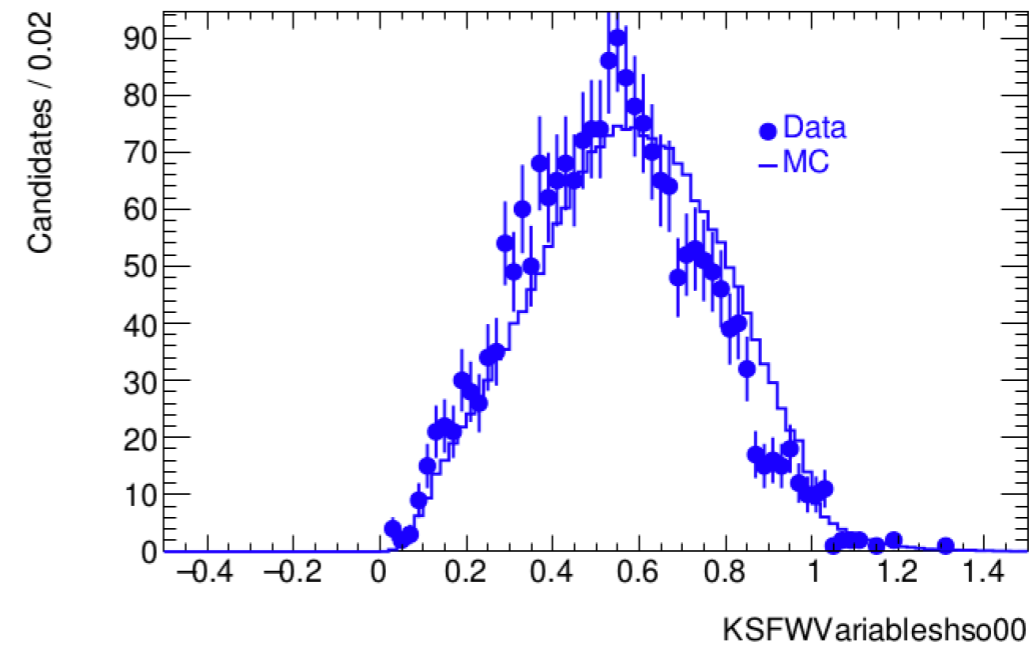
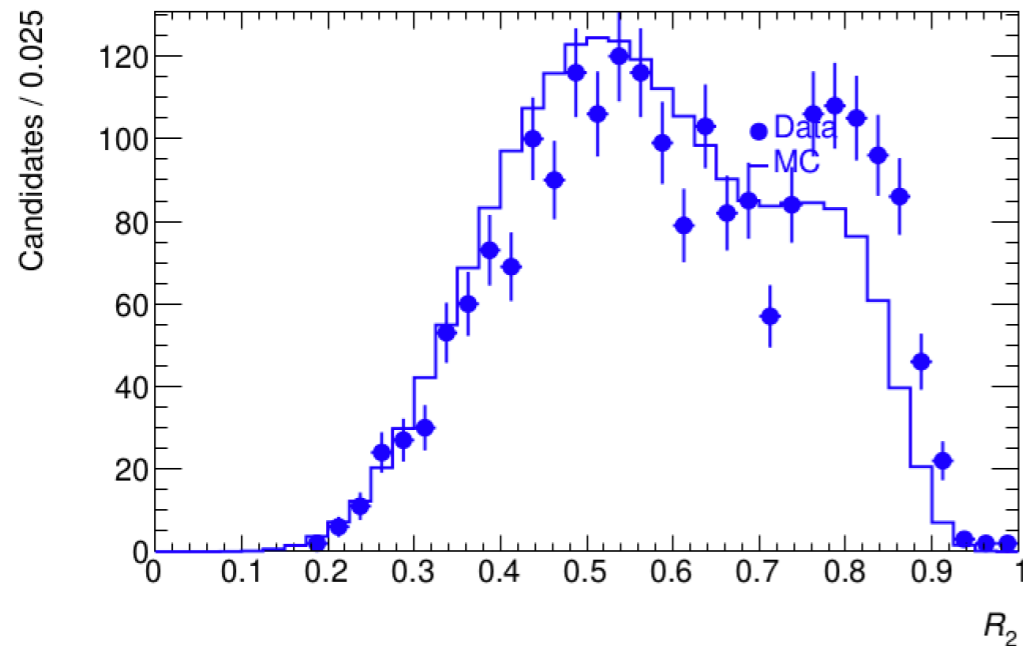
Use $B \rightarrow D(K\pi\pi^0)\pi$ sideband-subtracted data (proc13) and sideband-subtracted $B \rightarrow D(K\pi\pi^0)\pi$ MC15.



Sample has poor statistics, but do not observe any large discrepancy.

Inputs validation — Background only

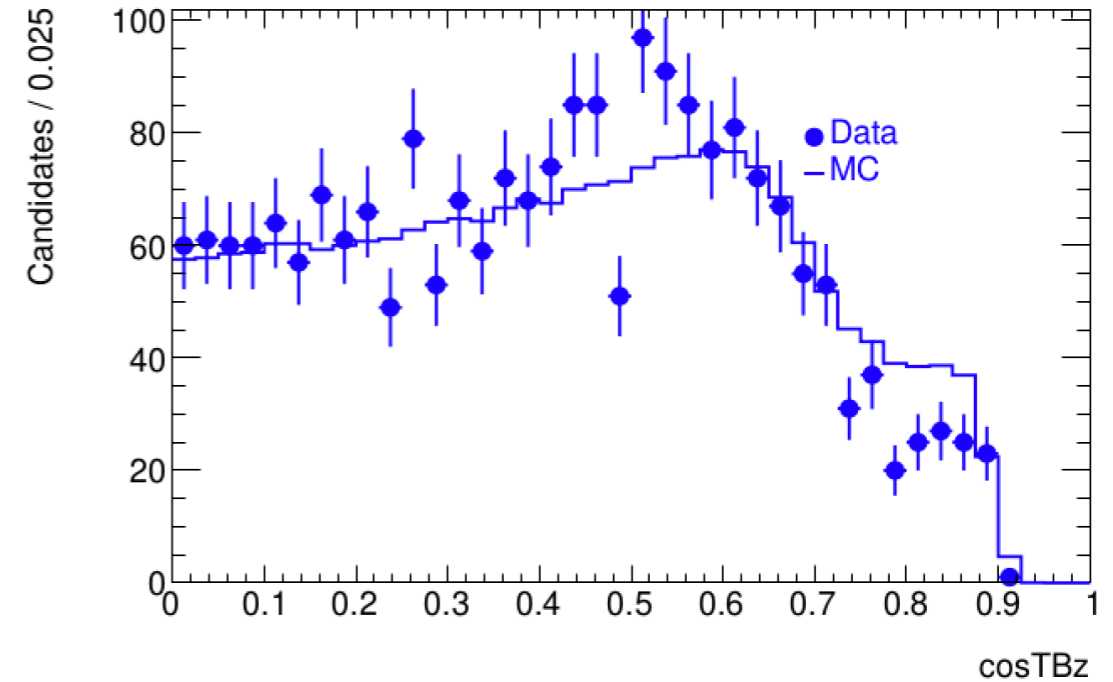
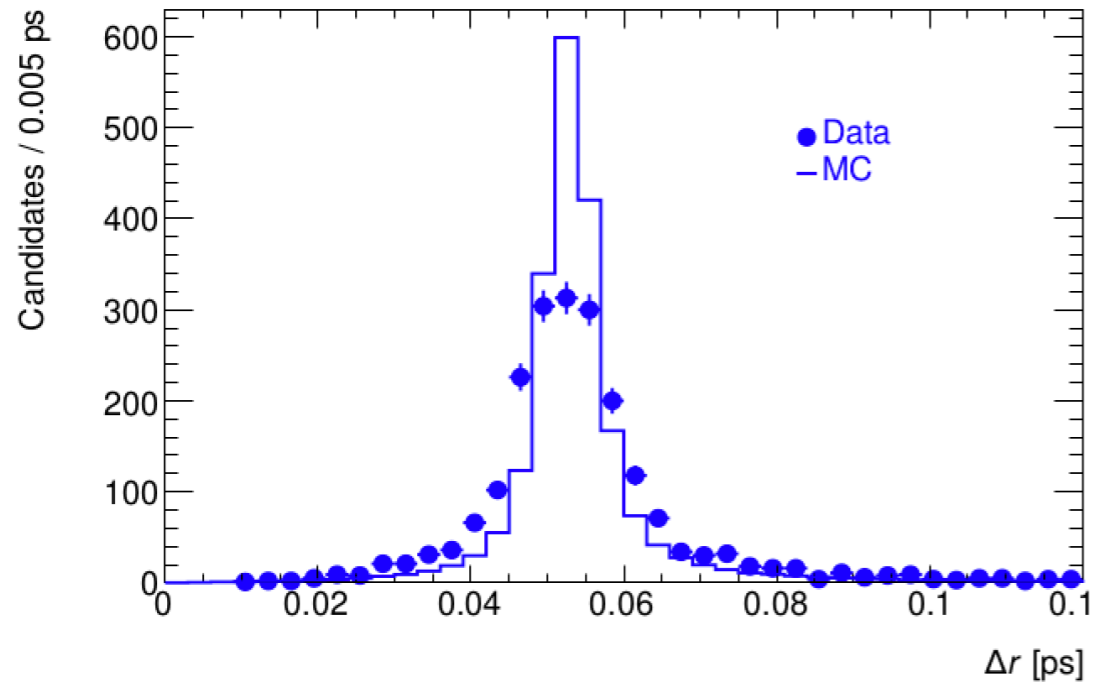
Use $B^0 \rightarrow \pi^0\pi^0$ sideband data (proc13) and $B^0 \rightarrow \pi^0\pi^0$ sideband MC15



Observe variables with some discrepancies.

Inputs validation — Background only

Use $B^0 \rightarrow \pi^0\pi^0$ sideband data (proc13) and $B^0 \rightarrow \pi^0\pi^0$ sideband MC15

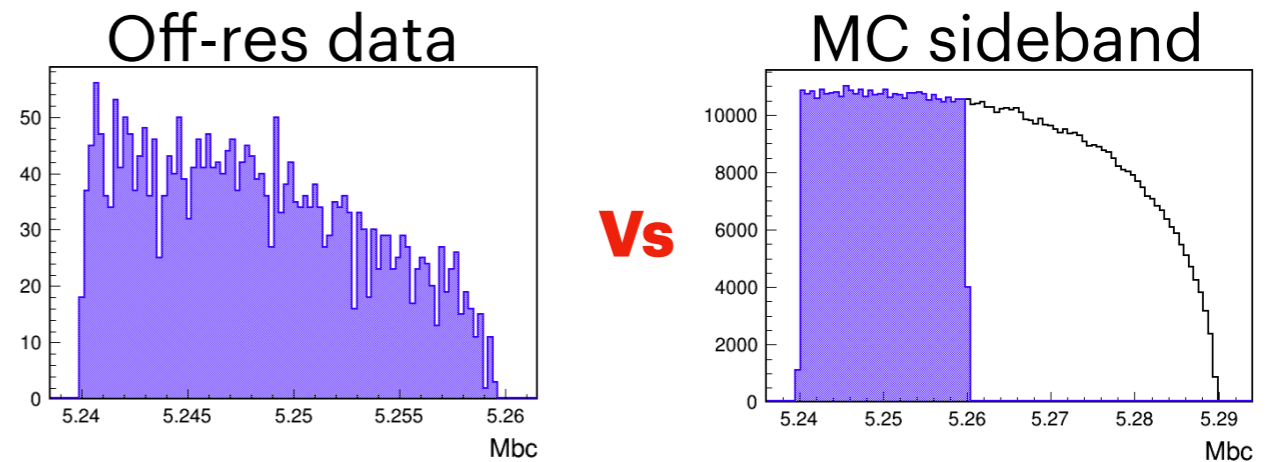


Observe variables with some discrepancies.

Better to use directly sideband data to train the CSBDT?

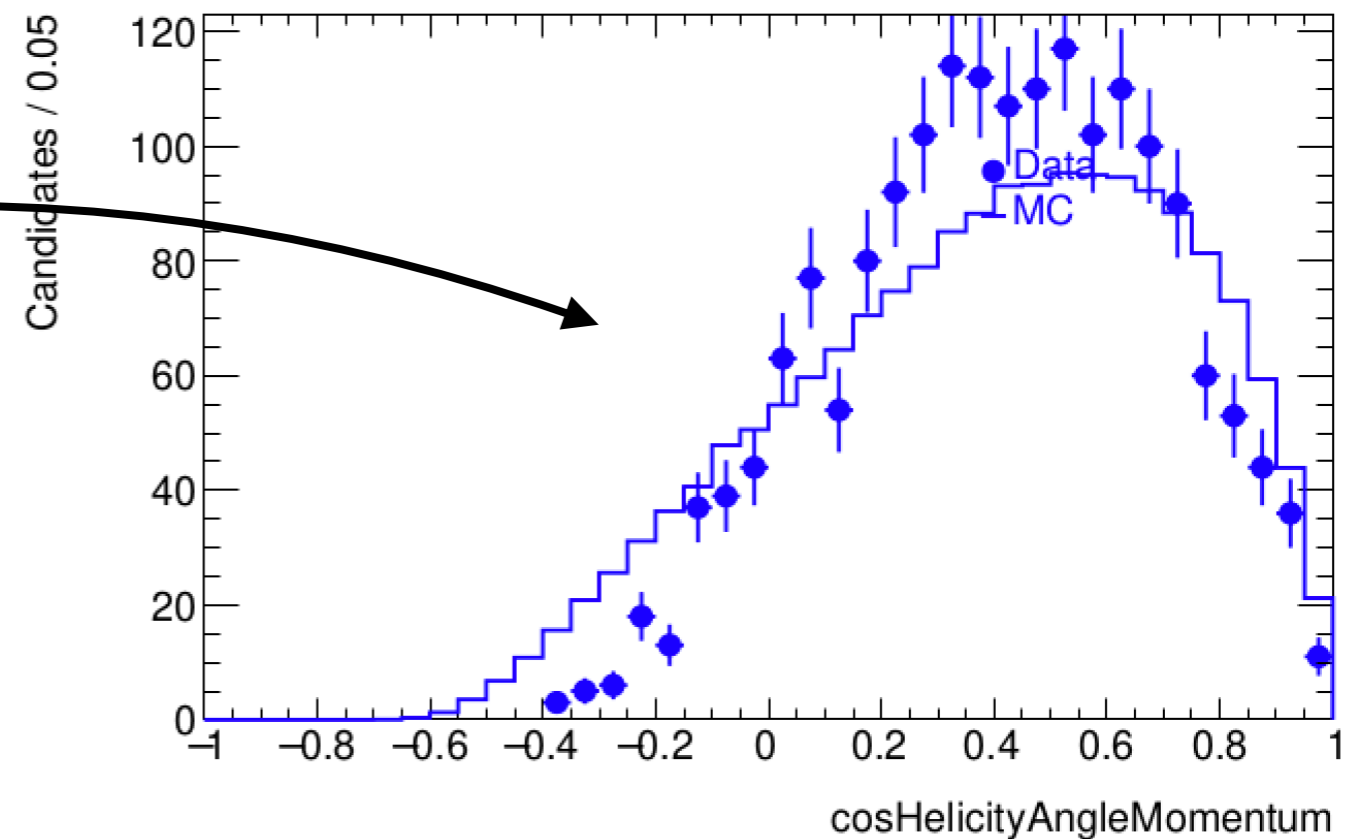
Check — Continuum in **off-res data** and **MC sideband**

Use $B^0 \rightarrow \pi^0 \pi^0$ continuum in **off-res data** and in **MC15 sideband**.



Observe discrepancies not present in sideband data/MC comparison.

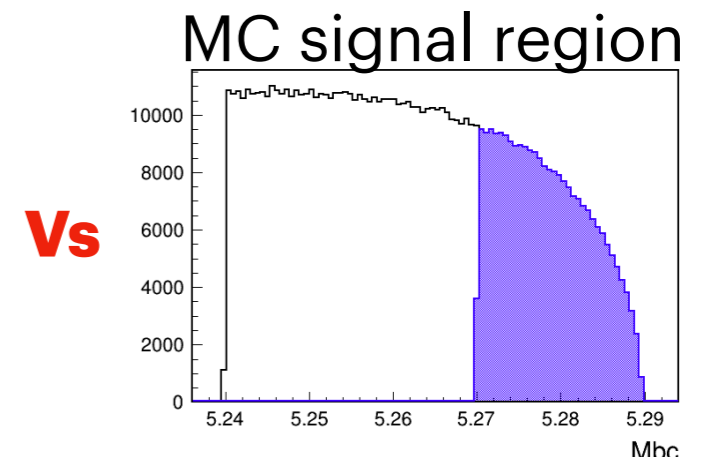
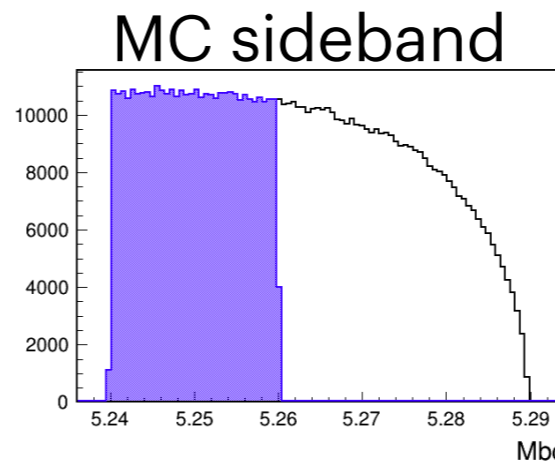
Data/MC discrepancies or different kinematic distributions?



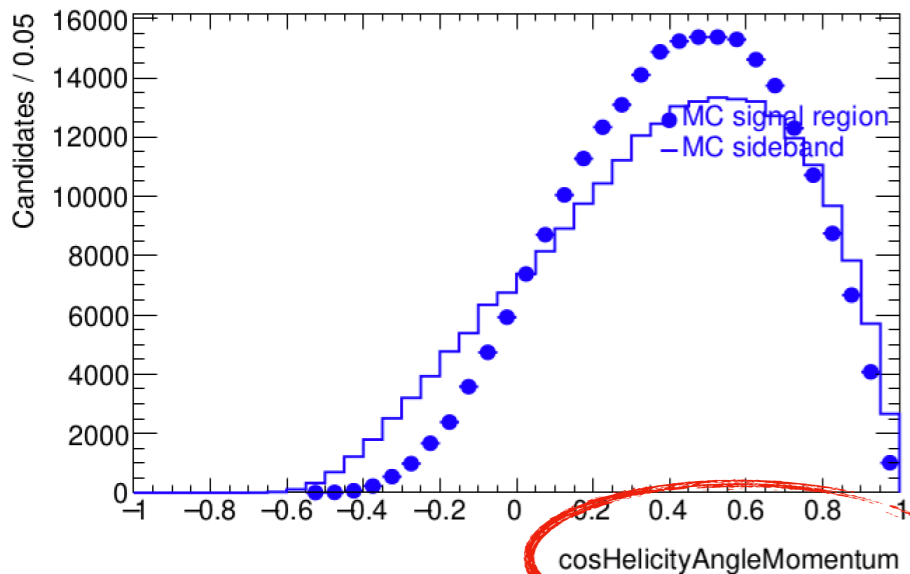
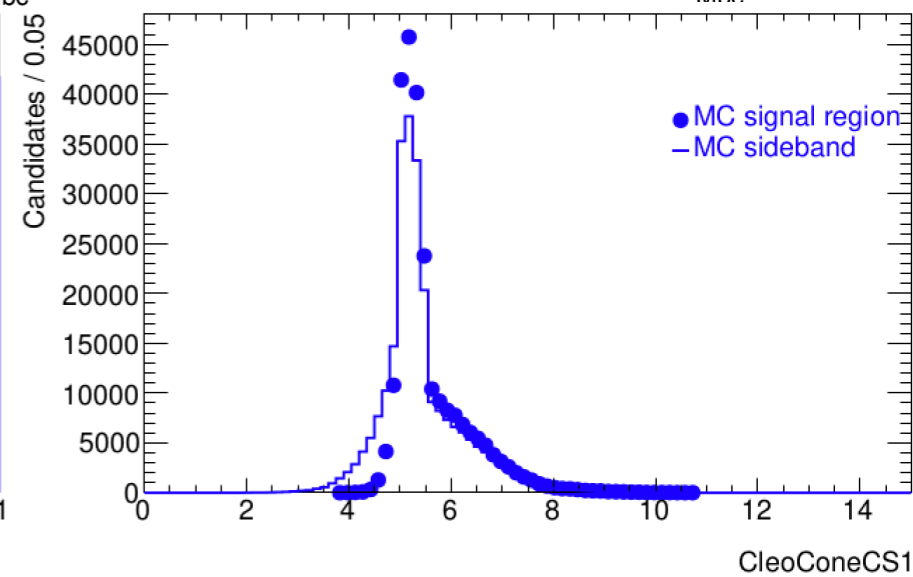
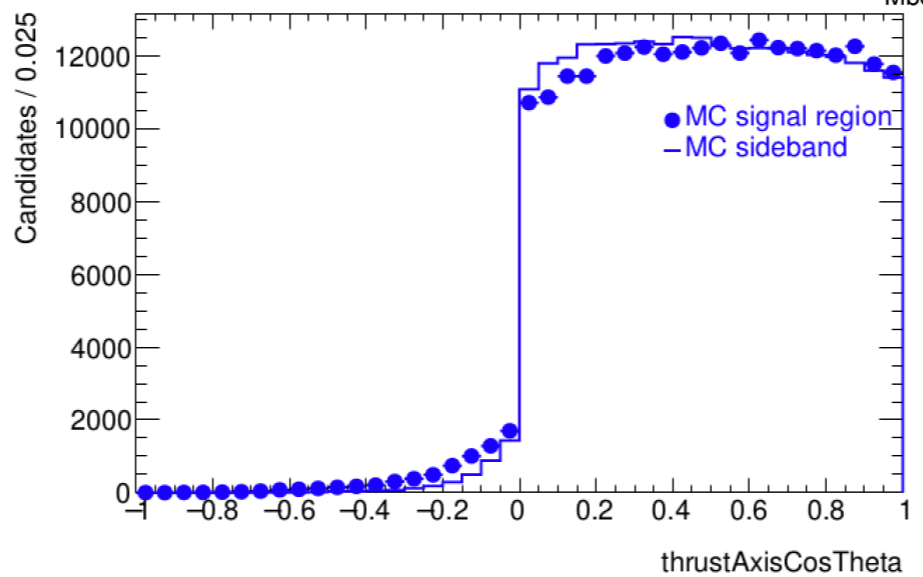
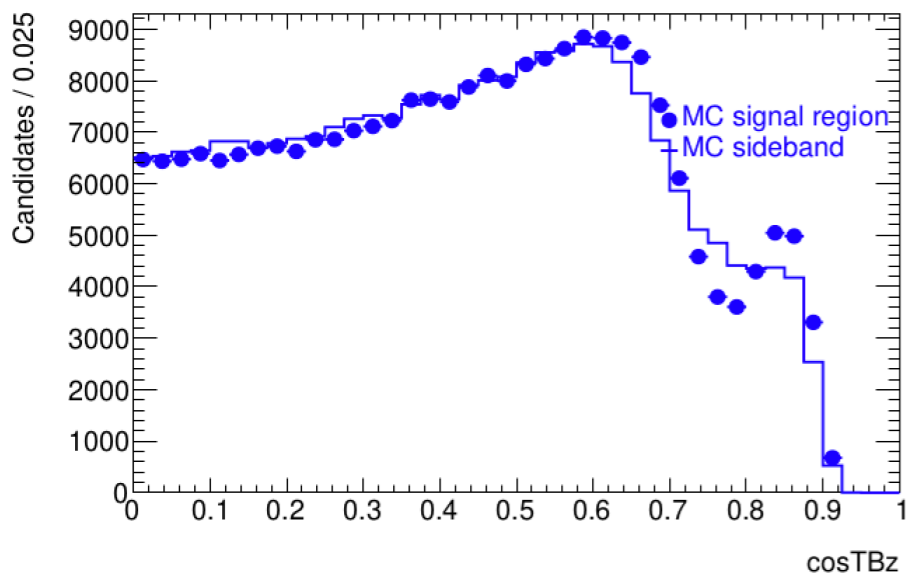
Must check if continuum in sideband and signal region have same distributions.

Check – Continuum in MC sideband and in MC signal region

Use MC15 $B^0 \rightarrow \pi^0 \pi^0$ continuum in sideband and in signal region.



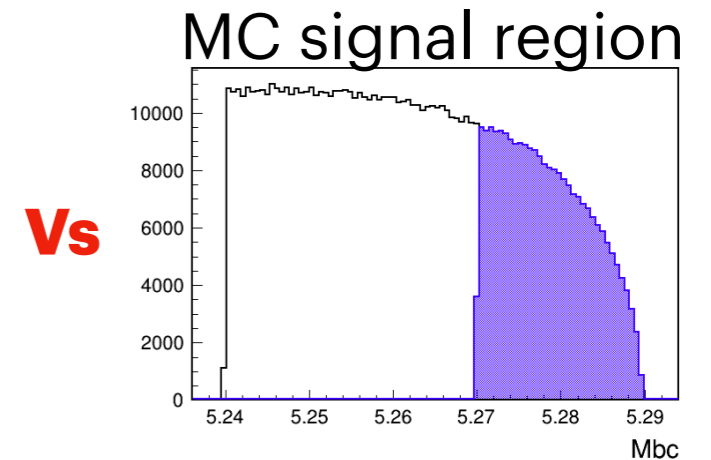
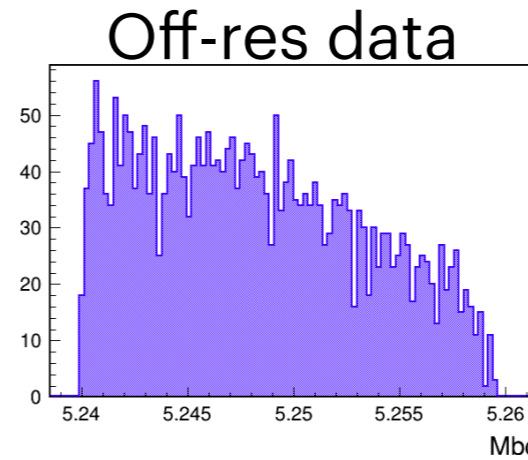
Vs



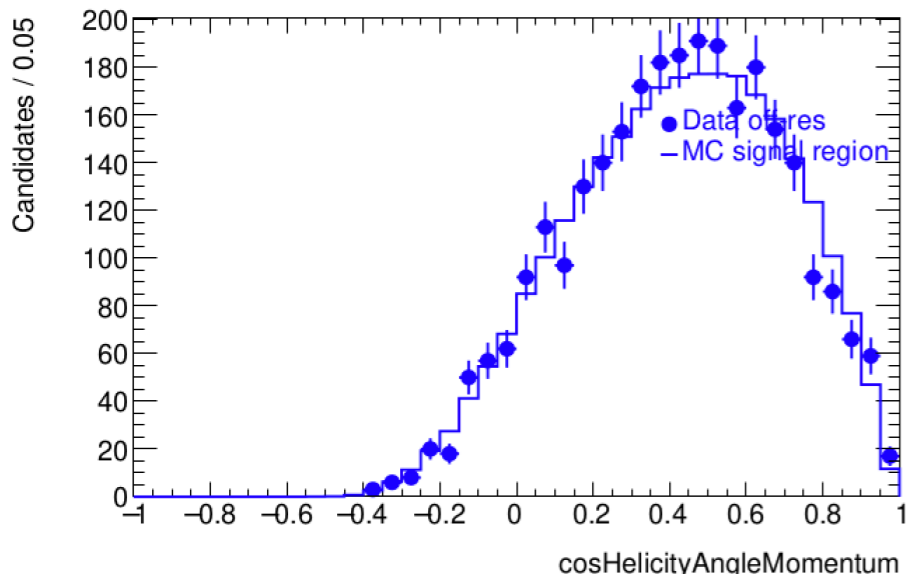
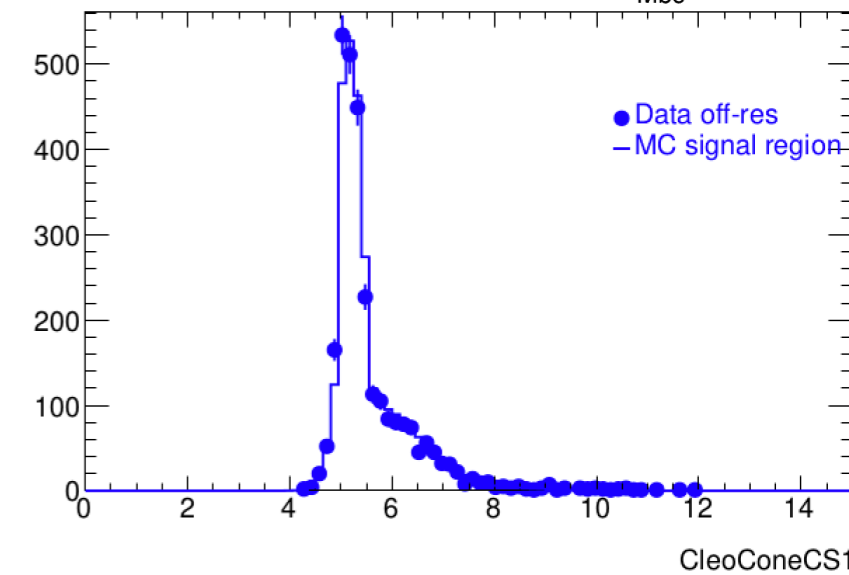
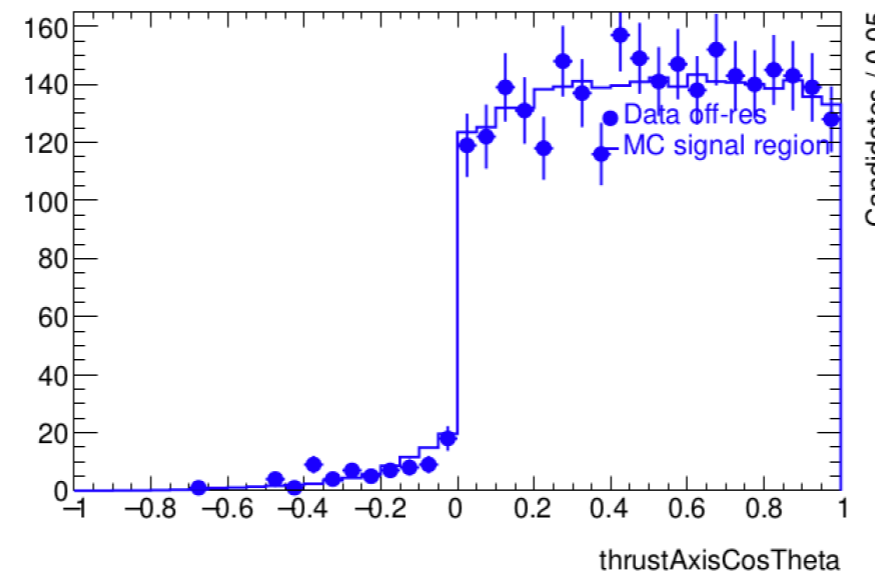
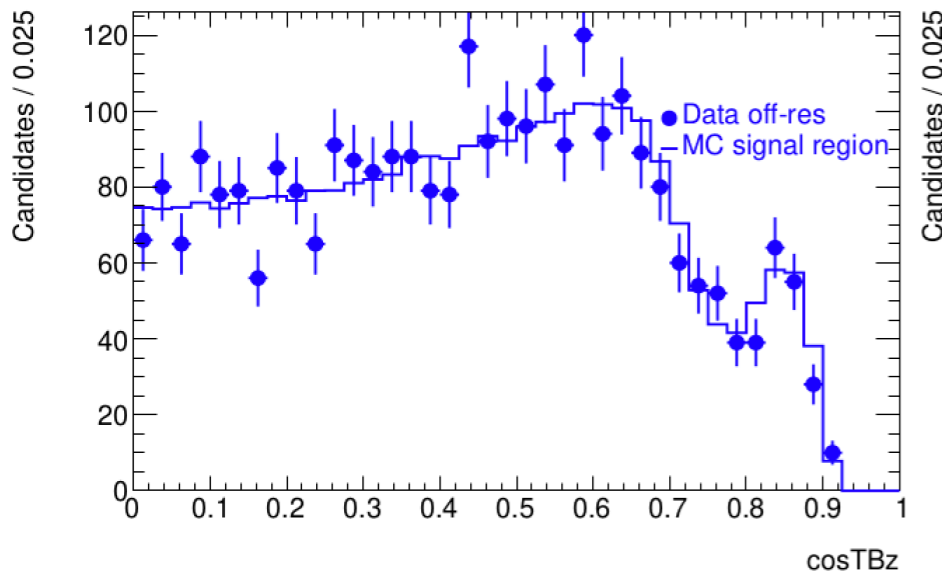
Few important discrepancies: maybe off-resonance data is more reliable wrt sideband data?

Check — Continuum in off-res data and in MC signal region

Use $B^0 \rightarrow \pi^0 \pi^0$ continuum in off-res data and in MC15 signal region.



Vs



Discrepancies disappeared: off-resonance data **is** more reliable wrt sideband data.

Note: input validation using off-res data shows anyway many data/MC discrepancies → best is to use off-res data for CSBDT training.

CSMVA inputs

Train on **MC sample** after applying all π^0 selections.

Inputs (after pruning)

7 Kakuno-Super-Fox-Wolfram moments

cosTBTO

1 CleoCone

cosTheta*

R2

thrustOm

ΔZ (BTag)

Δr (BTag)

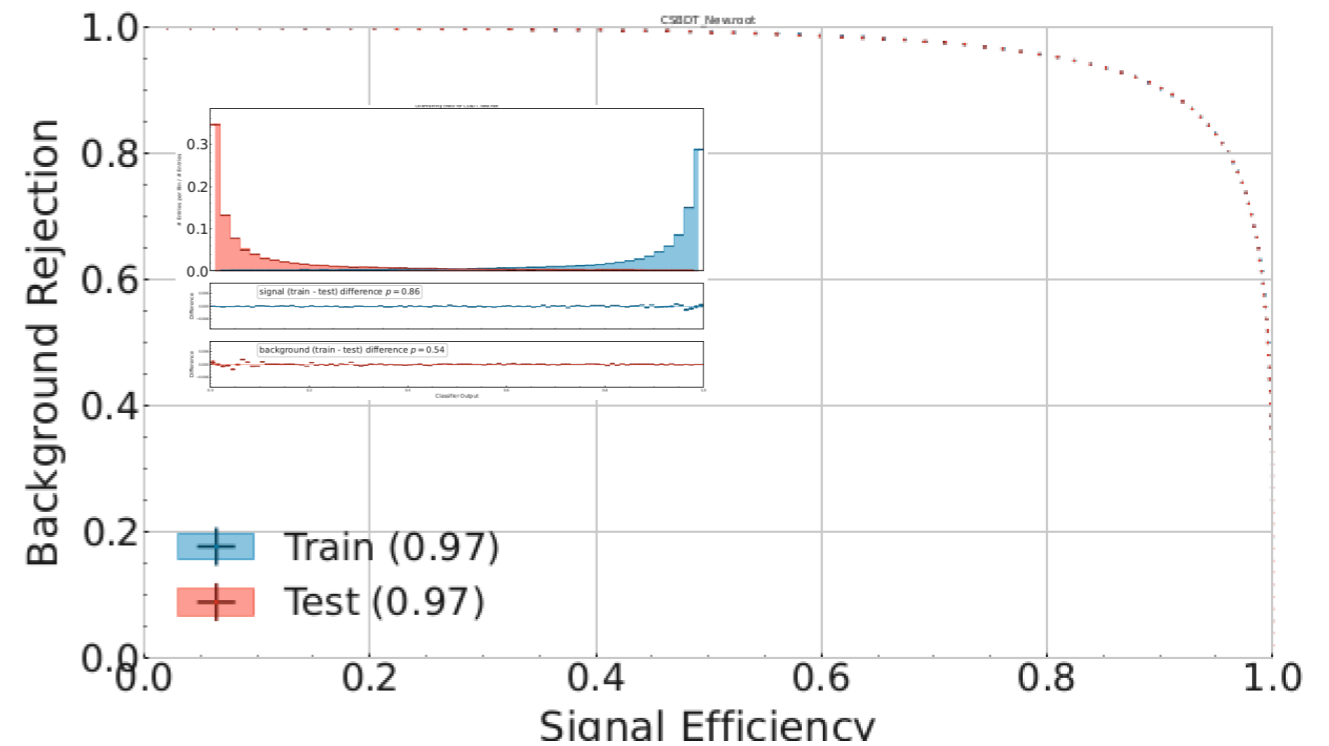
thrustAxisCosTheta

angle between π^0 's

cosHelicityAngle

KSFVVariableset

KSFVVariablesmm2

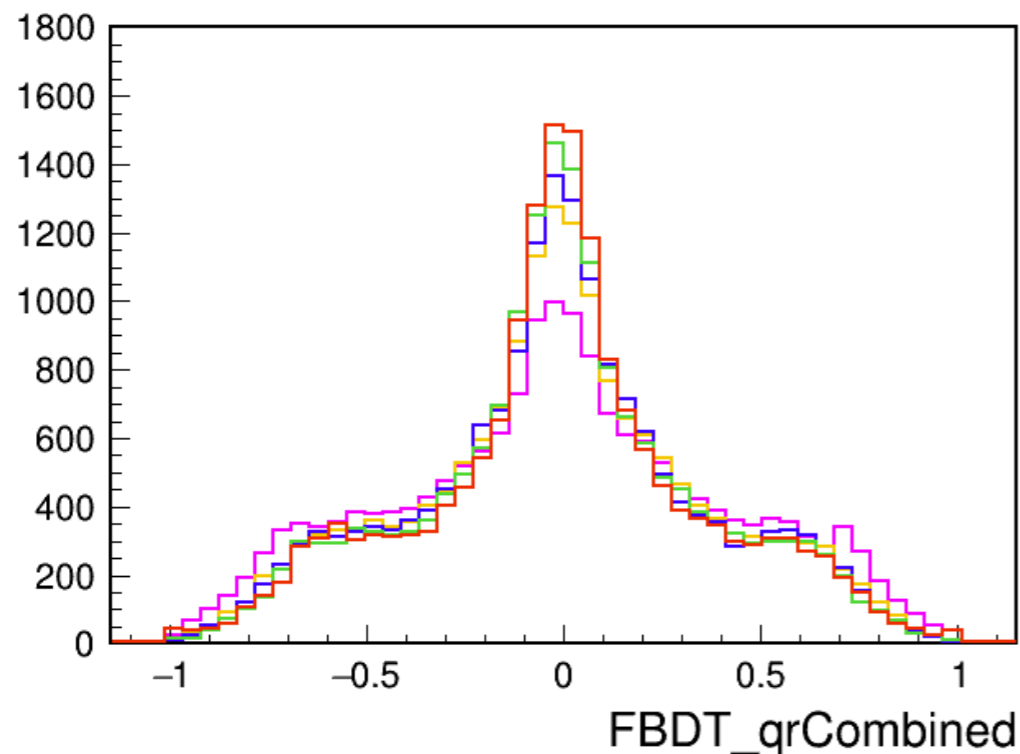
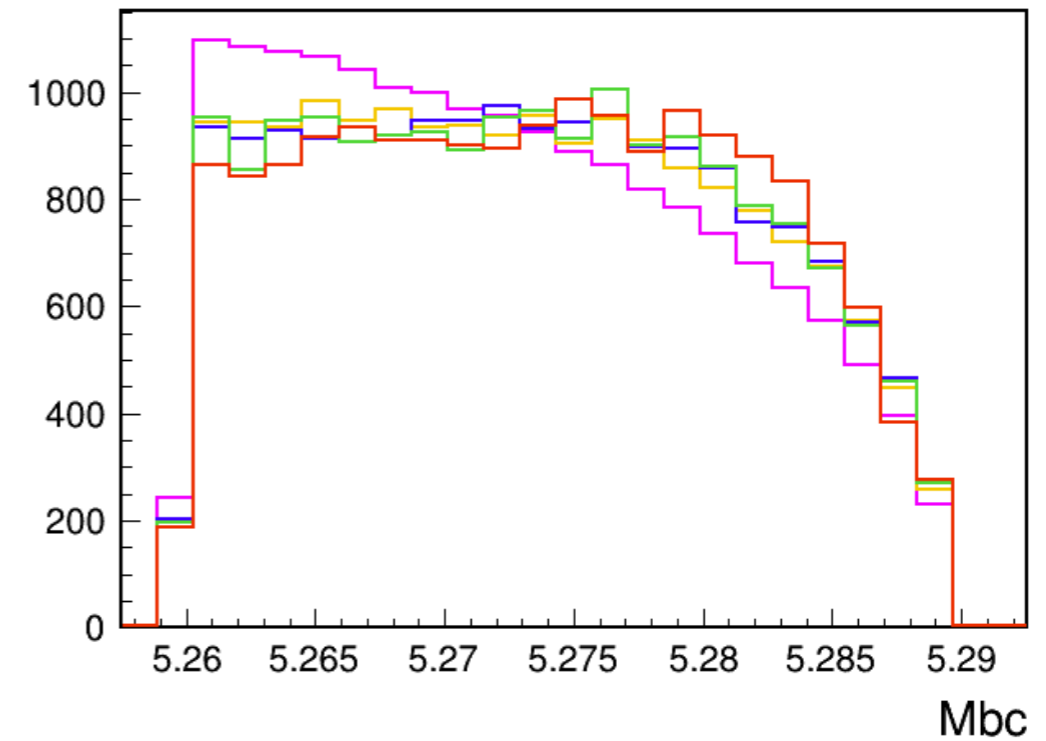
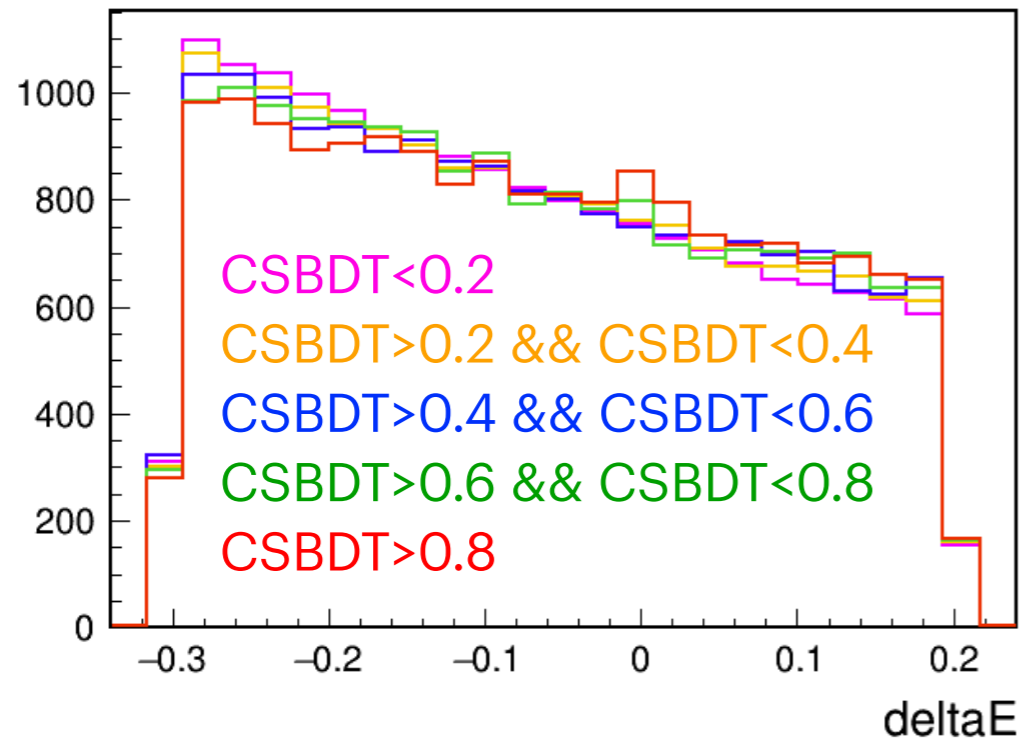


Better performance wrt old BDT
(AUC=0.95).

Will repeat this using off-res data
for the bkg training.

CSBDT dependences with fit variables

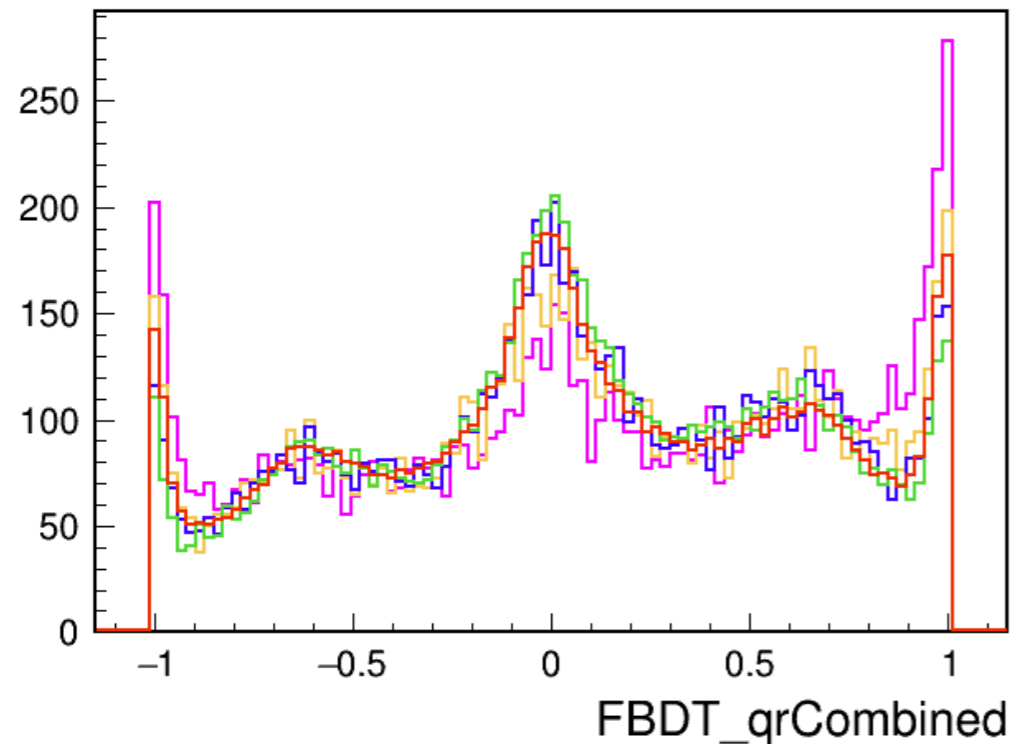
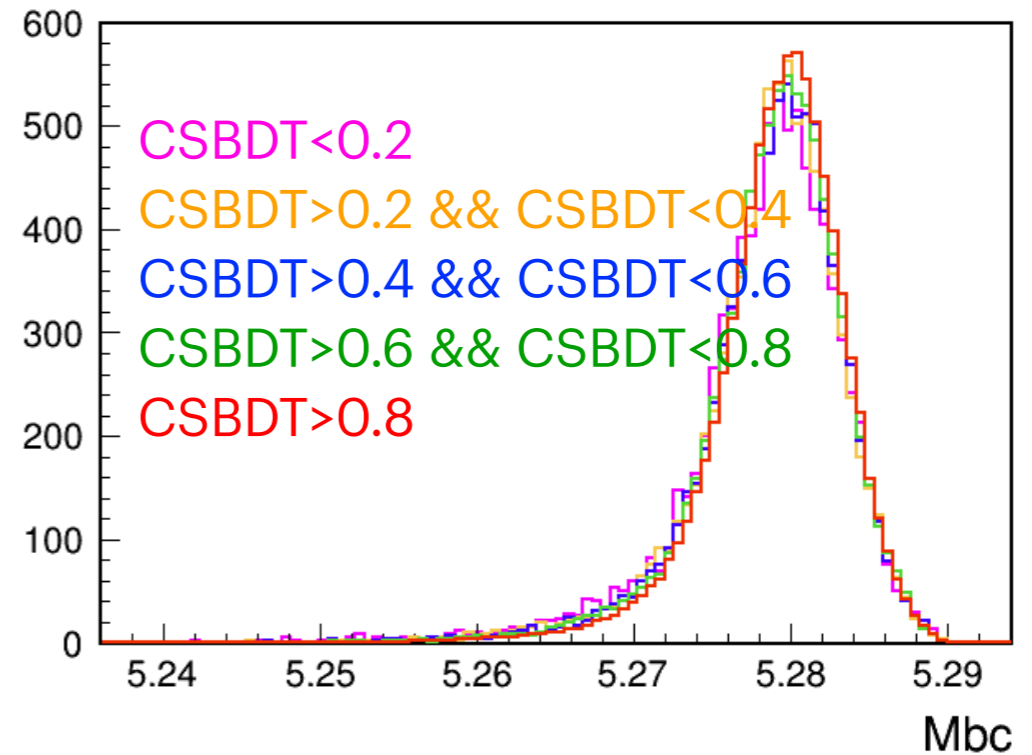
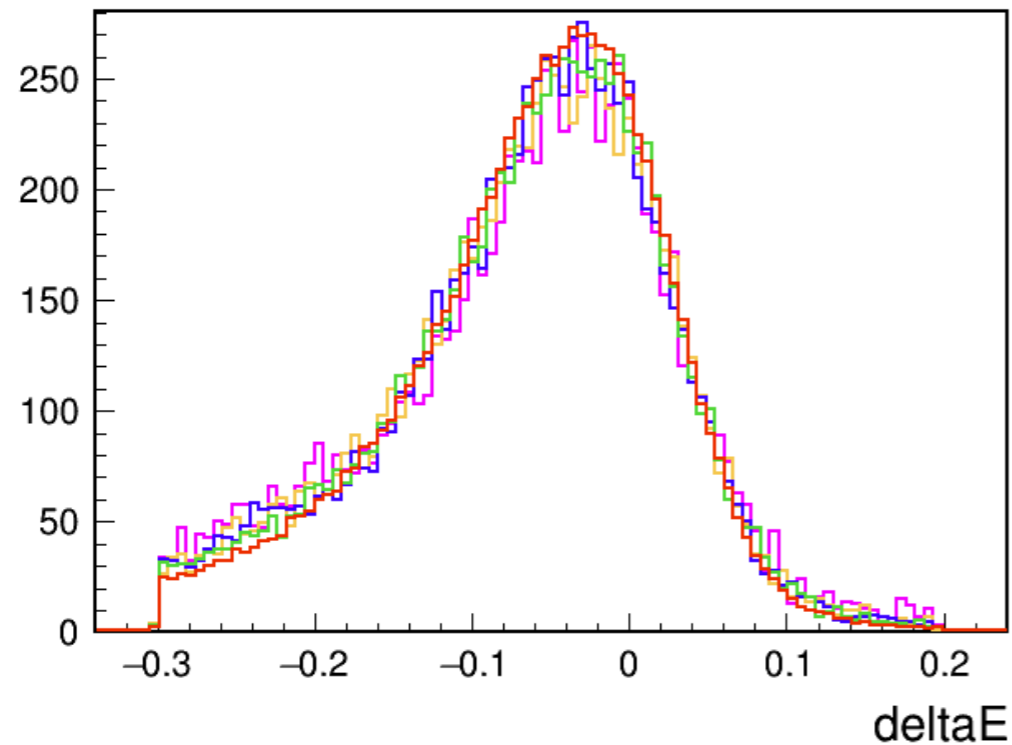
Draw fit variables in slices of CSBDT (background+signal).



No sculpting \rightarrow good.

CSBDT dependences with fit variables

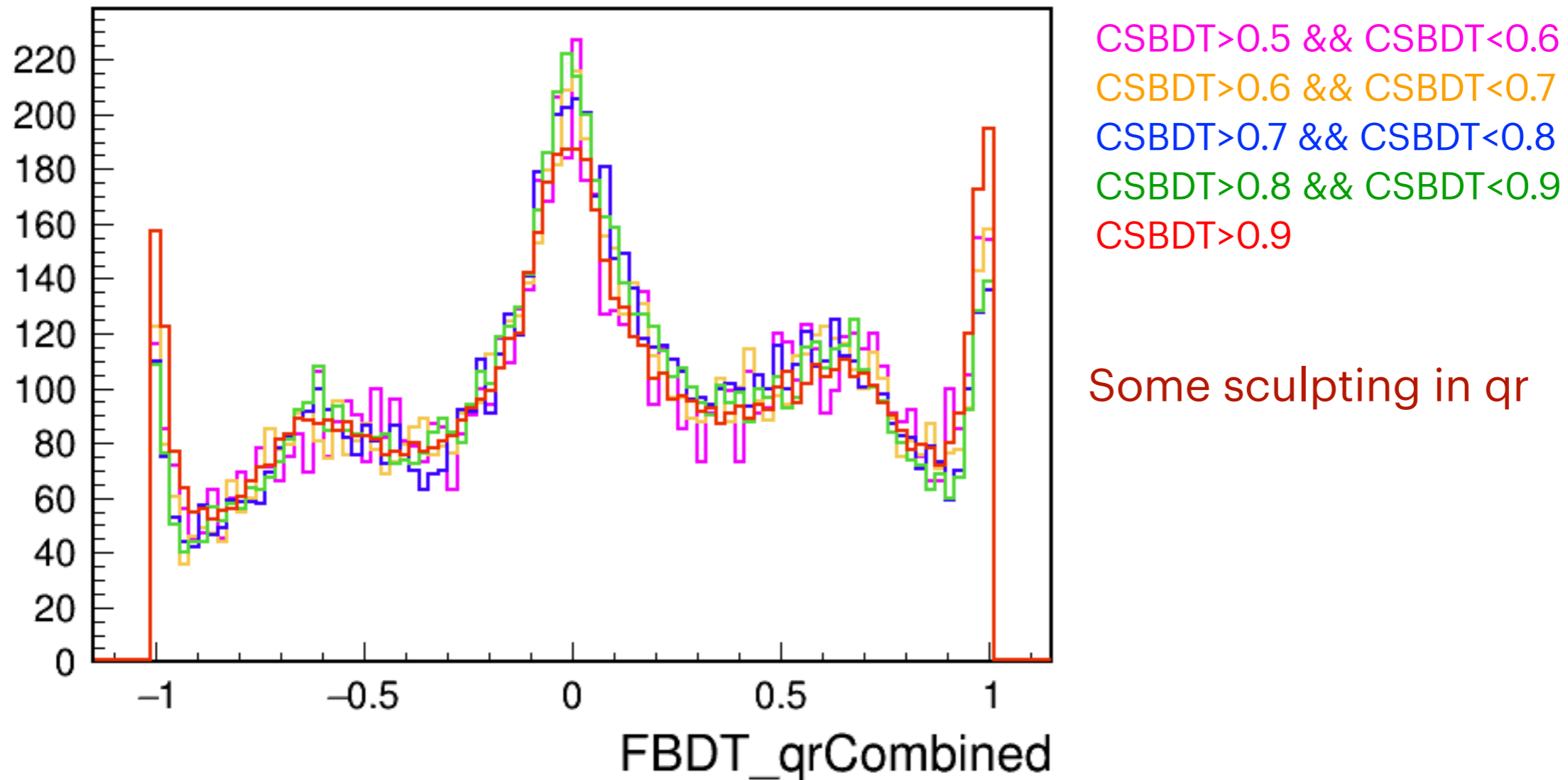
Draw fit variables in slices of CSBDT (signalMC only).



Some sculpting in qr

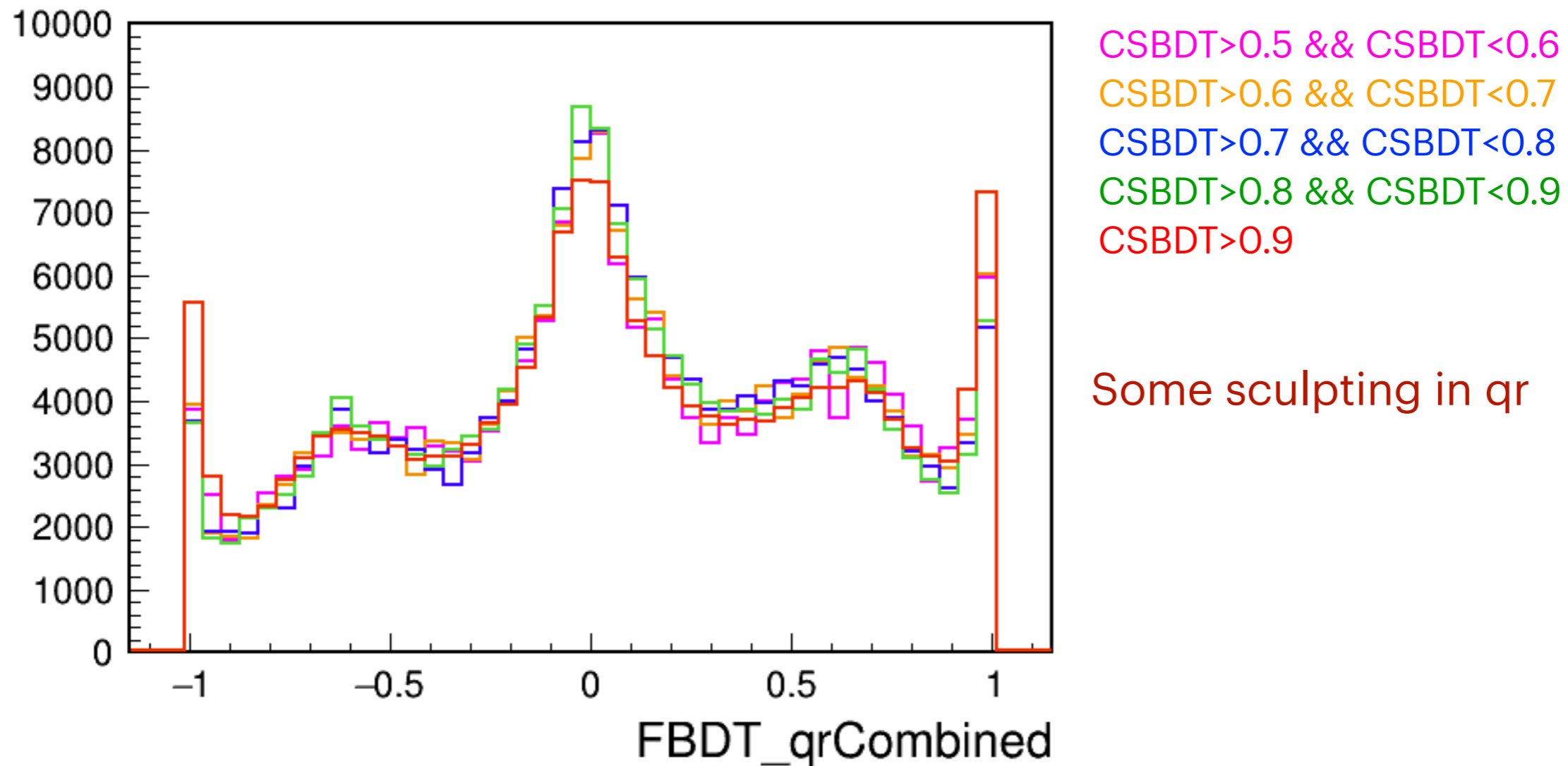
CSBDT dependences with fit variables

Check qr after CS selection (similar to what I'll apply in the end).



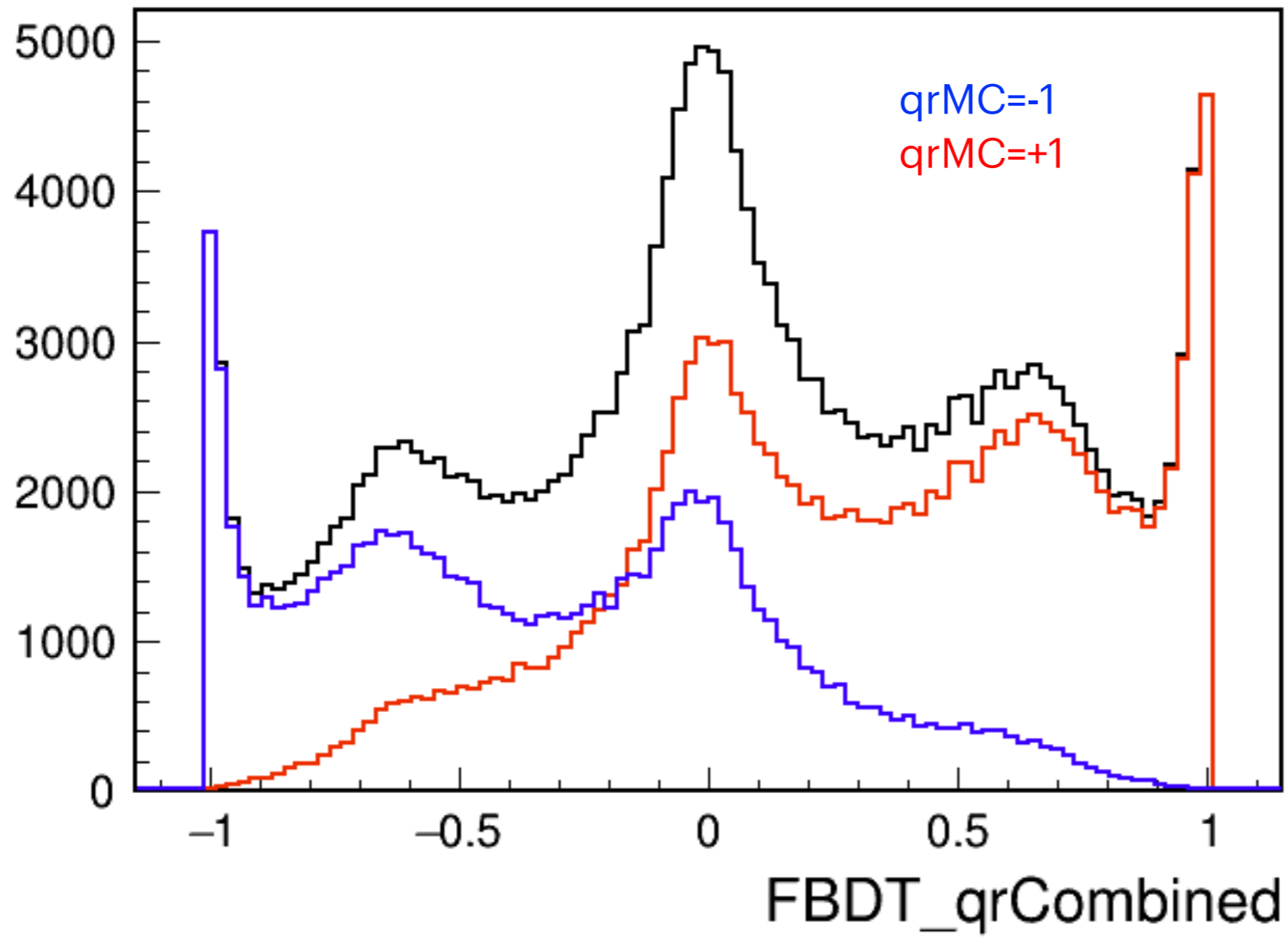
CSBDT dependences with fit variables

Check qr after CS selection (similar to what I'll apply in the end).



qr

Check qr after CS selection (>0.7).



qr variables on SignalMC

$B^0 \rightarrow \pi^0 \pi^0$

r - Interval	ε_i	$\Delta\varepsilon_i$	$w_i \pm \delta w_i$	$\Delta w_i \pm \delta \Delta w_i$	$\varepsilon_{eff,i} \pm \delta \varepsilon_{eff,i}$	$\Delta \varepsilon_{eff,i} \pm \delta \Delta \varepsilon_{eff,i}$	
0.000 – 0.100	16.9	0.17	47.51 ± 0.24	2.47 ± 0.47	0.0420 ± 0.0080	-0.0829 ± 0.0193	
0.100 – 0.250	16.6	0.12	41.01 ± 0.24	0.85 ± 0.47	0.5356 ± 0.0283	-0.0981 ± 0.0571	
0.250 – 0.500	21.1	0.75	29.90 ± 0.20	0.64 ± 0.39	3.4112 ± 0.0675	-0.0948 ± 0.1350	
0.500 – 0.625	11.7	-0.26	20.87 ± 0.23	1.87 ± 0.46	3.9781 ± 0.0667	-0.5973 ± 0.1346	
0.625 – 0.750	11.5	-0.03	15.09 ± 0.21	0.81 ± 0.41	5.5976 ± 0.0728	-0.2756 ± 0.1460	
0.750 – 0.875	8.7	0.06	8.27 ± 0.18	0.56 ± 0.37	6.0356 ± 0.0653	-0.1225 ± 0.1307	
0.875 – 1.000	13.5	-0.81	1.74 ± 0.07	0.21 ± 0.14	12.5839 ± 0.0721	-0.8637 ± 0.1445	
Total	$\varepsilon_{eff} = \sum_i \varepsilon_i \cdot \langle 1 - 2w_i \rangle^2 = 32.18 \pm 0.16$						$\Delta \varepsilon_{eff} = -2.13 \pm 0.32$

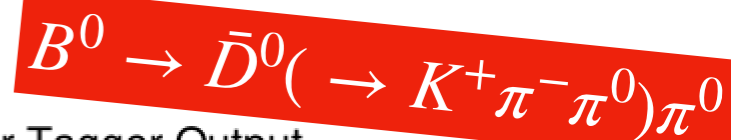
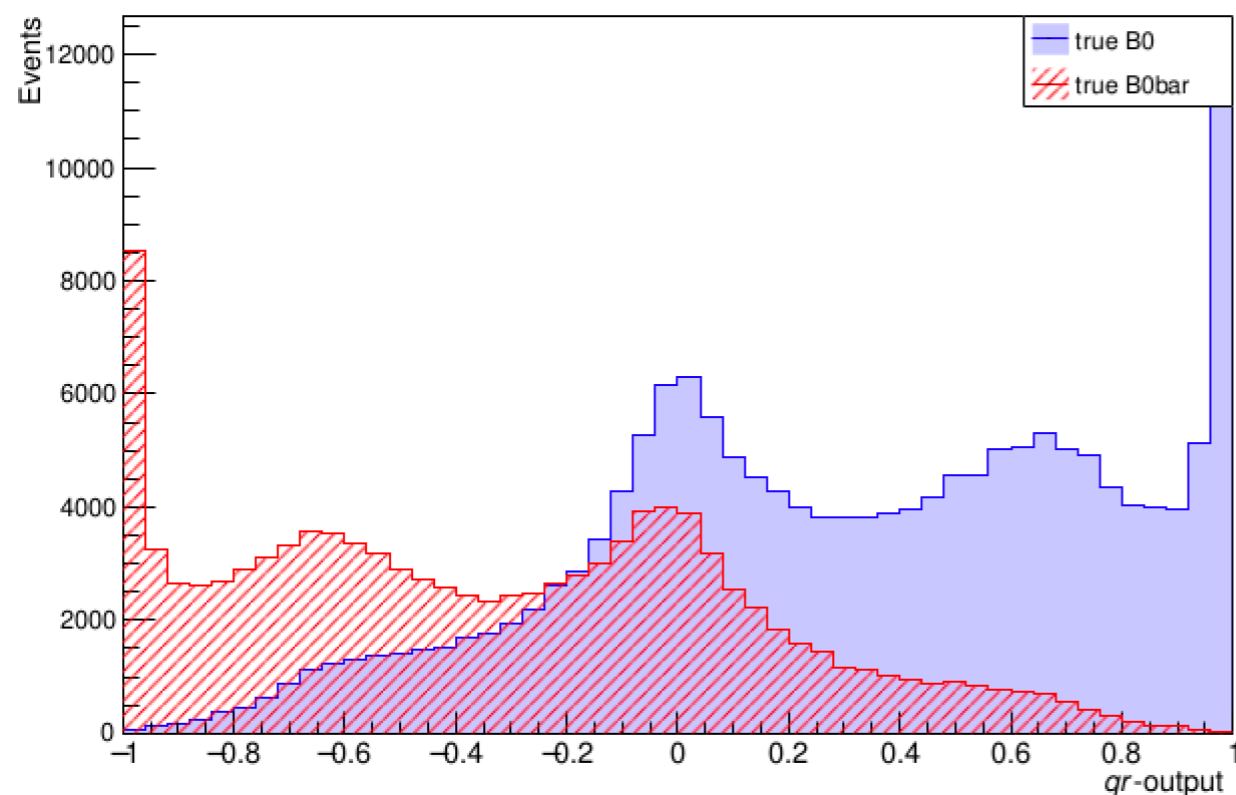
$B^0 \rightarrow \bar{D}^0 (\rightarrow K^+ \pi^- \pi^0) \pi^0$

r - Interval	ε_i	$\Delta\varepsilon_i$	$w_i \pm \delta w_i$	$\Delta w_i \pm \delta \Delta w_i$	$\varepsilon_{eff,i} \pm \delta \varepsilon_{eff,i}$	$\Delta \varepsilon_{eff,i} \pm \delta \Delta \varepsilon_{eff,i}$	
0.000 – 0.100	17.7	0.27	47.27 ± 0.26	2.58 ± 0.51	0.0526 ± 0.0099	-0.0986 ± 0.0217	
0.100 – 0.250	16.8	0.26	41.09 ± 0.26	1.07 ± 0.52	0.5322 ± 0.0311	-0.1189 ± 0.0623	
0.250 – 0.500	21.1	0.60	30.37 ± 0.22	0.24 ± 0.43	3.2461 ± 0.0728	0.0138 ± 0.1457	
0.500 – 0.625	11.5	0.06	21.40 ± 0.26	1.66 ± 0.52	3.7607 ± 0.0723	-0.4175 ± 0.1445	
0.625 – 0.750	11.2	-0.13	15.23 ± 0.23	0.97 ± 0.46	5.4192 ± 0.0794	-0.3649 ± 0.1588	
0.750 – 0.875	8.6	-0.22	8.46 ± 0.20	0.46 ± 0.41	5.9675 ± 0.0721	-0.2828 ± 0.1442	
0.875 – 1.000	13.1	-0.85	1.62 ± 0.08	0.47 ± 0.15	12.2983 ± 0.0783	-1.0309 ± 0.1565	
Total	$\varepsilon_{eff} = \sum_i \varepsilon_i \cdot \langle 1 - 2w_i \rangle^2 = 31.28 \pm 0.17$						$\Delta \varepsilon_{eff} = -2.30 \pm 0.34$

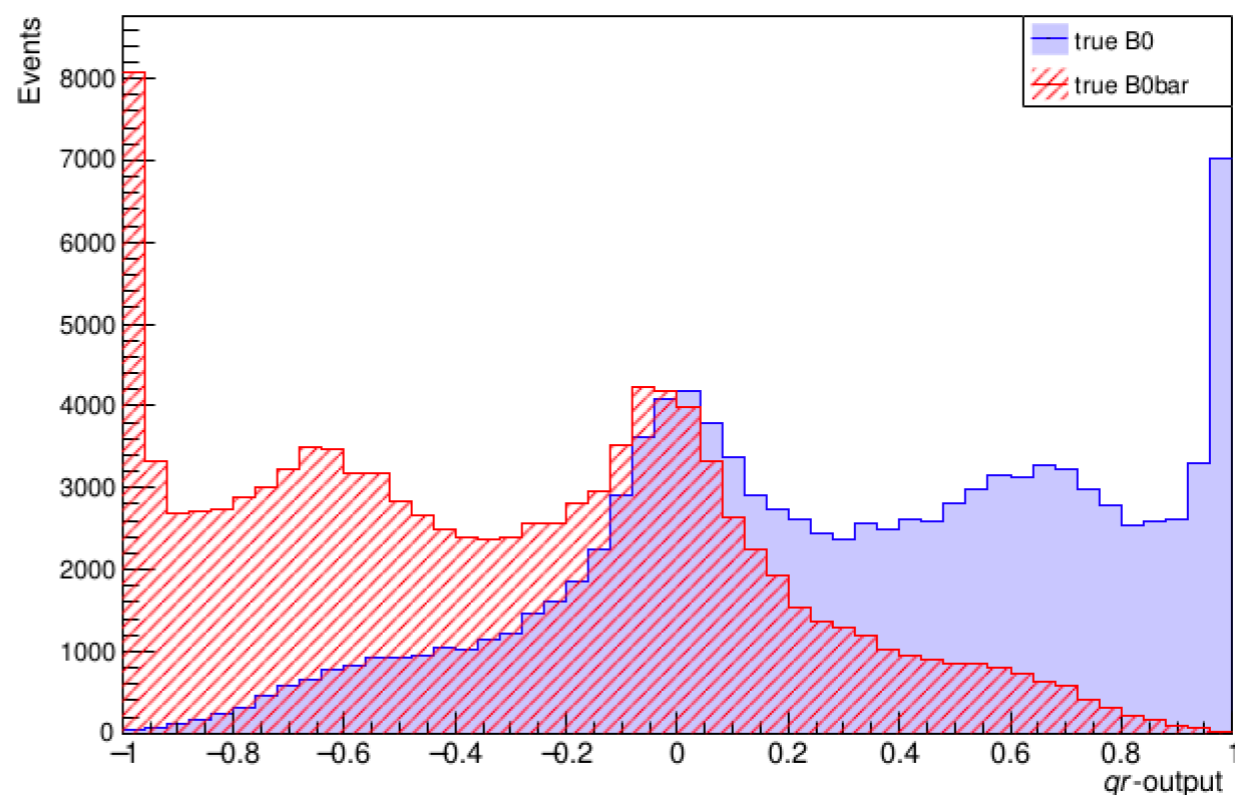
qr variables on SignalMC



Final Flavor Tagger Output



Final Flavor Tagger Output



```

TOTAL NUMBER OF TAGGED EVENTS = 273256
TOTAL AVERAGE EFFICIENCY (q=+-1)= 100.00 +- 0.00 %
TOTAL AVERAGE EFFECTIVE EFFICIENCY (q=+-1)= 32.184039 +- 0.156912 %
TOTAL AVERAGE EFFECTIVE EFFICIENCY ASYMMETRY (q=+-1)= -2.134811 +- 0.315022 %
B0-TAGGER TOTAL EFFECTIVE EFFICIENCIES: 31.14 +- 0.20 % (q=+1) 33.27 +- 0.25 % (q=-1)
FLAVOR PERCENTAGE (MC): 60.52 % (q=+1) 39.48 % (q=-1) Diff=21.03 %
    
```

```

TOTAL NUMBER OF TAGGED EVENTS = 214462
TOTAL AVERAGE EFFICIENCY (q=+-1)= 100.00 +- 0.00 %
TOTAL AVERAGE EFFECTIVE EFFICIENCY (q=+-1)= 31.276453 +- 0.171015 %
TOTAL AVERAGE EFFECTIVE EFFICIENCY ASYMMETRY (q=+-1)= -2.299739 +- 0.341994 %
B0-TAGGER TOTAL EFFECTIVE EFFICIENCIES: 30.15 +- 0.24 % (q=+1) 32.45 +- 0.24 % (q=-1)
FLAVOR PERCENTAGE (MC): 49.59 % (q=+1) 50.41 % (q=-1) Diff=-0.82 %
    
```

CSMVA using off-res data for the bkg

Train on **off-res data and signalMC** after applying all π^0 , ΔE , and M_{bc} selections.

Inputs (after pruning)

7 Kakuno-Super-Fox-Wolfram moments

cosTBTO

1 CleoCone

cosTheta*

R2

thrustOm

ΔZ (BTag)

Δr (BTag)

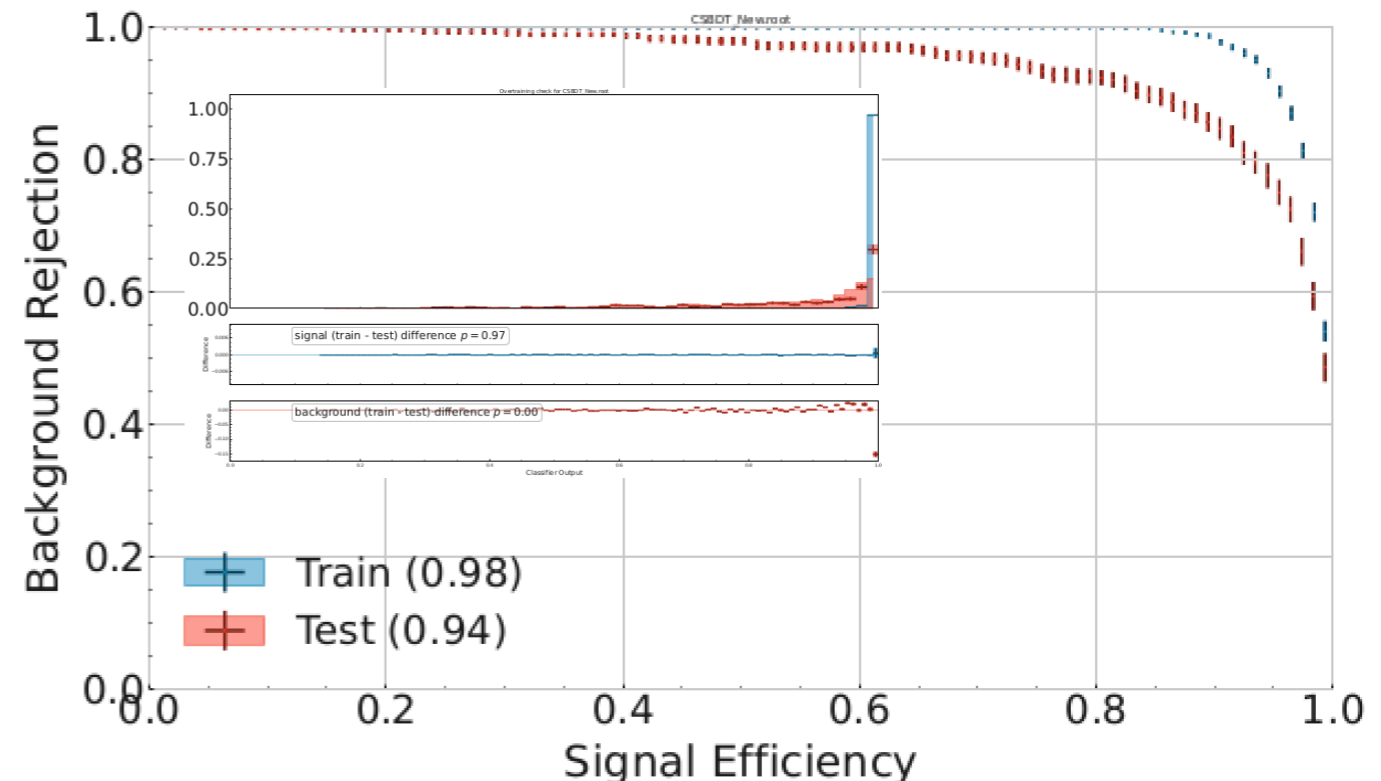
thrustAxisCosTheta

angle between π^0 's

cosHelicityAngle

KSFVVariableset

KSFVVariablesmm2



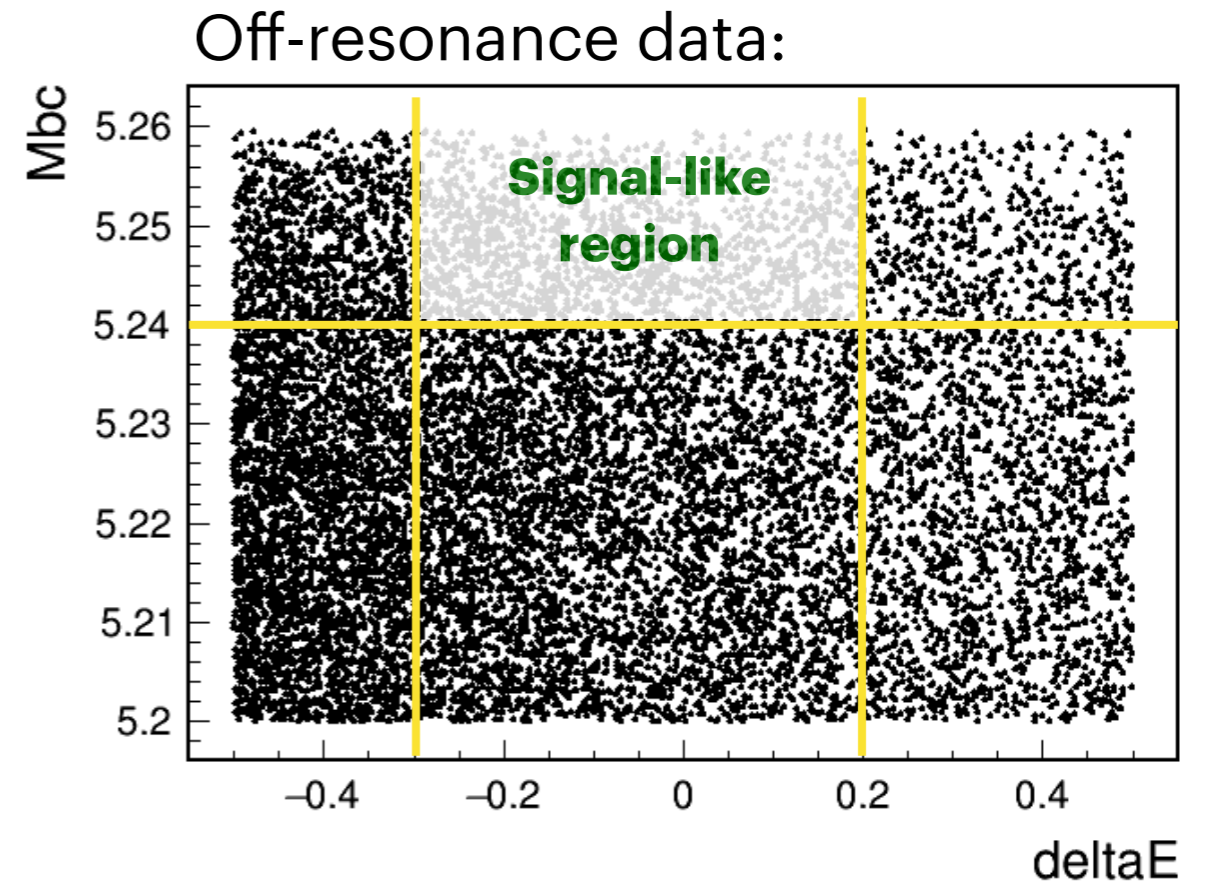
Train bkg sample (from offres): 1000 events
Train sig sample (from MC): 180000 events
Test bkg sample (from offres): 500 events
Test sig sample (from MC): 90000 events

Very small off-res sample (9fb^{-1})
→ poor BDT (total off-res sample
will be 18fb^{-1})

CSMVA using off-res data for the bkg

But what off-resonance data can I use?

In previous result I was using only the **signal-like region**



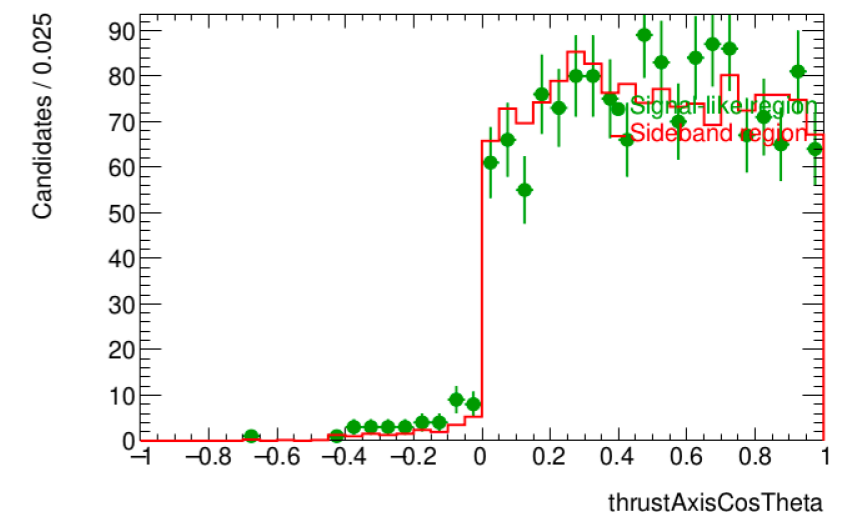
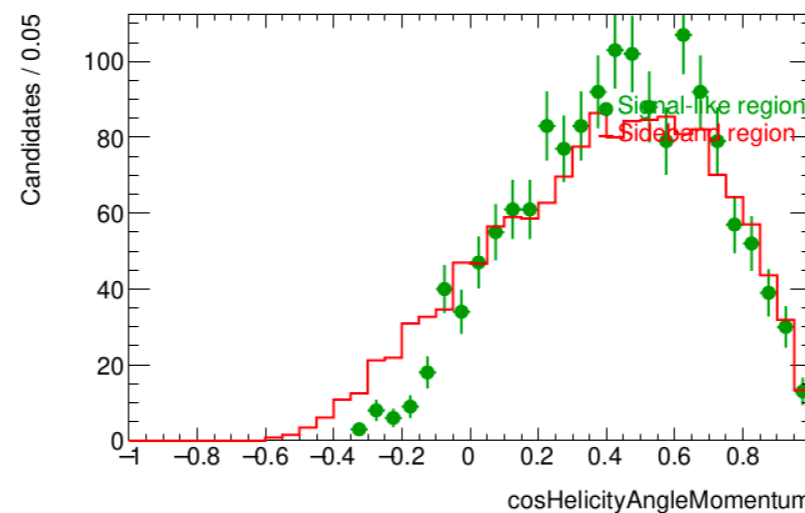
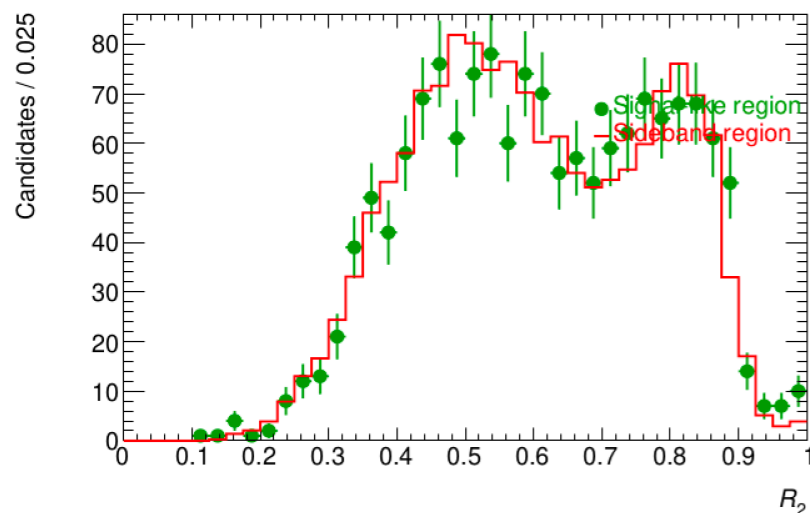
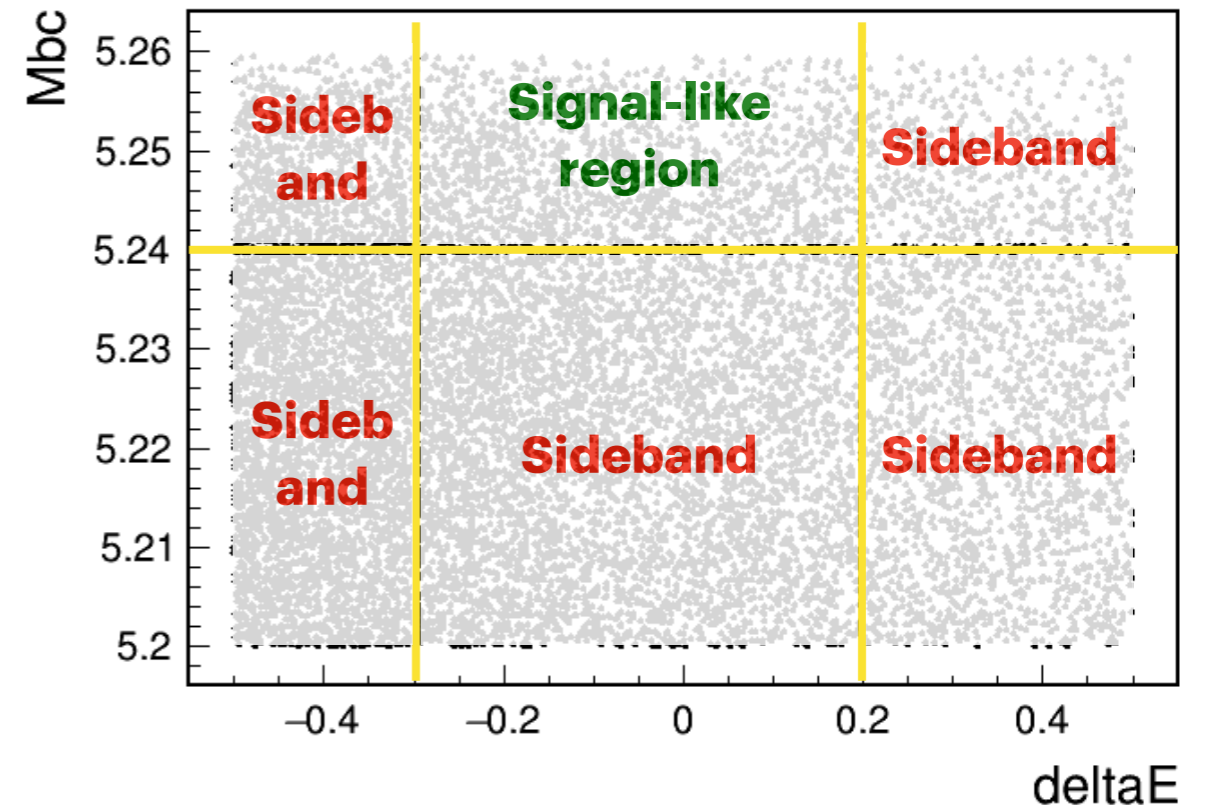
CSMVA using off-res data for the bkg

But what off-resonance data can I use?

In previous result I was using only the **signal-like region**

Compare CS input distributions in **signal-like** and **sideband** regions (in off-resonance data):

Off-resonance data:



All variables that have discrepancies are not CS inputs anymore (after pruning), except **cosHelAngleMomentum**.

CSMVA using off-res data for the bkg

Train on **off-res data and signalMC** after applying all π^0 selections.
Use all off-resonance data (including sidebands). Exclude cosHelicityAngle.

Inputs (after pruning)

7 Kakuno-Super-Fox-Wolfram moments

cosTBTO

1 CleoCone

cosTheta*

R2

thrustOm

ΔZ (BTag)

Δr (BTag)

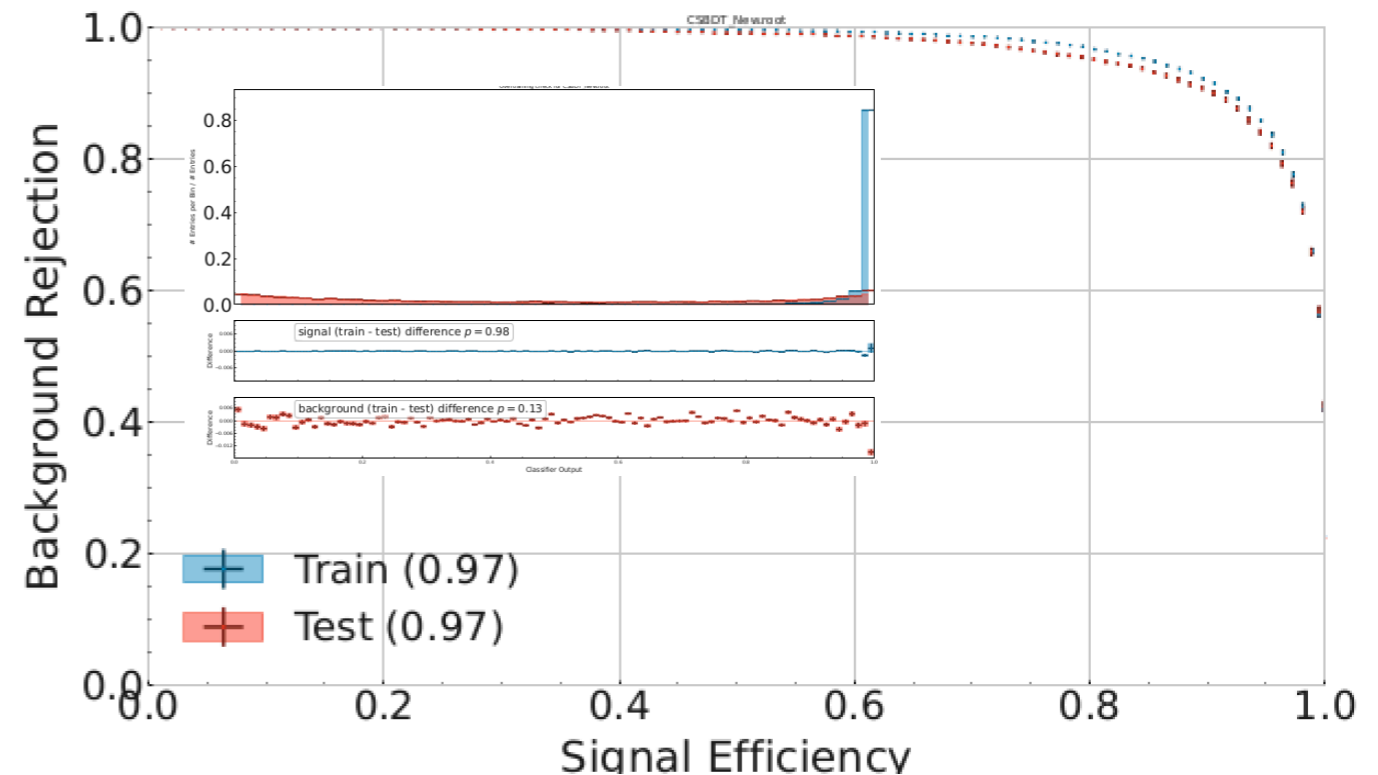
thrustAxisCosTheta

angle between π^0 's

~~cosHelicityAngle~~

KSFVVariableset

KSFVVariablesmm2

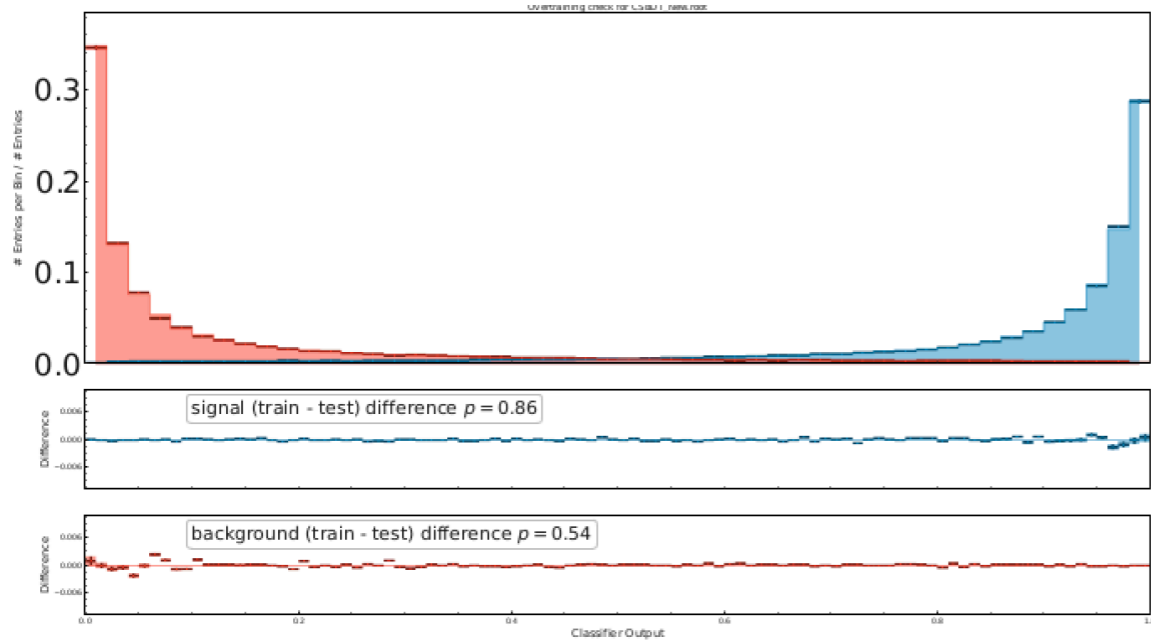
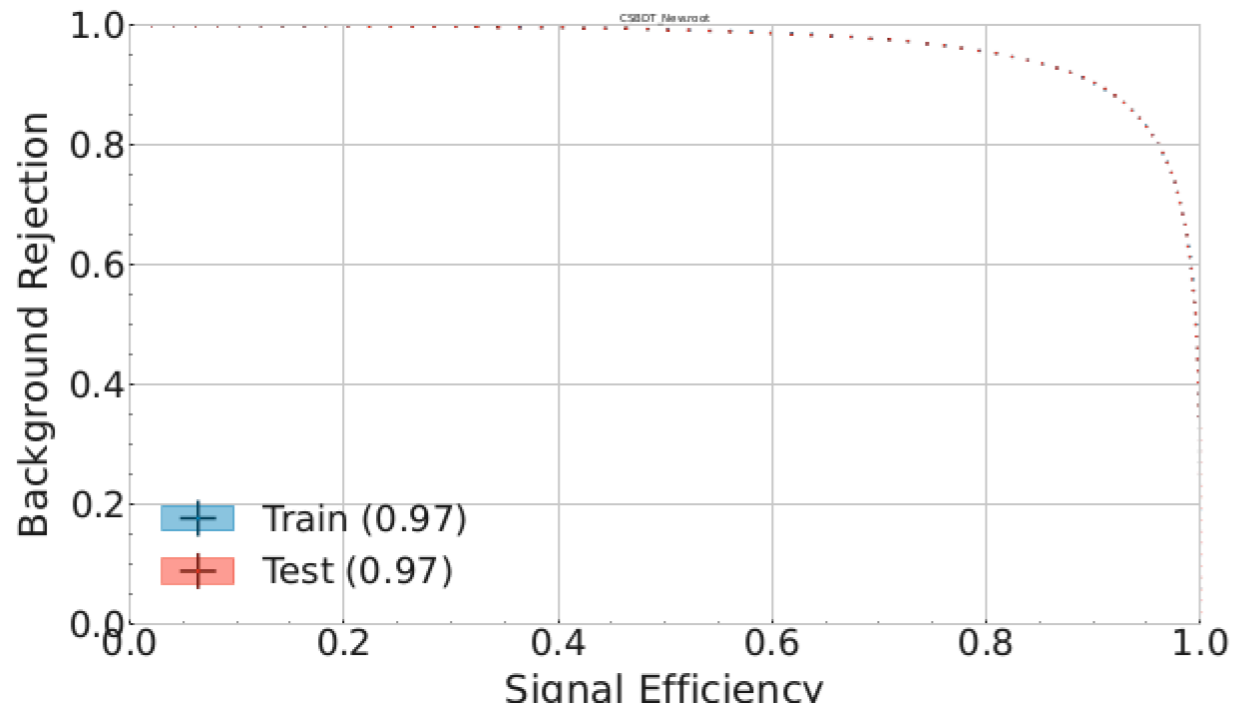


Train bkg sample (from offres): 8000 events
Train sig sample (from MC): 180000 events
Test bkg sample (from offres): 4000 events
Test sig sample (from MC): 90000 events

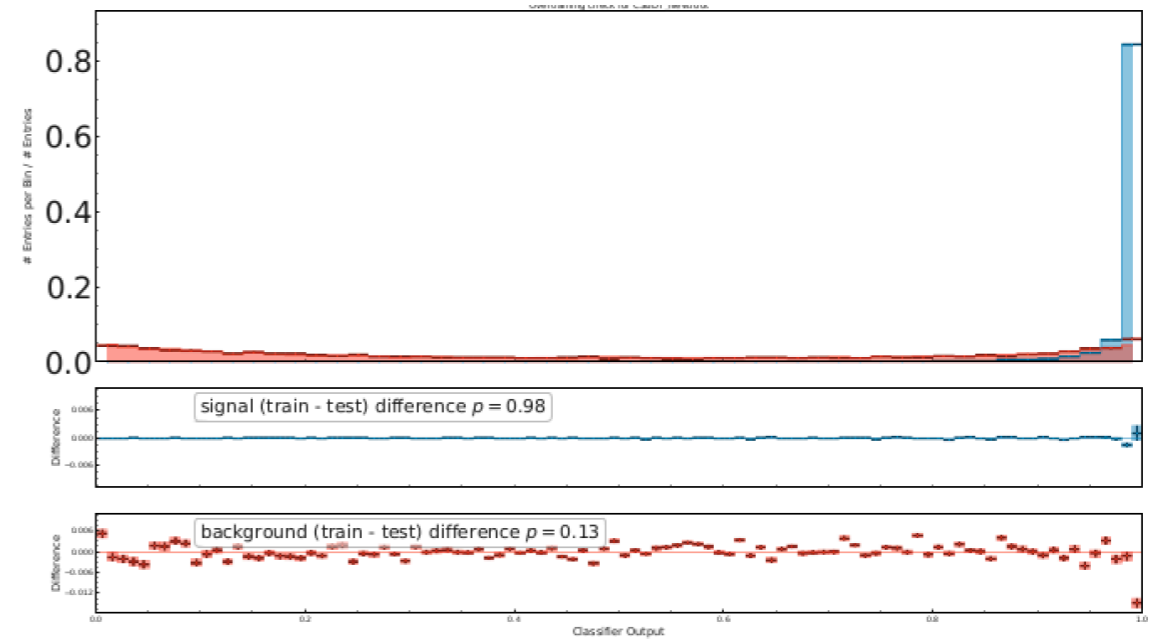
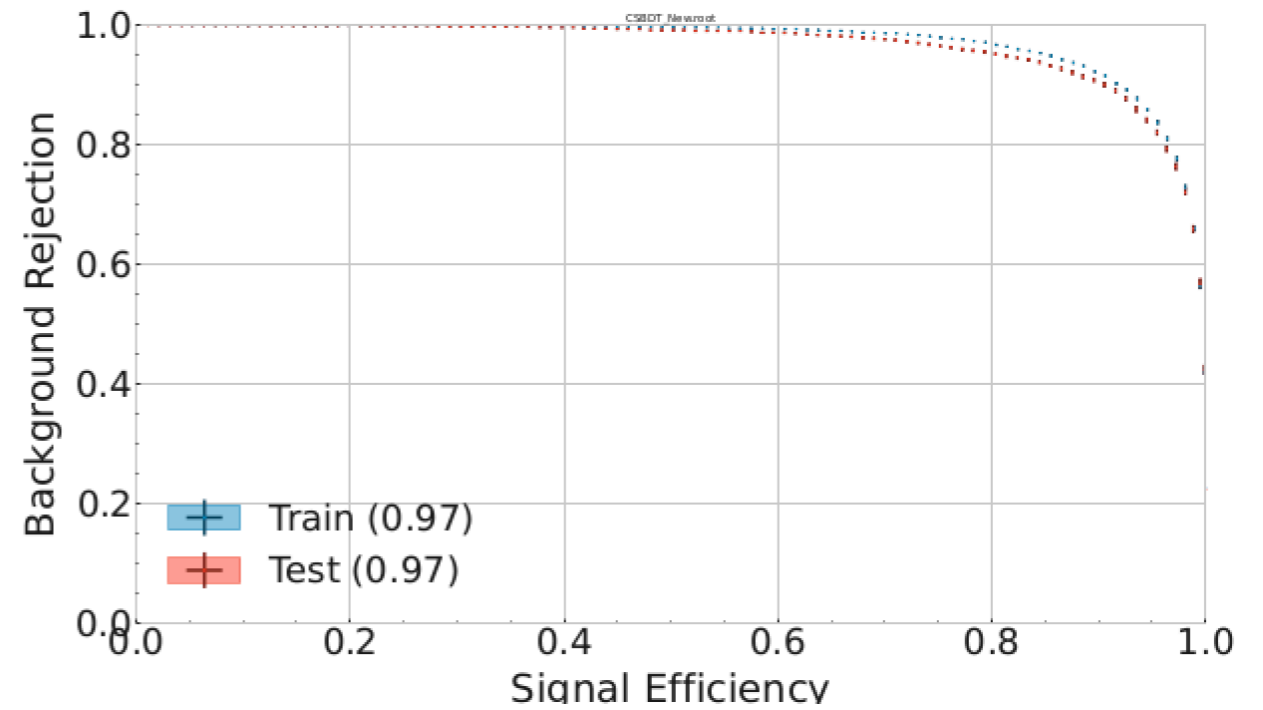
Better result wrt previous one

CSMVAs comparison

All MC



Off-res data + signalMC



ROC curves are the same, but distributions are quite different

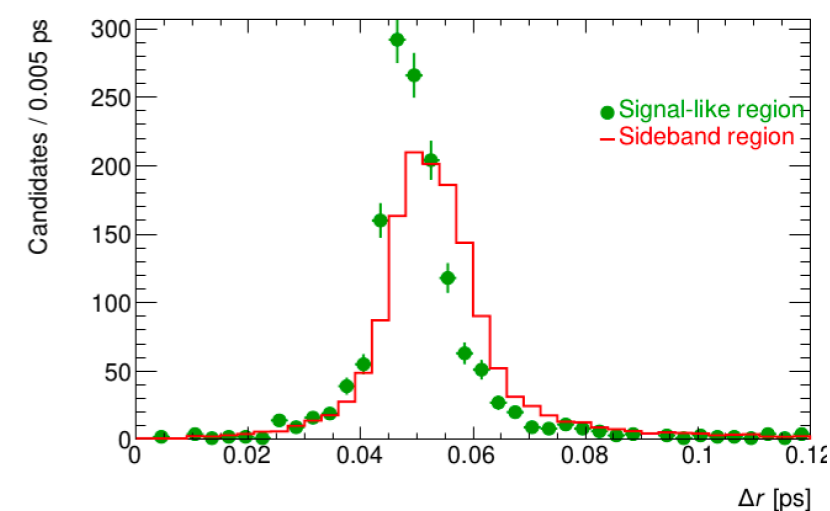
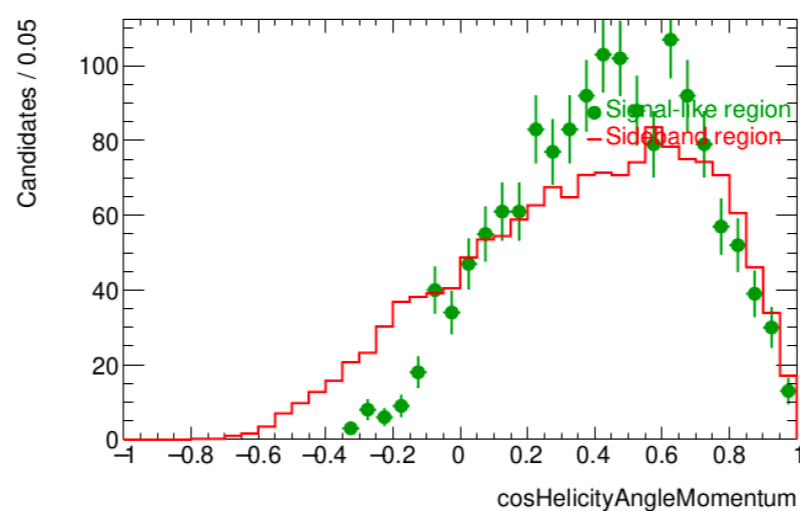
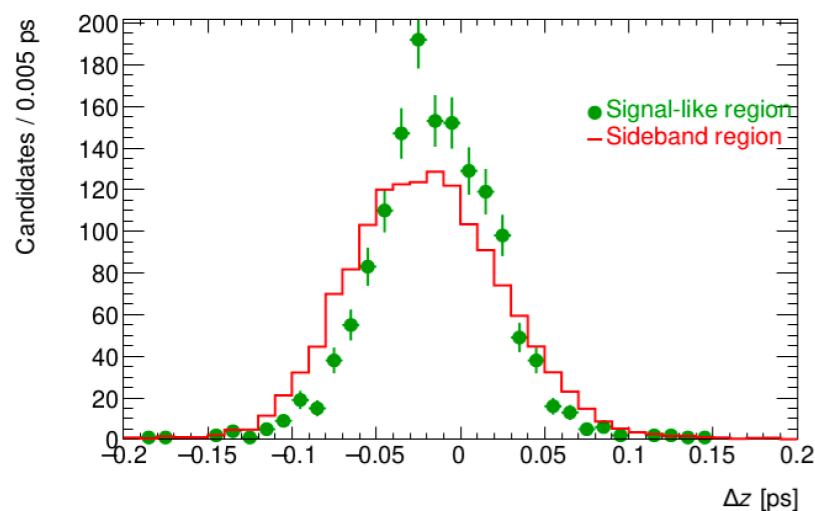
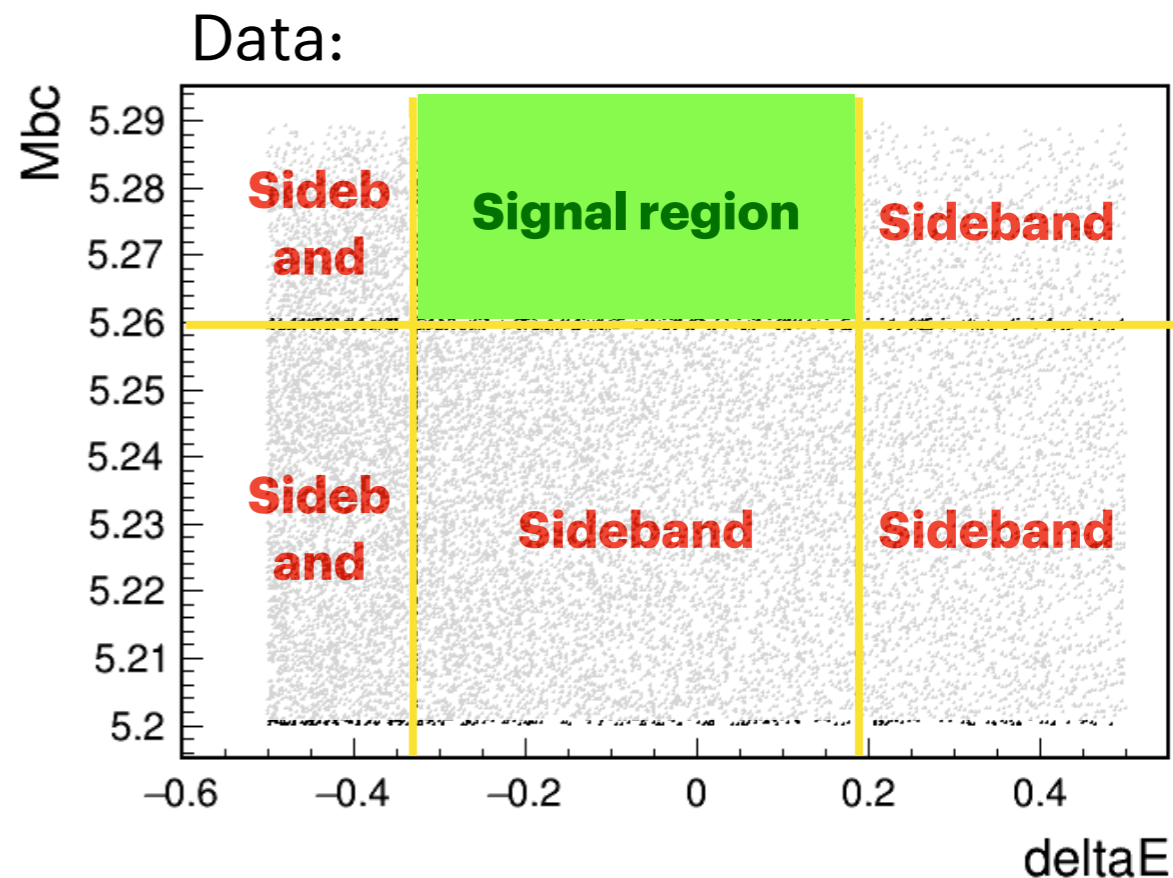
Sideband in off-res data is fine.

But then what about the
on-resonance sideband data?

Sideband data vs off-res data

Sideband distributions seem good in off-res.

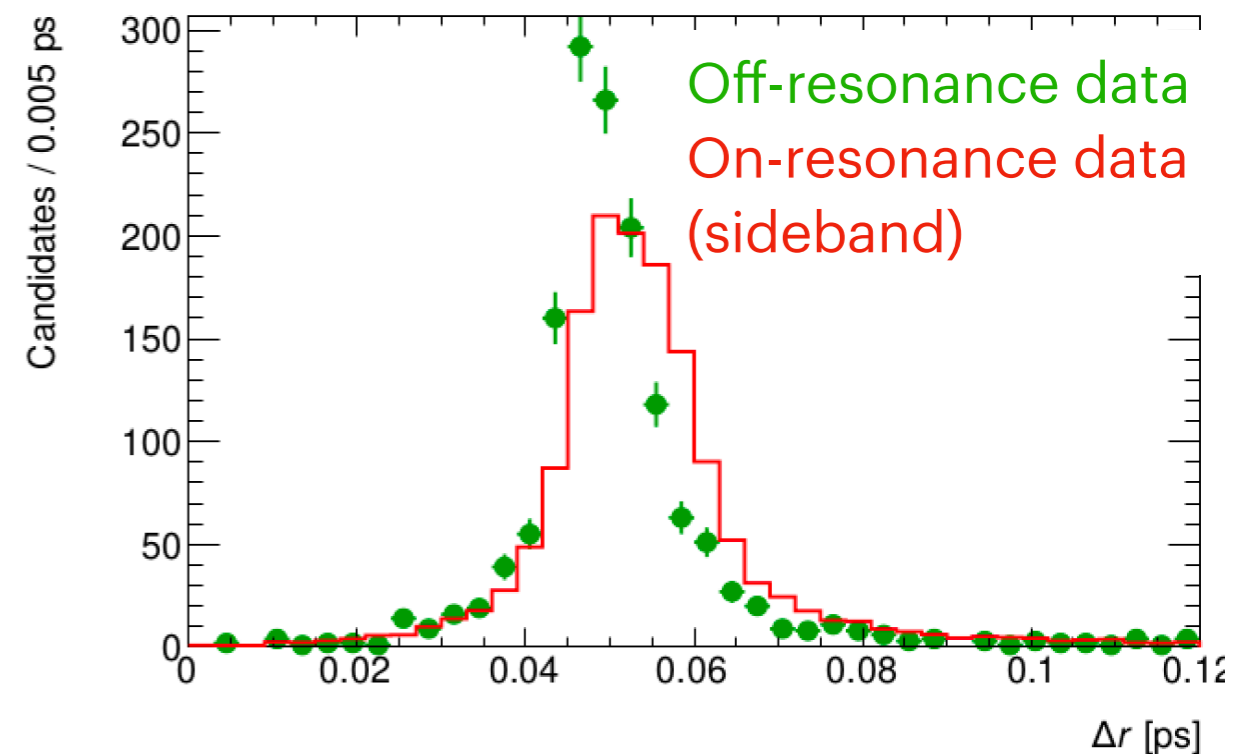
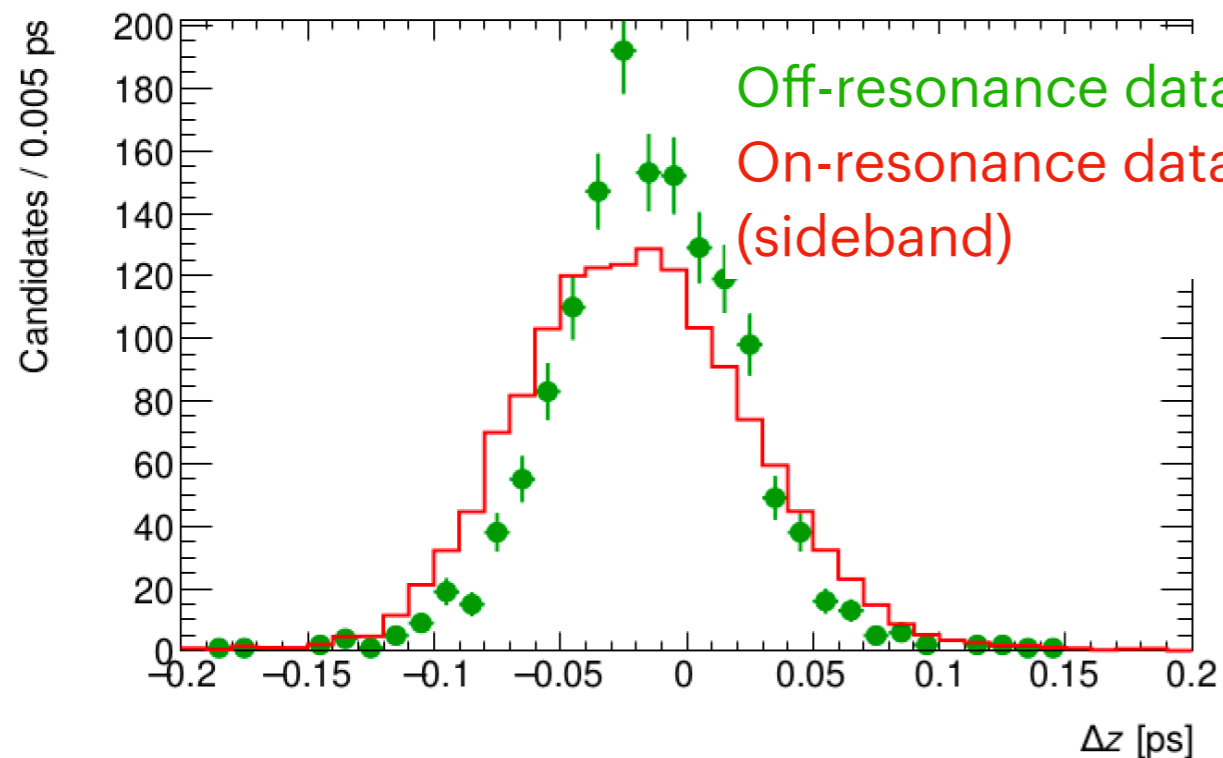
Compare CS inputs distributions in **signal-like region** (in off-resonance data) and **sideband regions** (in on-resonance data):



Observe large discrepancies also in Δr and ΔZ .

Which are the correct Δr and ΔZ distributions?

Sideband data and off-resonance data have different r and ΔZ distributions.



I need to understand which one reproduces correctly the signal region:

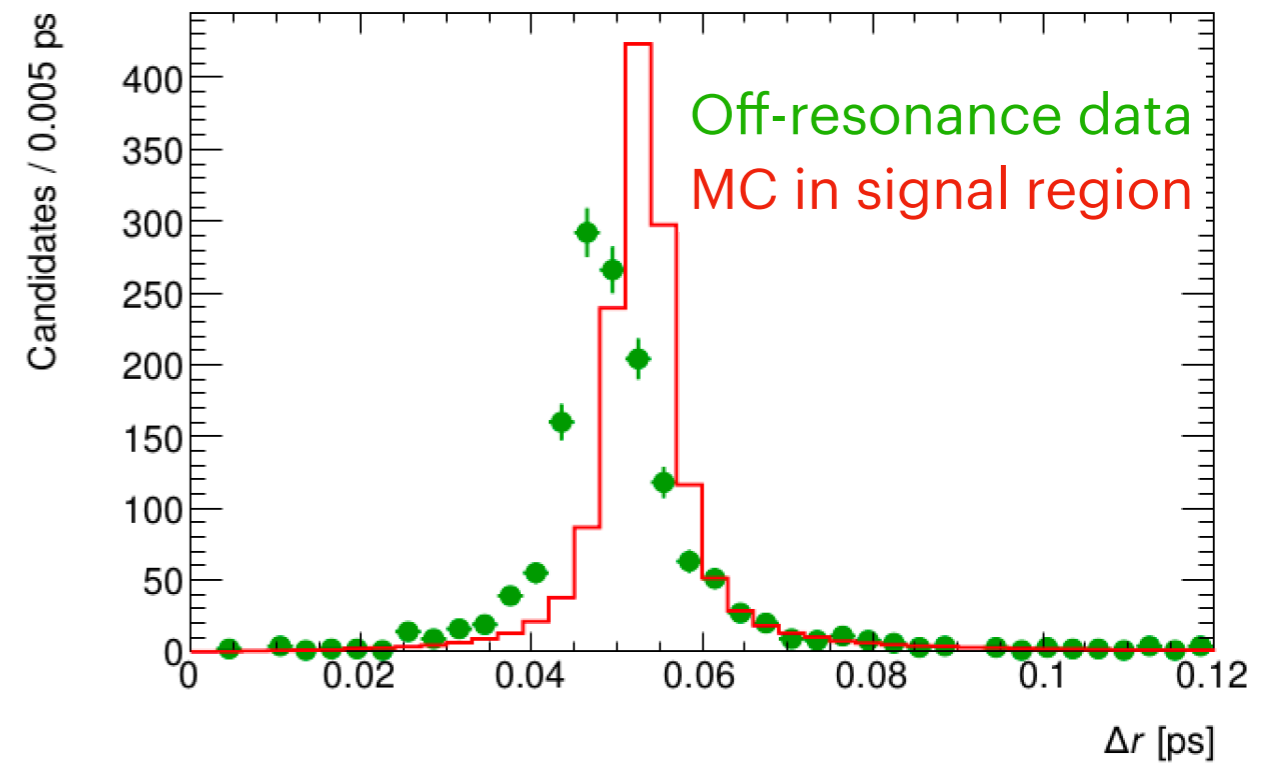
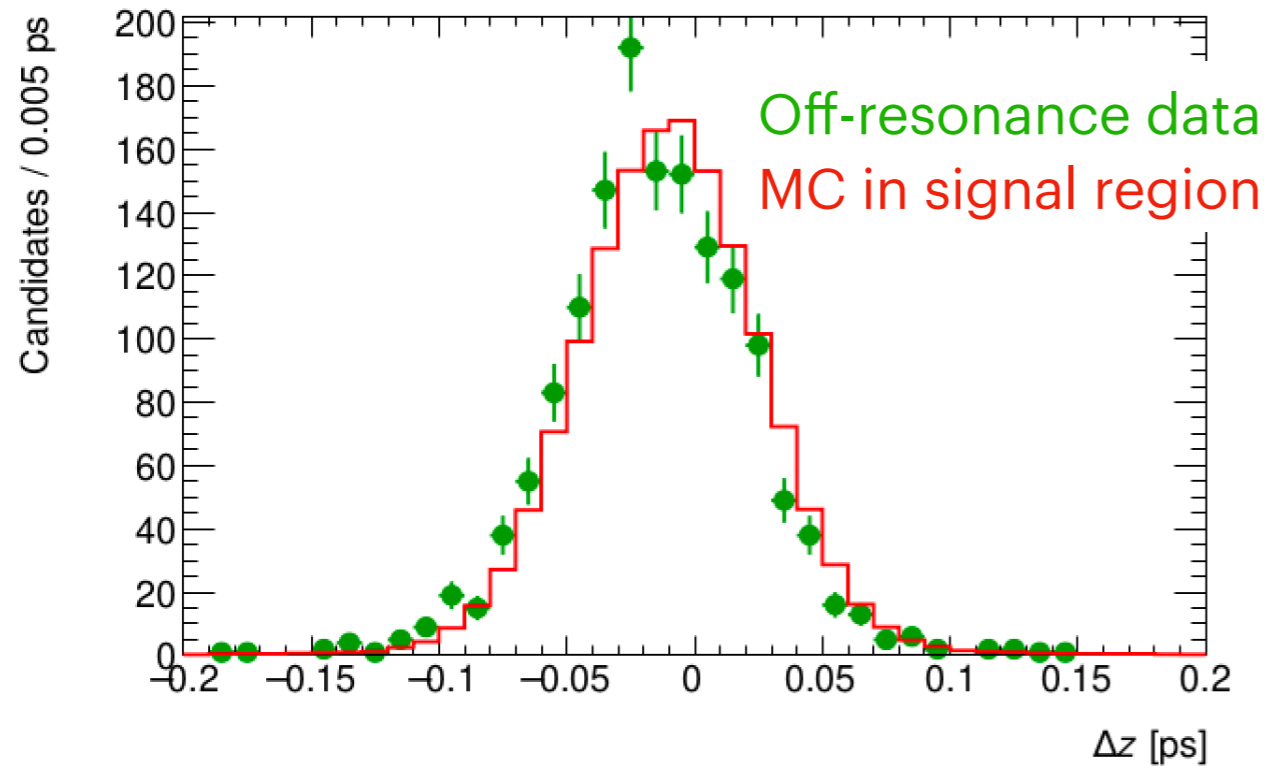
Red is correct → use sideband data (and exclude $\cos\theta_{\text{HelAngle}}$ from the inputs)

Green is correct → use off-res data (and exclude $\cos\theta_{\text{HelAngle}}$ from the inputs)

Stiil thinking how to do this.

Which are the correct Δr and ΔZ distributions?

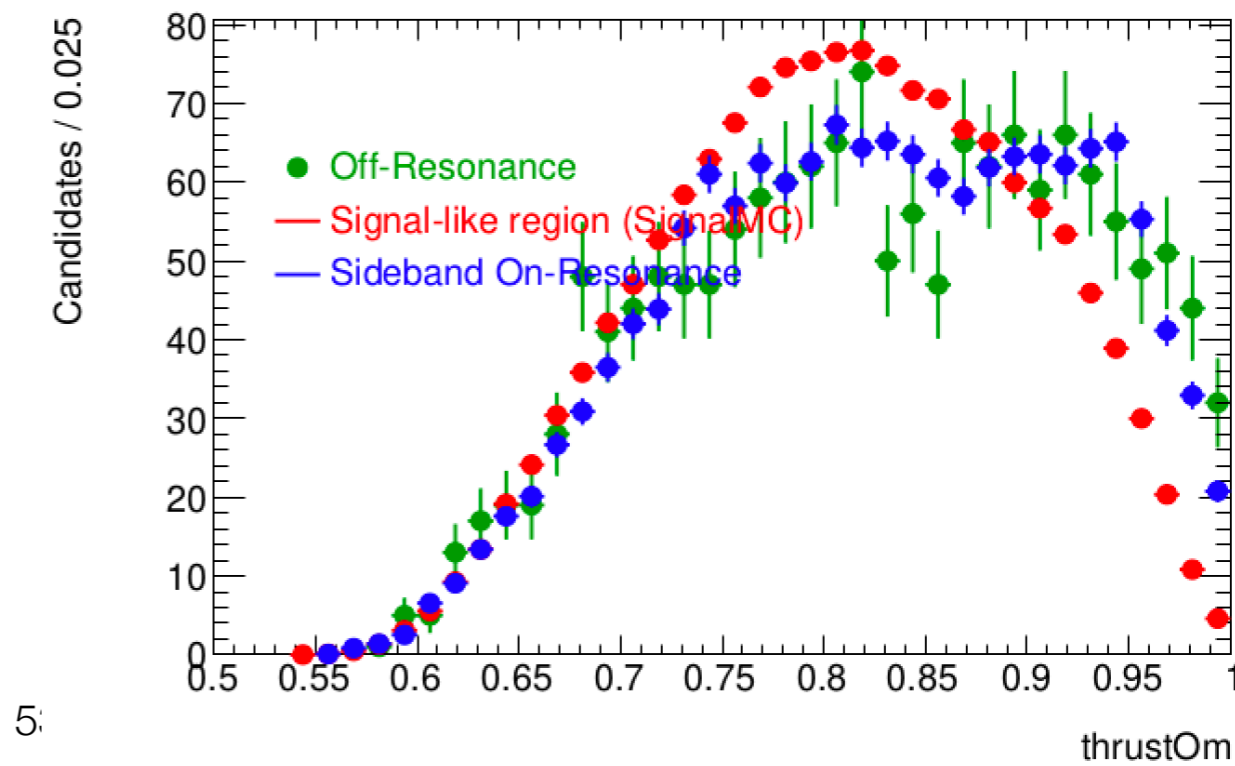
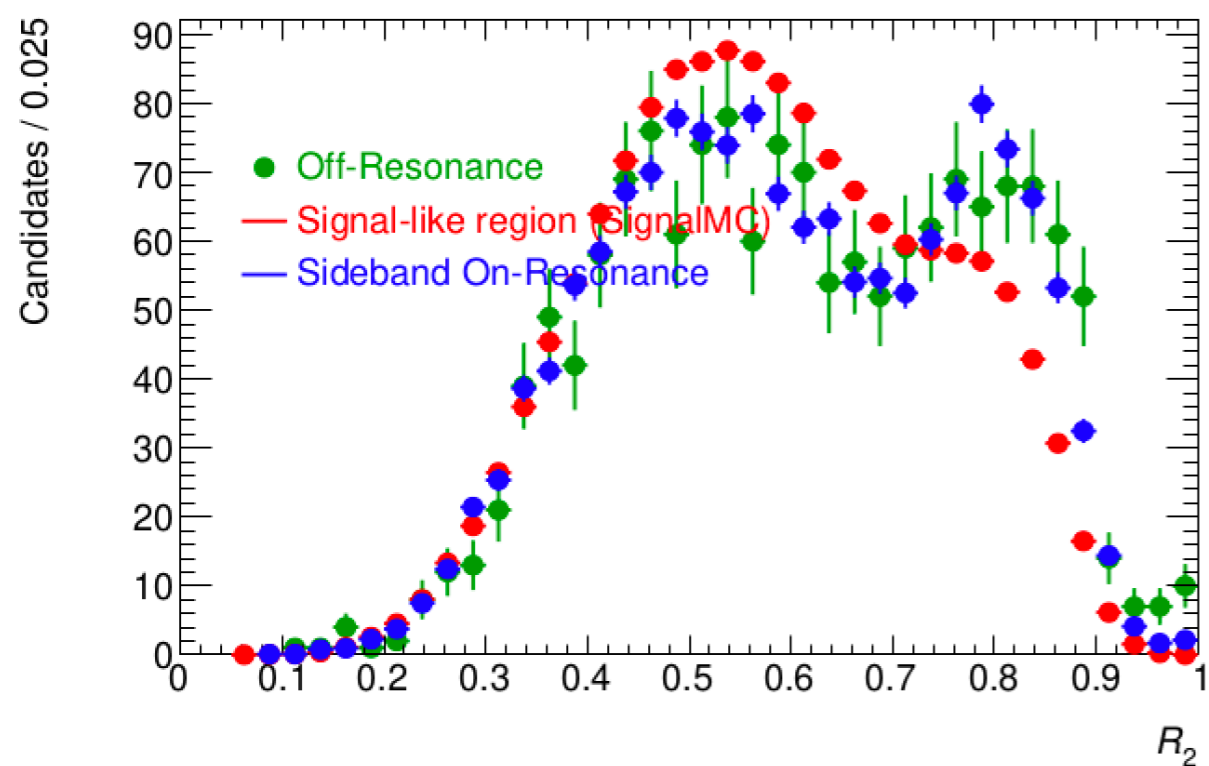
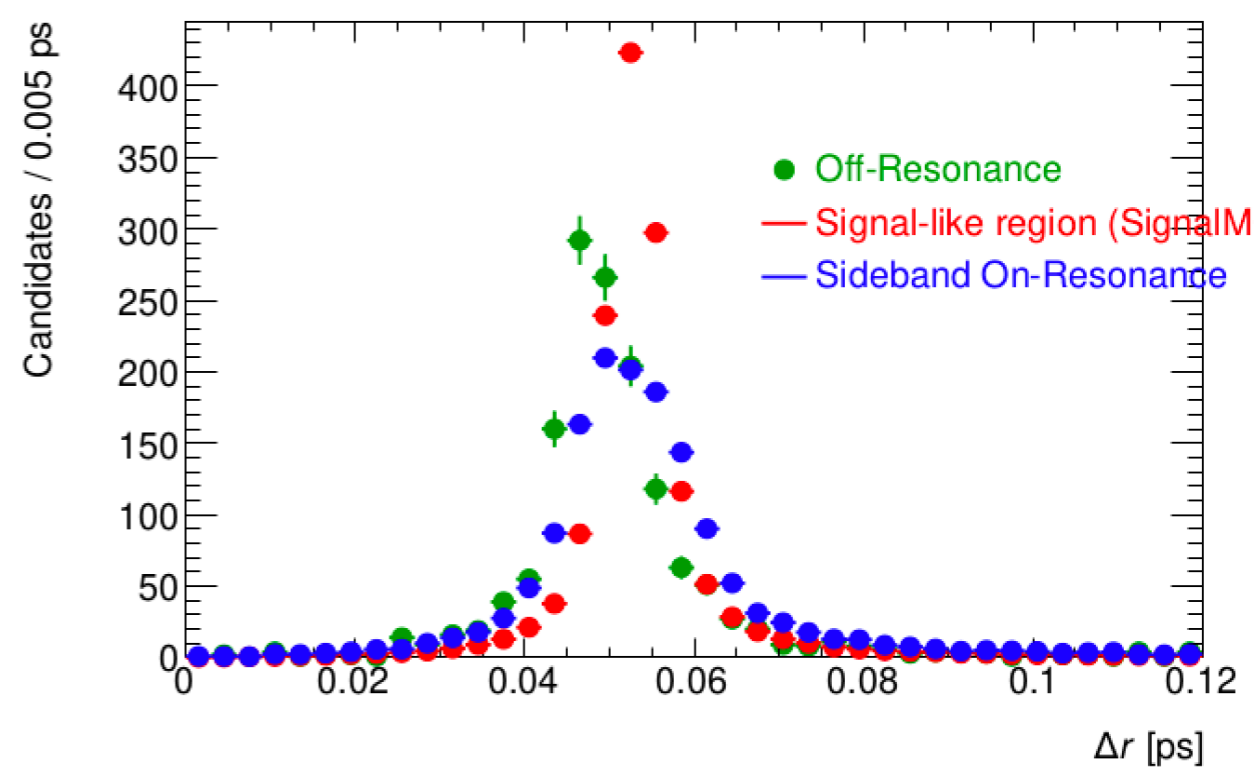
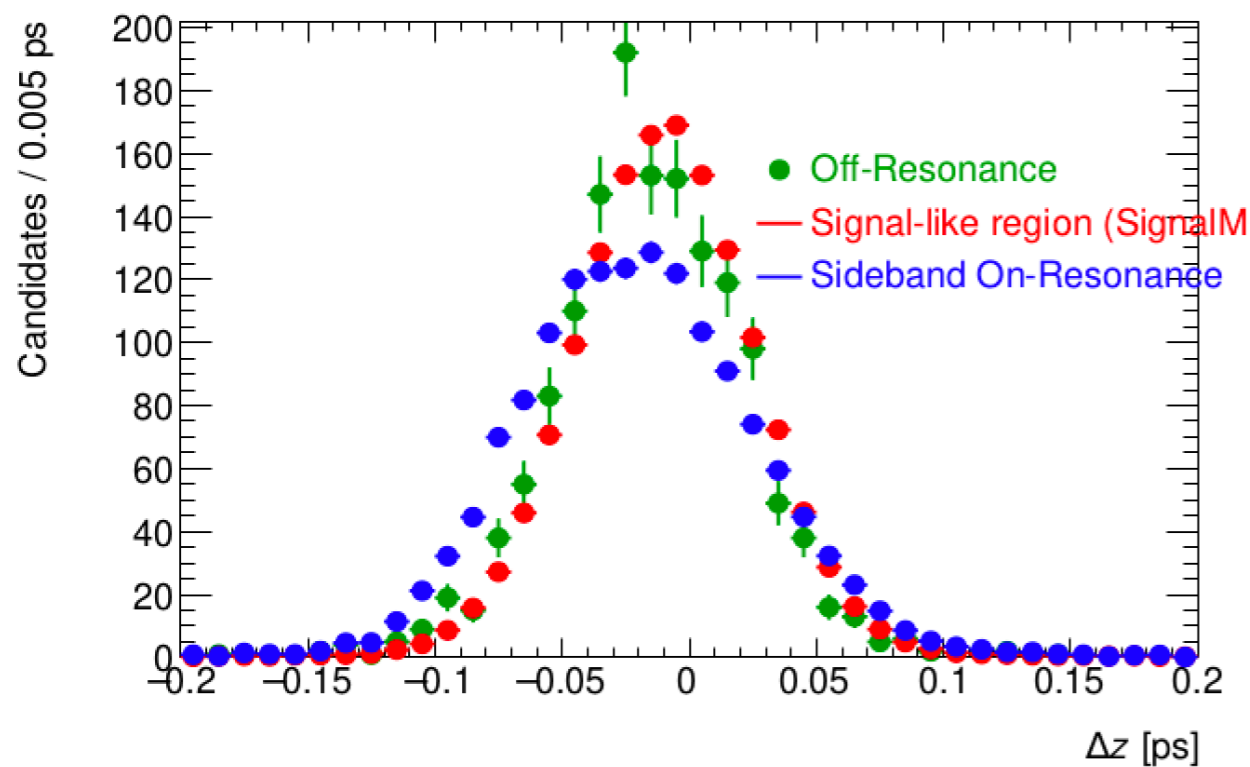
Sideband data and off-resonance data have different r and ΔZ distributions.



Metti tutti e tre insieme dai

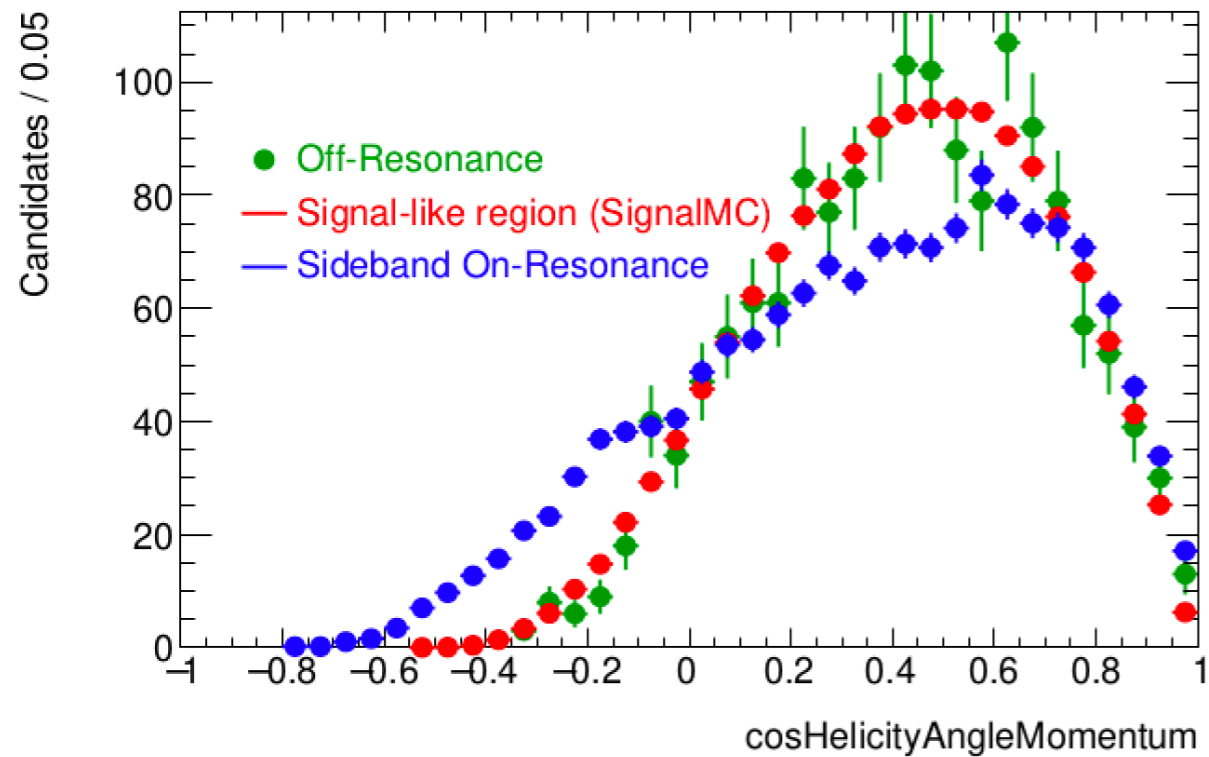
Which are the correct Δr and ΔZ distributions?

Sideband data and off-resonance data have different r and ΔZ distributions.



Which are the correct Δr and ΔZ distributions?

Sideband data and off-resonance data have different r and ΔZ distributions.



Off-Resonance data correction

On-resonance and off-resonance data have different beam energies.

Take this into account for M_{bc} and ΔE using

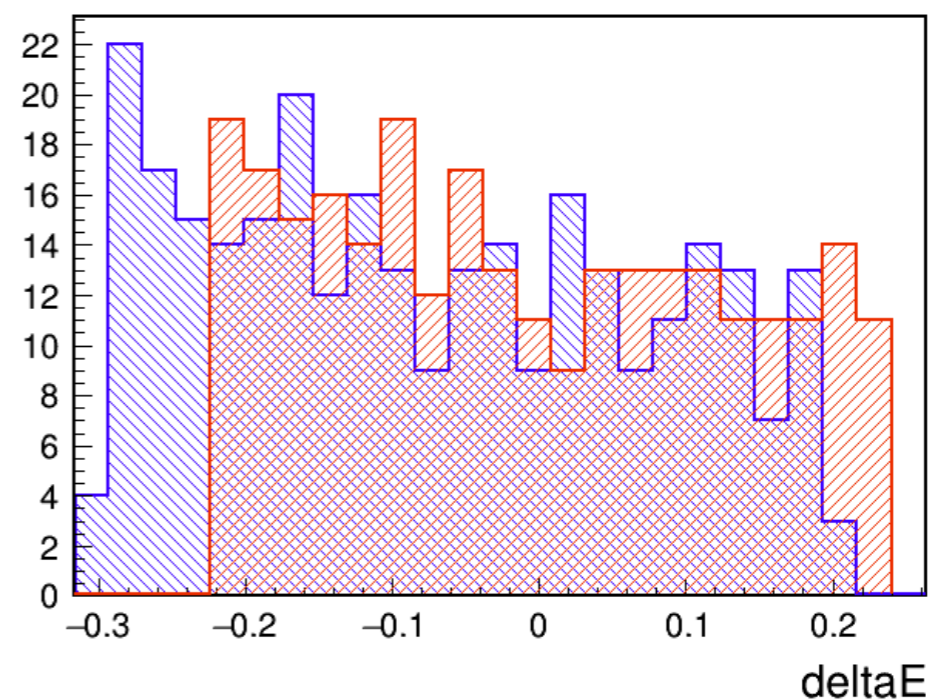
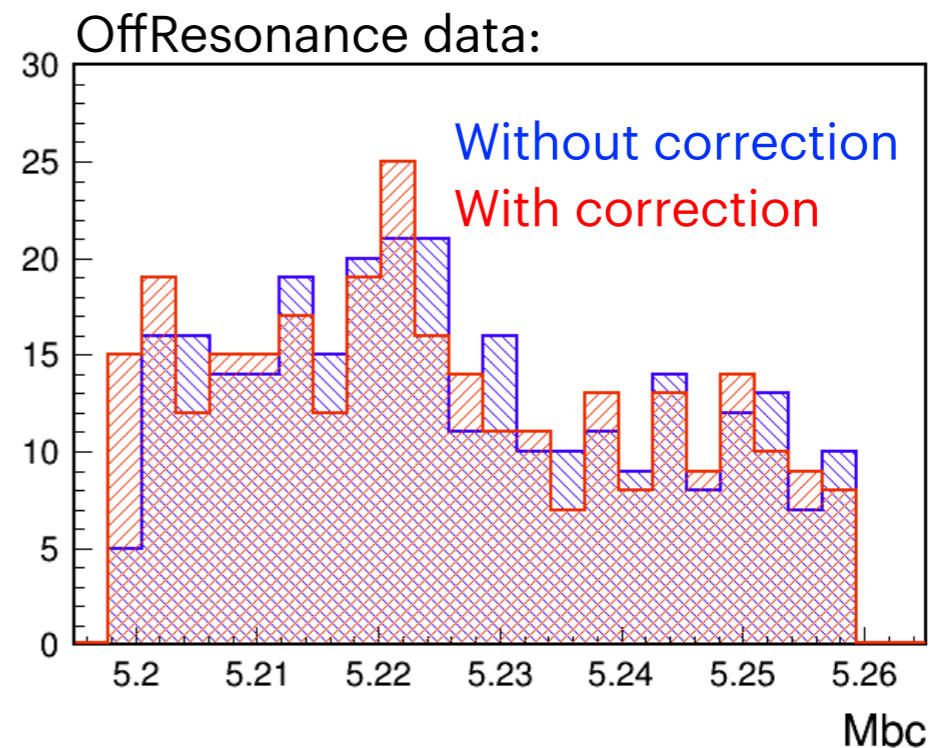
$$C_{\text{offres}}^{\text{total}} = \left(\frac{E_{\text{CMS}}^{\text{offres}}}{E_{\text{CMS}}^{\gamma(4S)}} \right)^2 \approx 0.98$$

$$M_{bc}^{\text{corr}} = \sqrt{\left(\frac{E_{\text{CMS}}^{\gamma(4S)}}{2} \right)^2 - \left(p_{\text{CMS}}^{B_{\text{tag}}} \cdot C_{\text{offres}}^E \right)^2}$$

$$\Delta E^{\text{corr}} = \frac{E_{\text{CMS}}^{\gamma(4S)}}{2} - E_{\text{CMS}}^{B_{\text{tag}}} \cdot C_{\text{offres}}^E$$

Did we expect this?

After the correction I expected the M_{bc} endpoint to be 5.29. Need to think about this.



ρ MVA

ρ MVA

Beyond the CS: identify the principal bkg components.

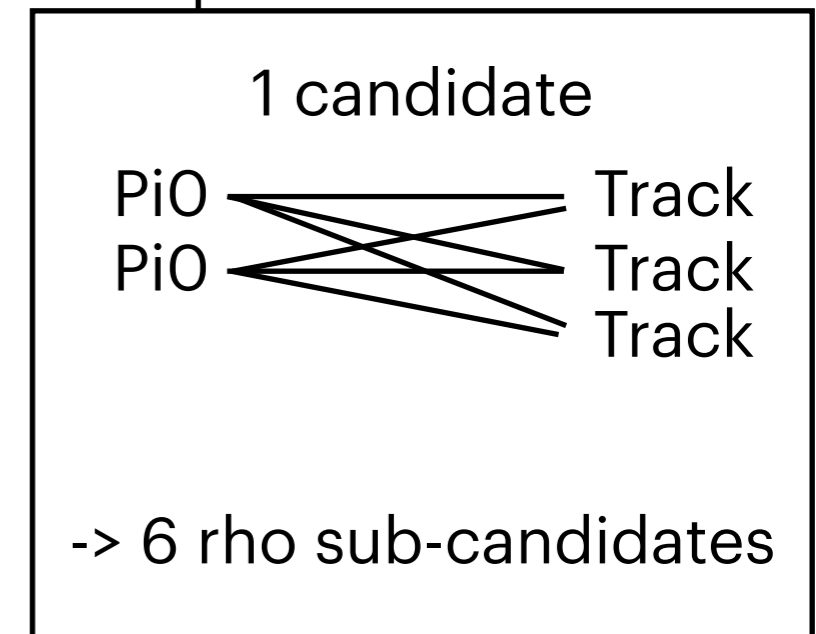
Events that have at least a π^0 from ...	
$\rho(770)^+$	47.1%
Z^0 (direct from e^+e^-)	75.0%

Large number of continuum π^0 's come from a $\rho \rightarrow$
develop a specific BDT (in addition to the default CS BDT).

Combine each track in the event with each π^0 .

Use kinematic and angular variables to distinguish
between ρ 's and other particles.

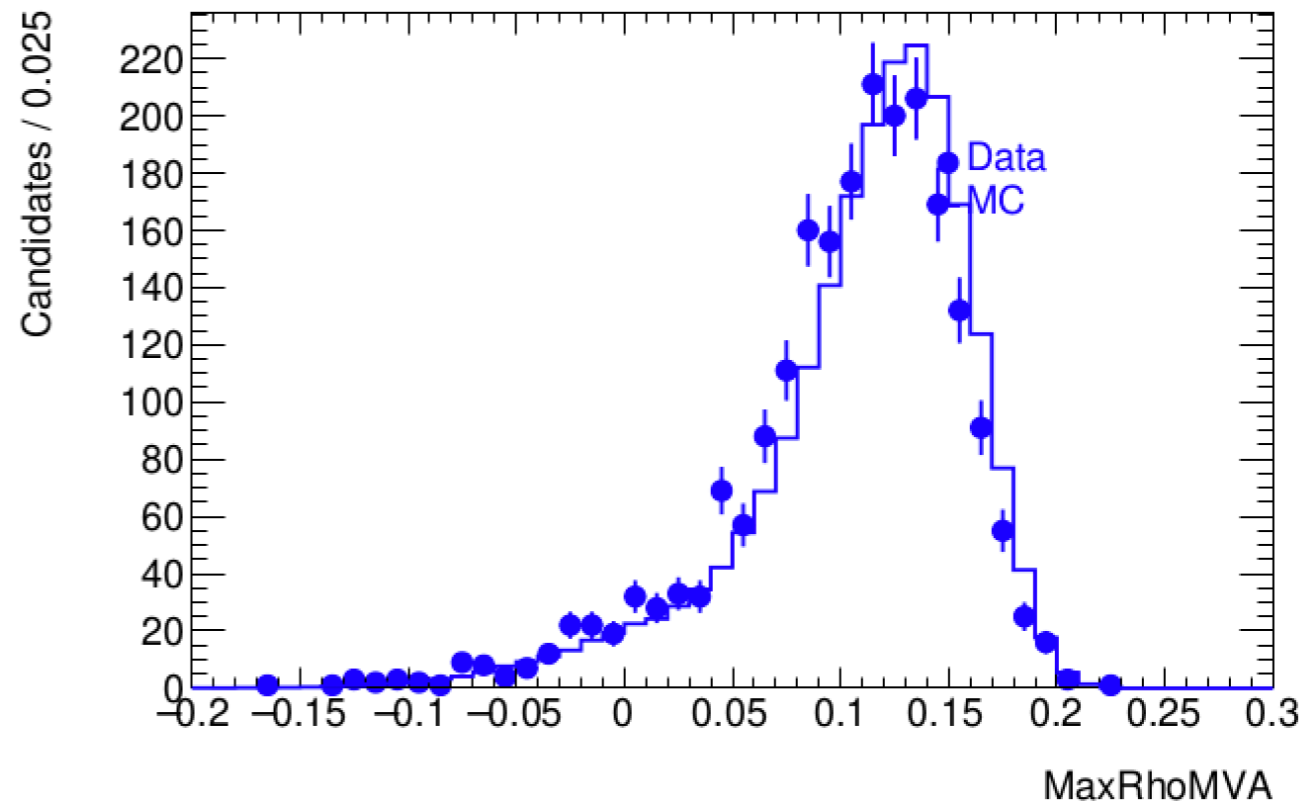
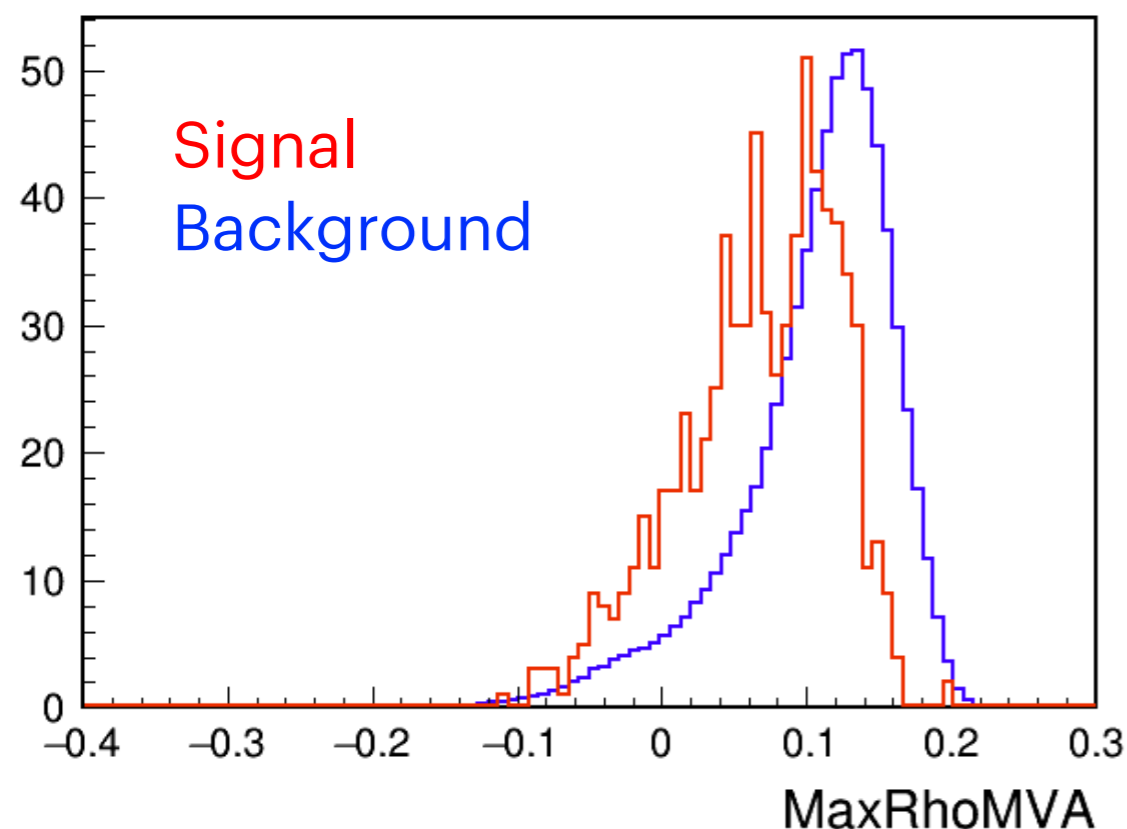
Example:



Max ρ MVA distribution

Each candidate has for example 20 ρ sub-candidates. Take the one with largest rhoMVA (the one more similar to a ρ).

Validation: use $B^0 \rightarrow \pi^0\pi^0$ sideband (inclusive sample of true and false ρ).



Variable gives separation, and discrepancy is acceptable

Total candidates

Candidates with at least one rho

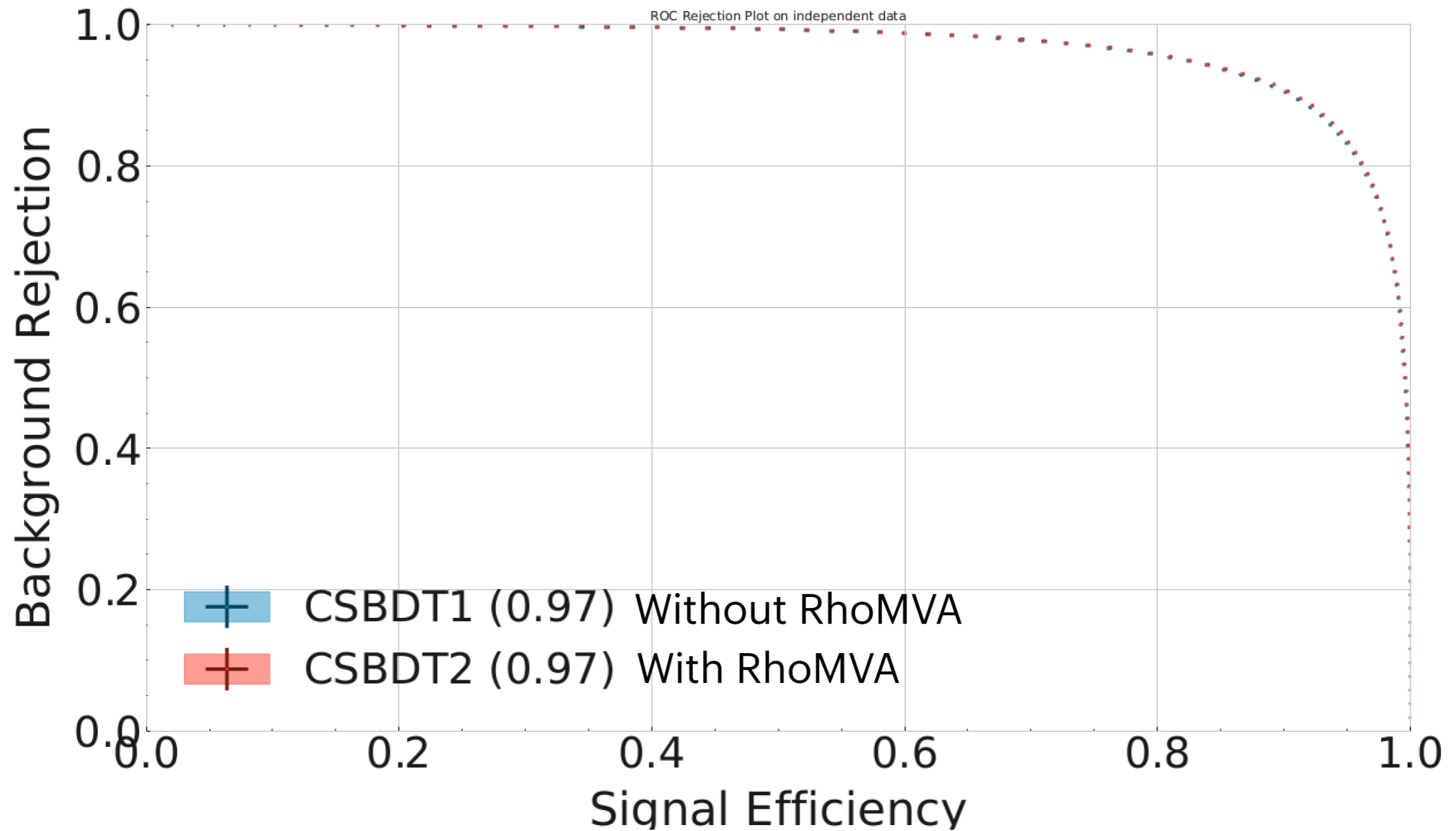
Candidates where the rho has been correctly identified

788473

285585

158393

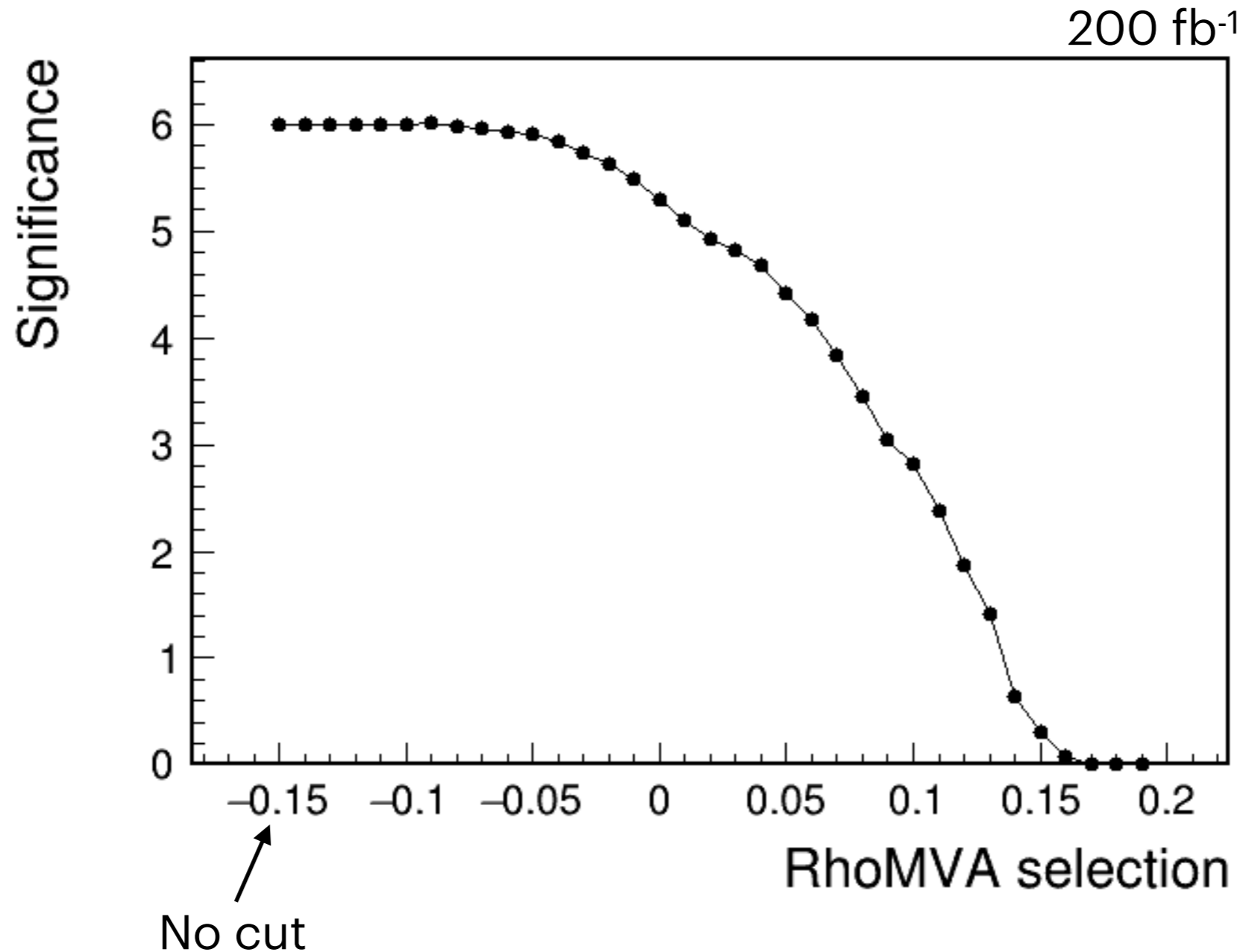
Use ρ MVA as input of the CSBDT



Inclusion of ρ MVA gives no improvement

Other possibility: ρ MVA after the CSBDT

Apply first the selection on the CSBDT (>0.8), $-0.2 < \Delta E < 0.1$ and $M_{bc} > 5.27$, then various selections on ρ MVA and calculate significance $S/\sqrt{S+B}$.



No gain in significance after selection on ρ MVA.

Summary

Prepare $B^0 \rightarrow \pi^0 \pi^0$ analysis for pre-LS1 dataset.

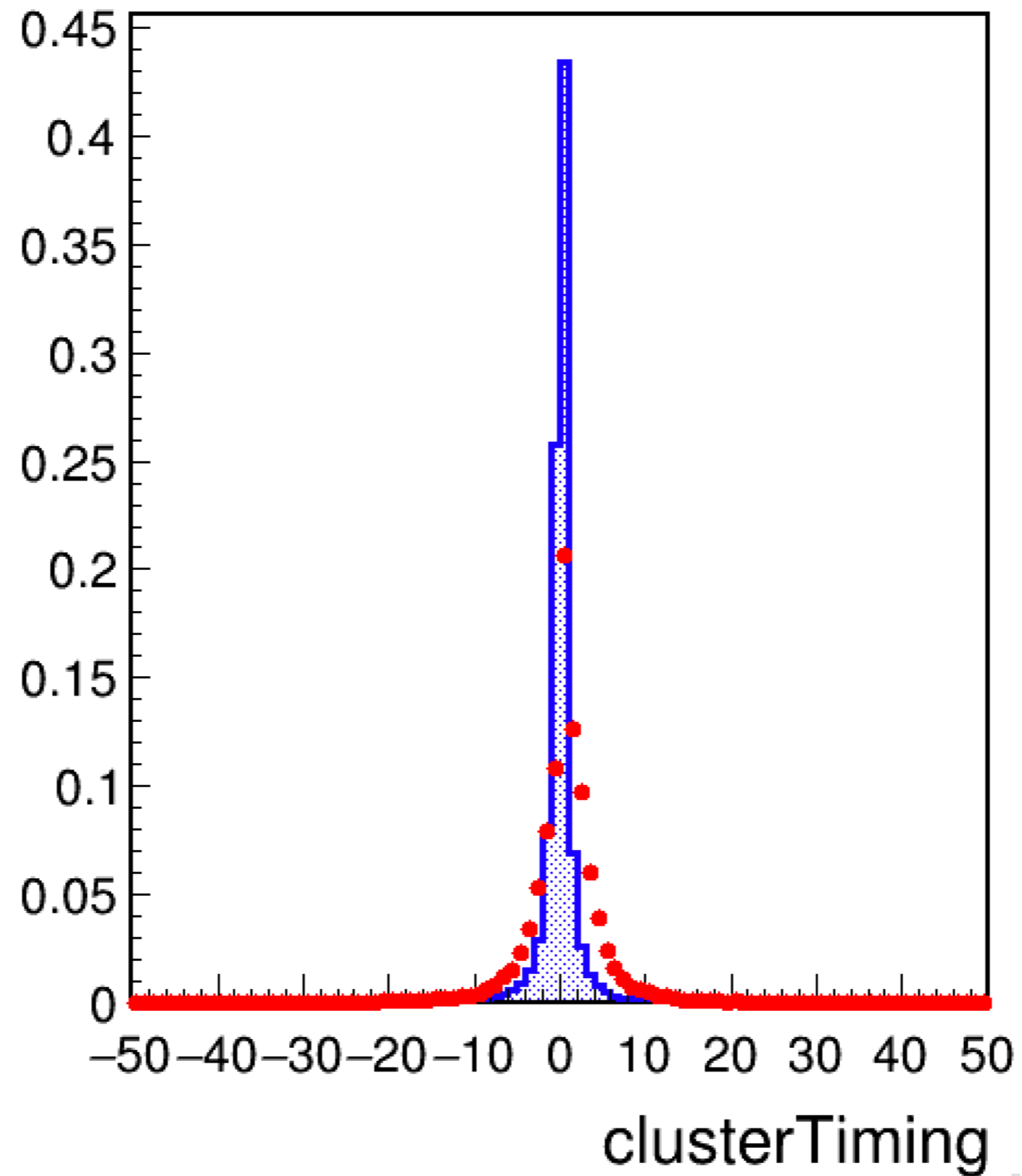
Revisited photonMVA: use new variables with good data/MC agreement.
Already validated on data.

Revisited CSBDT: add B_{Tag} variables to suppress even more continuum.
Variables are ready, but need to repeat training using off-res data (is it enough?). Check how the use of B_{Tag} variables impacts the flavour tagger.

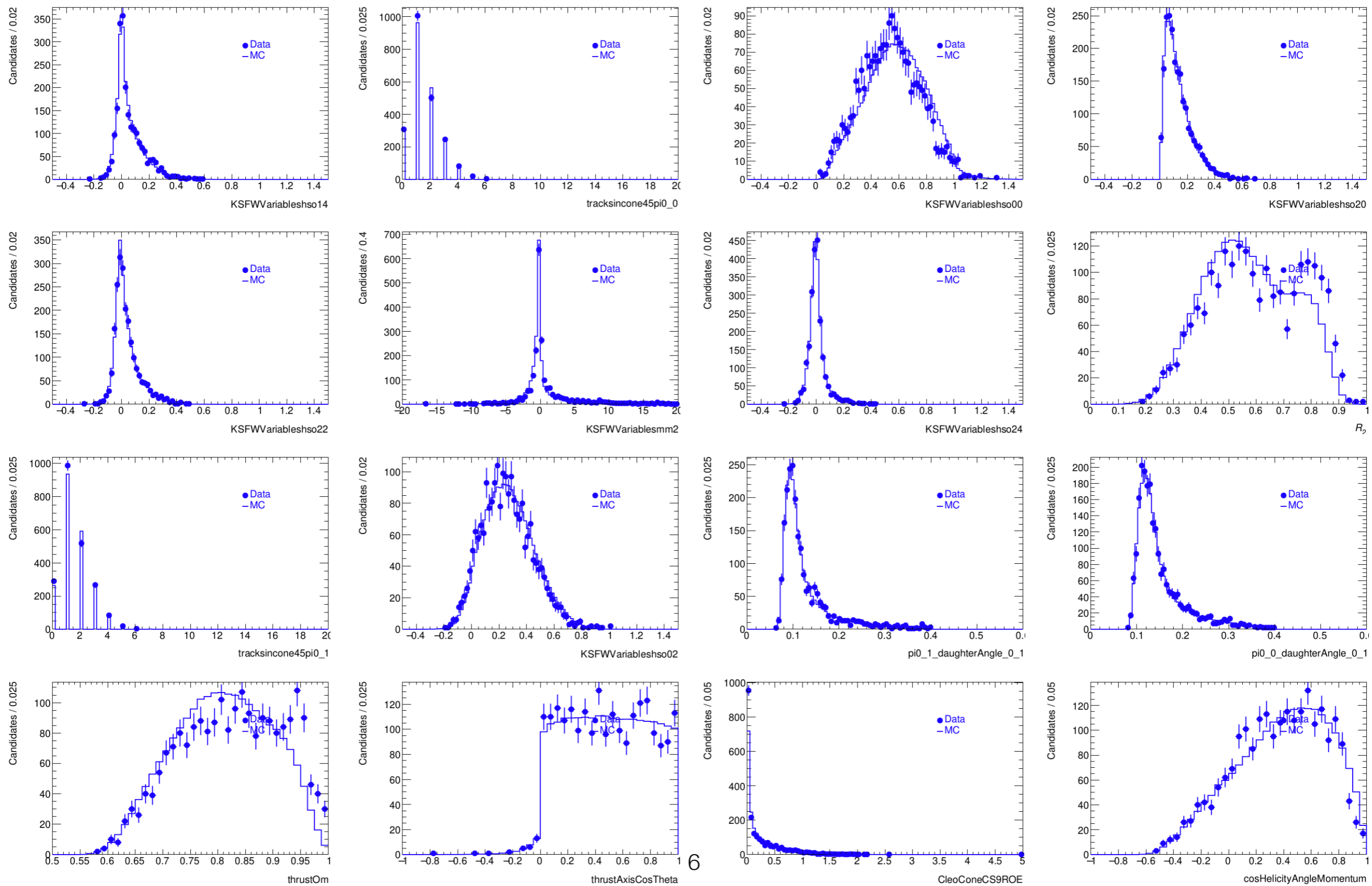
Introduced ρ BDT: improvement is negligible, maybe not useful to add it in the analysis.

Backup

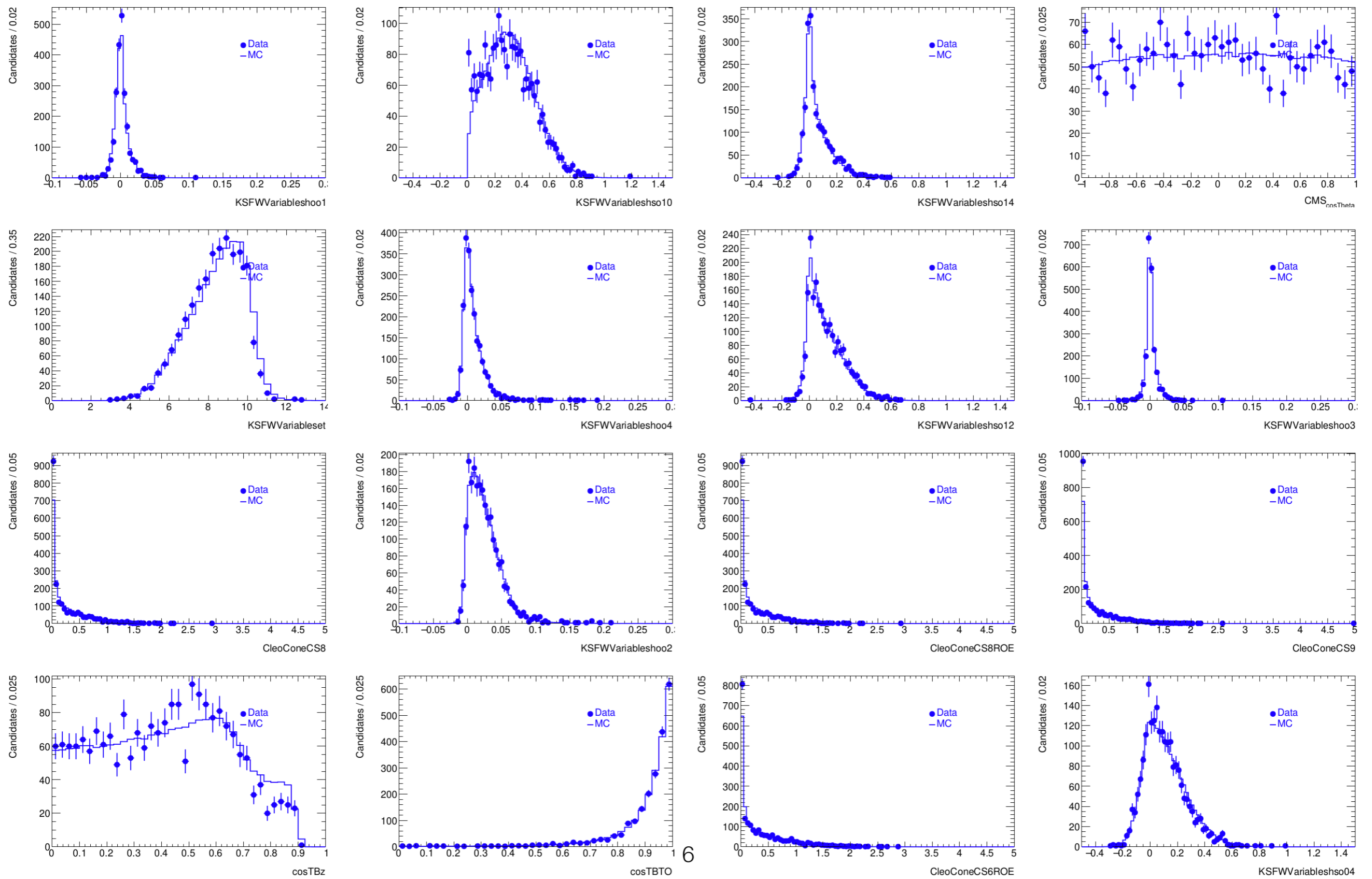
ClusterTiming (rel-06)



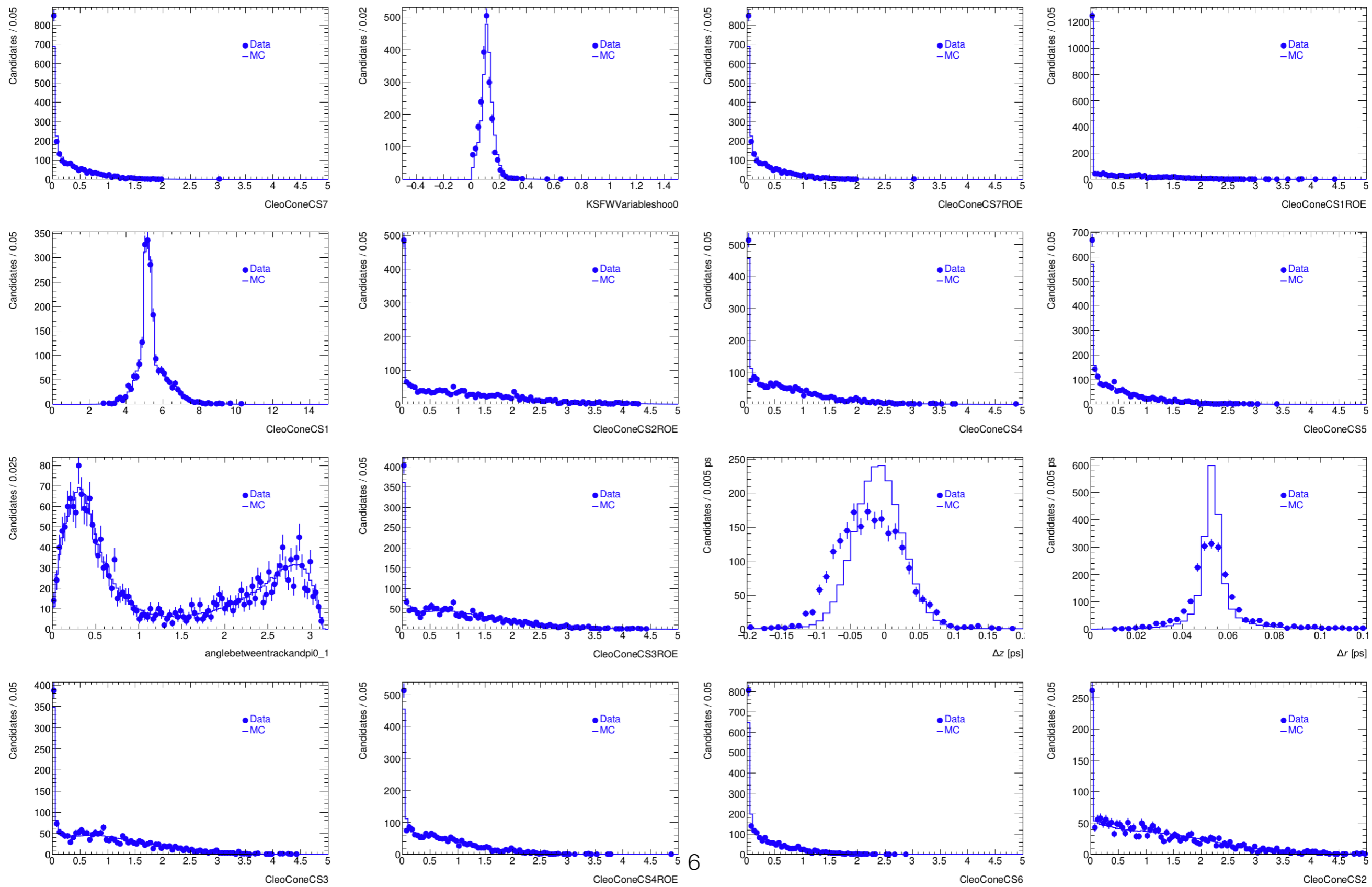
Inputs validation — Background only



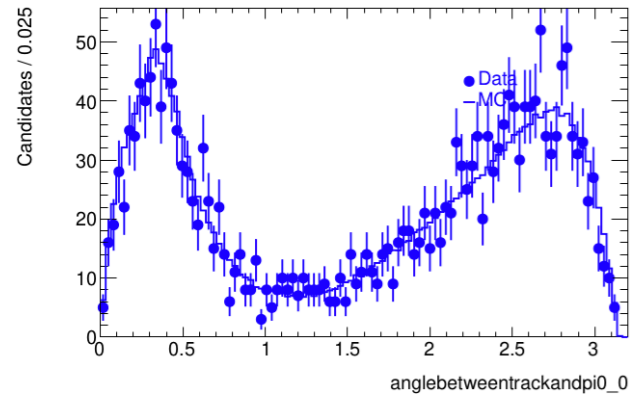
Inputs validation — Background only



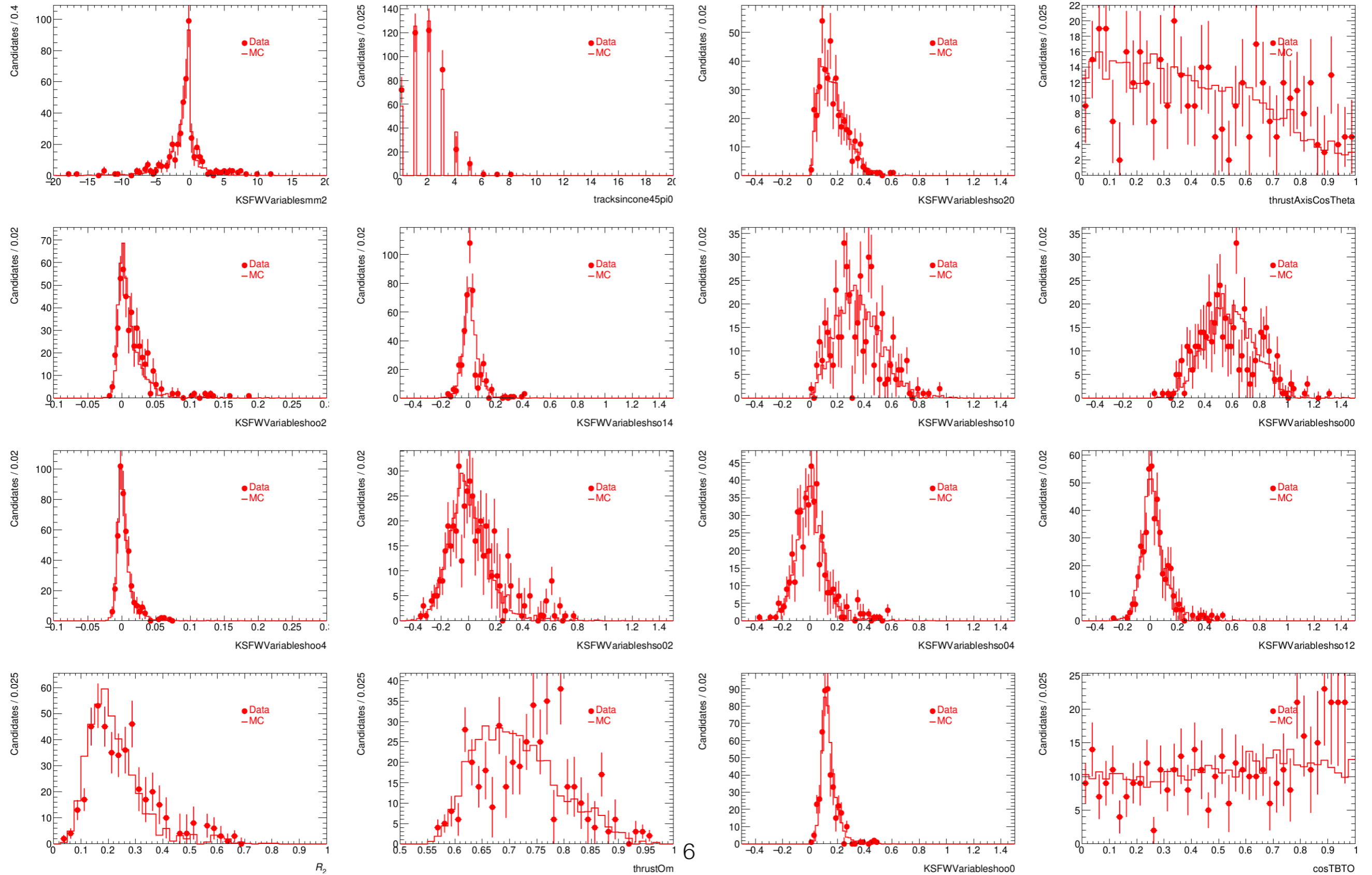
Inputs validation — Background only



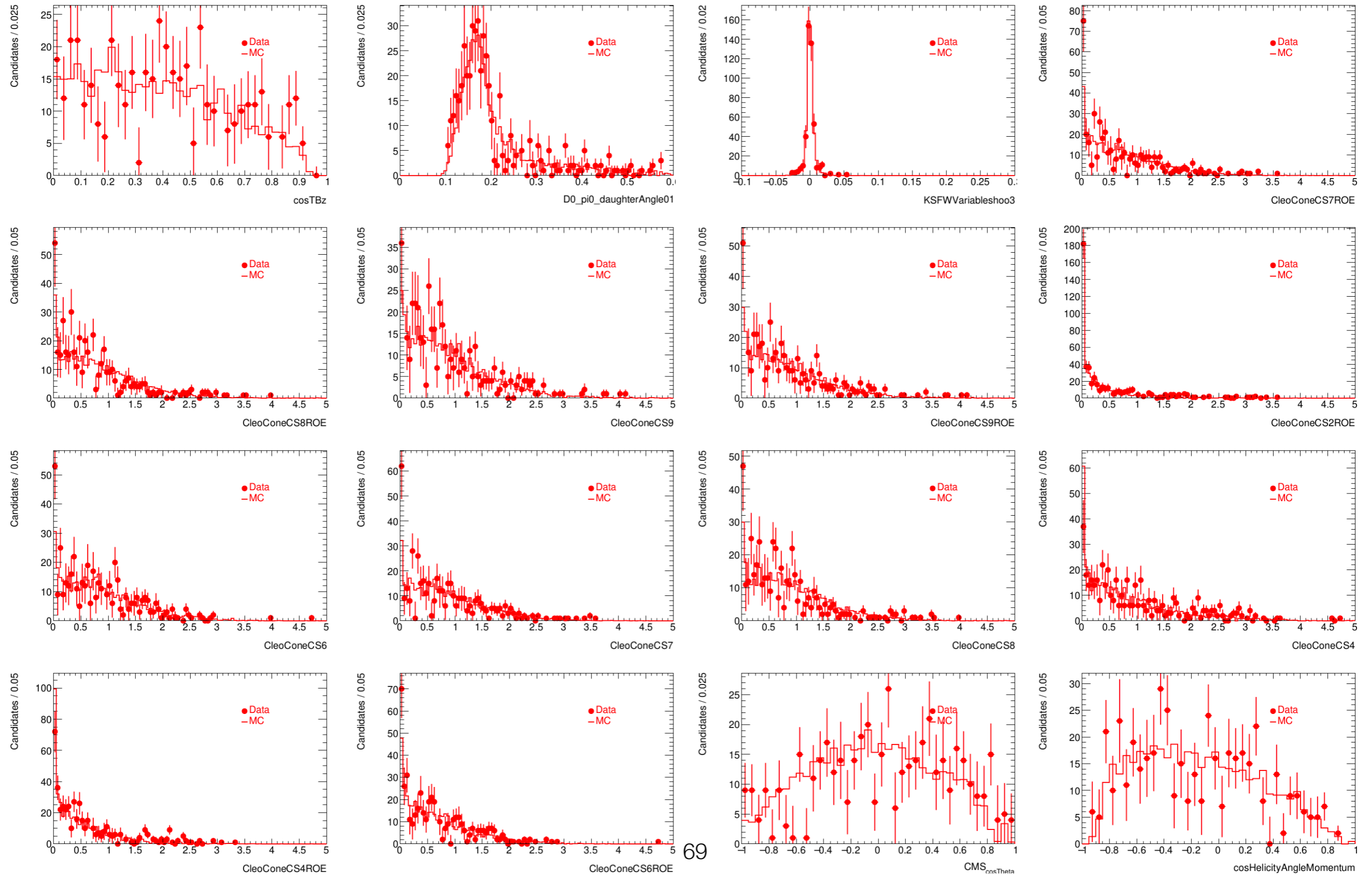
Inputs validation — Background only



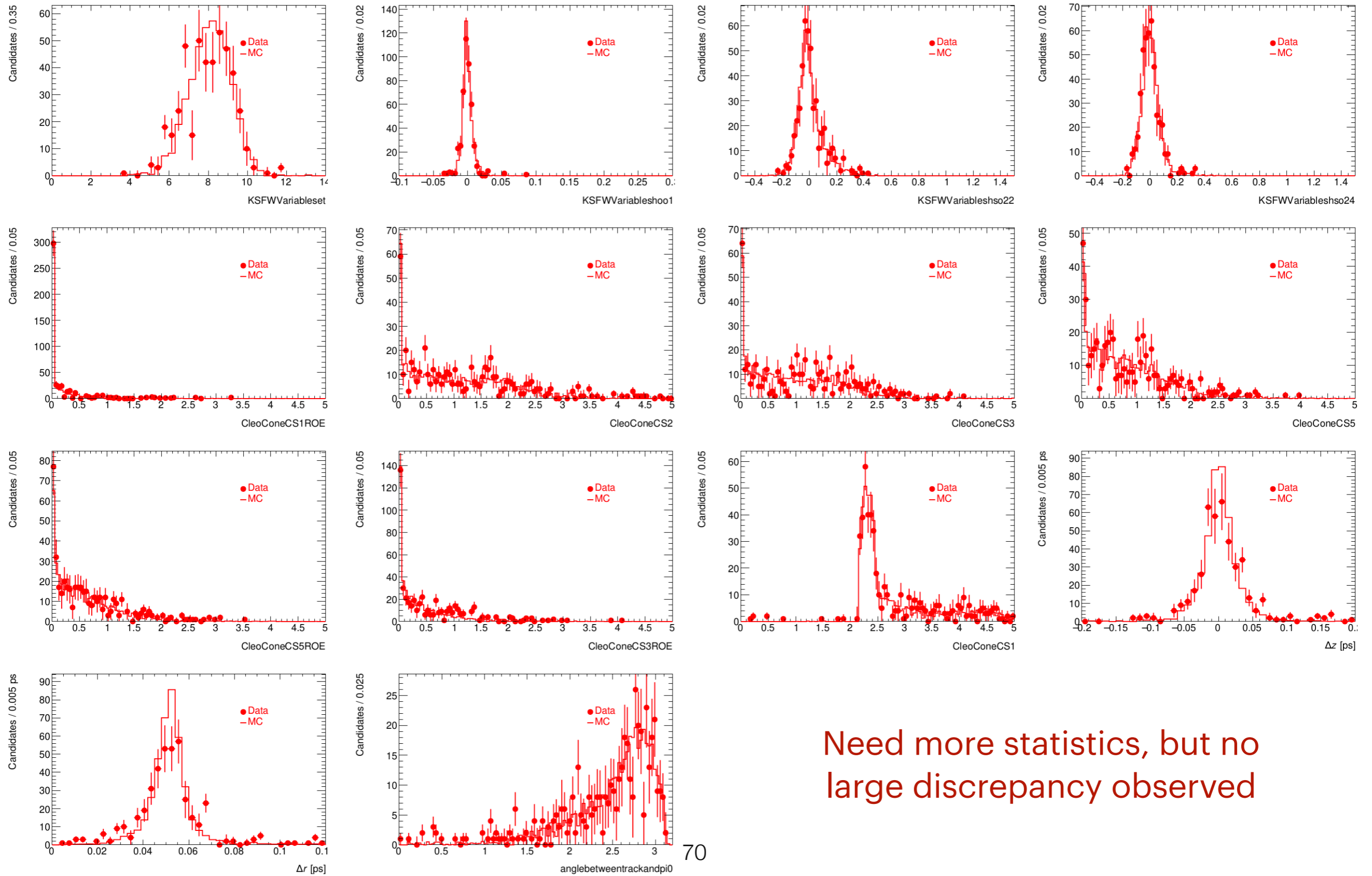
Inputs validation — Signal only



Inputs validation — Signal only



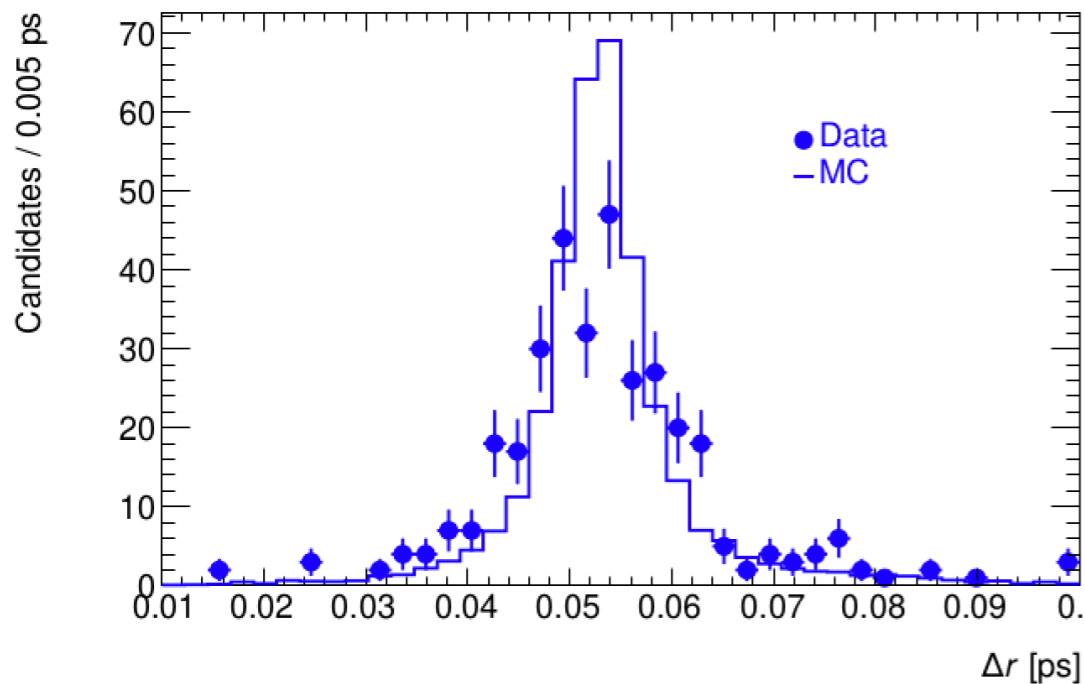
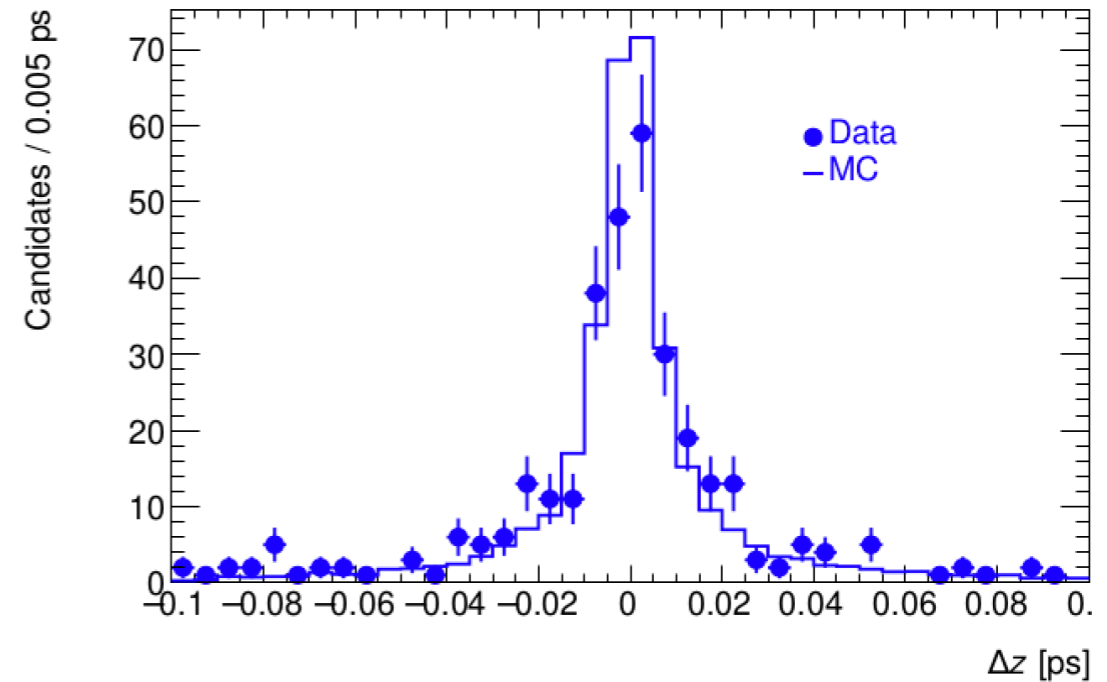
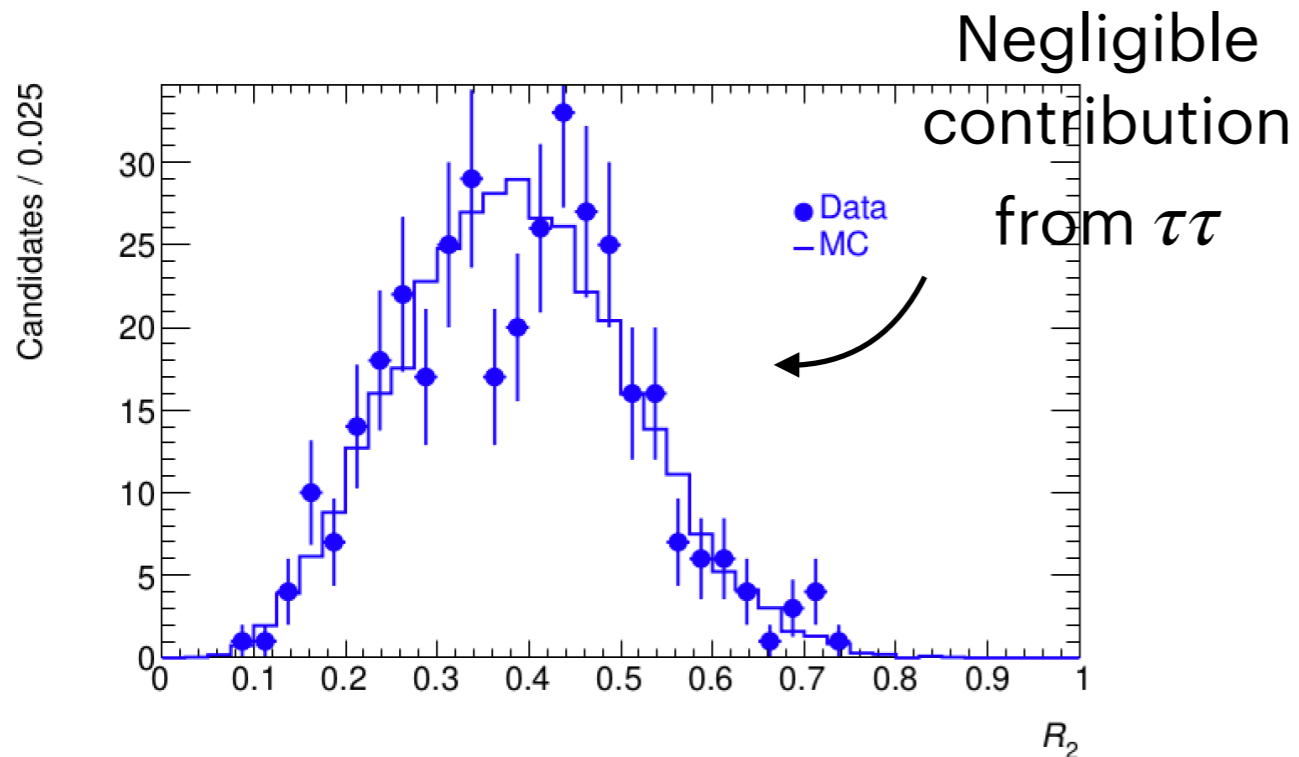
Inputs validation — Signal only



Need more statistics, but no large discrepancy observed

Check — Background only using

$$B \rightarrow D(K\pi\pi^0)\pi \text{ sideband}$$



Need more statistics — but observe smaller discrepancies in Δr and Δz wrt $B^0 \rightarrow \pi^0\pi^0$. Why?