



LCFIVertex Flavour Tagging Studies

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Introduction

- The aim of these studies is to extract the optimal LCFIVertex parameters for flavour tagging considering more recent and more realistic detector model and reconstruction software.
 - Parameters for joint probability;
 - Neural networks;
 - Track selection for vertex reconstruction and flavour tag inputs.
- Present LCFIVertex default parameters for flavour tagging are outdated (Tesla-SGV).

LCFIVertex package

- The LCFIVertex package is a software package that uses the vertex-detector information for high-level event reconstruction at the ILC. It provides:
 - The ZVTOP vertex finder;
 - A flavour tagging algorithm based on neural networks approach;
 - Vertex charge determination for b- and c-jets.
- The code is based on Marlin and uses LCIO for input and output.
- The code and the networks are available from a CVS repository under *marlinreco* and *tagnet*, respectively:
<http://www-zeuthen.desy.de/lc-cgi-bin/cvsweb.cgi>

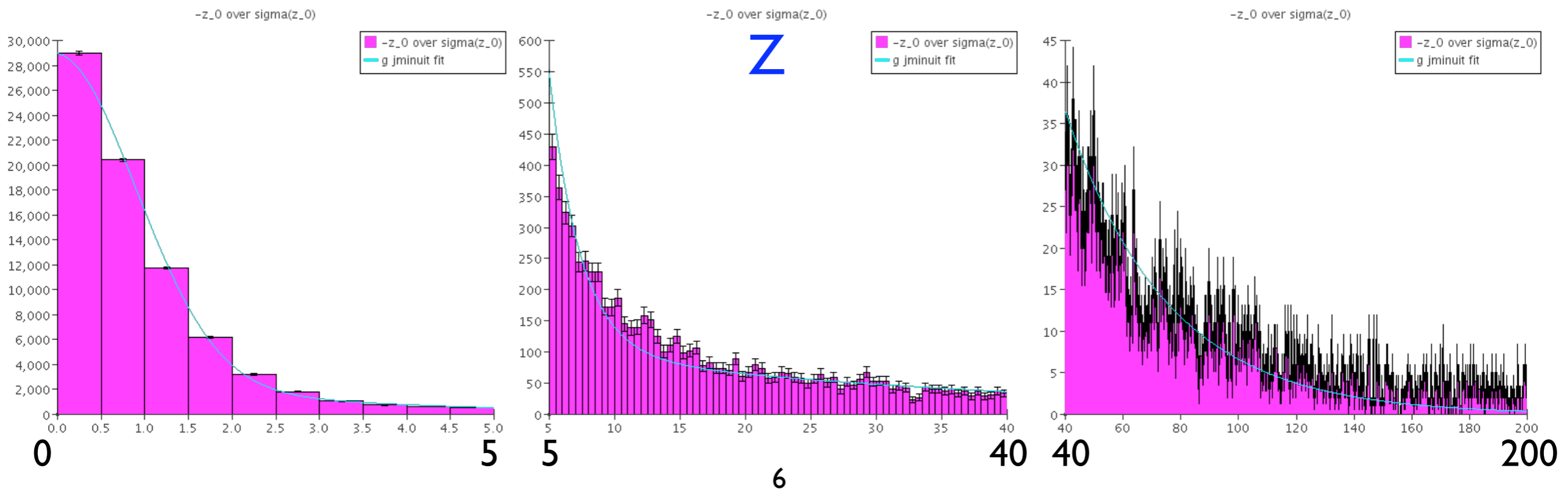
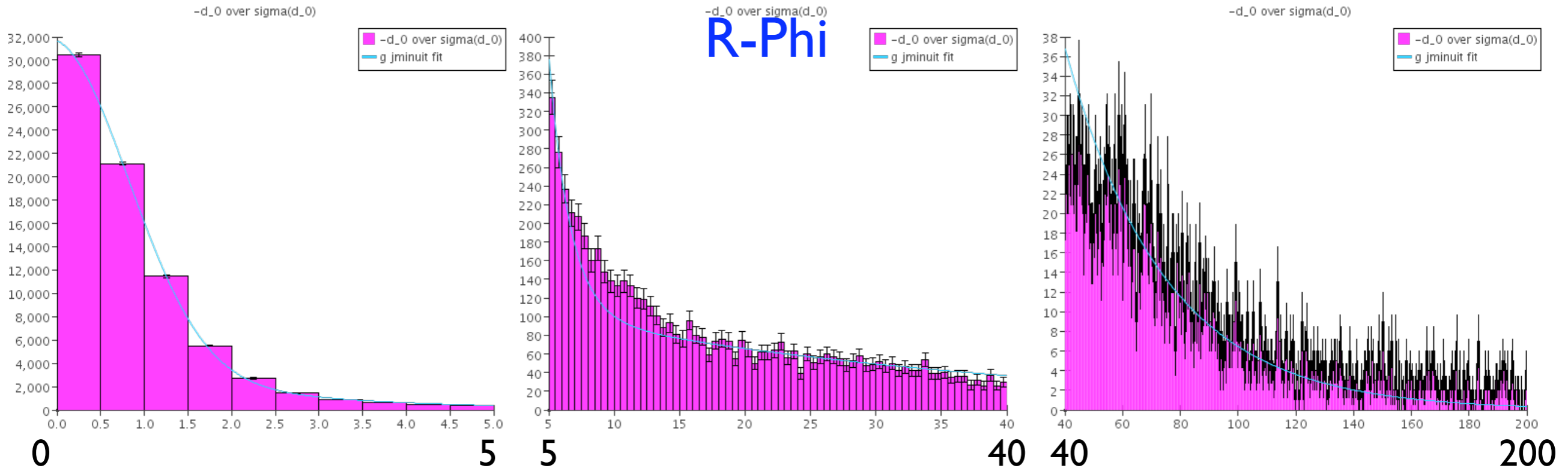
Parameters for Joint Probability

Joint probability

- Joint probability is the probability that a track in a jet comes from the primary vertex.
- The parameters used in the joint probability are obtained by fitting the negative impact parameters in $R\Phi$ and z to a gaussian+exponential+exponential function. The SignificanceFit processor is used.

Joint probability parameters: SignificanceFit

Zooming the global fit



Joint probability parameters: SignificanceFit

default

PARAMETERS FOR RPHI Joint Probability

1.01313412
0.0246350896
0.102197811
0.0411203019
0.0157710761

PARAMETERS FOR Z Joint Probability

1.01629865
0.0271386635
0.0948112309
0.0410759225
0.0148685882

new (LDCPrime_02Sc)

PARAMETERS FOR RPHI Joint Probability

**0.843068
0.364774
0.619891
0.150243
0.0290308**

PARAMETERS FOR Z Joint Probability

**0.910629
0.305746
0.422501
0.139363
0.028365**

LDC01_05Sc

PARAMETERS FOR RPHI Joint Probability

1.01498
0.27984
0.561155
0.00601934
0.0476549

PARAMETERS FOR Z Joint Probability

1.04273
0.27057
0.468457
0.00631426
0.0479484

Flavour tag inputs

Flavour tag inputs

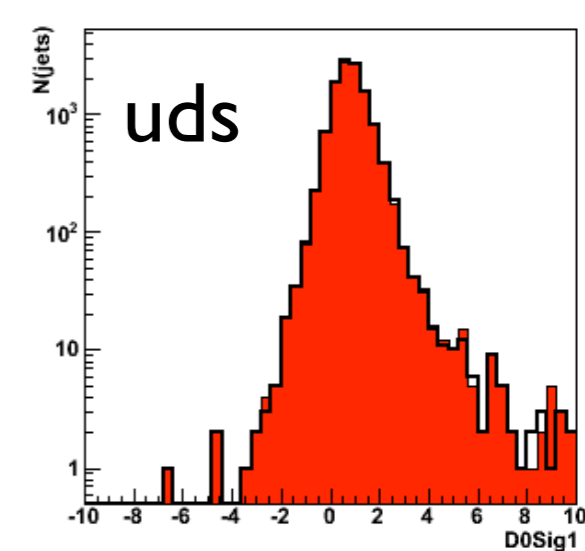
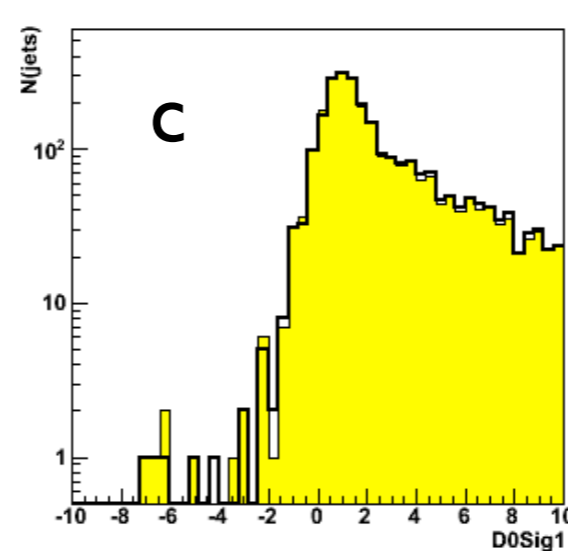
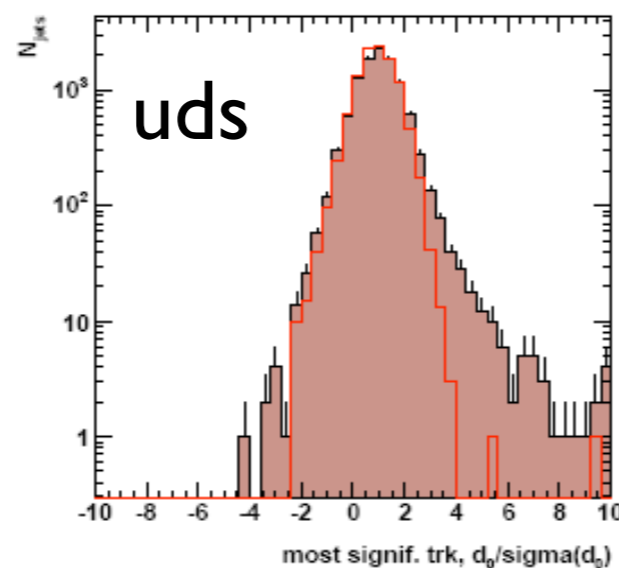
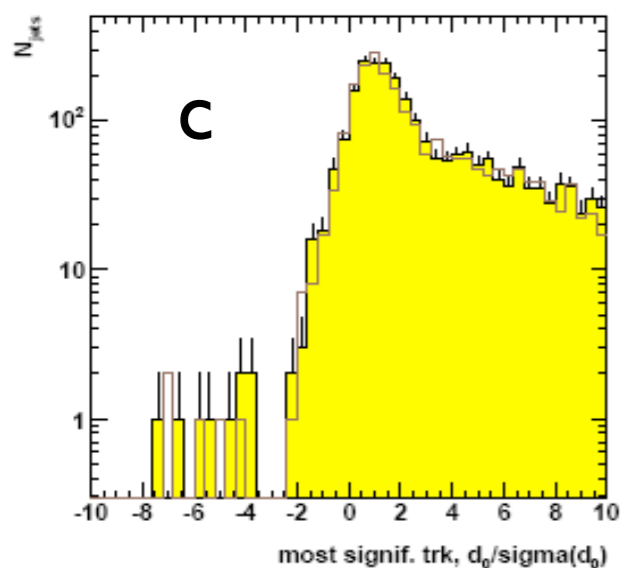
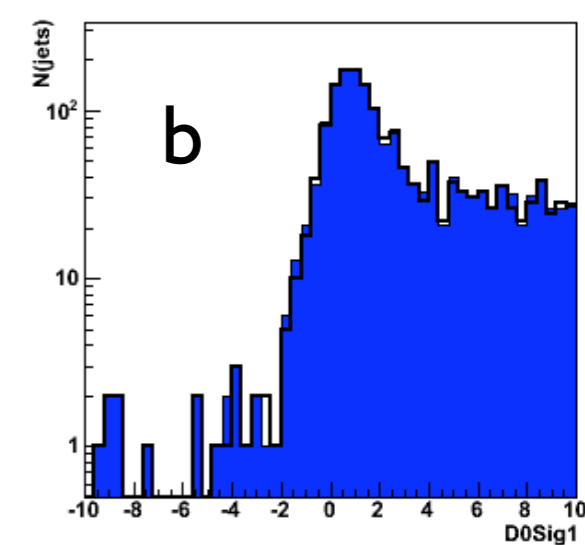
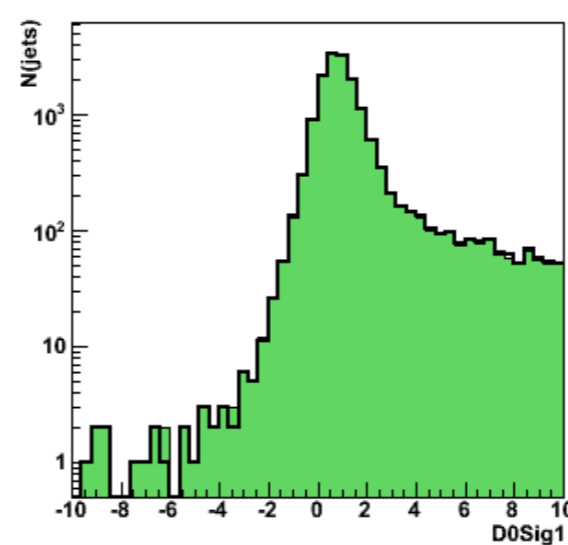
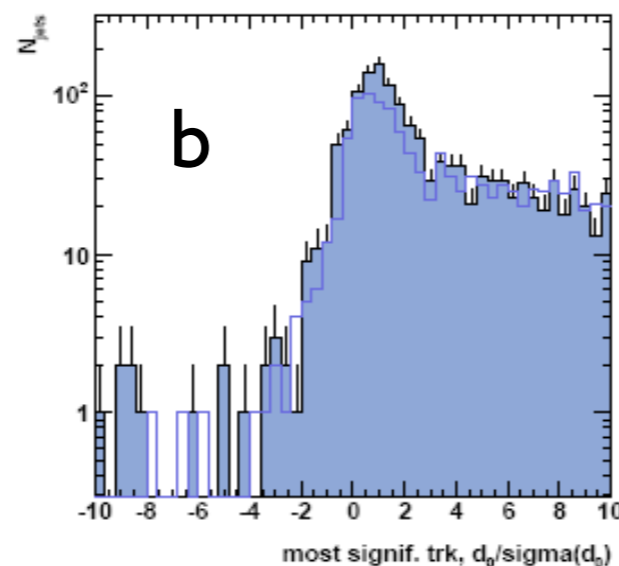
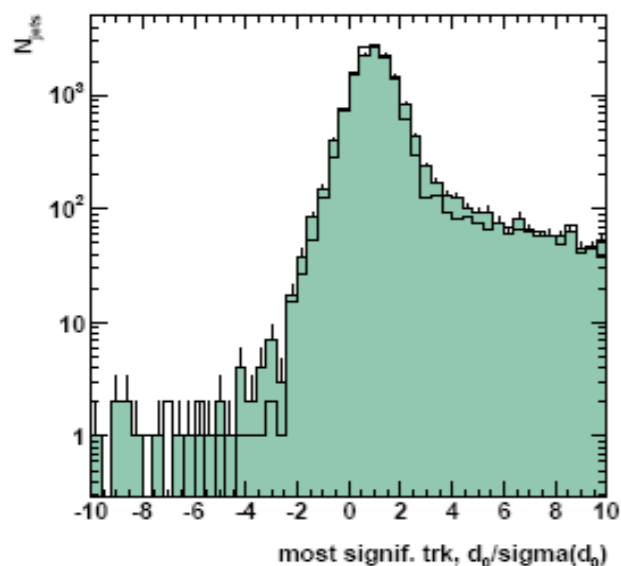
- **Monte Carlo sample:**
 - $e^+e^- \rightarrow Z \rightarrow qq$, $\sqrt{s} = 91.2$ GeV;
 - 10000 events;
 - Detector model: LDCPrime_02Sc.
- **Reconstruction:**
 - ilcsoft v01-03-06-p02; LCFIVertex HEAD;
 - tracking: FullLDCTracking;
 - clustering + particle flow: PandoraPFA;
 - jets: Satoru jet finder, durhamjet, njet = 2;
 - vertexing and flavour tag inputs: ConversionTagger

Impact parameter significance $R\varphi$

Most significant track

SGV - line; Mokka - histo
(S. Hillert)

LDCPrime_02Sc
cheat - line; conv. tag - histo



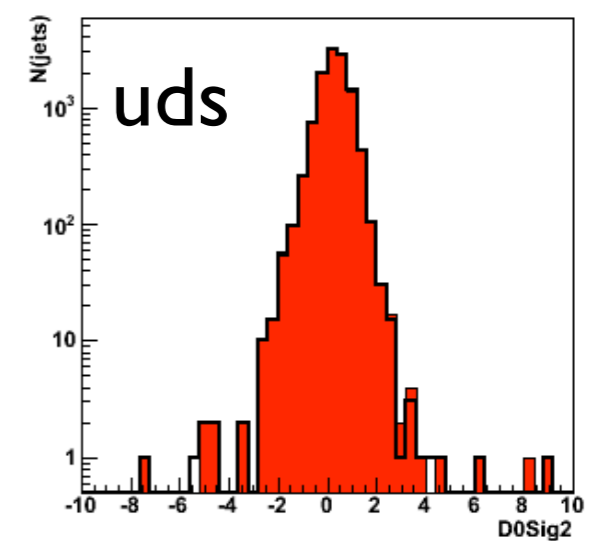
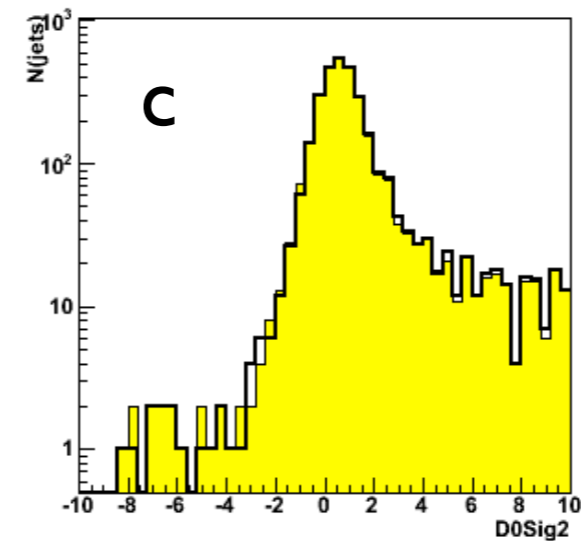
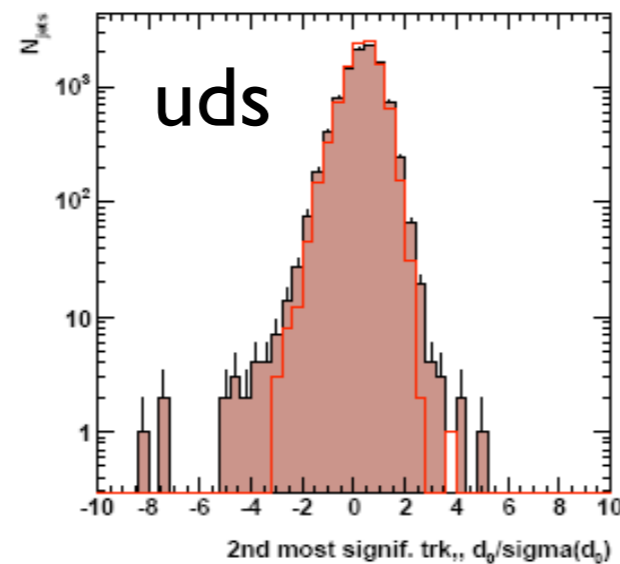
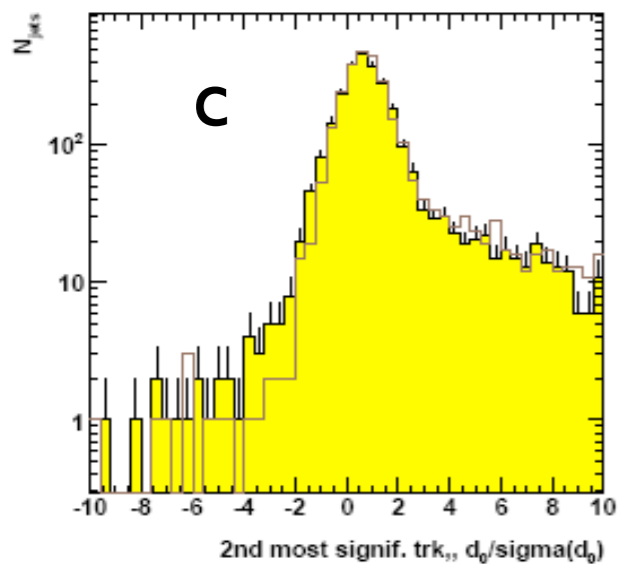
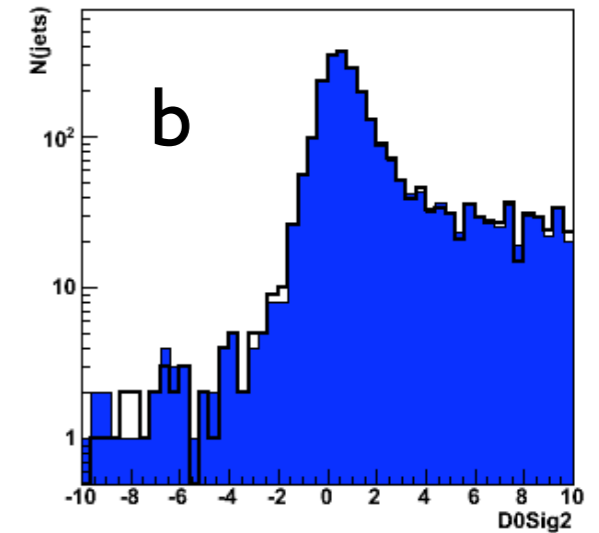
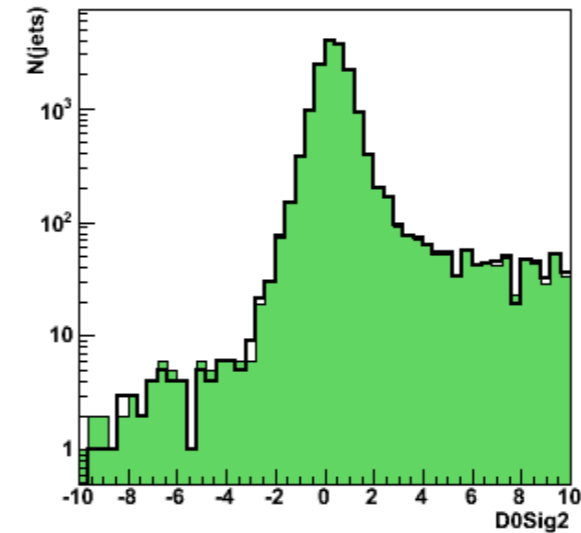
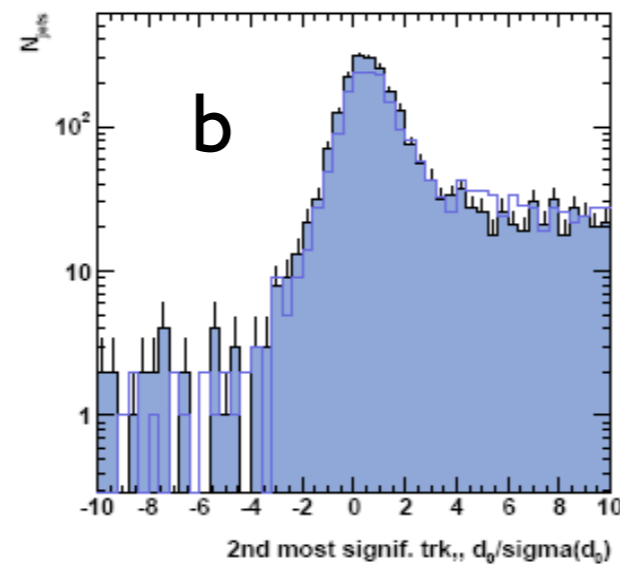
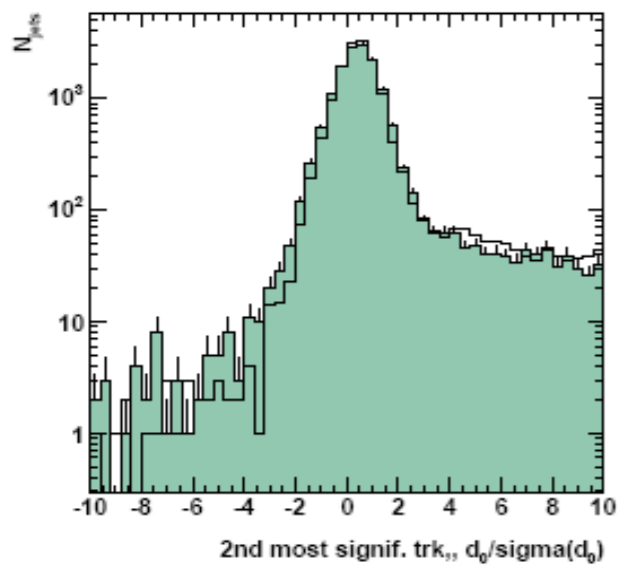
Notice the log scale!

Impact parameter significance $R\varphi$

Second most significant track

SGV - line; Mokka - histo
(S. Hillert)

LDCPrime_02Sc
cheat - line; conv. tag - histo



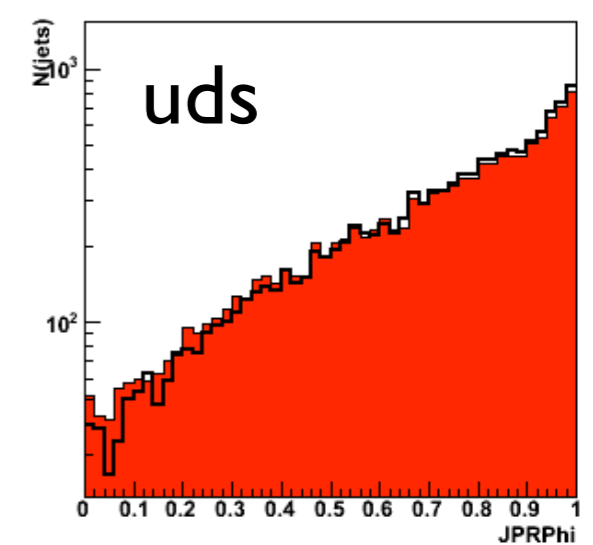
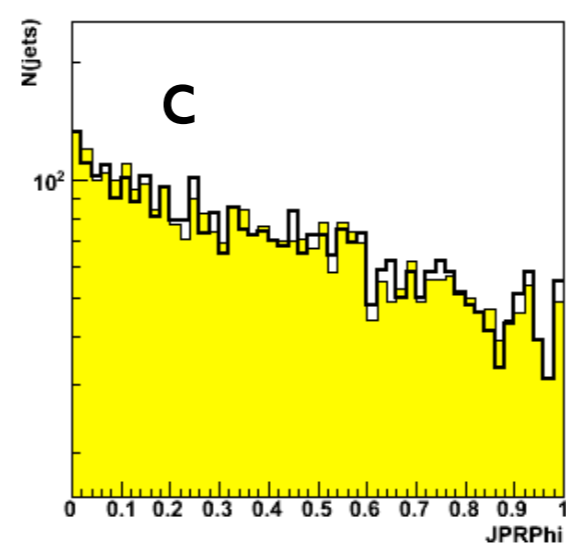
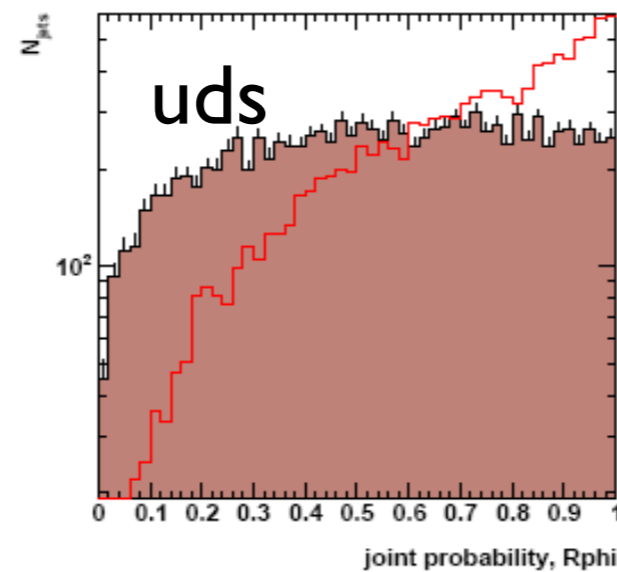
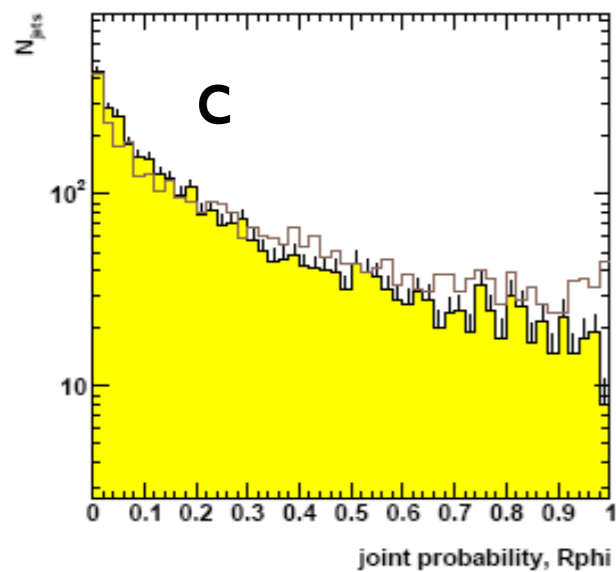
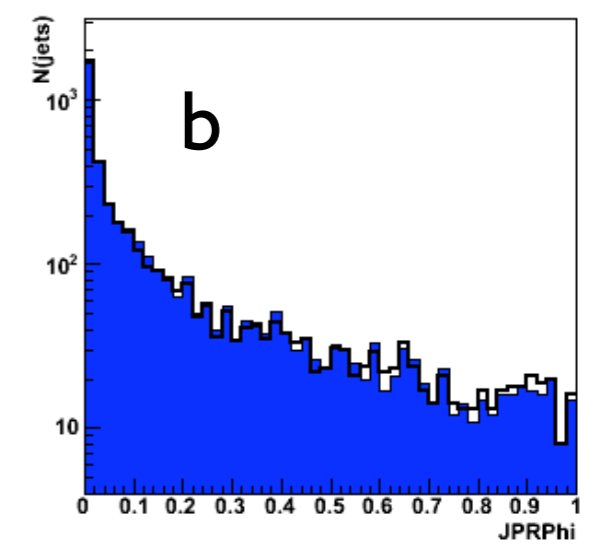
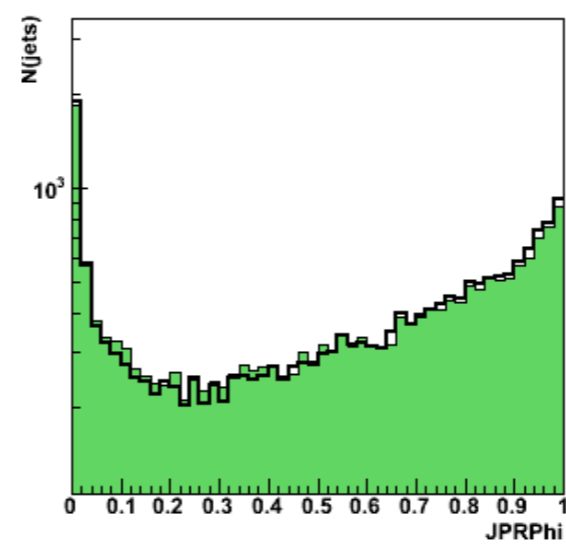
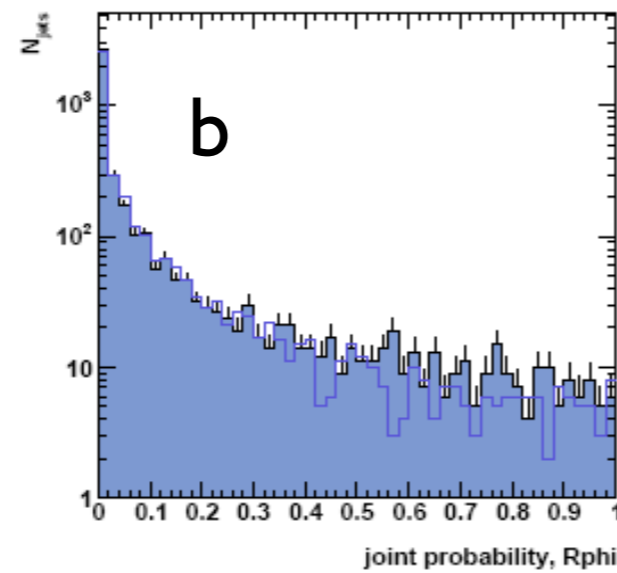
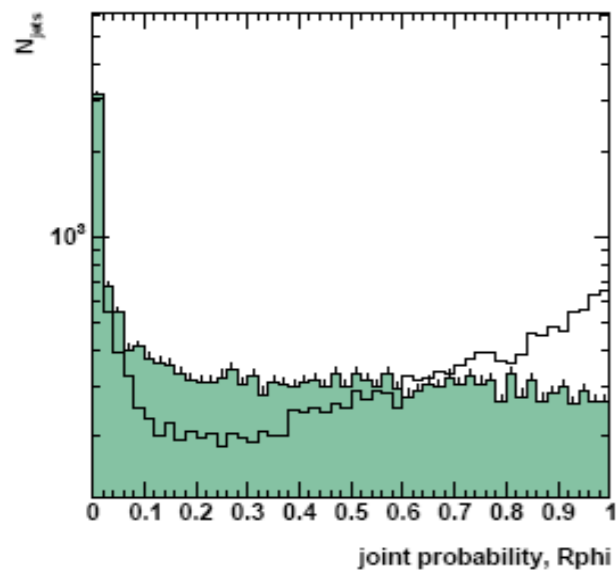
Notice the log scale!

||

Joint probability $R\phi$

SGV - line; Mokka - histo
(S. Hillert)

LDCPrime_02Sc
cheat - line; conv. tag - histo

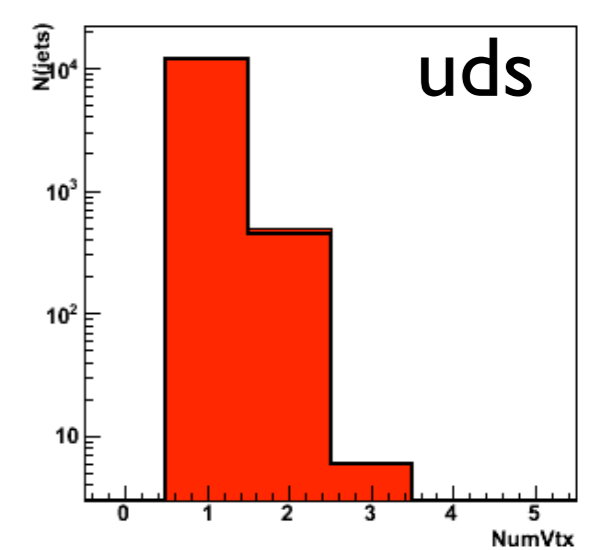
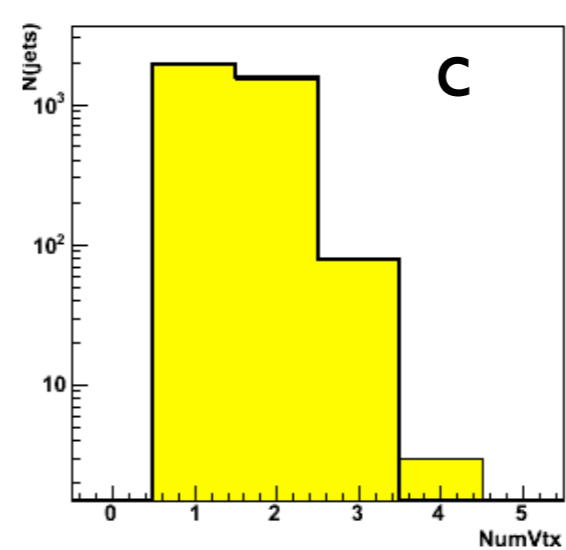
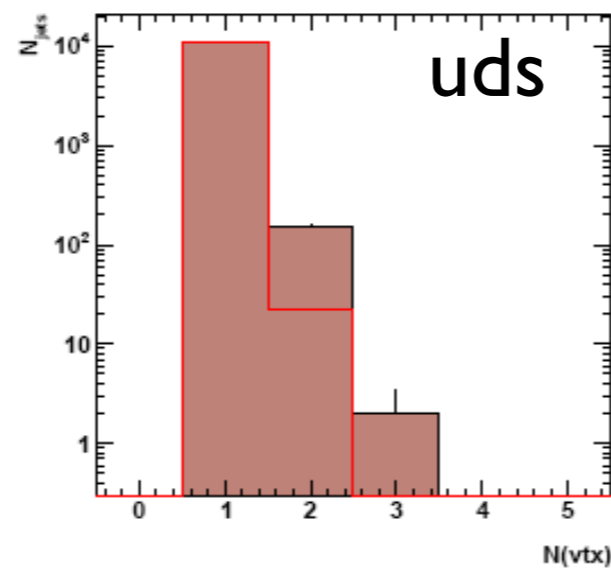
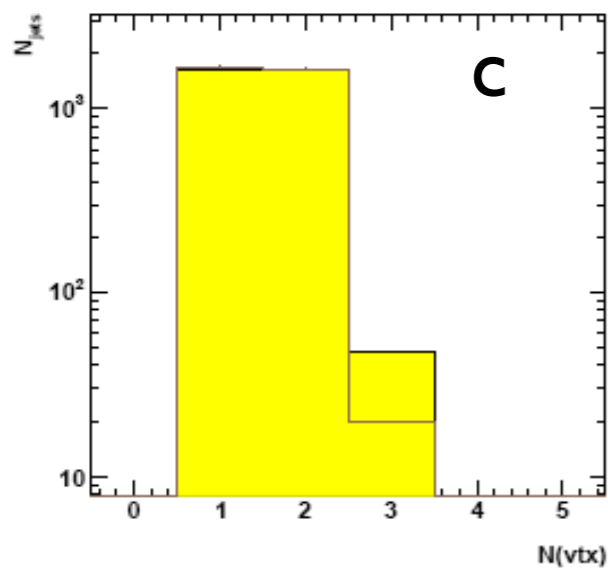
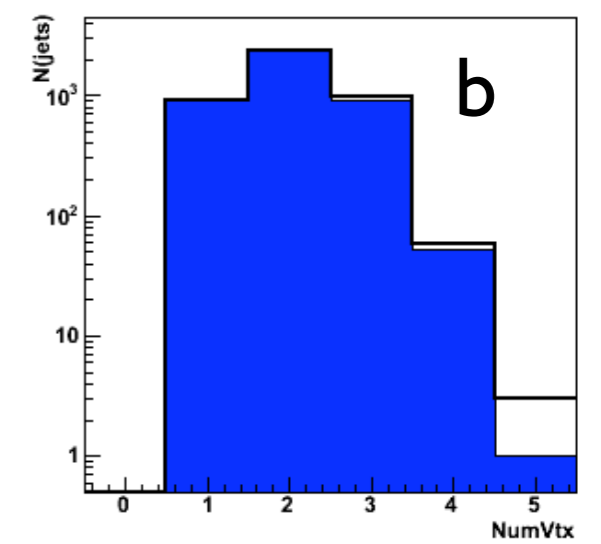
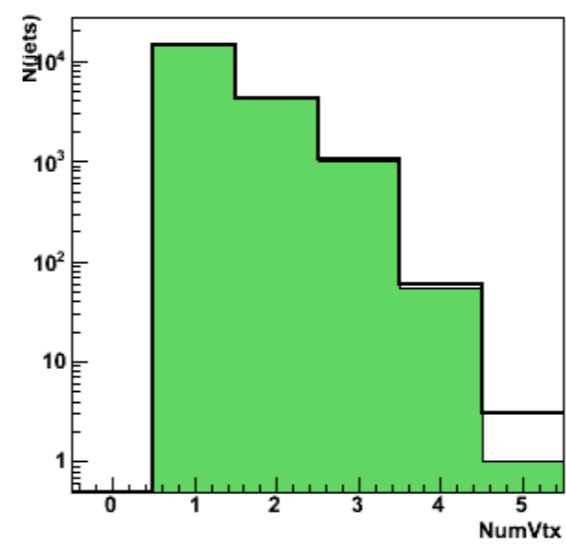
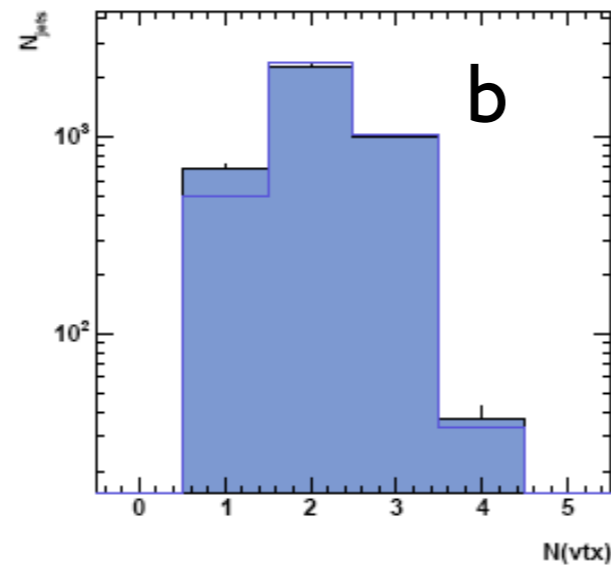
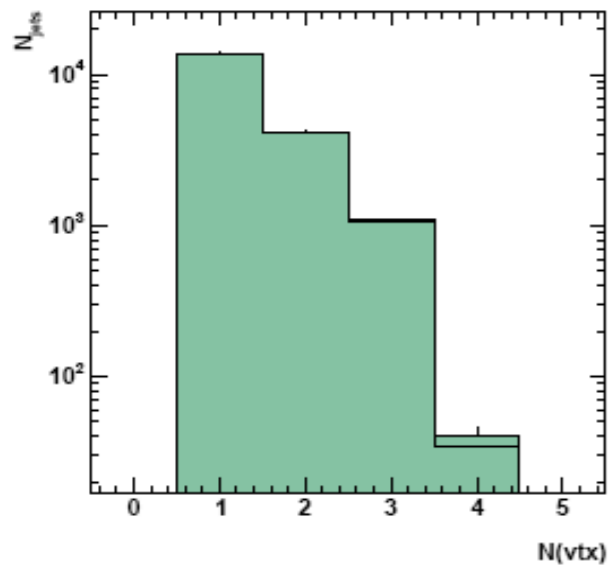


Notice the log scale!

Vertex multiplicity

SGV - line; Mokka - histo
(S. Hillert)

LDCPrime_02Sc
cheat - line; conv. tag - histo

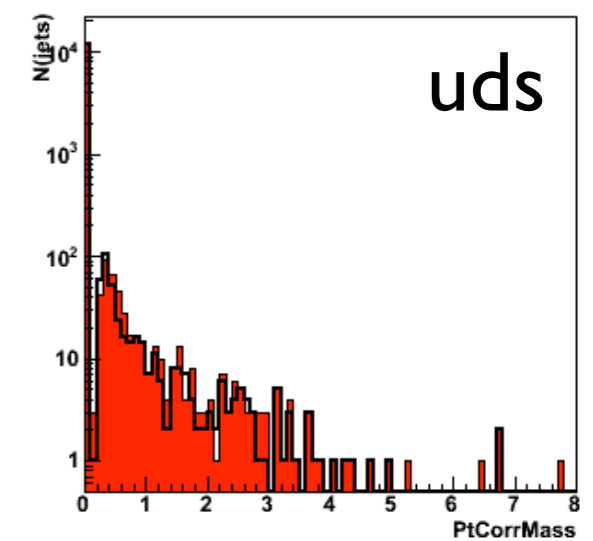
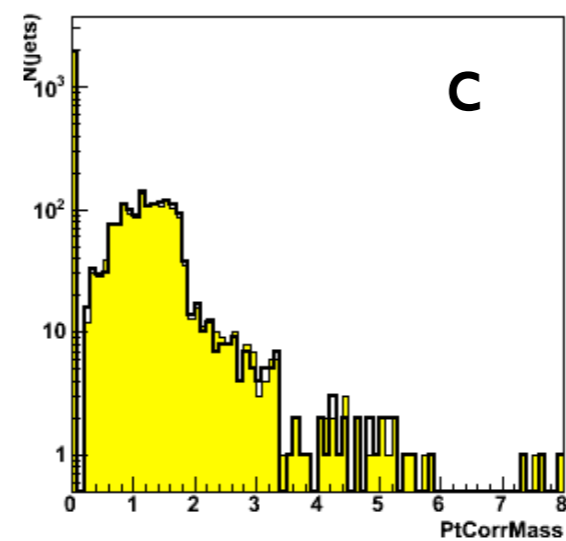
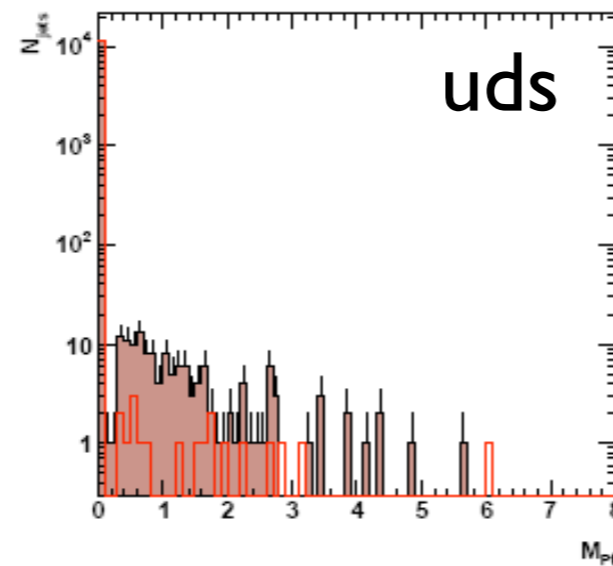
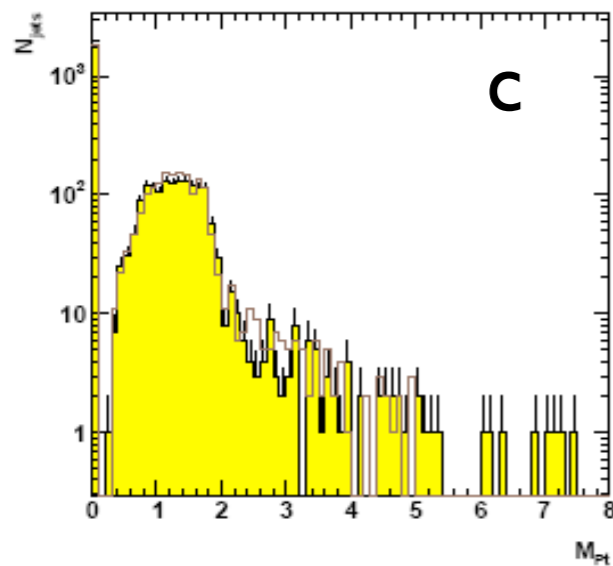
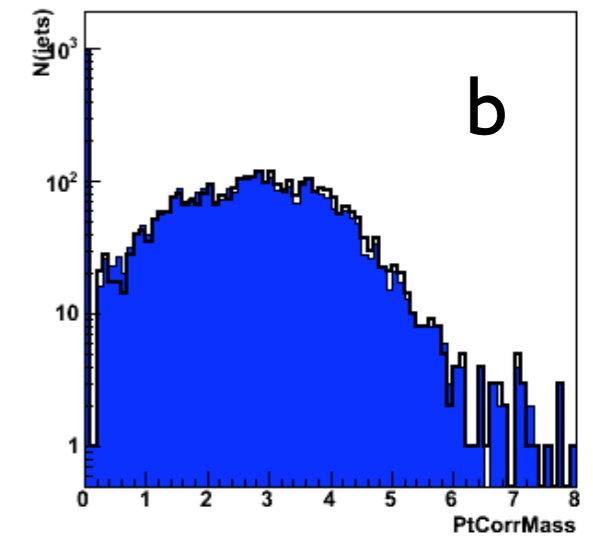
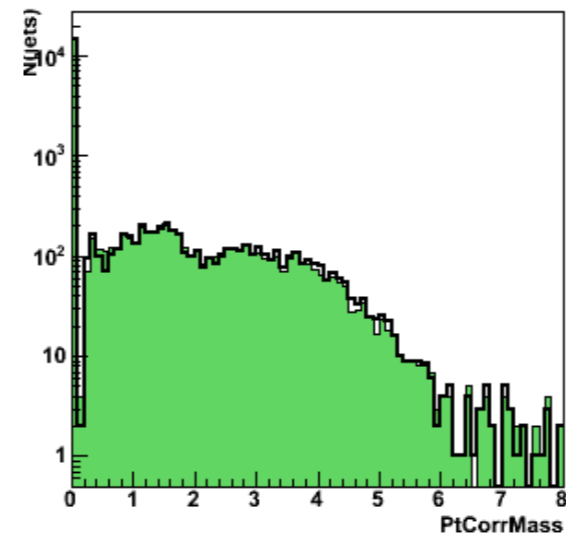
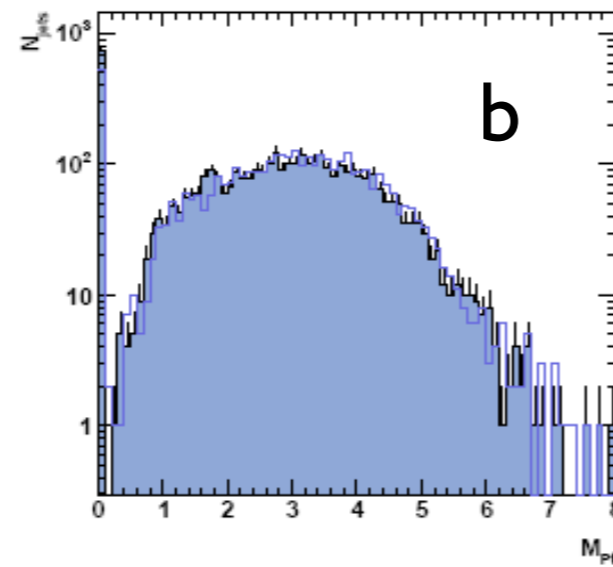
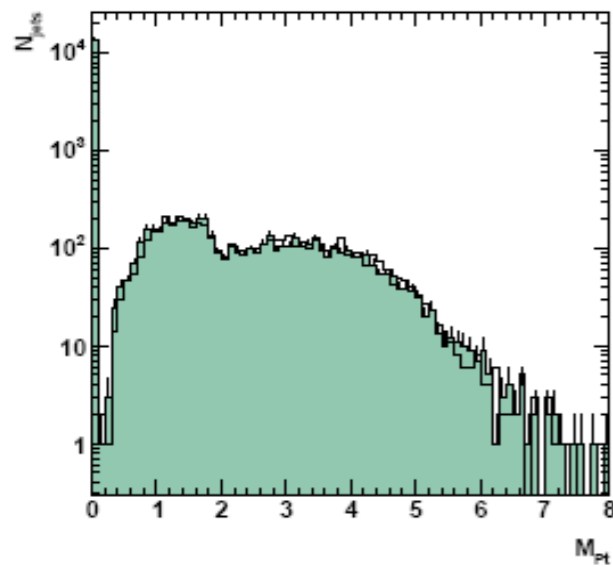


Notice the log scale!

Pt corrected mass

SGV - line; Mokka - histo
(S. Hillert)

LDCPrime_02Sc
cheat - line; conv. tag - histo



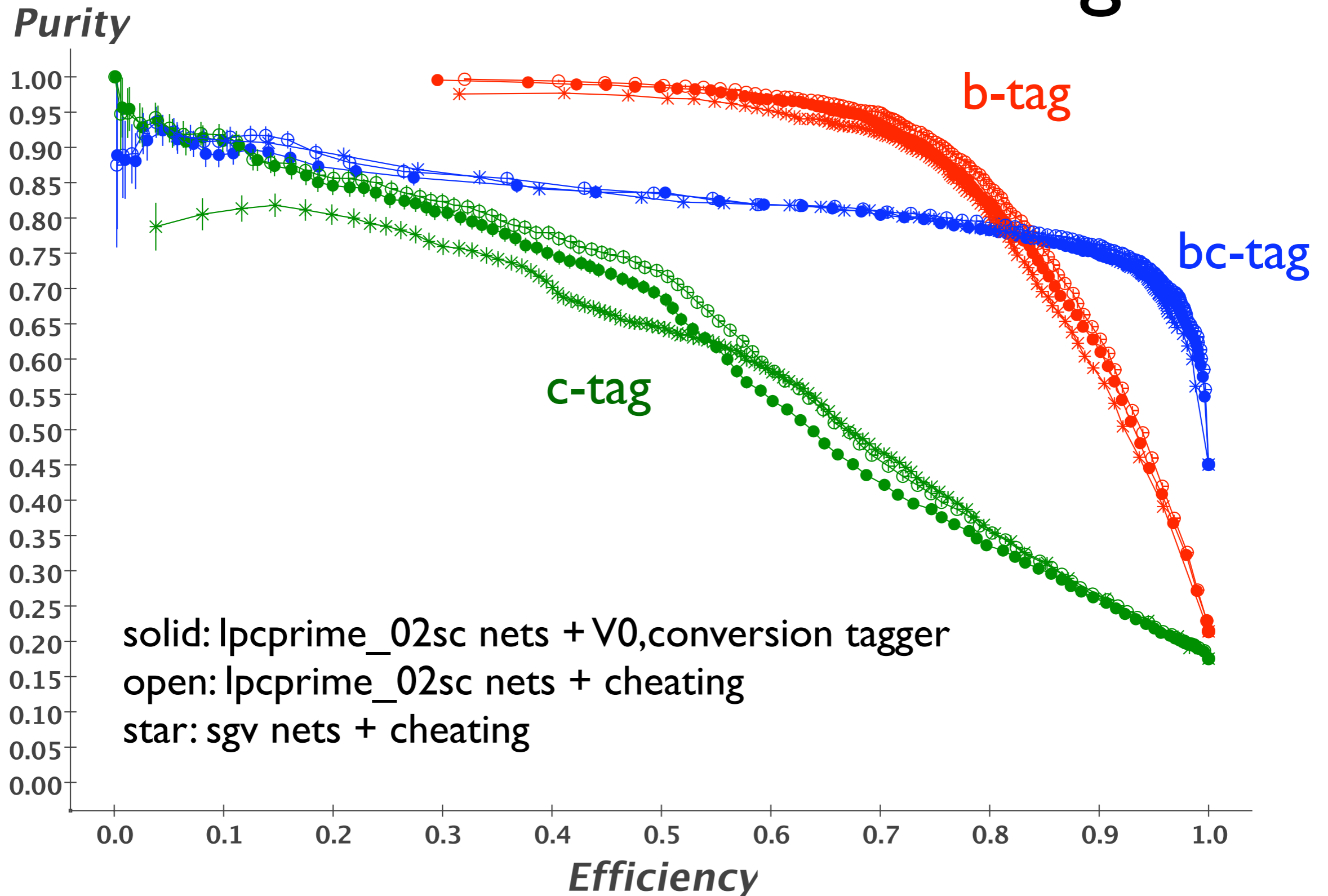
Notice the log scale!

Neural networks

Neural networks training

- Flavour tagging uses 9 neural networks for the b-, c- and bc-tag each for the cases where 1 vertex, 2 vertices or more than 3 vertices are found.
- Present LCFIVertex networks were trained using samples simulated with SGV.
- Samples (LDCPrime_02Sc, ilcsoft 01-03-06-p02):
 - 49000 events $Z \rightarrow bb$
 - 49000 events $Z \rightarrow cc$
 - 49000 events $Z \rightarrow uds$
- NeuralNetTrainer processor used.
- Durham_2Jets collection as input.

Neural networks training



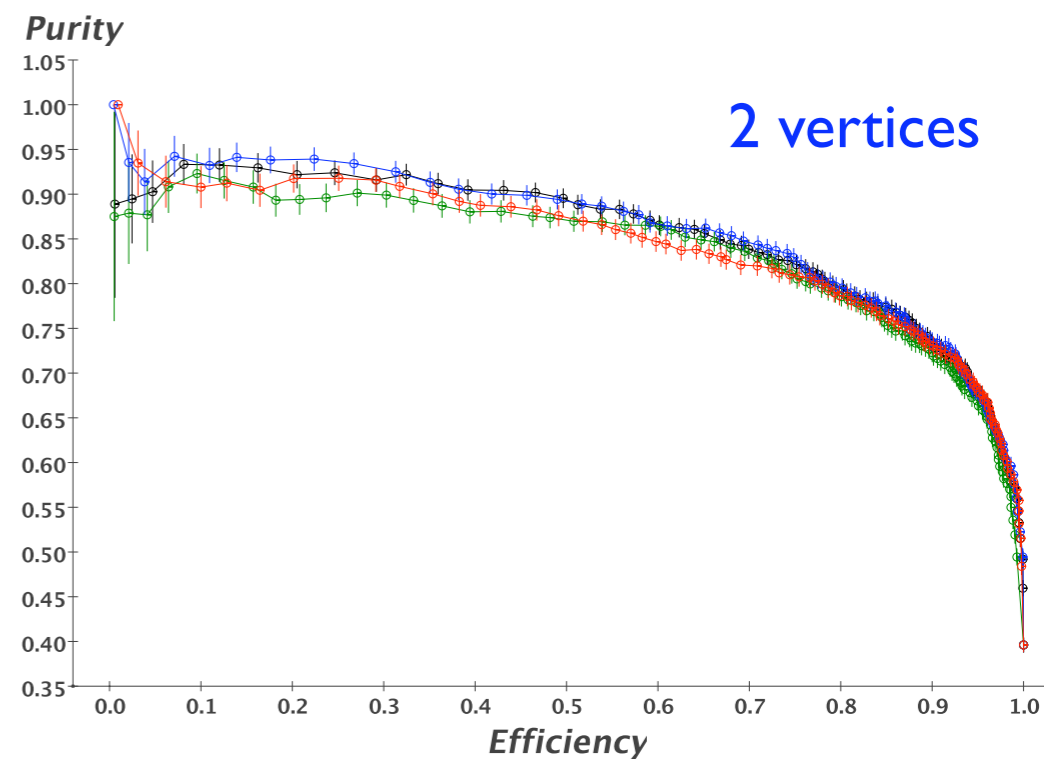
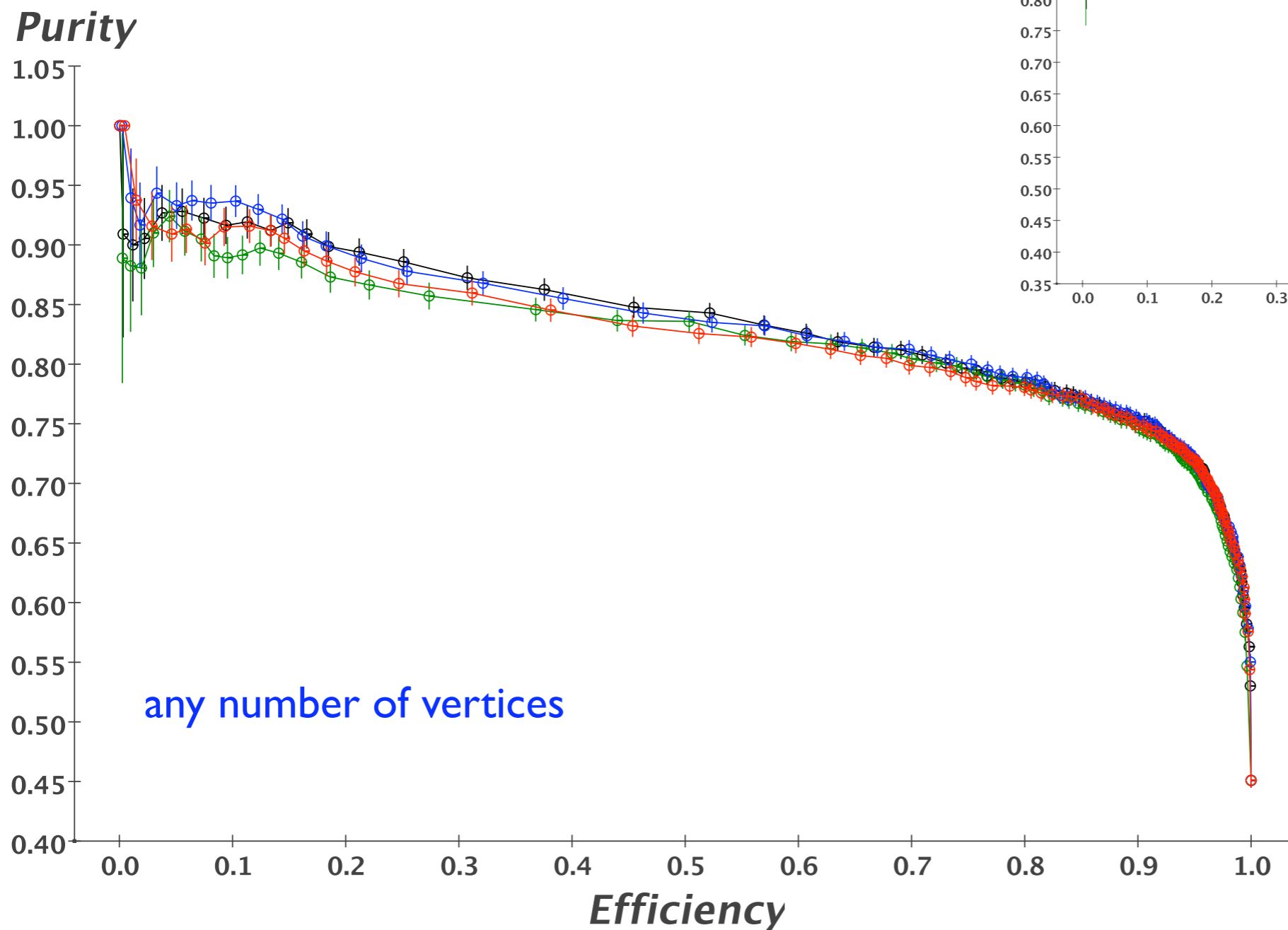
- lpcprime_02sc neural nets were trained with conversion tagger!

Neural networks training

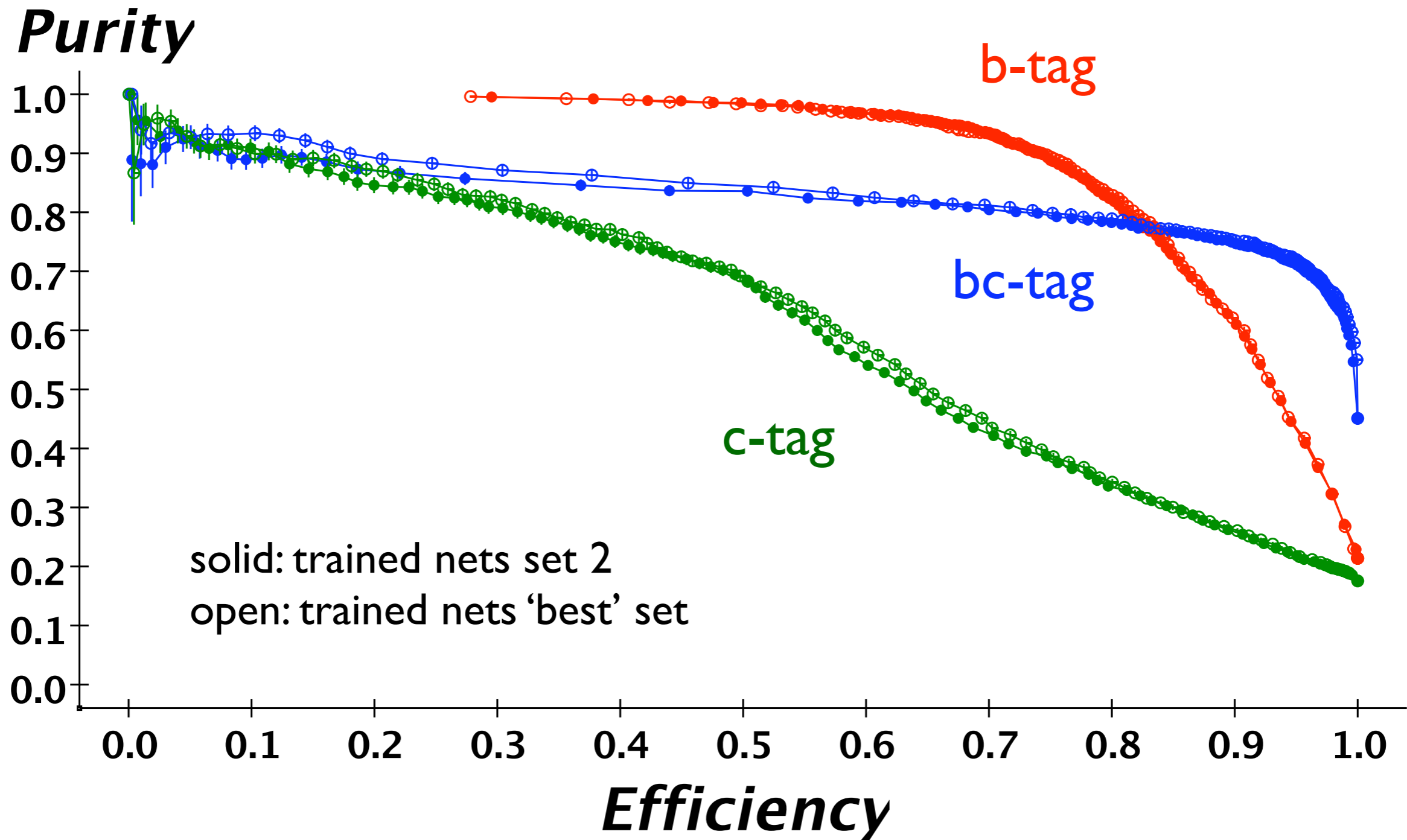
- The random nature of the networks training produces fluctuations (up to ~ 2 sigma) in the efficiency x purity plots.
- Perform a certain number of training and pick the networks that give the best results.
- Four sets for each network were trained.

Neural networks training

- BC-TAG



Neural networks training



Track selection tuning

Track selection tuning

- Vary the parameters of track selection for the vertices reconstruction (IPFIT, ZVRES) and for the flavour tag inputs (FTI).
- Aim: Improve flavour tagging keeping the performance of vertex reconstruction and flavour tag inputs.

Track selection tuning

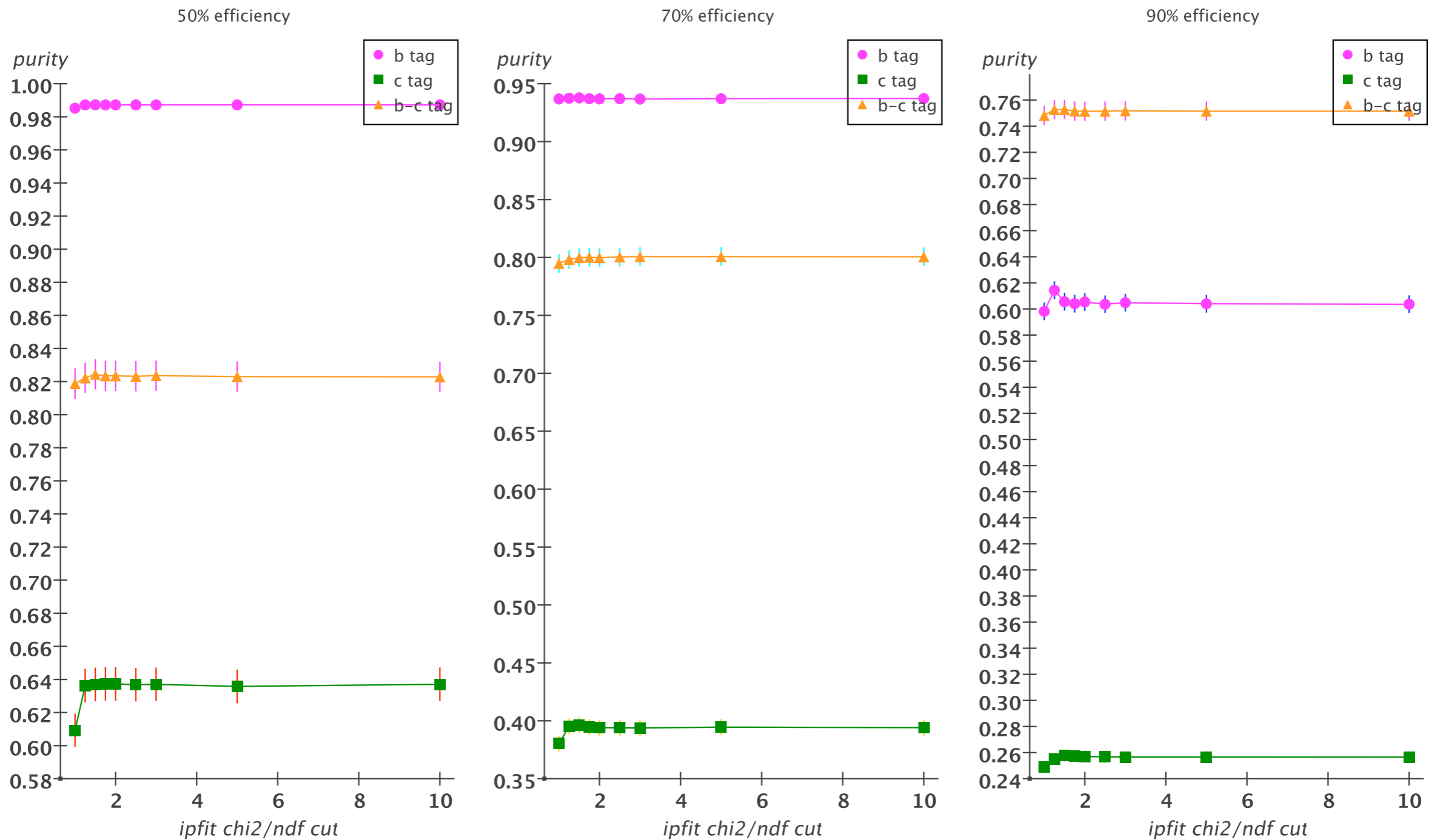
- **Monte Carlo sample:**
 - $e^+e^- \rightarrow Z \rightarrow qq$, $\sqrt{s} = 91.2$ GeV;
 - 10000 events;
 - Detector model: LDCPrime_02Sc.
- **Reconstruction:**
 - ilcsoft v01-03-06-p02; LCFIVertex HEAD;
 - tracking: FullLDCTracking;
 - clustering + particle flow: PandoraPFA;
 - jets: Satoru jet finder, durhamnjet, njet = 2;
 - vertexing and flavour tag inputs: ConversionTagger

Track selection parameters

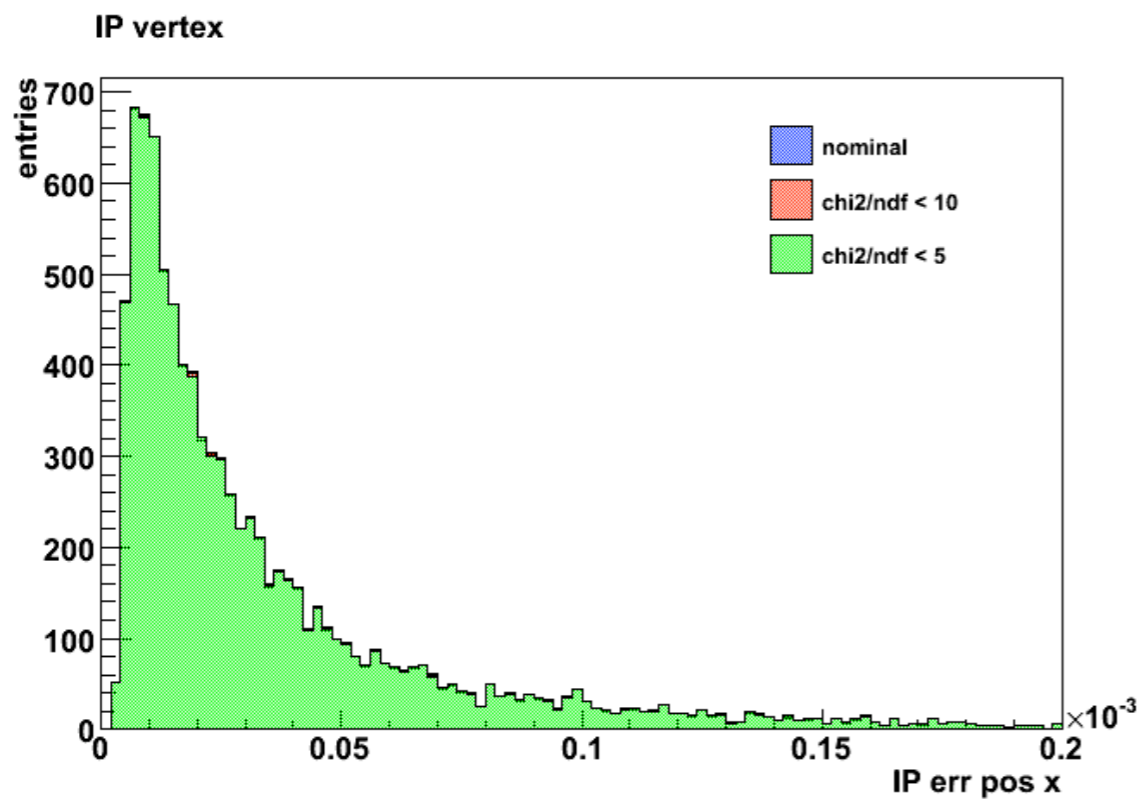
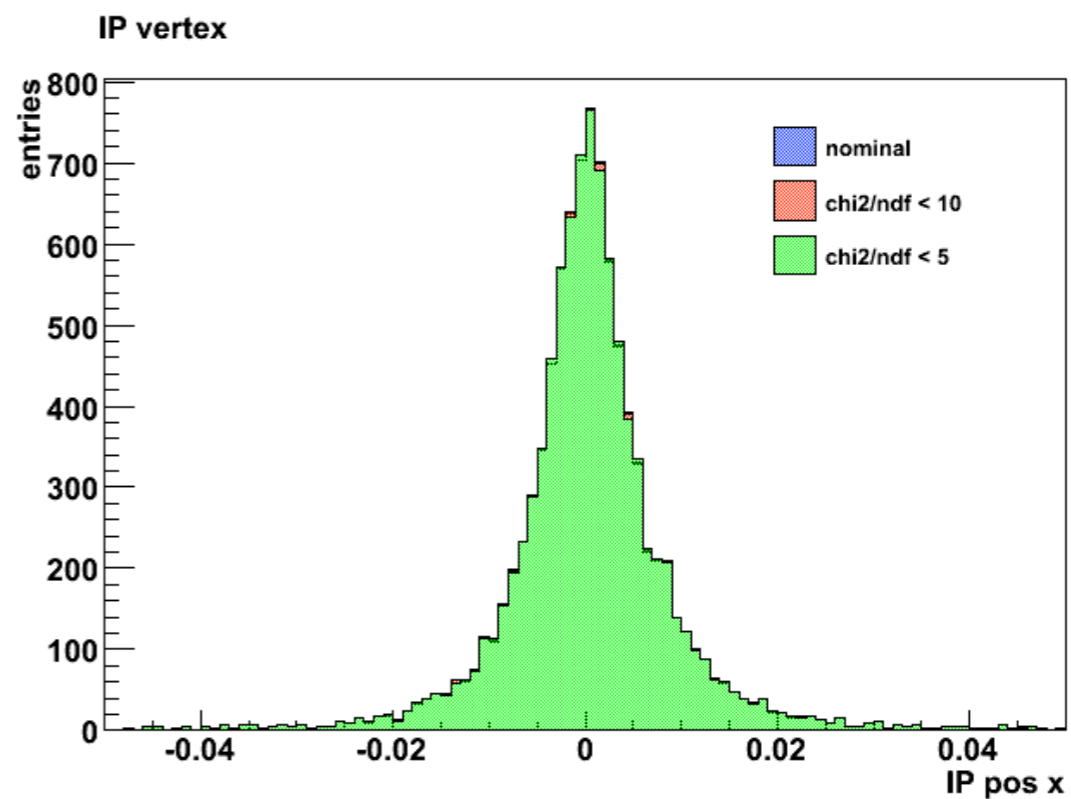
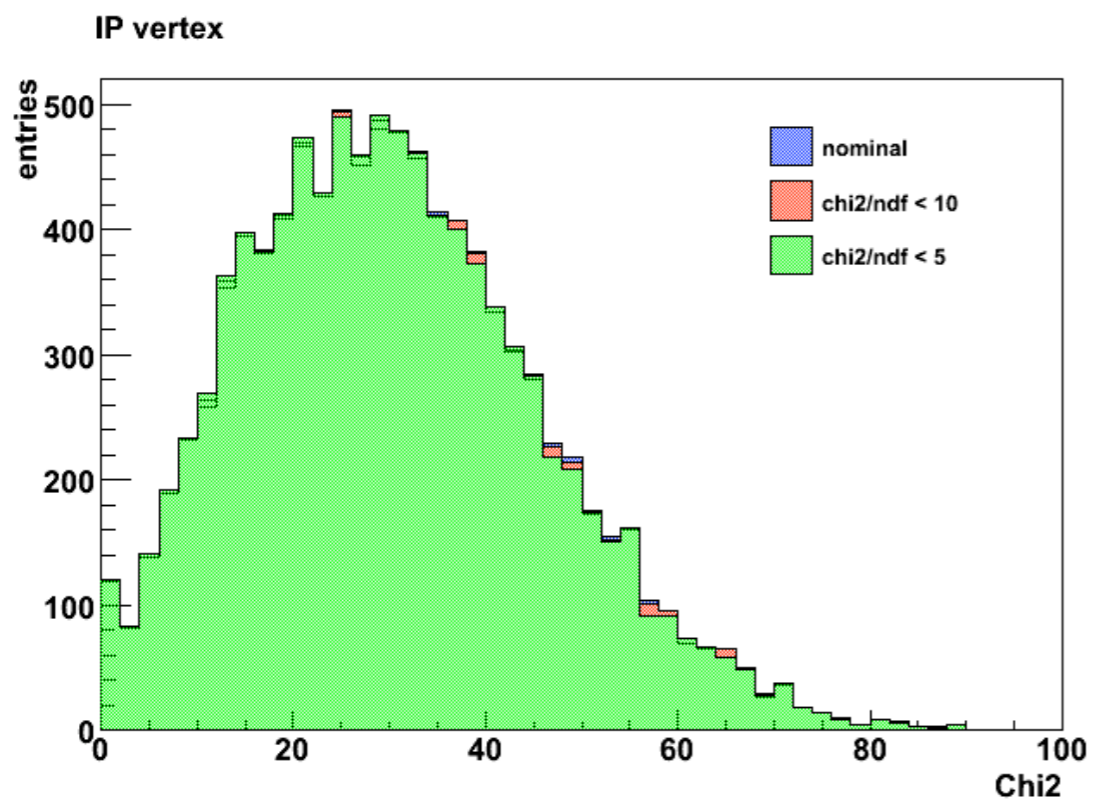
Description	xml parameter names	Code default	ipfit.xml	zvres.xml	fti.xml
Cut on χ^2/ndf of track fit	a1_Chi2OverDOFEnable a2_Chi2OverDOFCutLowerThan a3_Chi2OverDOFCutValue	10	10 X	10 X	10 X
Cut on d0 (R ϕ impact parameter)	b1_D0Enable b2_D0CutLowerThan b3_D0CutValue	20	50 ✓ (mm)	10 ✓ (mm)	20 ✓ (mm)
Cut on d0 error	c1_D0ErrEnable c2_D0ErrCutLowerThan c3_D0ErrCutValue	0.25	0.025 X (mm)	0.25 ✓ (mm)	0.025 X (mm)
Cut on z impact parameter	d1_Z0Enable d2_Z0CutLowerThan d3_Z0CutValue	20	50 ✓ (mm)	20 ✓ (mm)	20 ✓ (mm)
Cut on error on z imp param	e1_Z0ErrEnable e2_Z0ErrCutLowerThan e3_Z0ErrCutValue	0.25	0.025 X (mm)	0.025 X (mm)	0.025 X (mm)
Cut on pT of track	f1_PTEnable f2_PTCutLowerThan f3_PTCutValue	0.1	0.1 X (GeV/c)	0.1 ✓ (GeV/c)	0.1 ✓ (GeV/c)
cut on Ks, Λ decay tracks	h1_MCPIDEnable h2_CutPIDS h3_MonteCarloLCRelationCollection	0	X	✓ +- 310 +- 3122	✓ +- 310 +- 3122

X: disabled; ✓: enabled

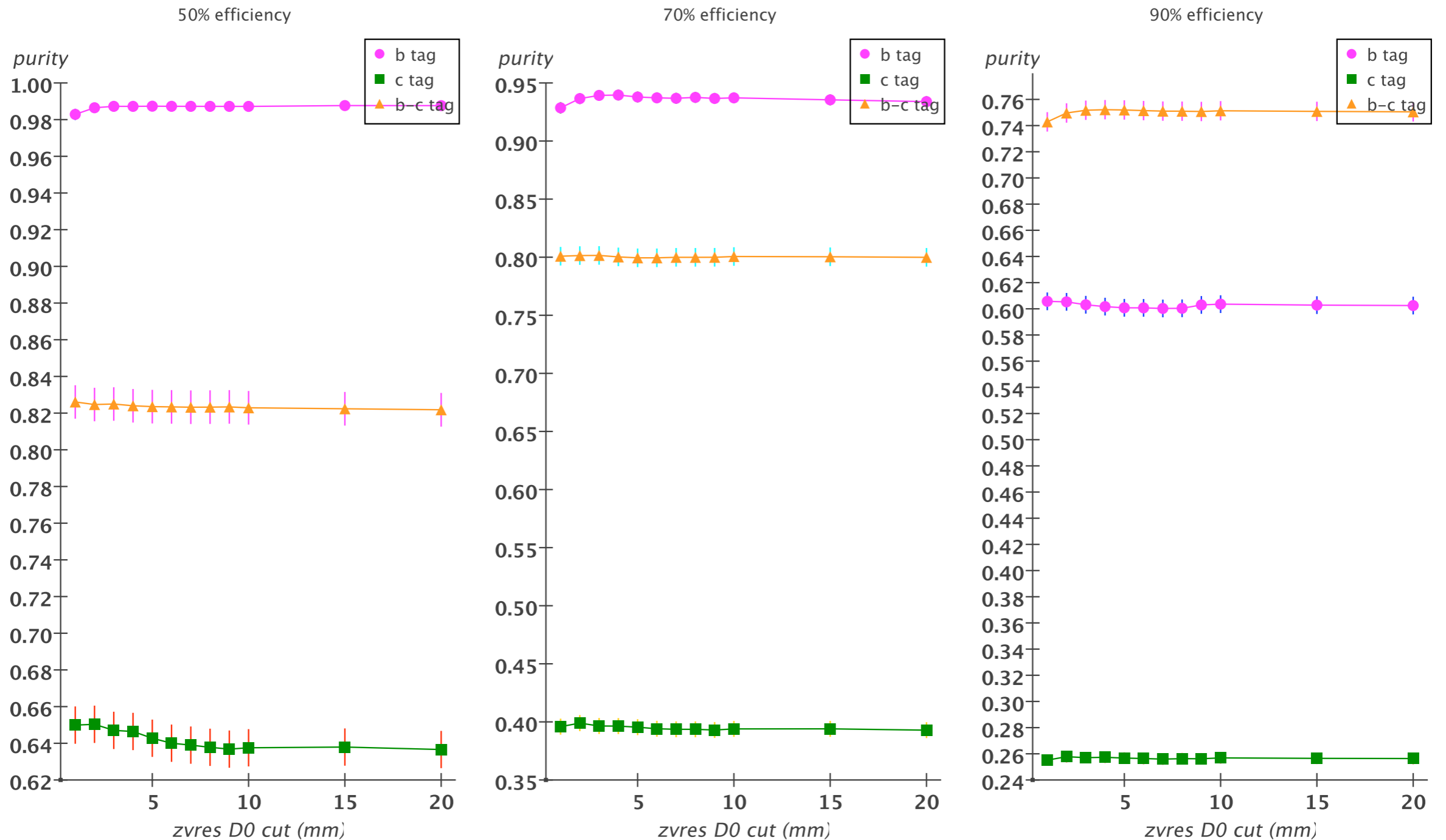
IPFIT: flavour tag purity as a function of the χ^2/ndf of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: no cut



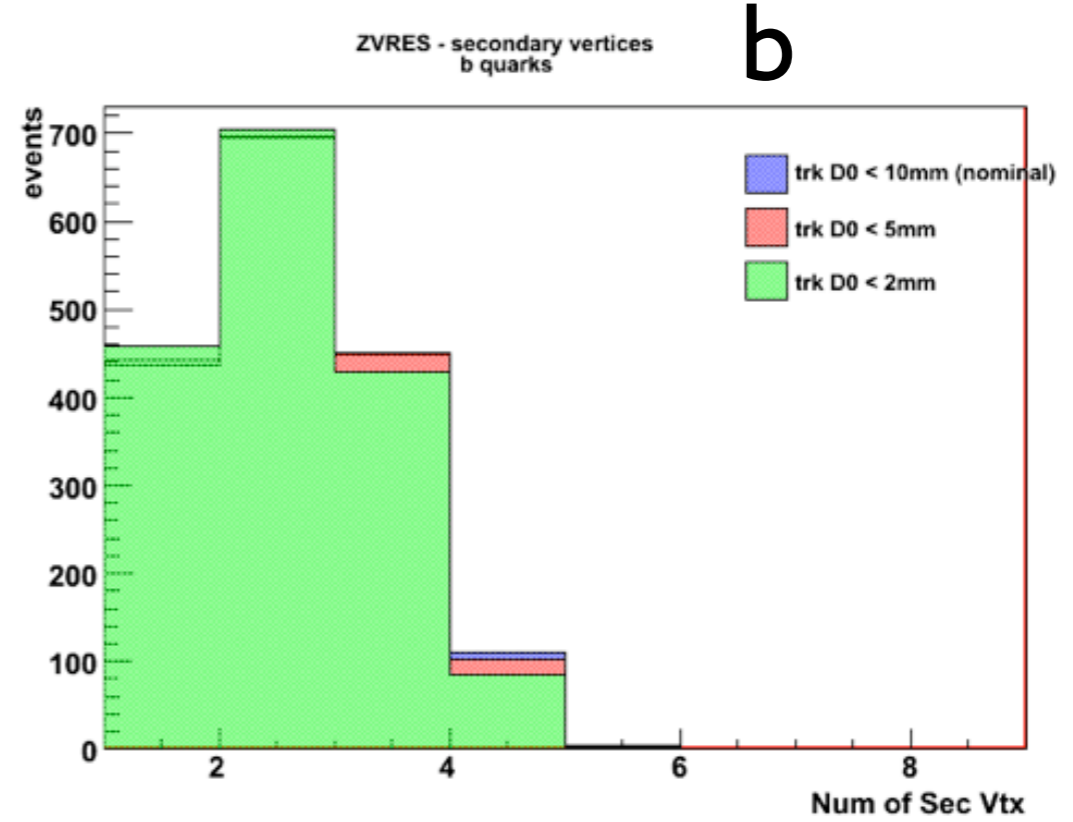
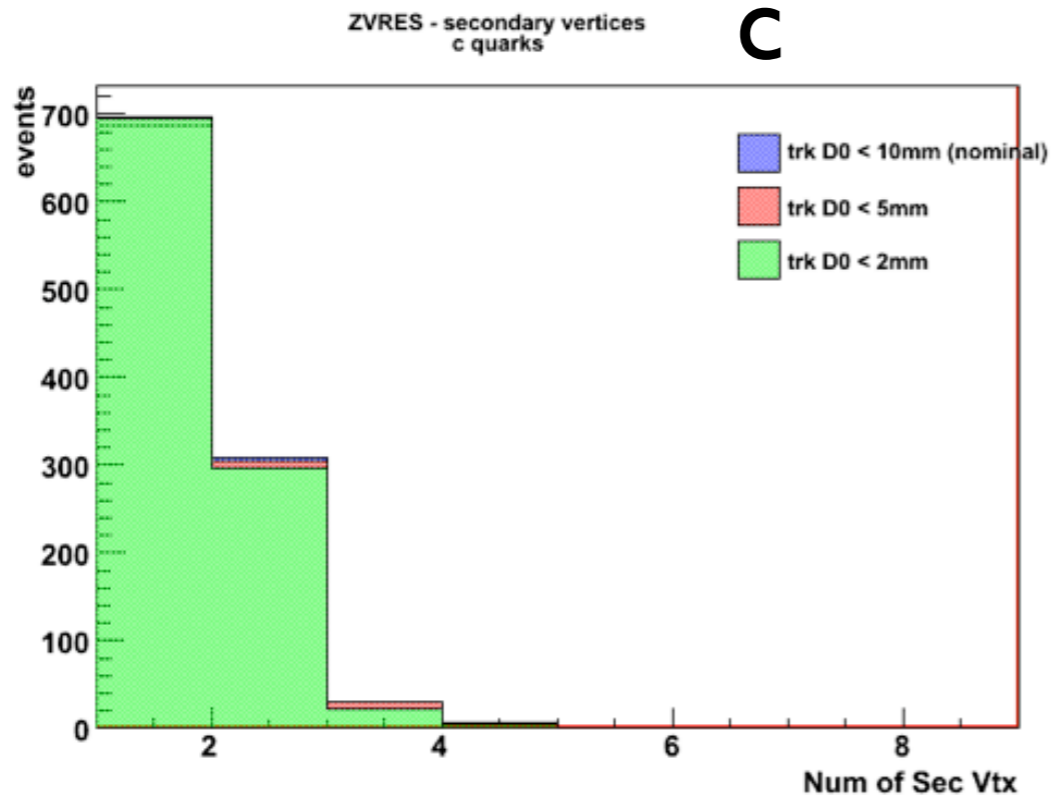
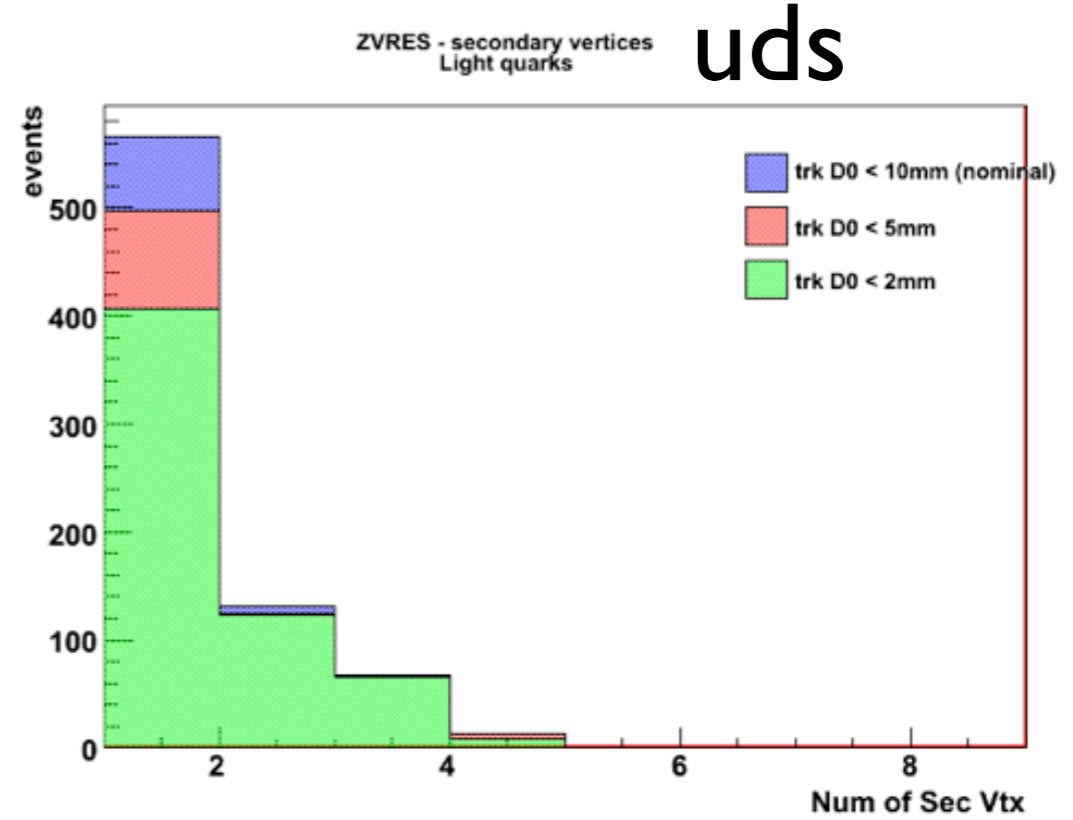
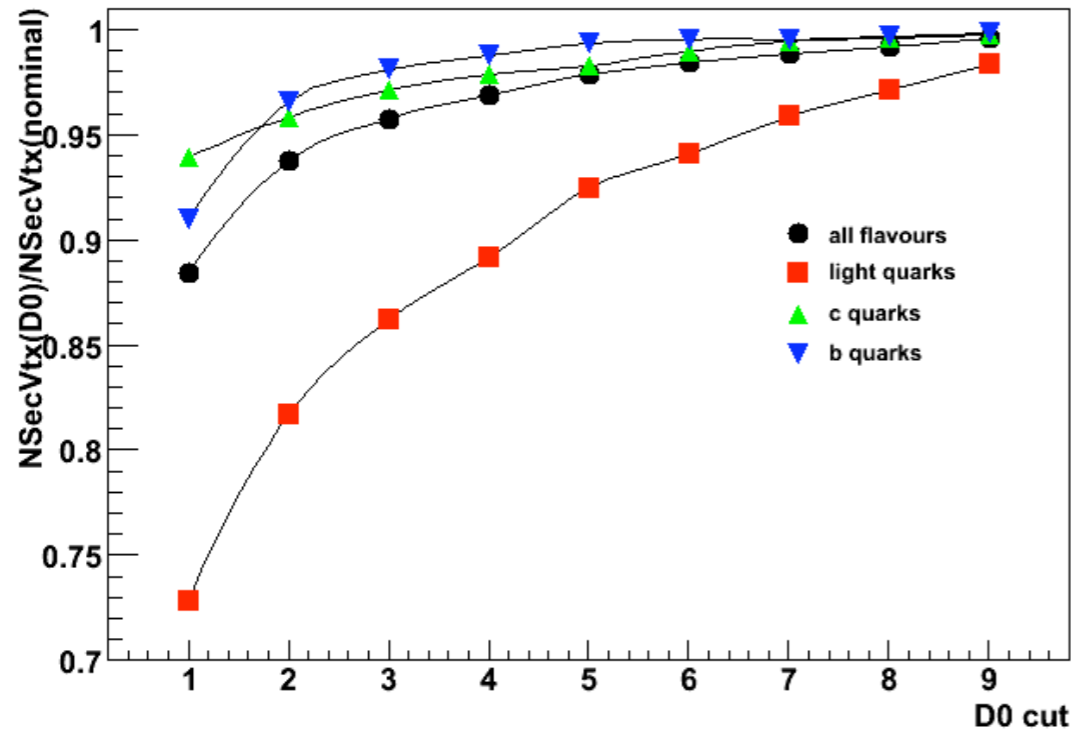
IP vertex (chi2/ndf cut)



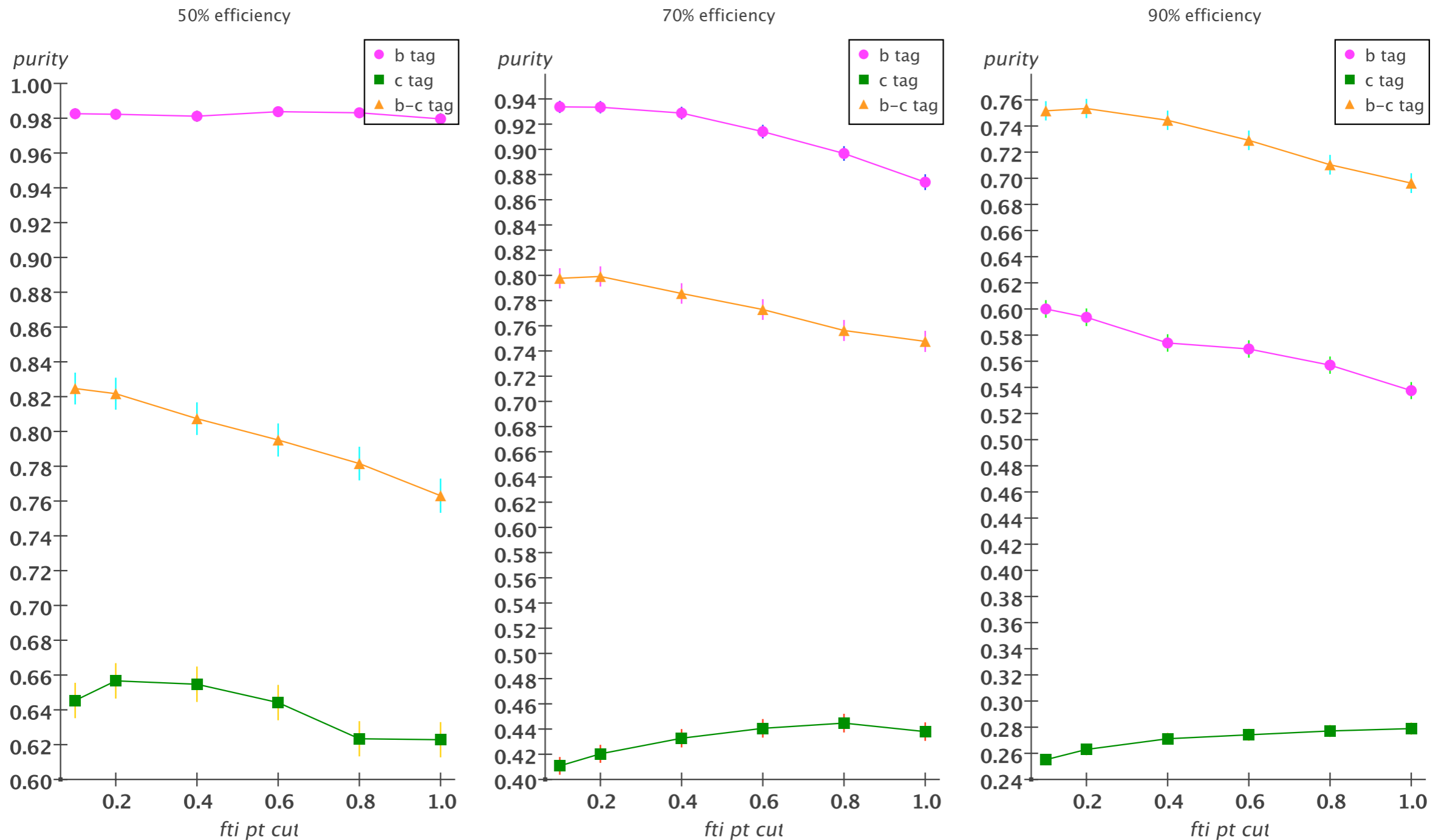
ZVRES: flavour tag purity as a function of the d_0 of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: 10mm



ZVRES secondary vertices (d0 cut)



FTI: flavour tag purity as a function of the pt of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: 0.1 GeV



Track selection tuning

One also has to be careful with correlations...

Considering...

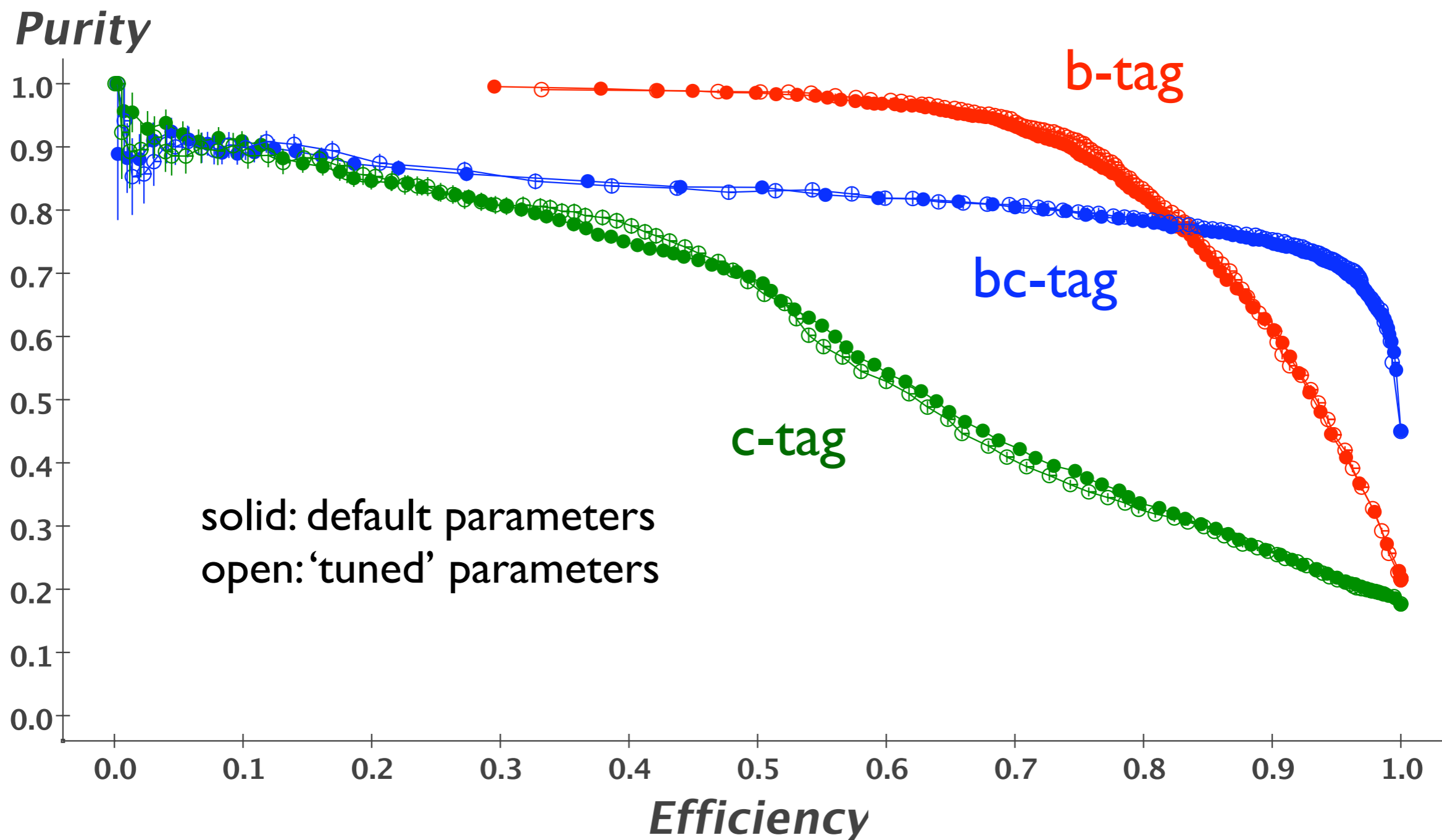
IPFIT track selection:

- $\chi^2/\text{ndf} < 5$;
- $d_0 < 20\text{mm}$;
- d_0 error: no cut;
- $z_0 < 20\text{mm}$;
- z_0 error: nocut;
- $p_T > 0.1 \text{ GeV}$;

ZVRES track selection:

- $\chi^2/\text{ndf} < 4$;
- $d_0 < 2\text{mm}$;
- d_0 error $< 0.007\text{mm}$;
- $z_0 < 5\text{mm}$;
- z_0 error $< 0.025\text{mm}$;
- $p_T > 0.2 \text{ GeV}$;

Track selection tuning



Note: Not using the 'best' set of neural nets.

Summary & Conclusions

- Tuning of flavour tag is performed for detector model LDCPrime_02Sc:
 - Realistic V0 and photon conversion tagger is used;
 - Parameters for the joint probabilities were recalculated;
 - Neural networks for flavour tagging were trained;
 - Studies on track selection for vertex reconstruction on flavour tag inputs are performed.

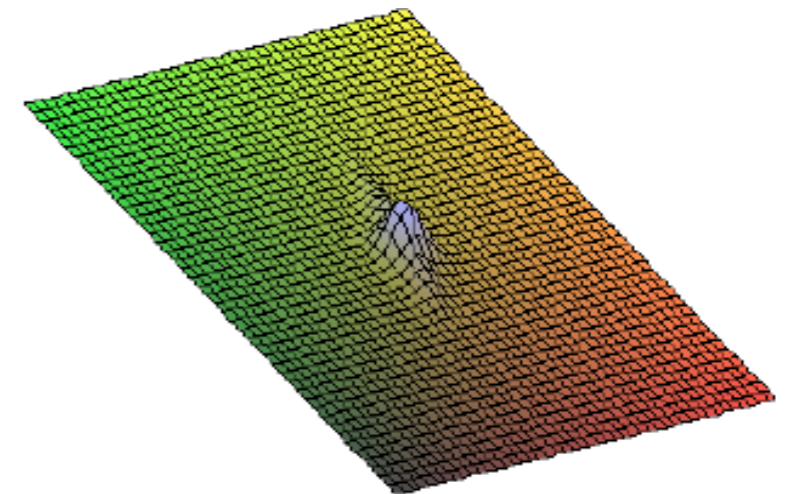
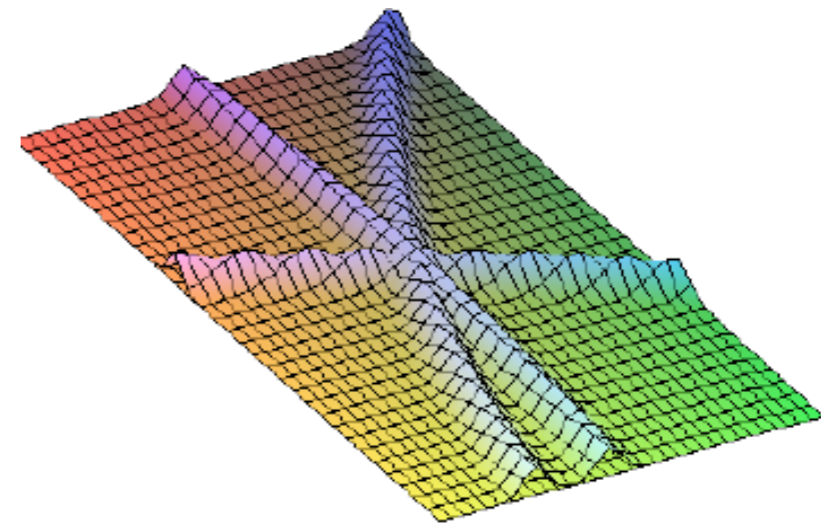
Summary & Conclusions

- Realistic V0 and photon conversion tagger and re-trained neural networks give good performance for b- and bc-tag and c-tag at low efficiencies, but c-tag for high efficiencies is lightly degraded.
- Track selection tuning still ongoing. Need to be careful with correlations. It may not improve the efficiency x purity performance, but tighter cuts should be needed when one considers beam backgrounds.
- More realistic status of the detector description and of the reconstruction software does not imply degradation in flavour tagging performances.

Extra slides

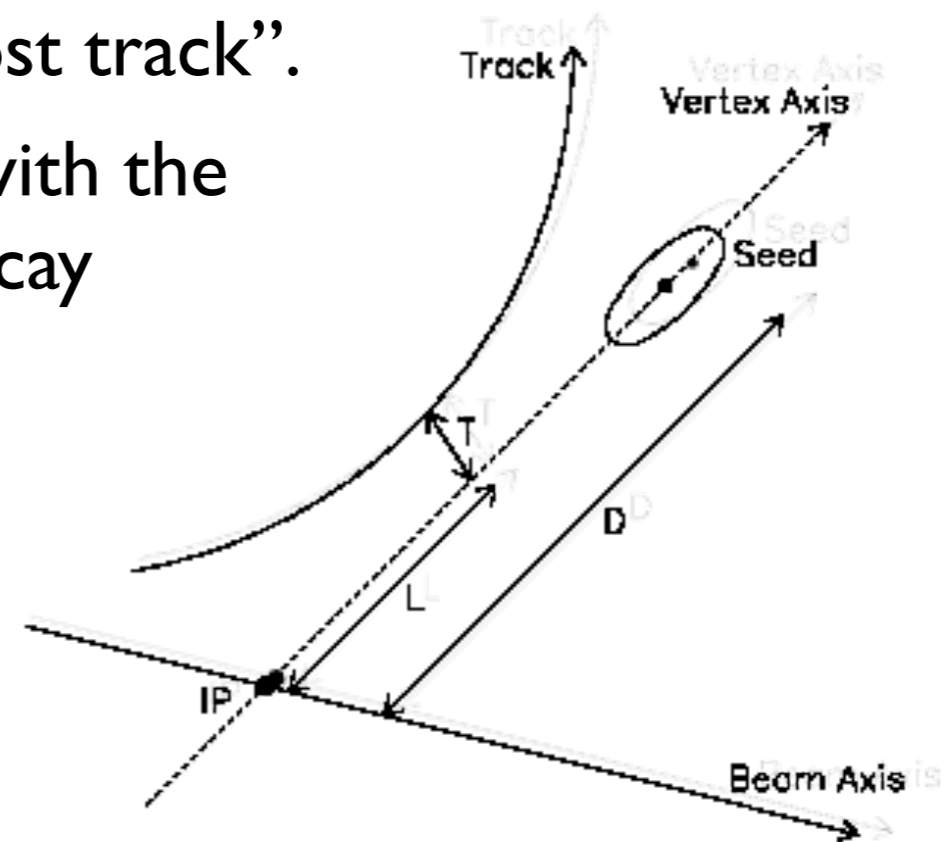
ZVTOP vertex finder: ZVRES

- ZVTOP (*D. Jackson, NIM A388, 247*) consists of two branches:
 - **ZVRES**: General algorithm that can handle arbitrary multi-prong topologies.
 - Each track is represented as a gaussian “probability tube” in 3D where the width of the tube is the error of the track.
 - The track “tubes” are combined into a “vertex probability function”. A maximum of this function is searched in 3D-space and the χ^2 of the vertex fit is minimised iteratively.



ZVTOP vertex finder: ZVKIN

- **ZVKIN** (ghost track): Specialised algorithm to reconstruct vertices in b-jets with two subsequent one-prong decays. e.g. in a $IP \rightarrow B \rightarrow D$ decay chain.
 - Tracks not associated to any vertex but with small values ($\sim 50\mu\text{m}$) of the 3D transverse impact parameter T are likely to come from a B decay chain.
 - The best estimate of the direction of flight of the B-hadron is found and a finite width is assigned \rightarrow “ghost track”.
 - The tracks in the jet are combined with the ghost track the IP to build up the decay chain along the ghost direction.
 - Should improve flavour tagging efficiency, as well as the vertex charge determination.

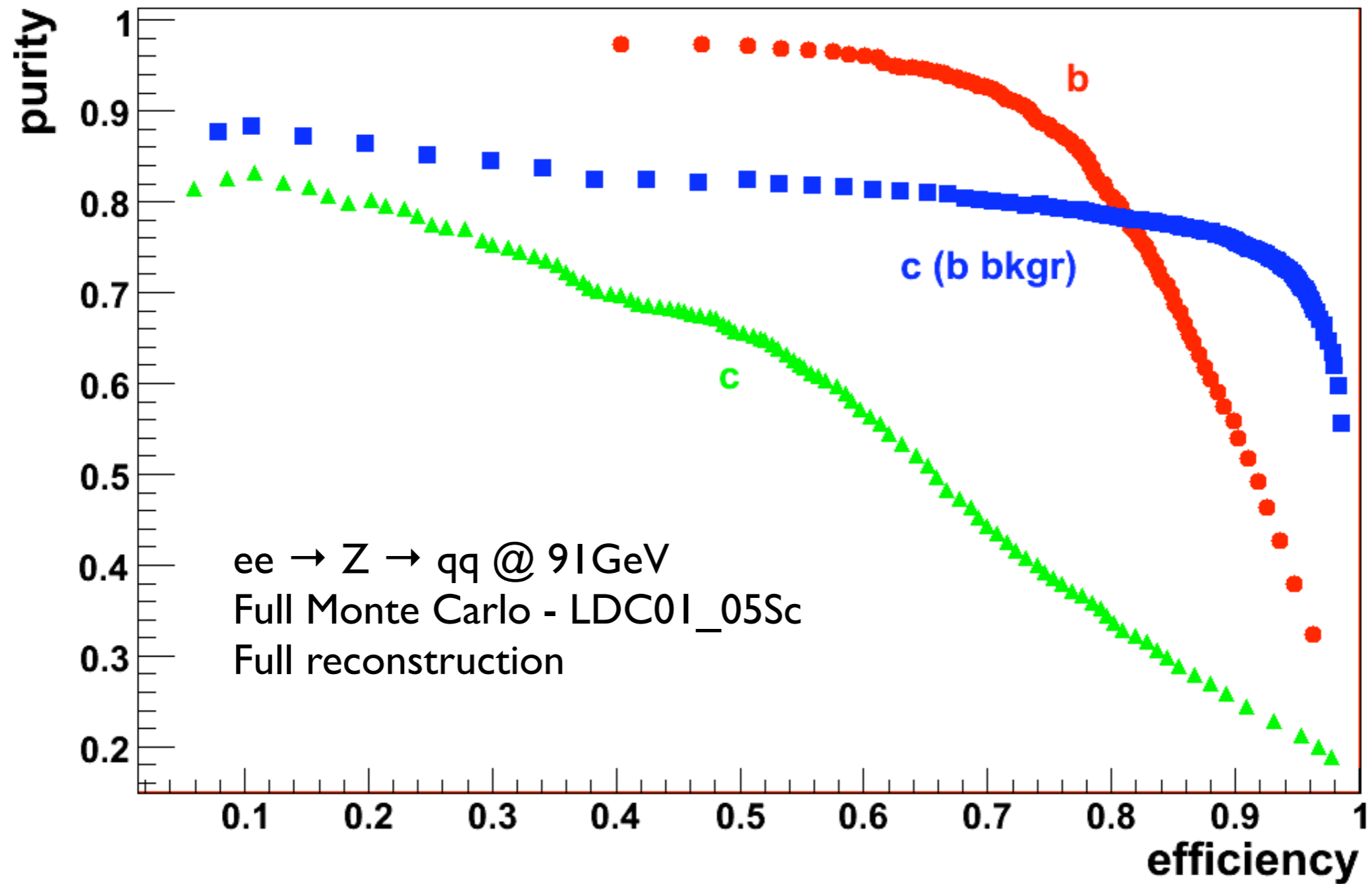


Flavour tagging

- The LCFIVertex package also incorporates a flavour tagging code (*R. Hawkings, LC-PHSM-2000-021*) using a neural network approach to discriminate between b-, c- and uds- or gluon-jets.
- Neural network input variables:
 - secondary vertex is found: p_T corrected mass, momentum, decay length and decay-length significance;
 - only IP is found: momentum and impact parameter significance in $R-\Phi$ of the two most significant tracks in the jet;
 - in all cases: joint probability in $R-\Phi$ and z (combined probability that each track of a jet comes from the primary vertex) .

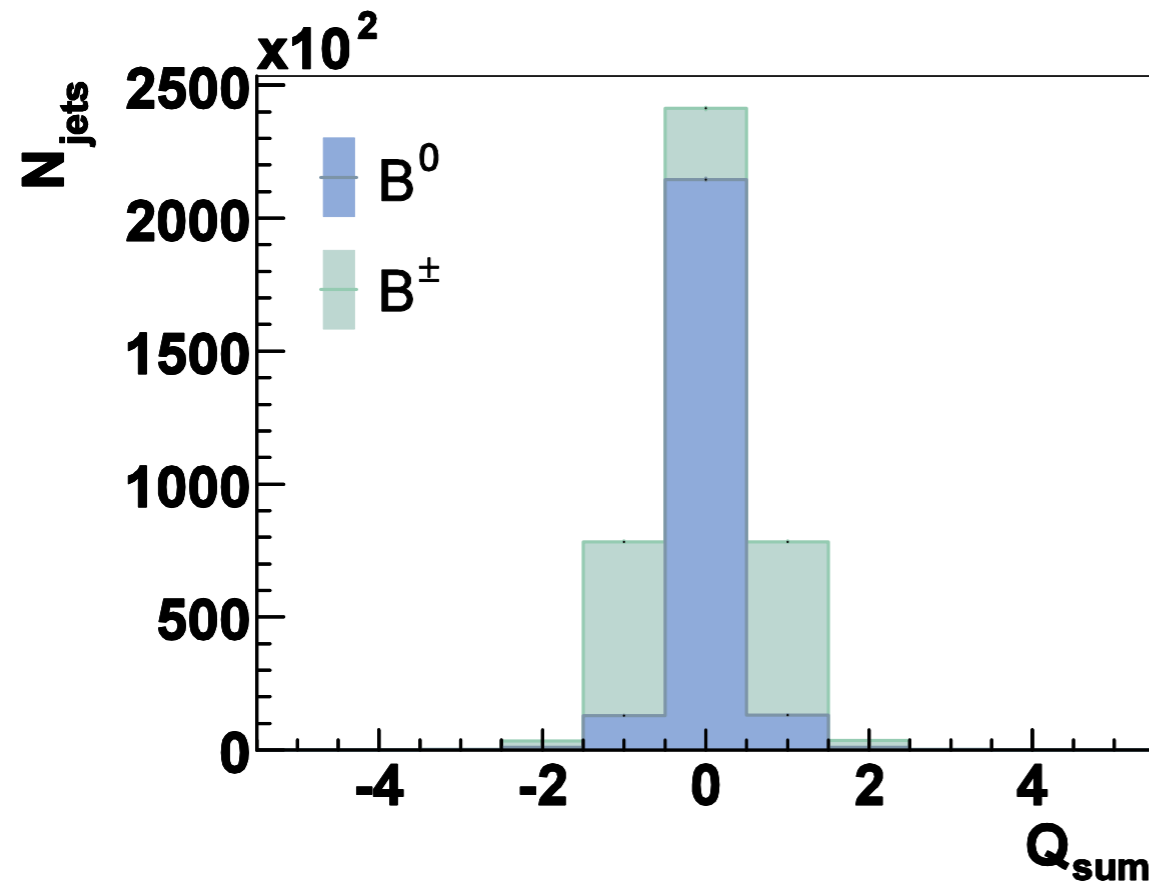
Flavour tagging

Efficiency-purity for FlavourTag



Vertex charge

- The vertex charge limited to charged B-hadrons (40% of the b-jets); to reconstruct the vertex charge it is necessary to find all stable tracks from the B-decay chain.
- Probability of mis-reconstruction of the vertex charge is small for both charged and neutral hadrons.
- To extend for neutral B-hadrons, a charge dipole technique is required (under development).



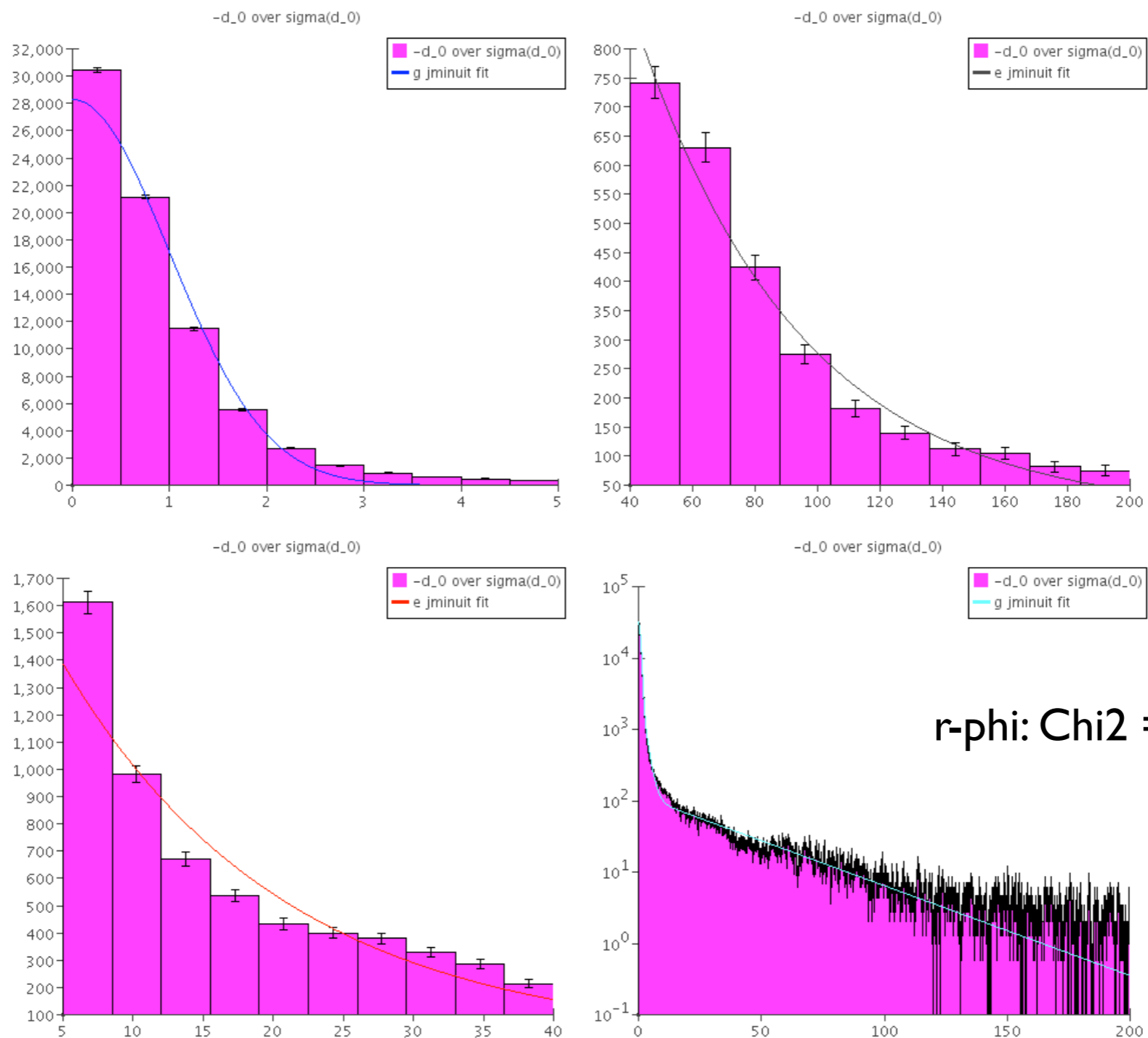
(SGV fast monte carlo sample
 $ee \rightarrow Z \rightarrow qq @ 91\text{GeV}$)

Parameters for Joint Probability

SignificanceFit (Durham_2Jets)

Z->qq, 10000 events
LDCPrime_02Sc

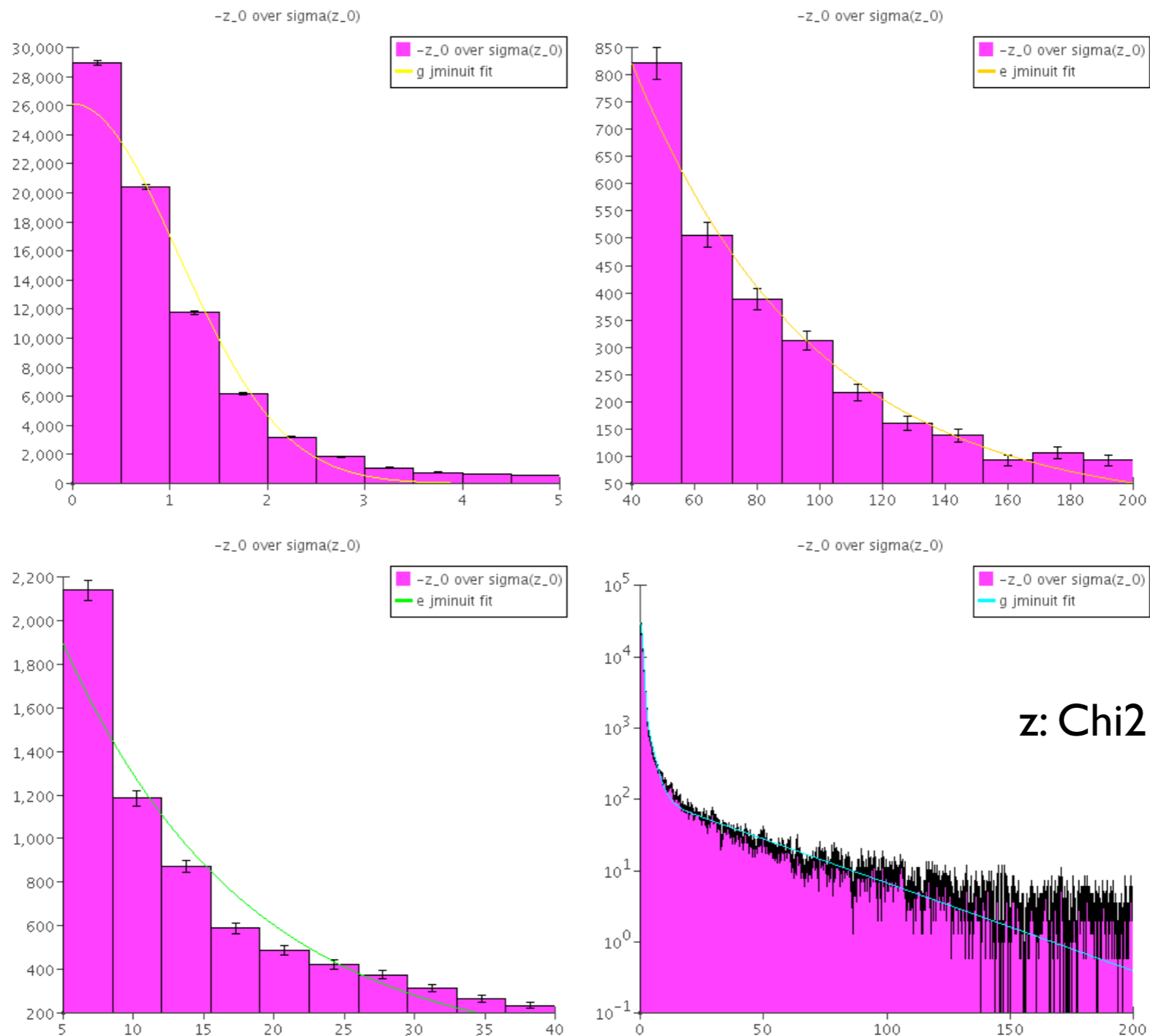
ILC SOFT v01-03-06-p02, LCFIVertex HEAD,
V0+conversion tagger



SignificanceFit (Durham_2Jets)

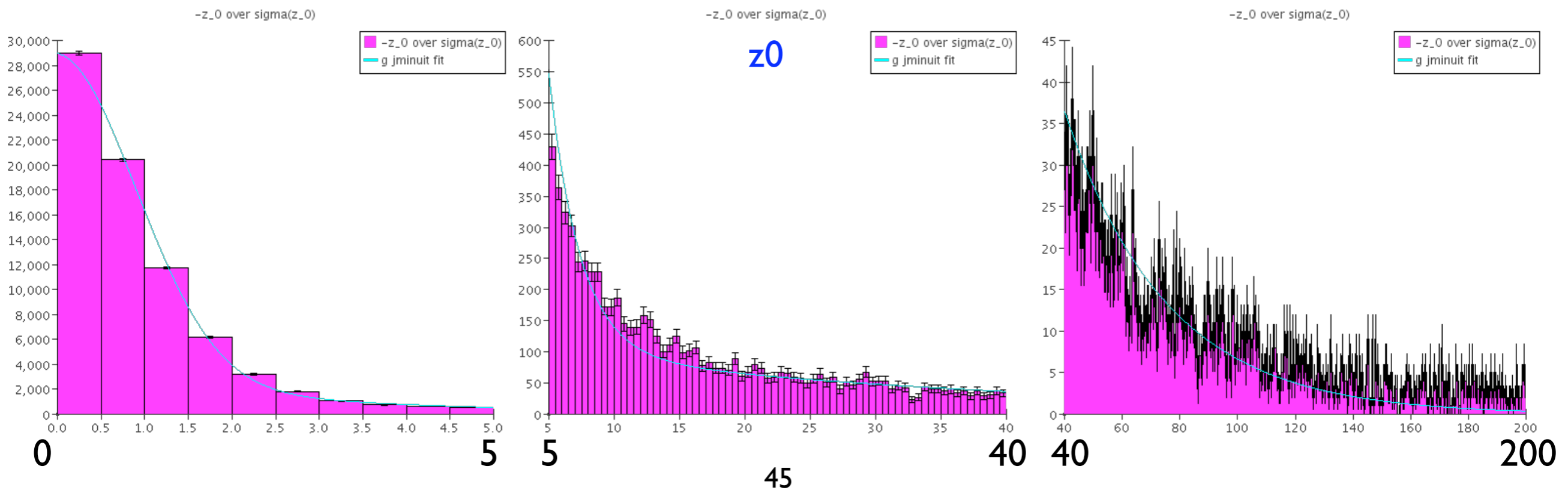
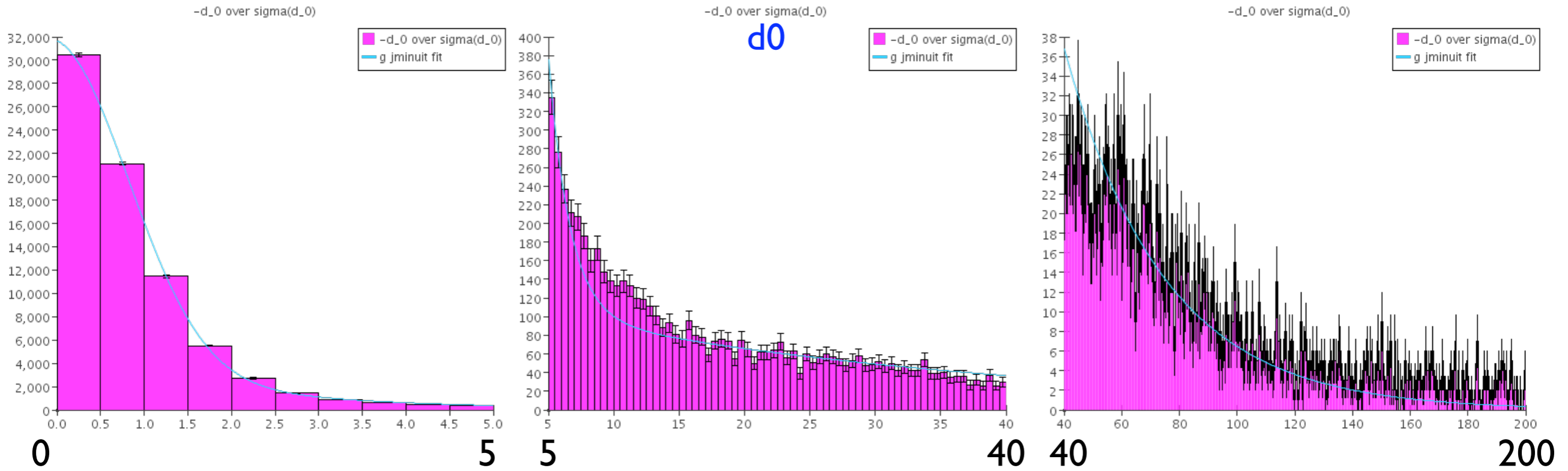
Z->qq, 10000 events
LDCPrime_02Sc

ILC SOFT v01-03-06-p02, LCFIVertex HEAD,
V0+conversion tagger



SignificanceFit (Durham_2Jets)

Zooming the global fit



SignificanceFit (Durham_2Jets)

Fit results

```
RPHI - Plane

Local Fit of Gaussian
Parameter  amplitude      28282.4
Parameter  mean          -9.27832e-11
Parameter  sigma          0.992607
Chi^2/ndf:  460.266
Local Fit of first Exponential
Parameter  amplitude      1893.54
Parameter  exponent      -0.0624859
Chi^2/ndf:  23.2849
Local Fit of second Exponential
Parameter  amplitude      1876.12
Parameter  exponent      -0.0191247
Chi^2/ndf:  5.02206

Global Fit
Parameter  amplitude      25456.5
Parameter  mean          -1.41158e-11
Parameter  sigma          0.843068
Parameter  amplitude_1    6082.21
Parameter  exponent      -0.619891
Parameter  amplitude_2    117.321
Parameter  exponent_2     -0.0290308
Chi^2/ndf:  1.97361
```

```
Z - axis

Local Fit of Gaussian
Parameter  amplitude      26119.1
Parameter  mean          -8.87055e-11
Parameter  sigma          1.07717
Chi^2/ndf:  523.081
Local Fit of first Exponential
Parameter  amplitude      2767.23
Parameter  exponent      -0.0762036
Chi^2/ndf:  30.195
Local Fit of second Exponential
Parameter  amplitude      1642.4
Parameter  exponent      -0.0173515
Chi^2/ndf:  5.48903

Global Fit
Parameter  amplitude      25143.4
Parameter  mean          -6.88864e-11
Parameter  sigma          0.910629
Parameter  amplitude_1    3706.92
Parameter  exponent      -0.422501
Parameter  amplitude_2    113.437
Parameter  exponent_2     -0.028365
Chi^2/ndf:  2.49354
```

Parameters for Joint Probability

original



new (LDCPrime_02Sc, Durham_2Jets)

```
PARAMETERS FOR RPHI Joint Probability
1.01313412
0.0246350896
0.102197811
0.0411203019
0.0157710761
PARAMETERS FOR Z Joint Probability
1.01629865
0.0271386635
0.0948112309
0.0410759225
0.0148685882
```

```
PARAMETERS FOR RPHI Joint Probability
0.843068
0.364774
0.619891
0.150243
0.0290308
PARAMETERS FOR Z Joint Probability
0.910629
0.305746
0.422501
0.139363
0.028365
```

Talini's LDC01_05Sc

```
PARAMETERS FOR RPHI Joint Probability
1.01498
0.27984
0.561155
0.00601934
0.0476549
PARAMETERS FOR Z Joint Probability
1.04273
0.27057
0.468457
0.00631426
0.0479484
```

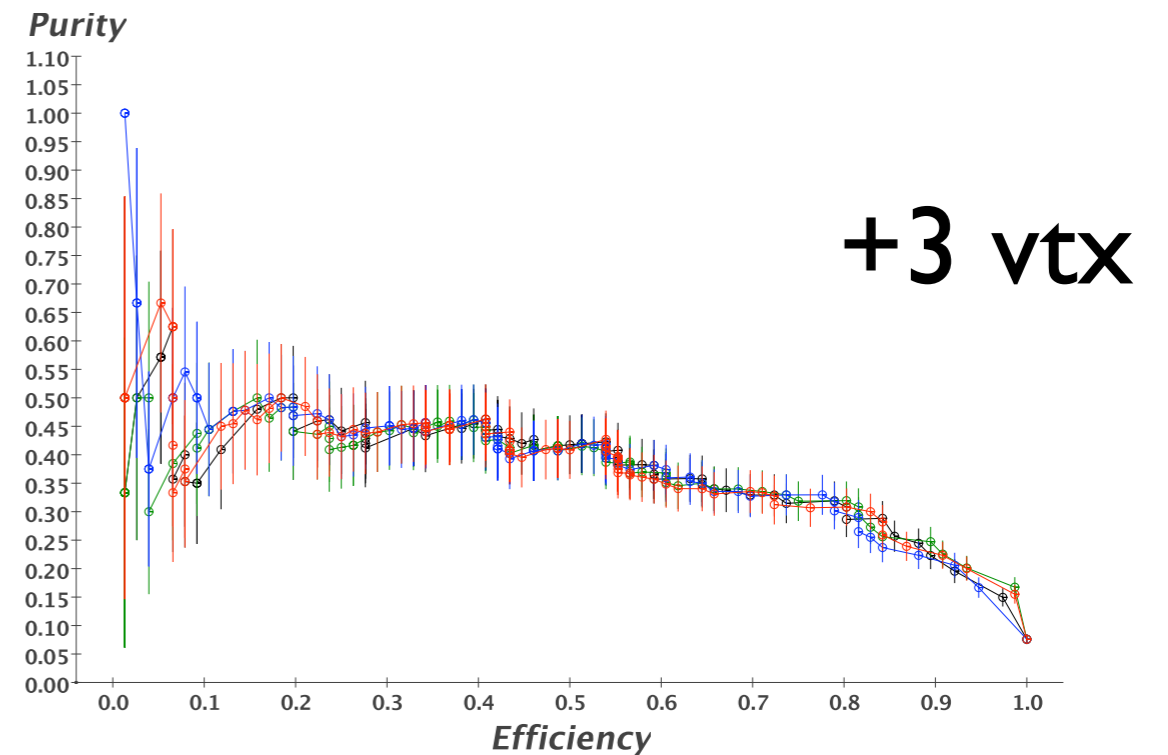
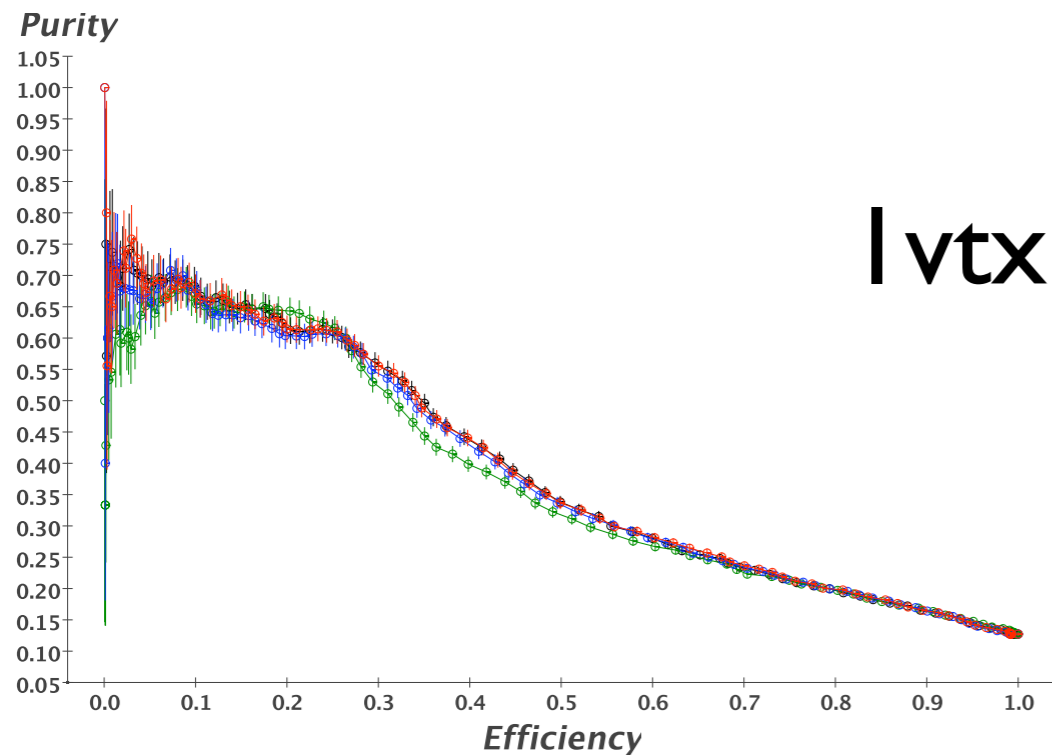
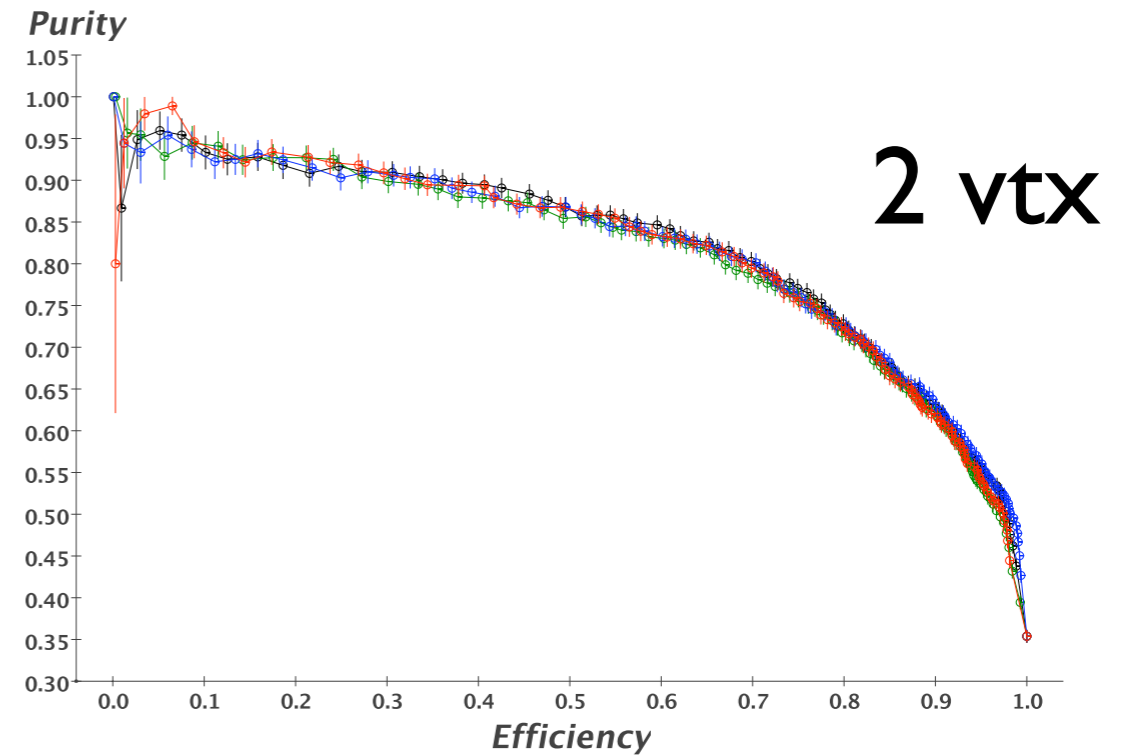
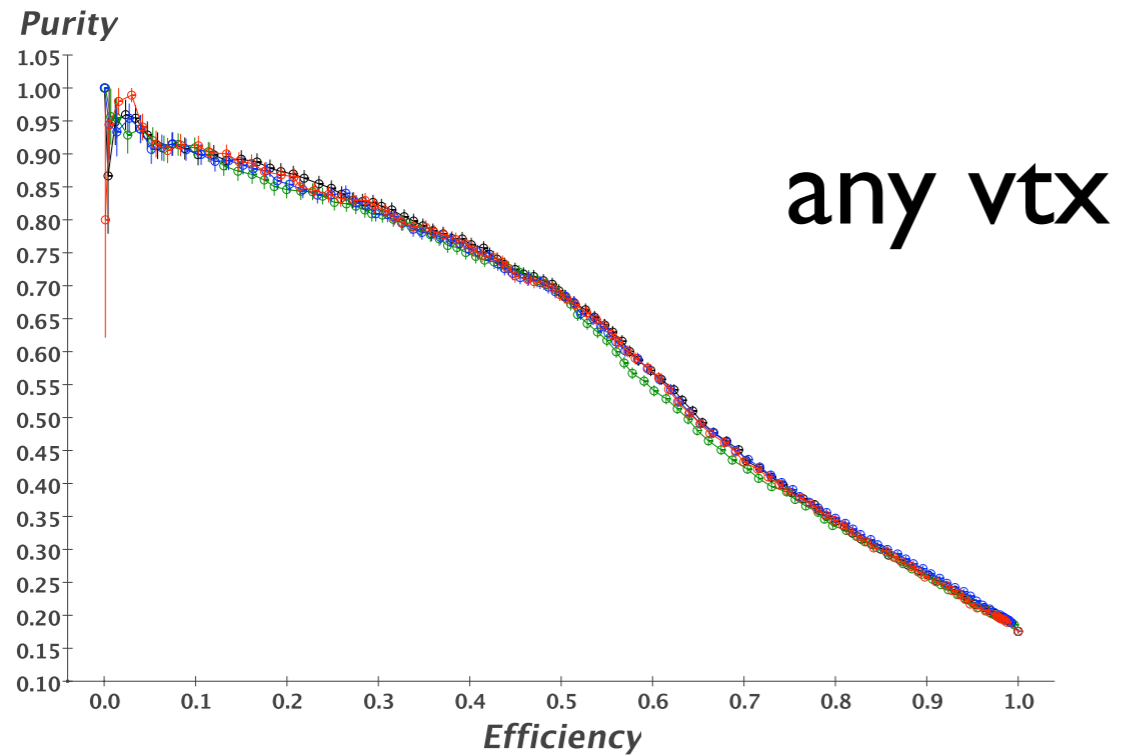
Neural networks training

Neural nets training

- The random nature of the networks training produces fluctuations (up to ~ 2 sigma) in the efficiency x purity plots.
- Perform a certain number of training and pick the networks that give the best results.
- Four sets for each network of training are ready. Reconstruction ongoing.

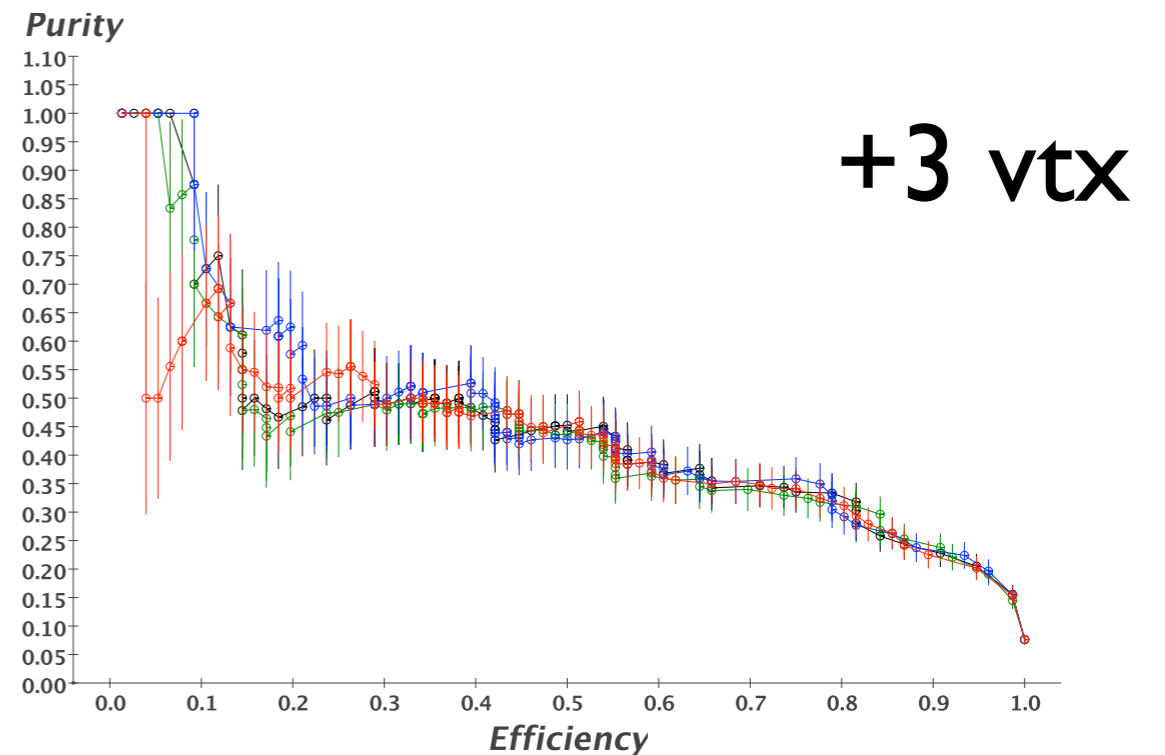
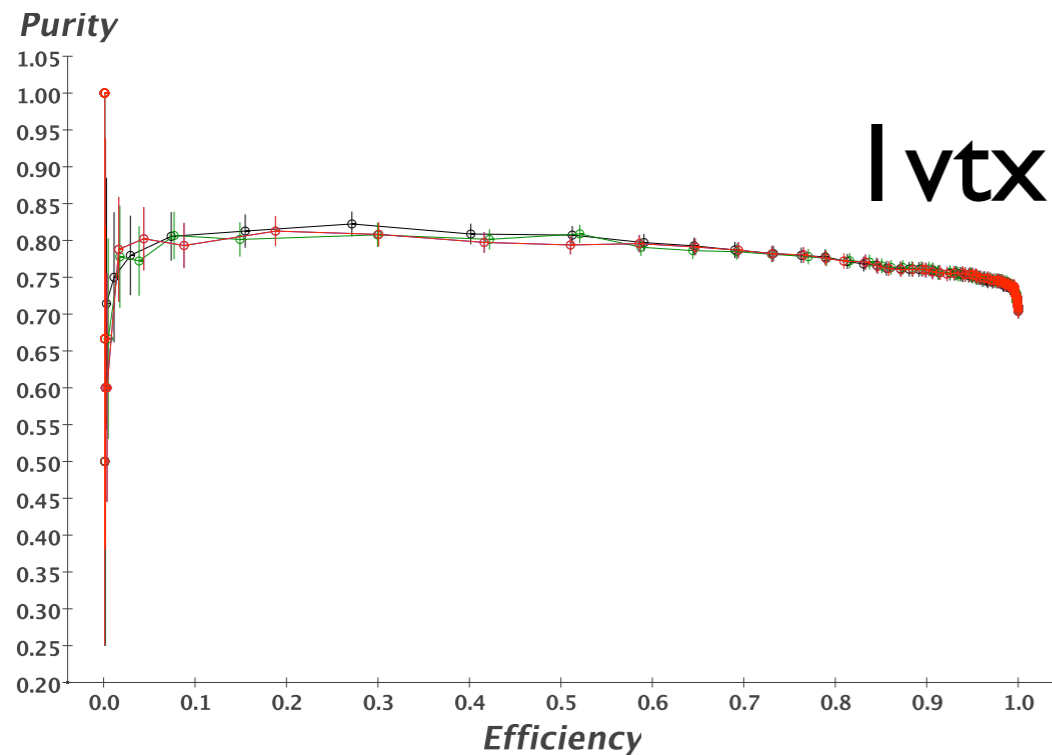
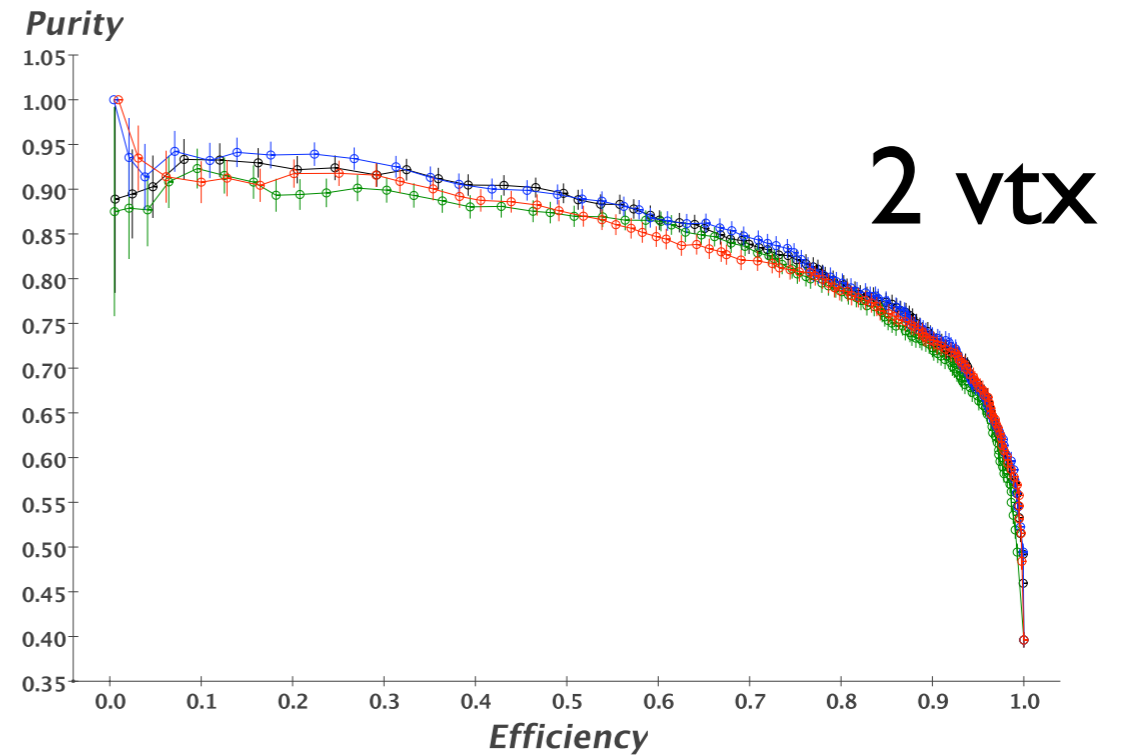
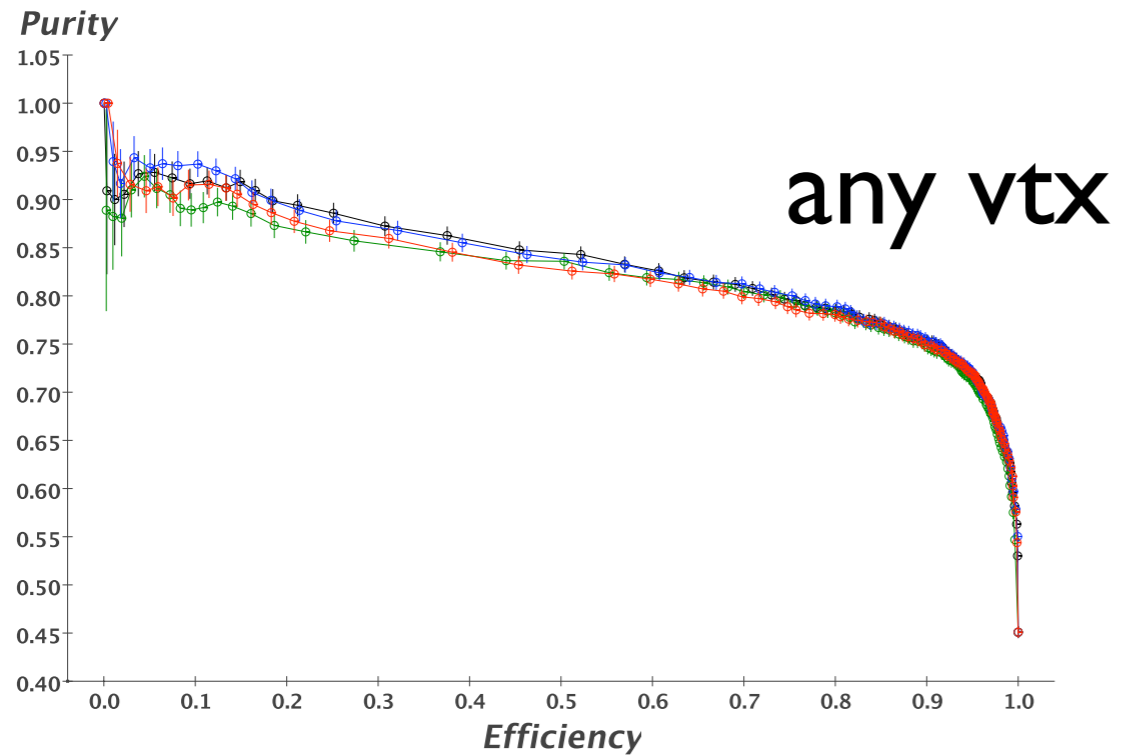
Neural nets training

- C-TAG



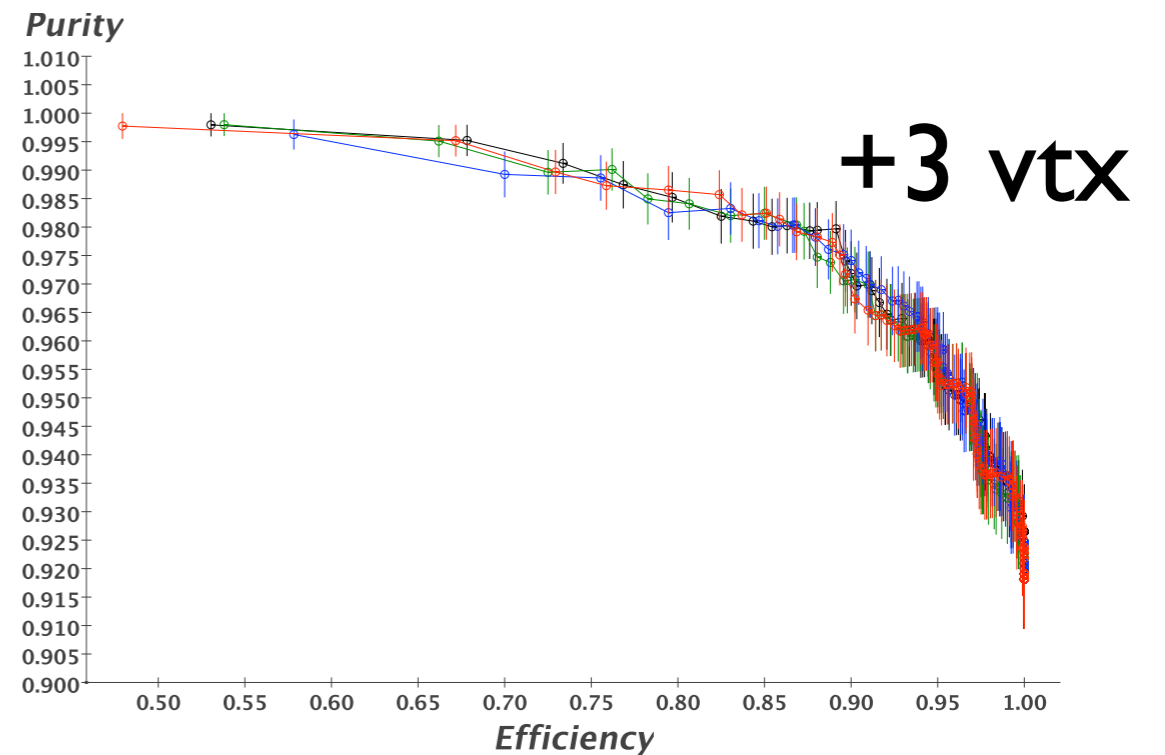
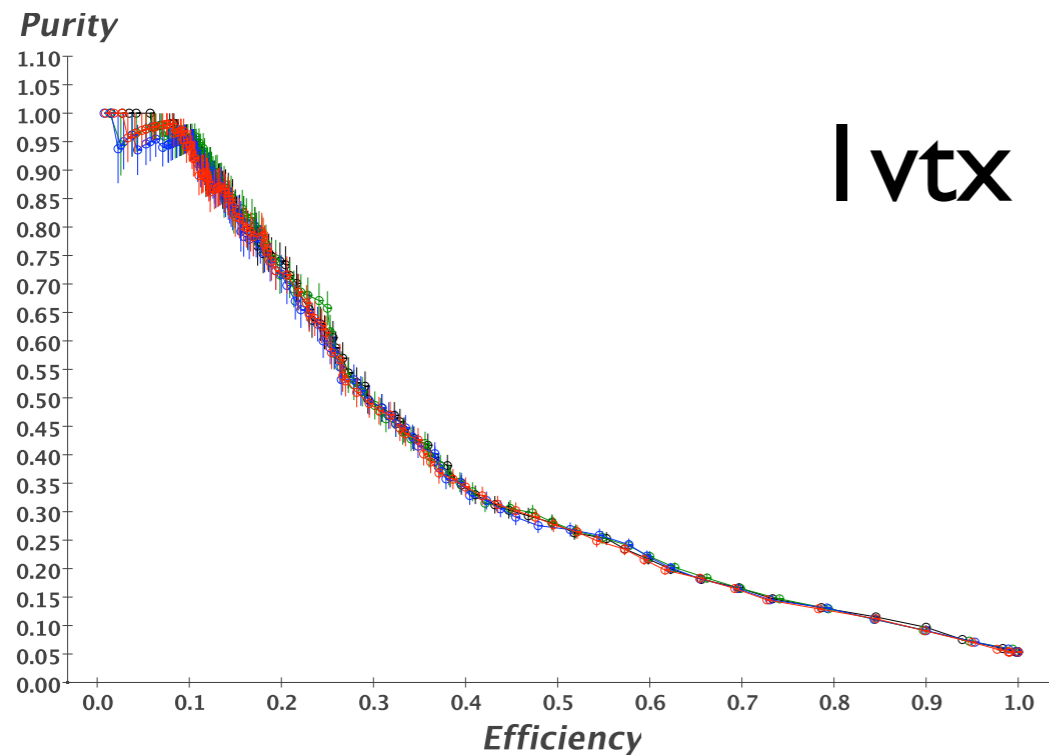
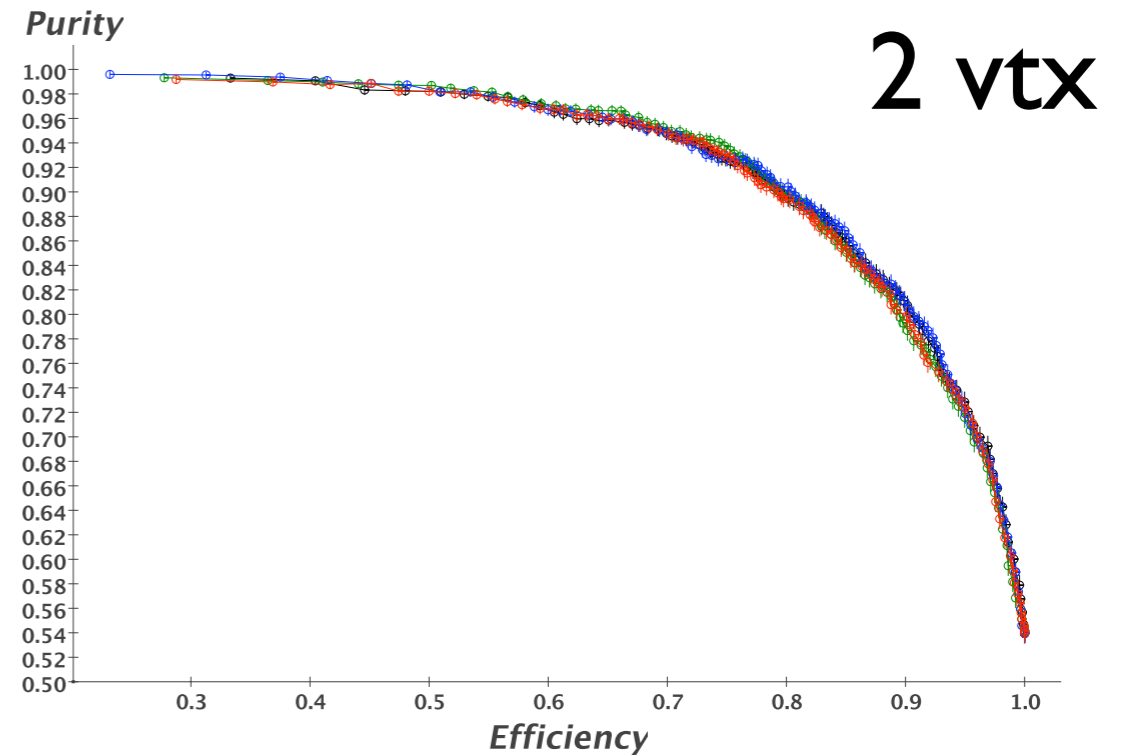
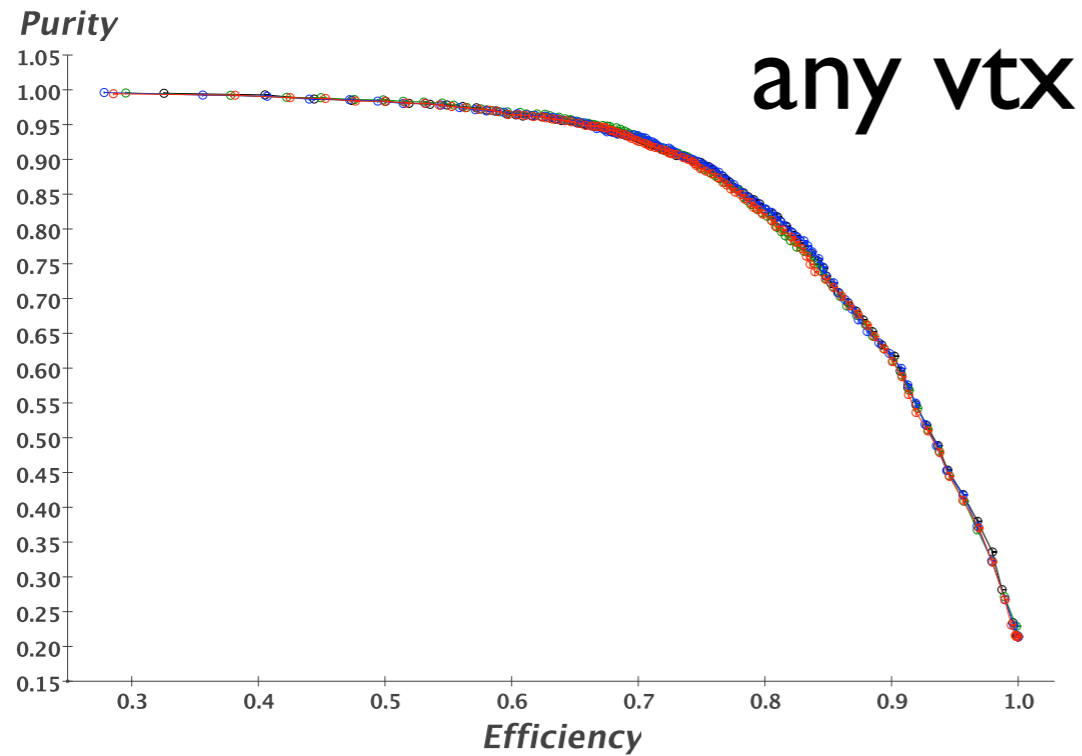
Neural nets training

- BC-TAG



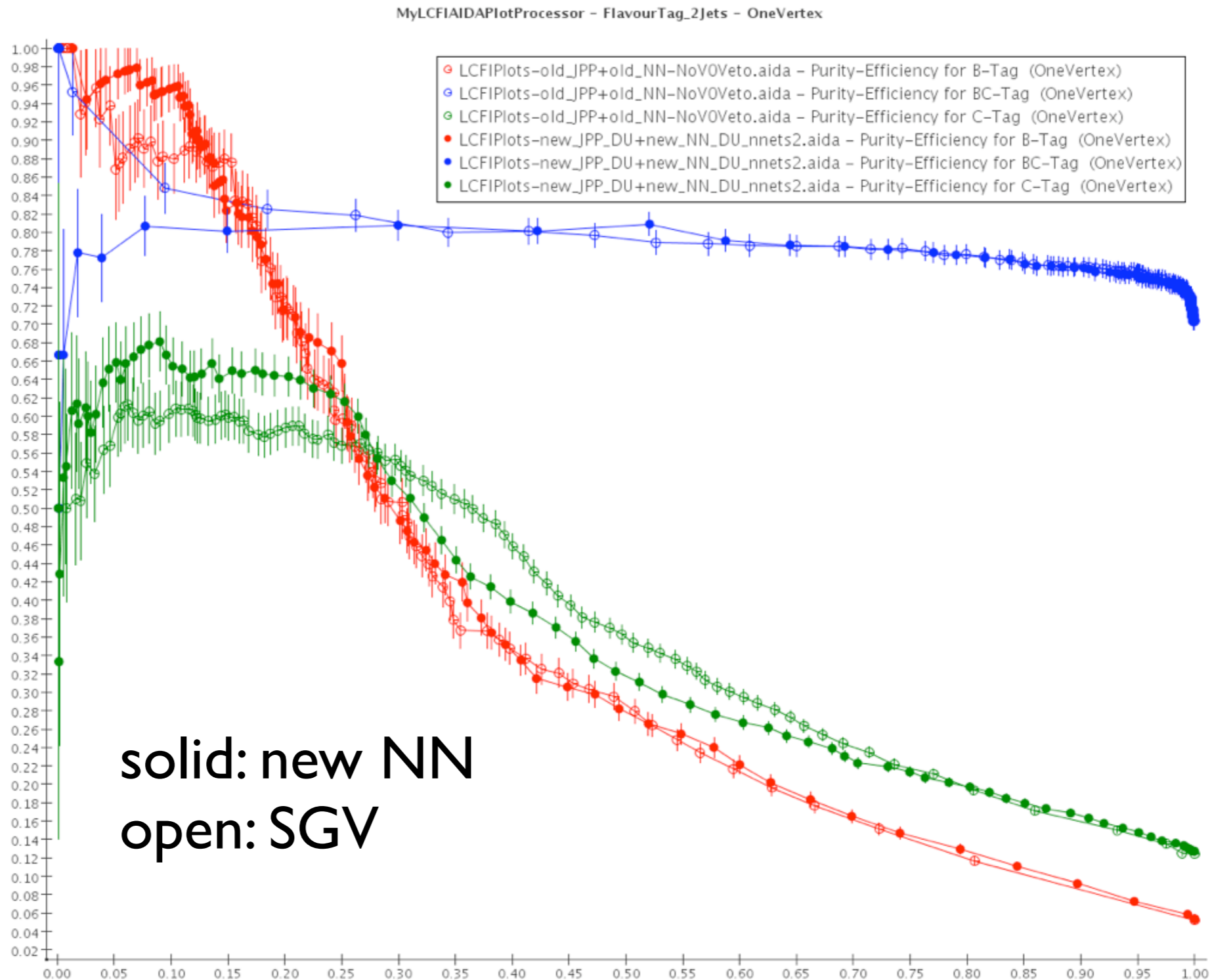
Neural nets training

- B-TAG



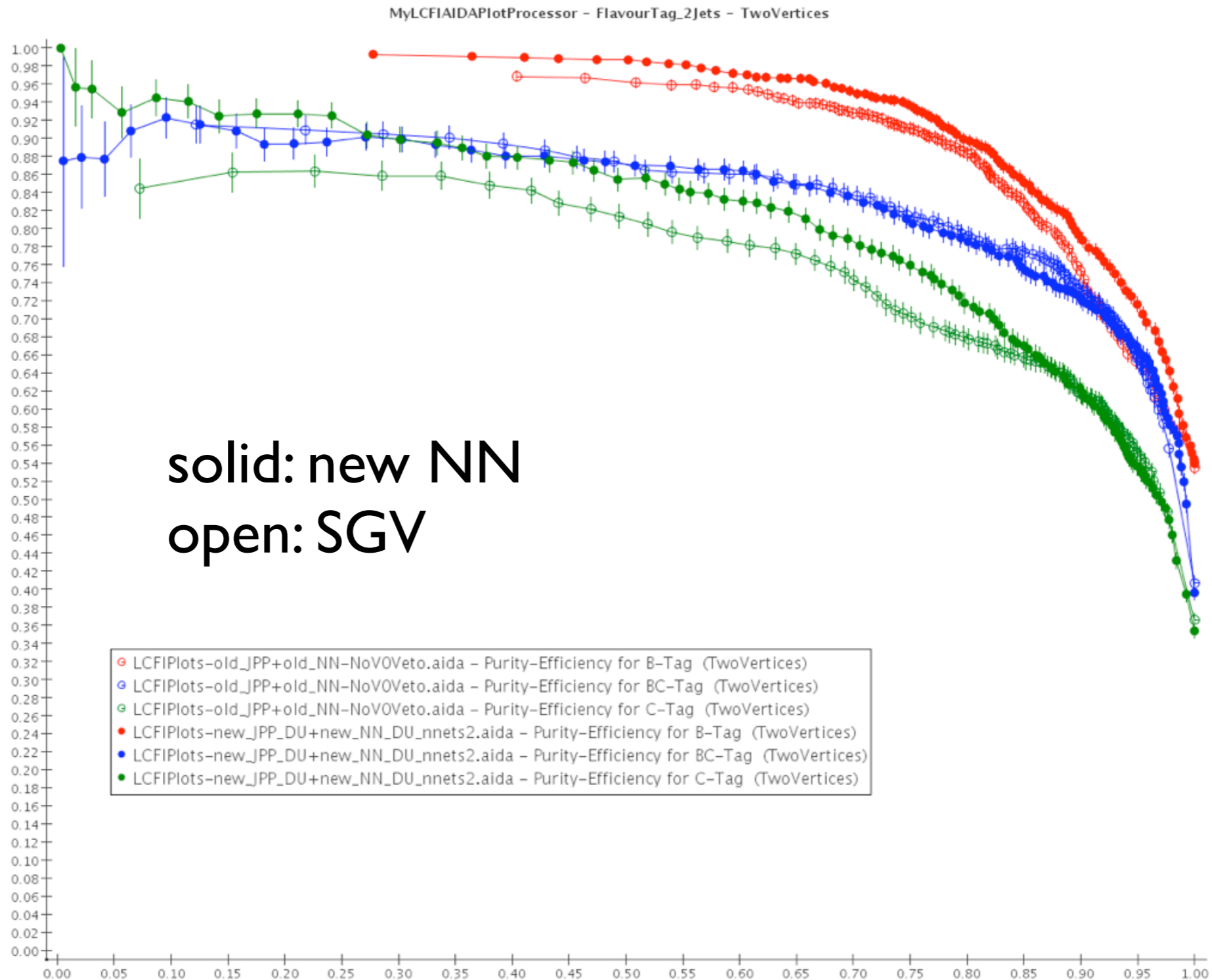
Neural Nets re-training

- Neural nets after retraining I-vtx



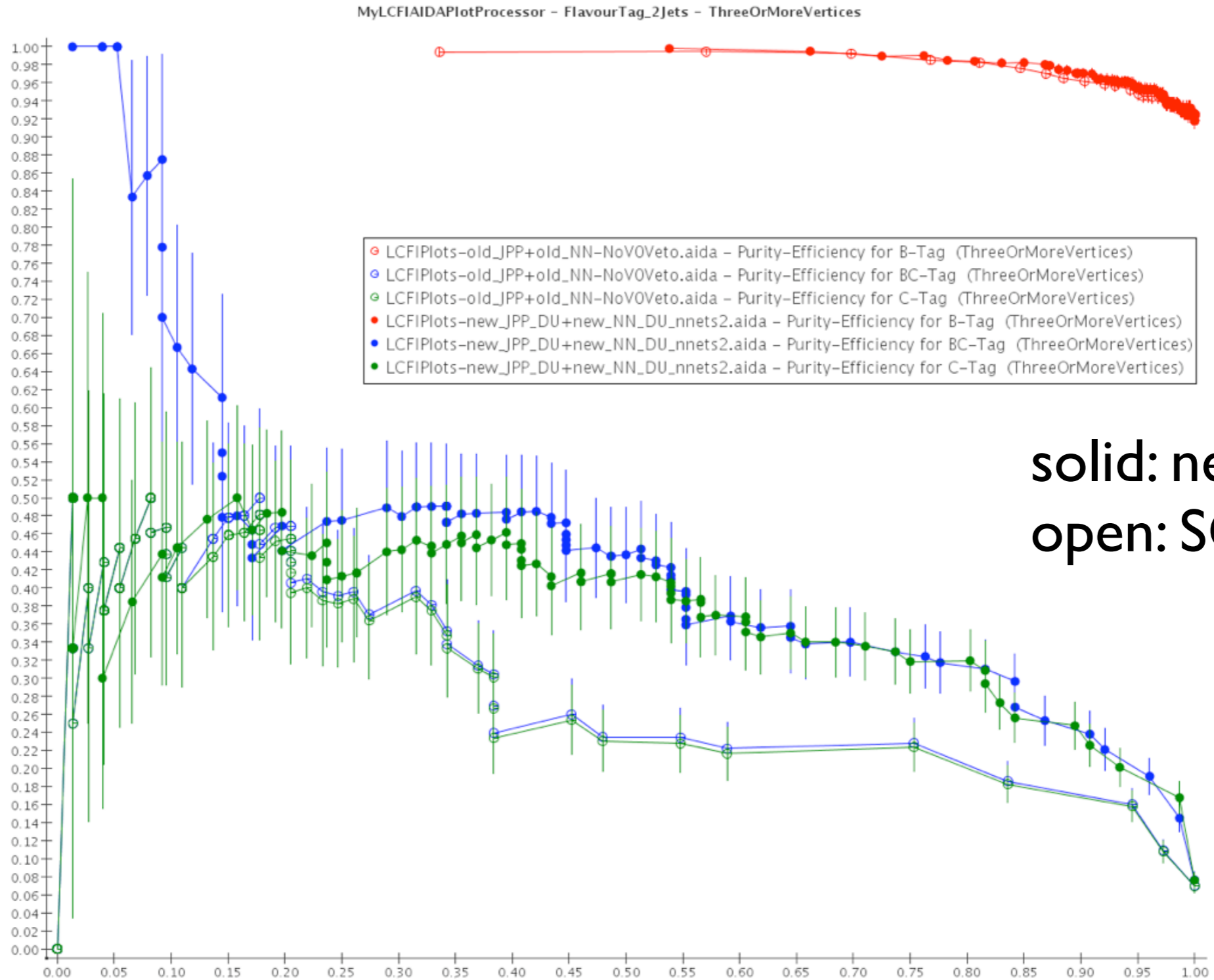
Neural Nets re-training

- Neural nets after retraining 2-vtx



Neural Nets re-training

- Neural nets after retraining +3-vtx



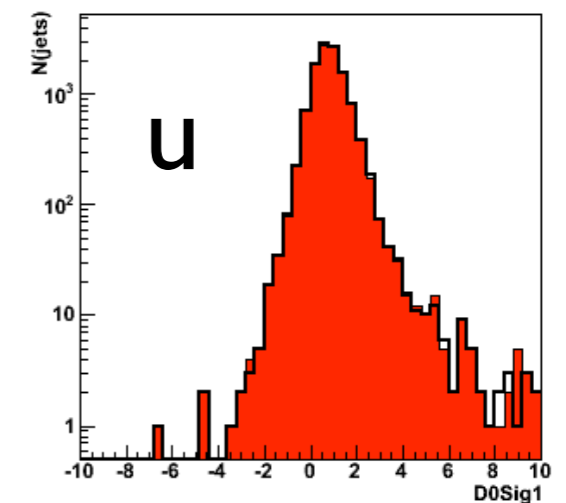
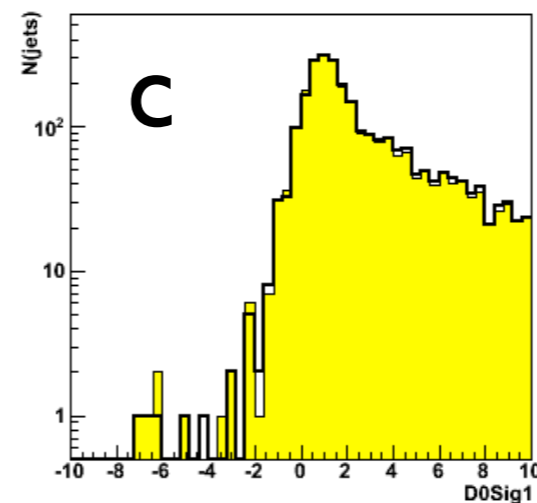
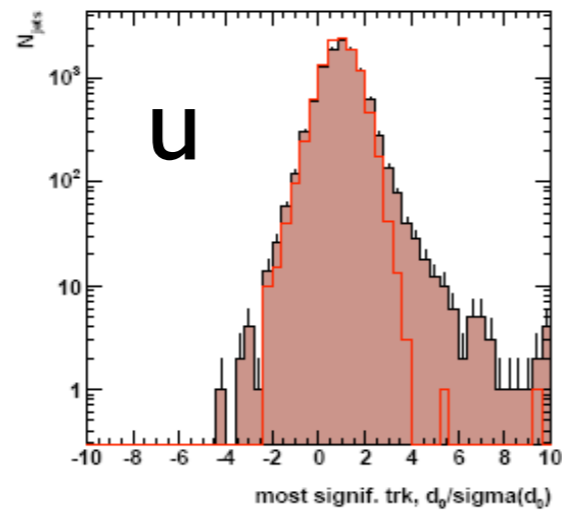
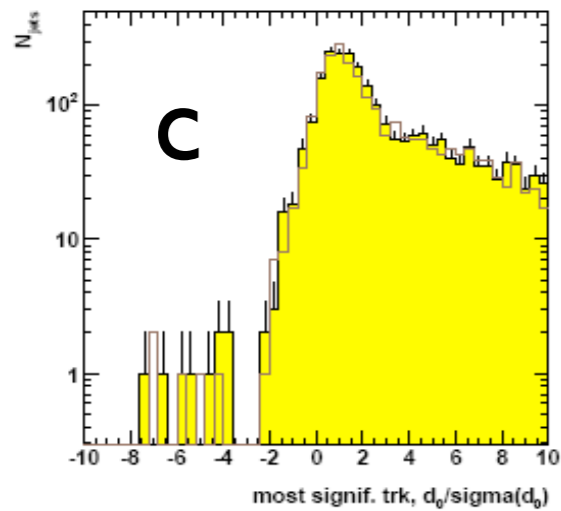
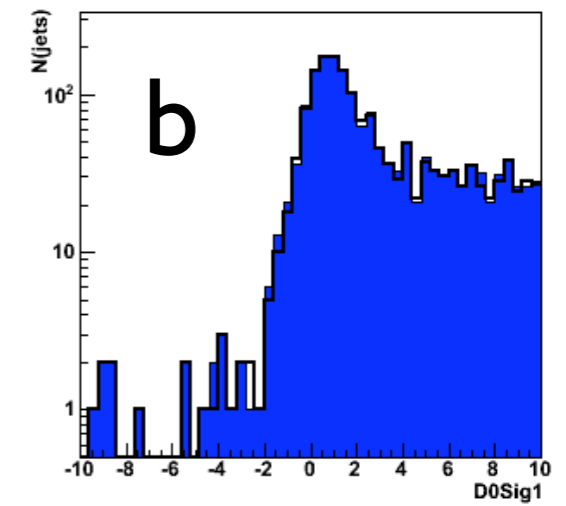
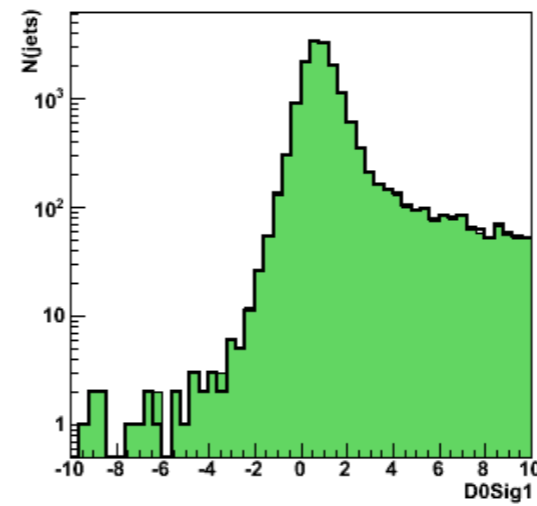
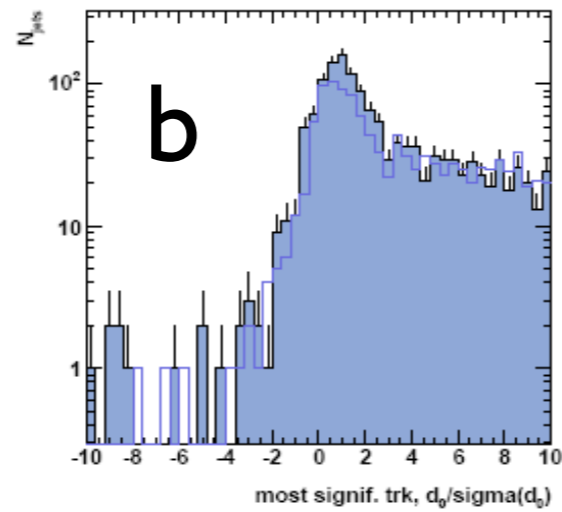
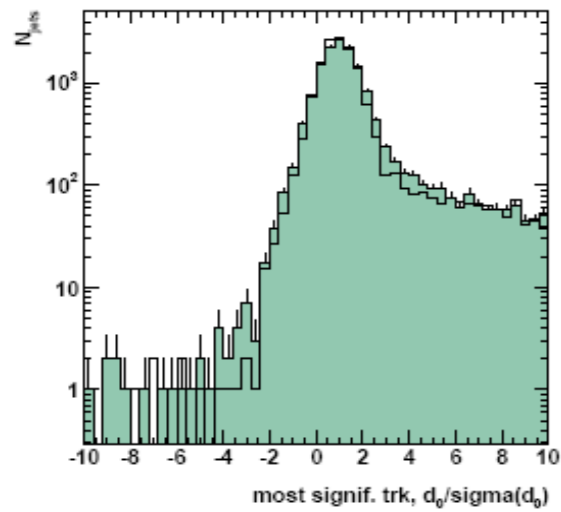
Flavour tag inputs

Impact parameter significance

- Most significant track

SGV - line; Mokka - histo

LDCPrime_02Sc
cheat - line; ConvTag - histo

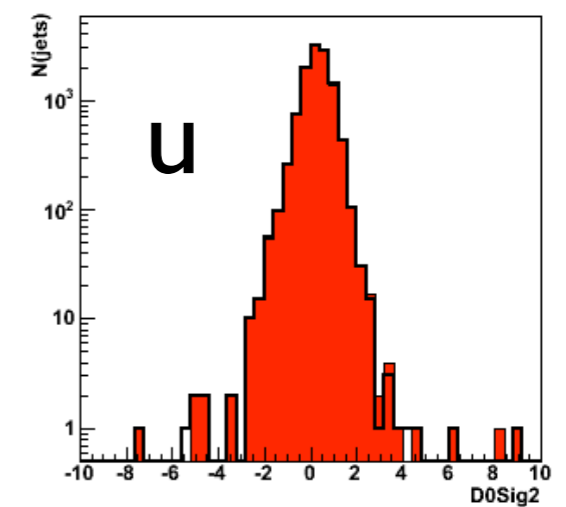
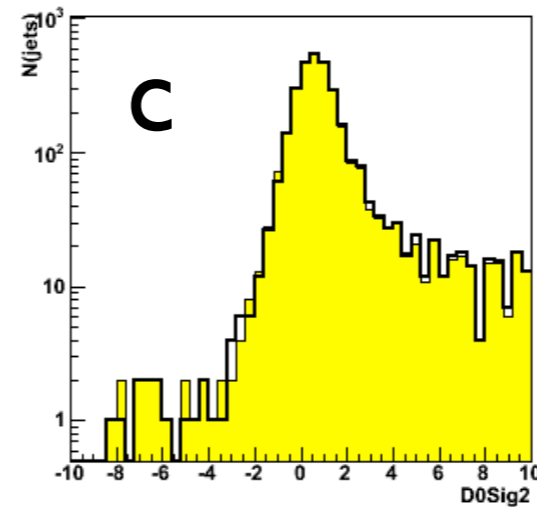
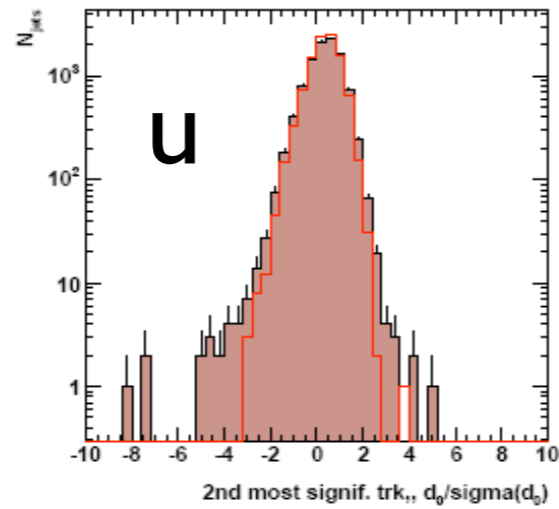
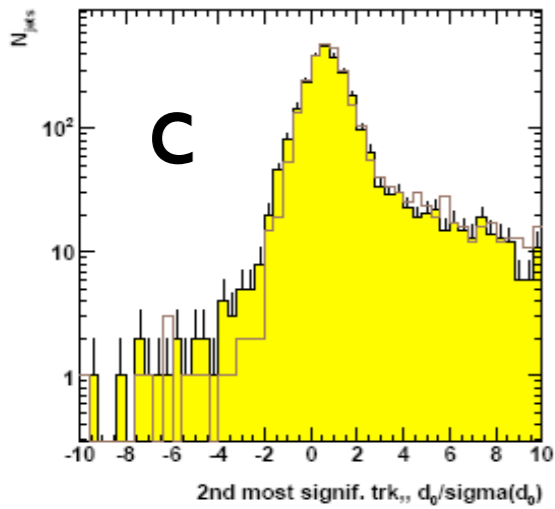
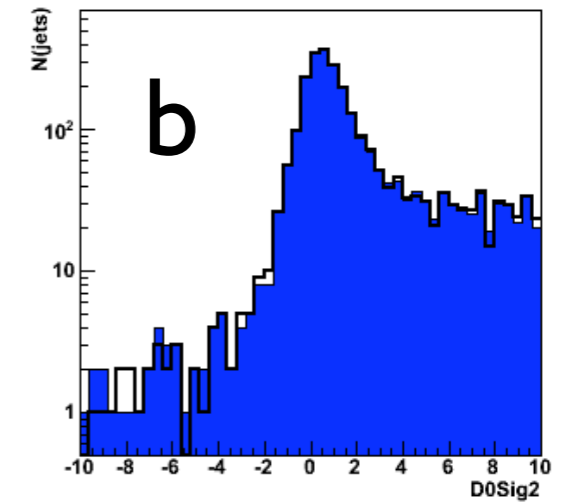
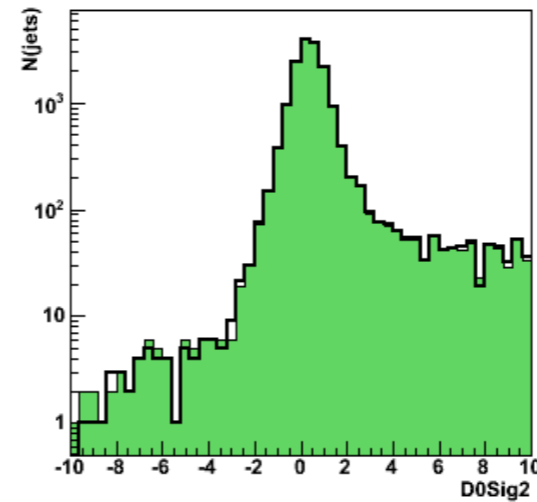
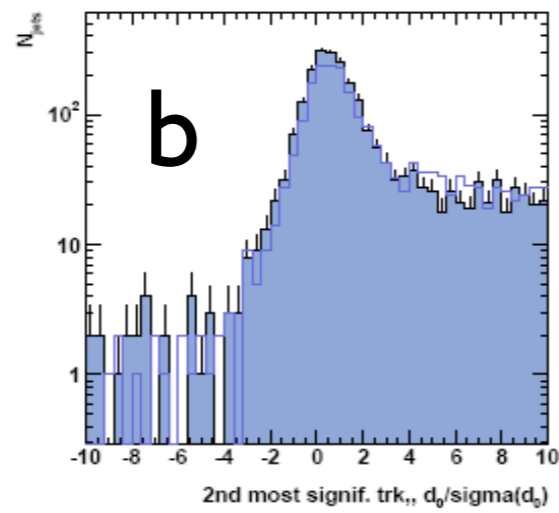
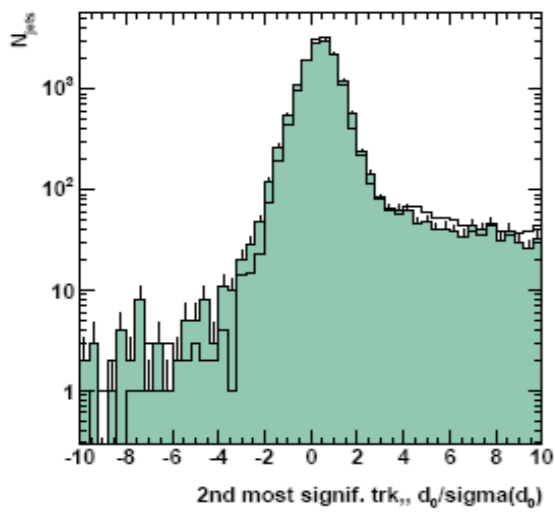


Impact parameter significance

- Second most significant track

SGV - line; Mokka - histo

LDCPrime_02Sc
cheat - line; ConvTag - histo

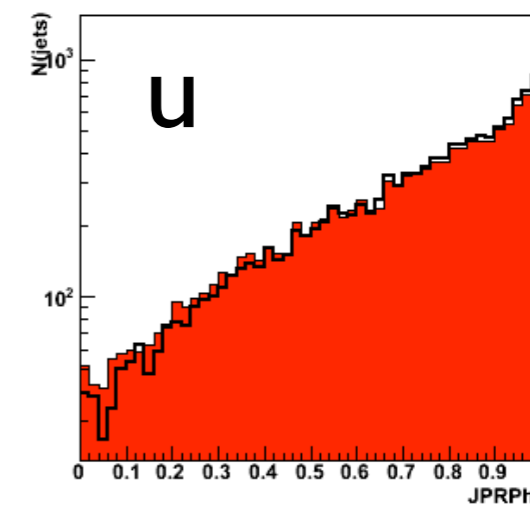
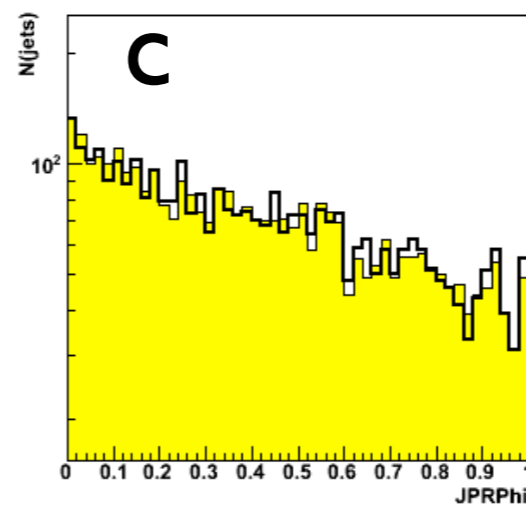
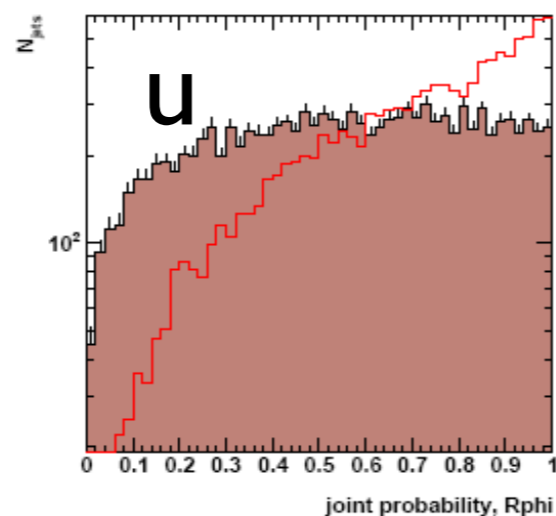
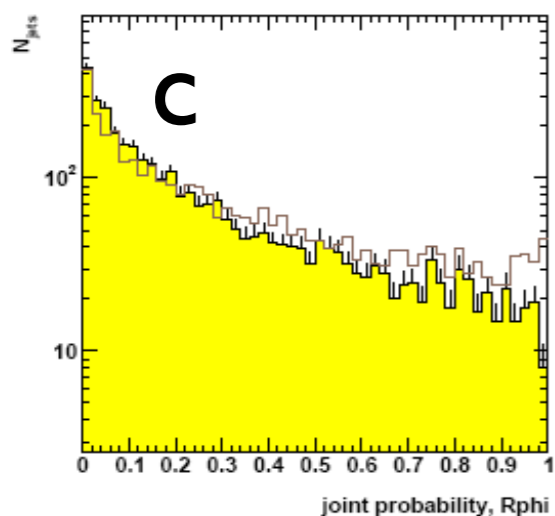
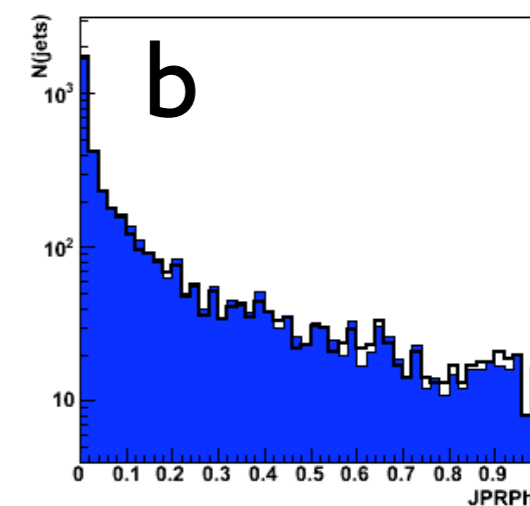
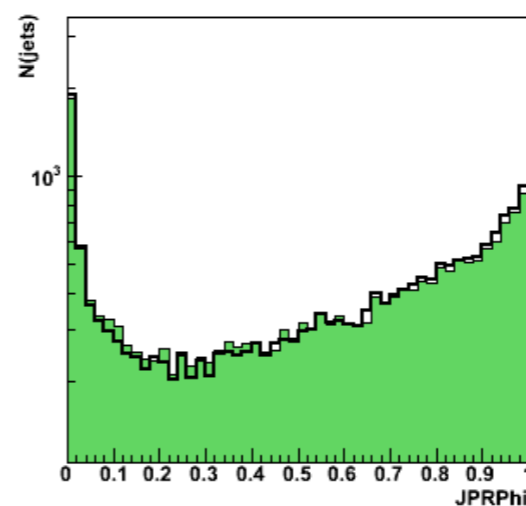
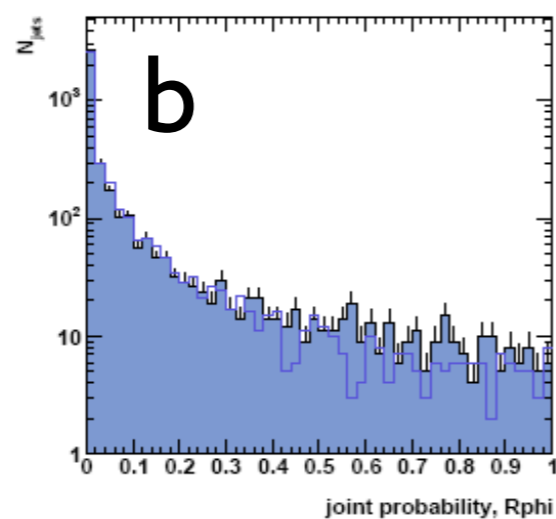
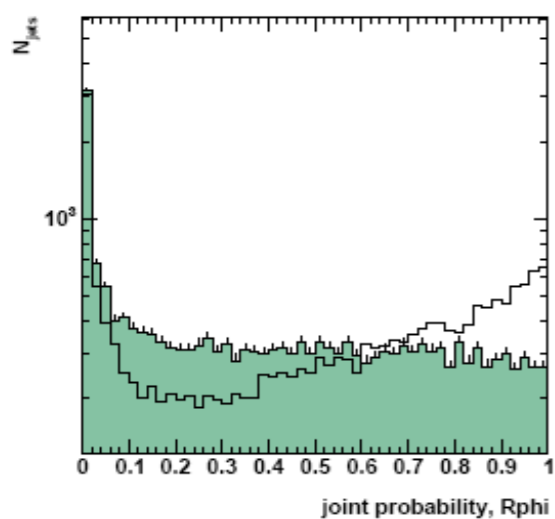


Joint probability

- R-Phi

SGV - line; Mokka - histo

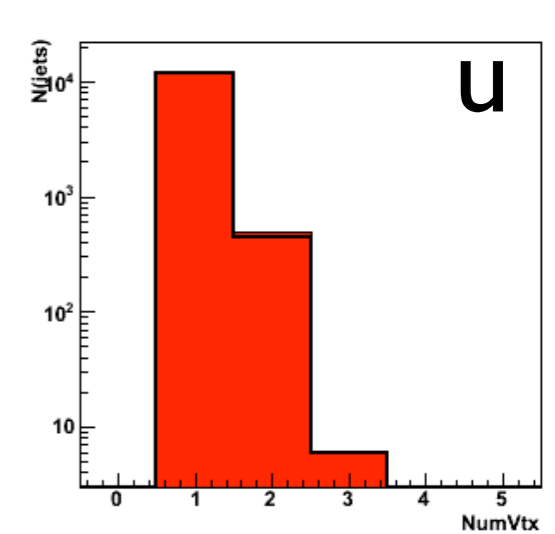
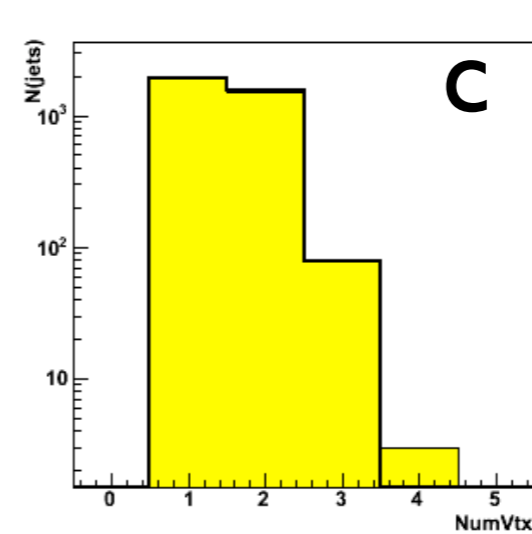
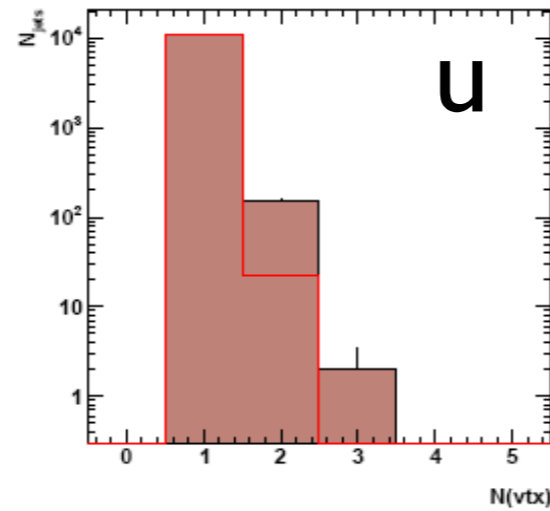
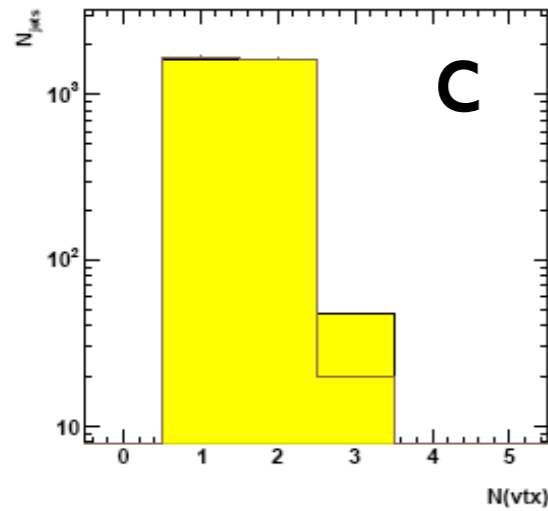
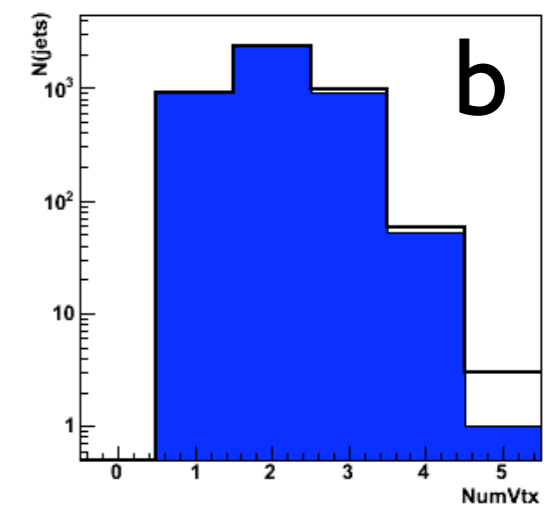
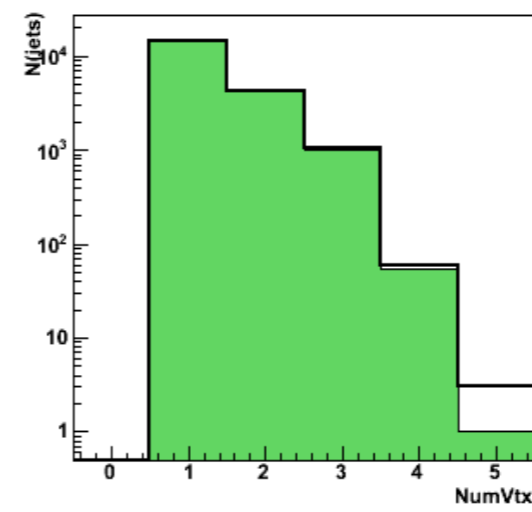
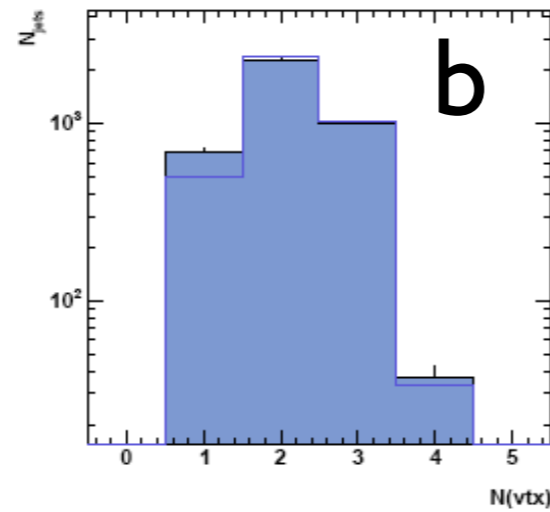
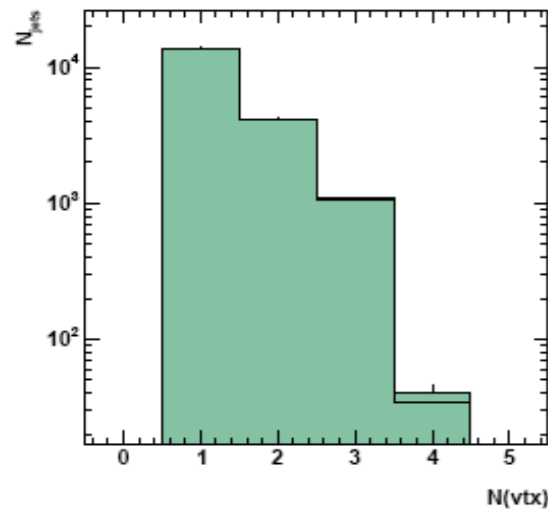
LDCPrime_02Sc
cheat - line; ConvTag - histo



Vertex multiplicity

SGV - line; Mokka - histo

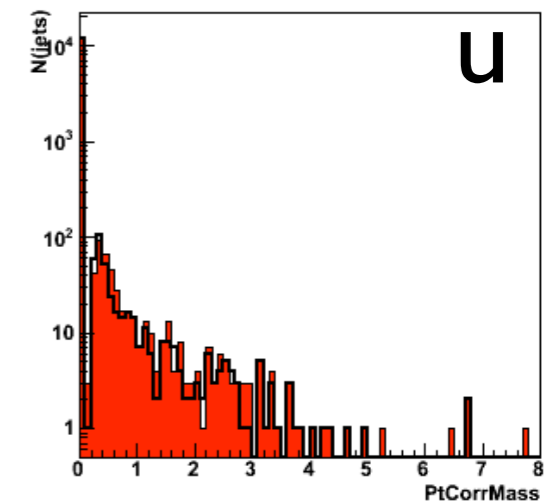
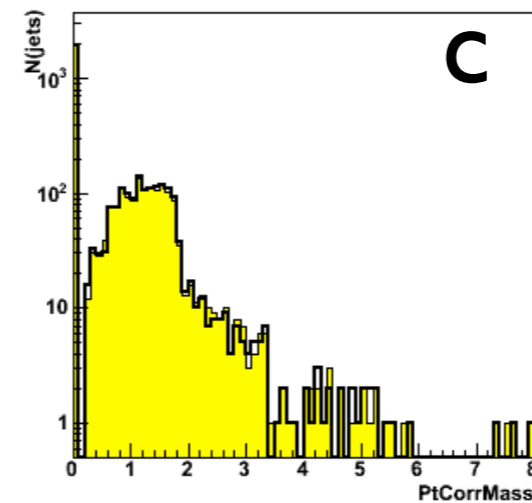
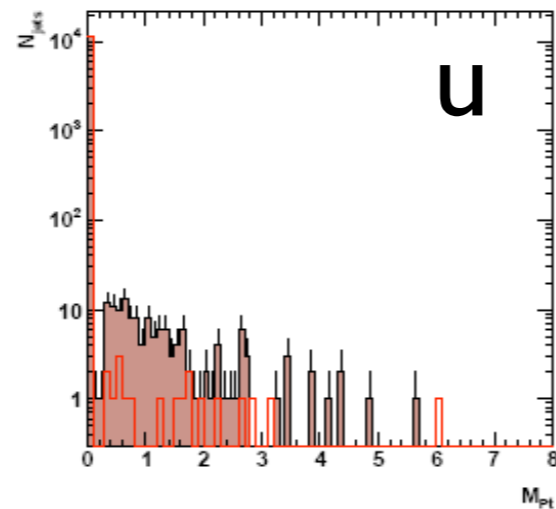
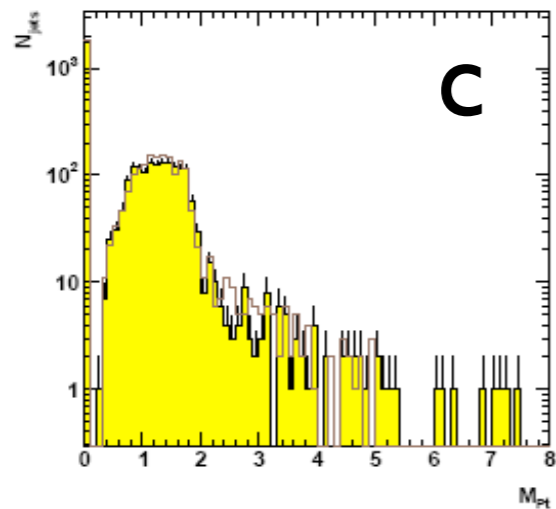
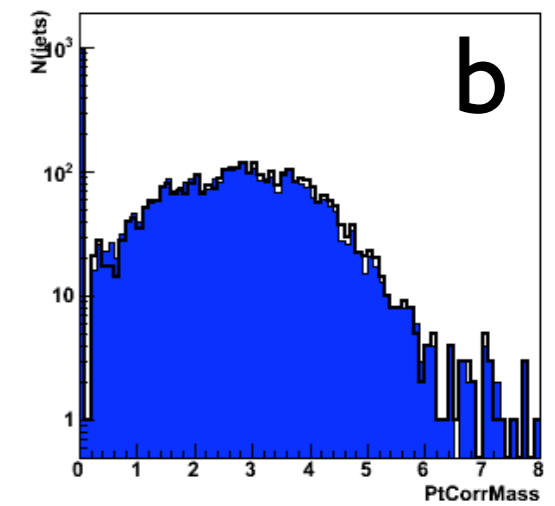
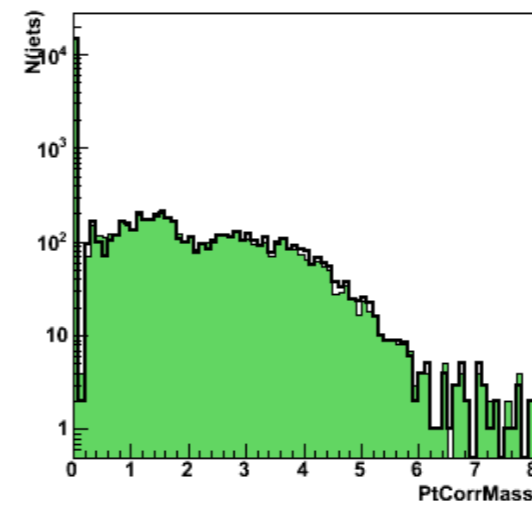
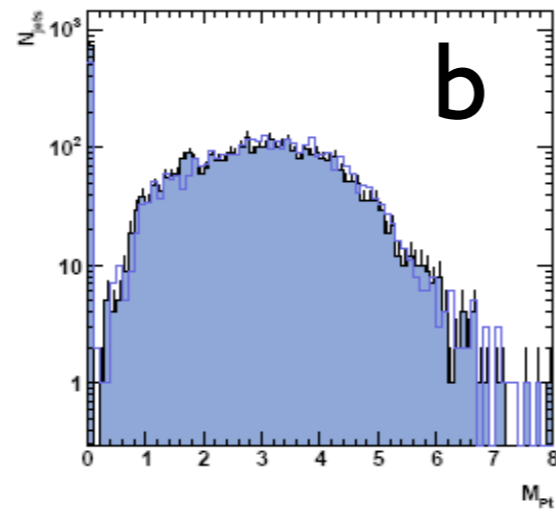
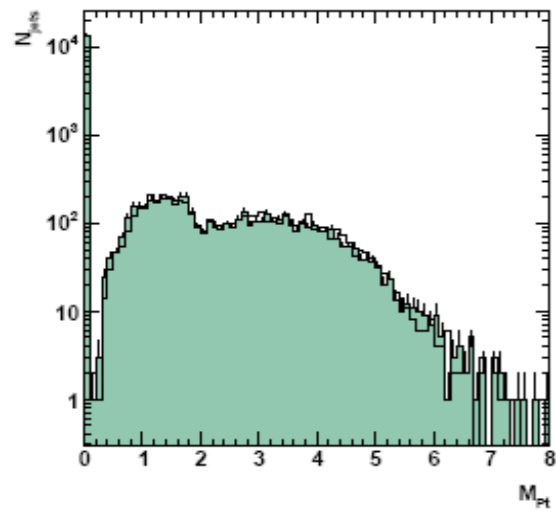
LDCPrime_02Sc
cheat - line; ConvTag - histo



Pt corrected mass

SGV - line; Mokka - histo

LDCPrime_02Sc
cheat - line; ConvTag - histo



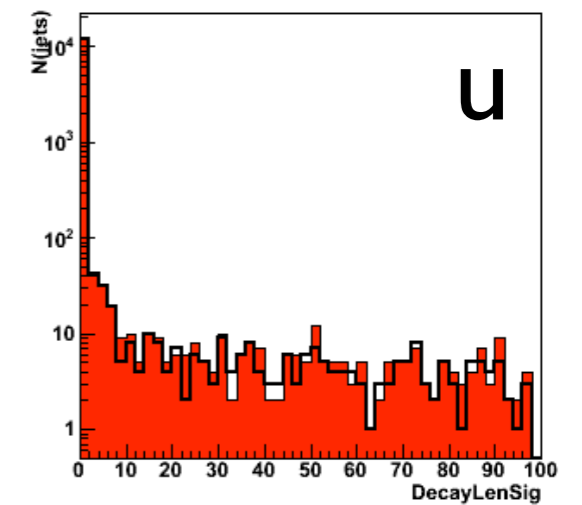
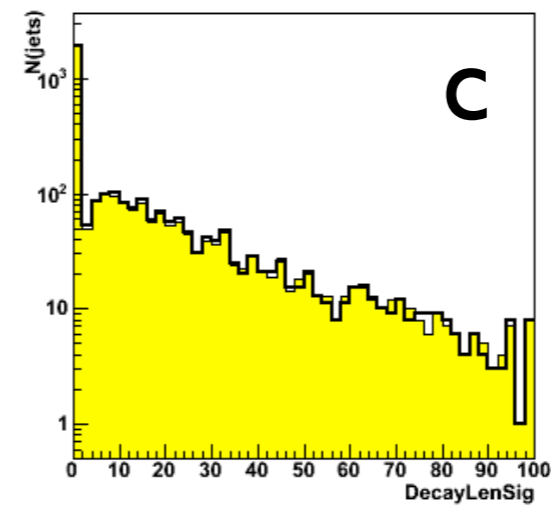
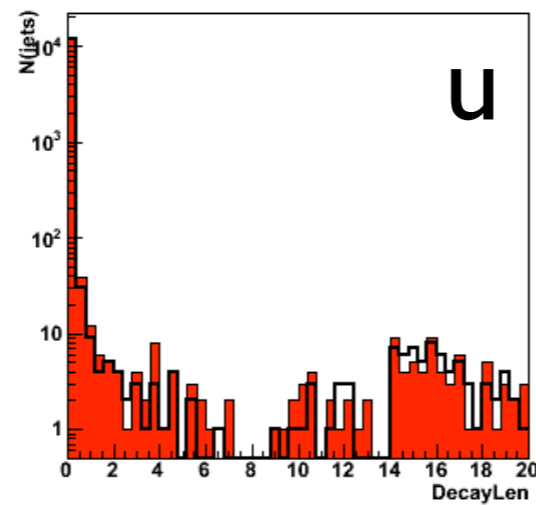
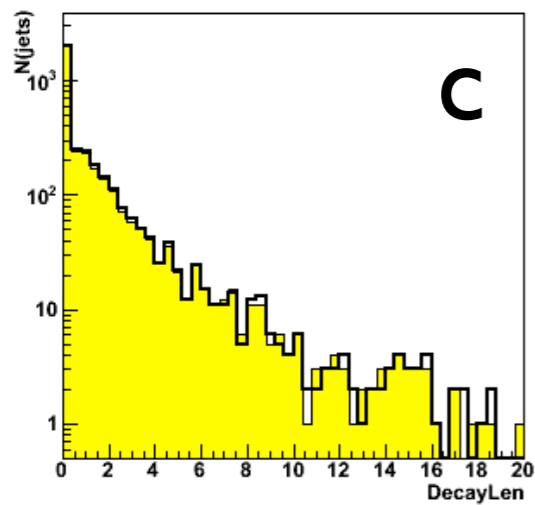
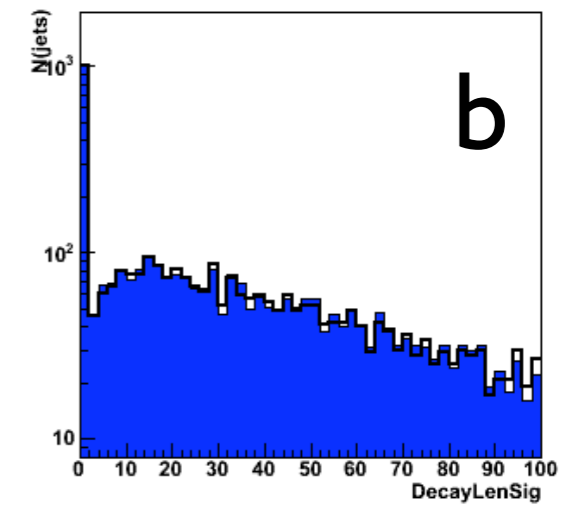
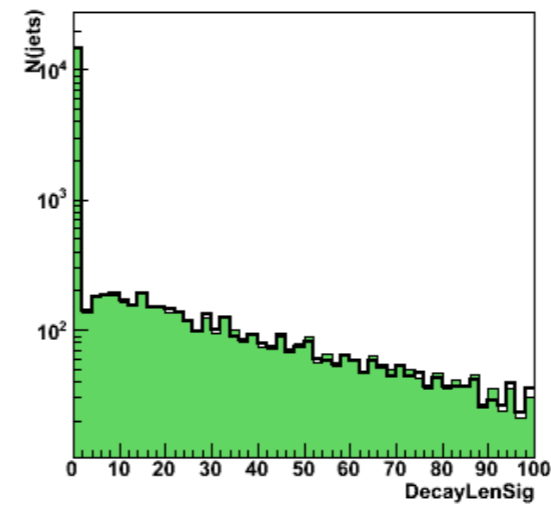
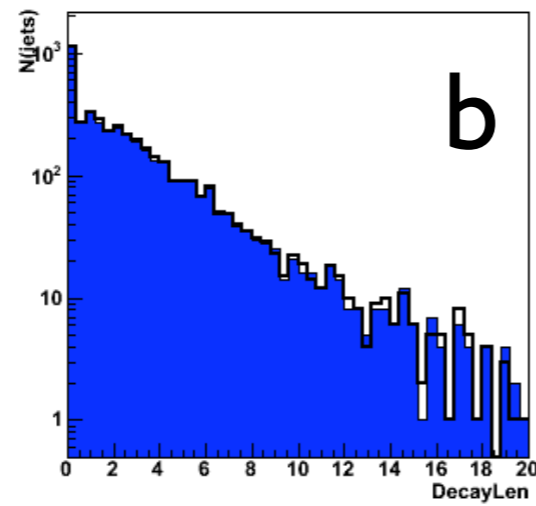
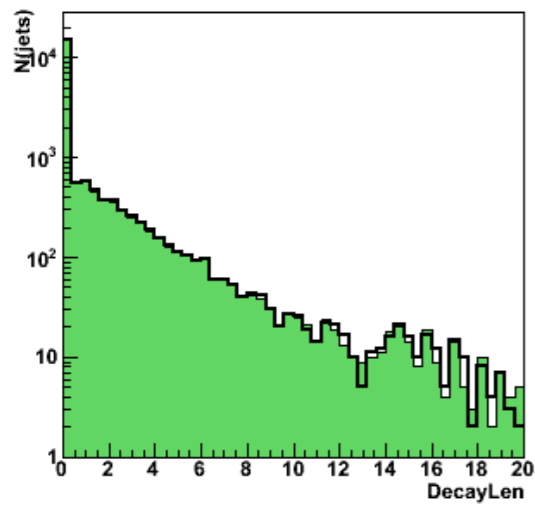
More flavour tag inputs...

LDCPrime_02Sc

cheat - line; ConvTag - histo

Decay length

Decay length significance



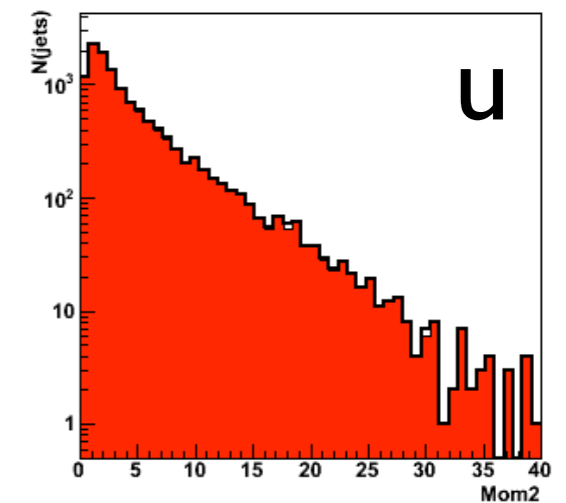
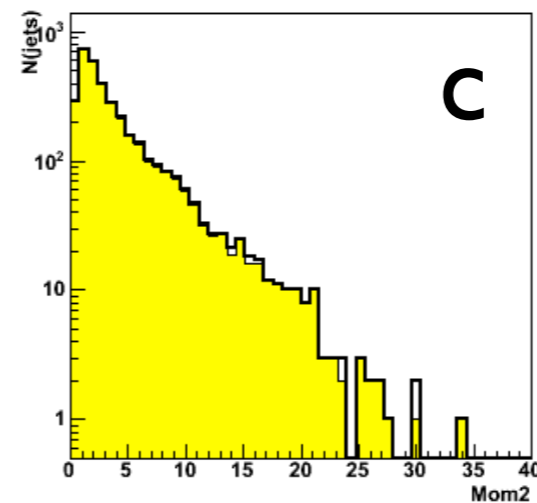
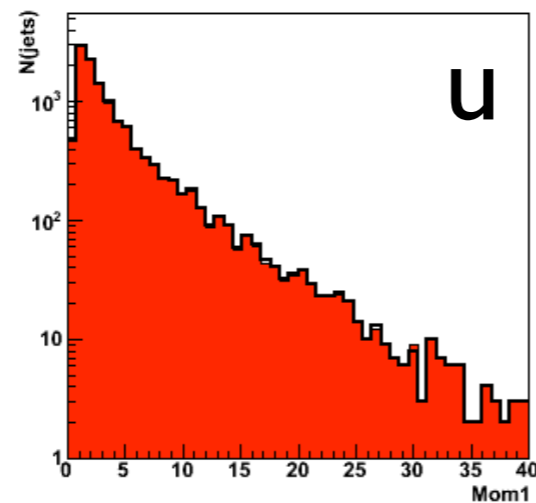
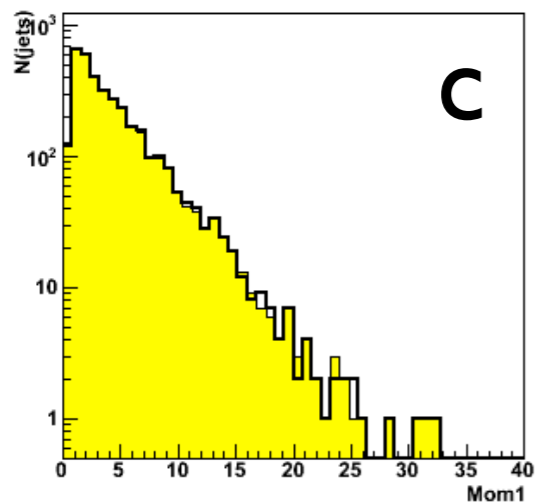
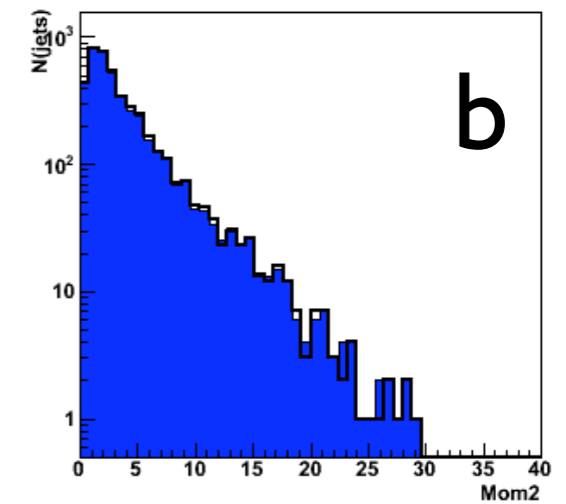
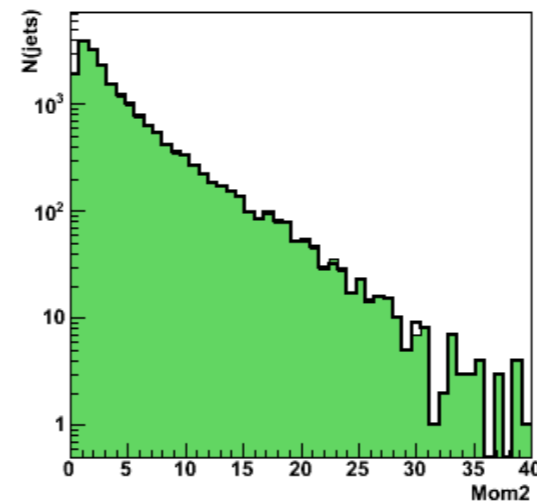
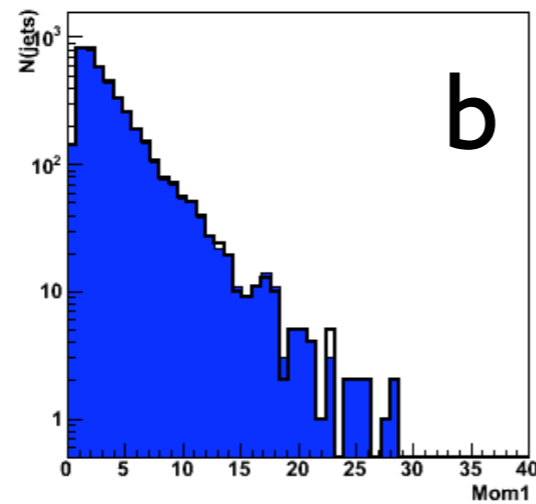
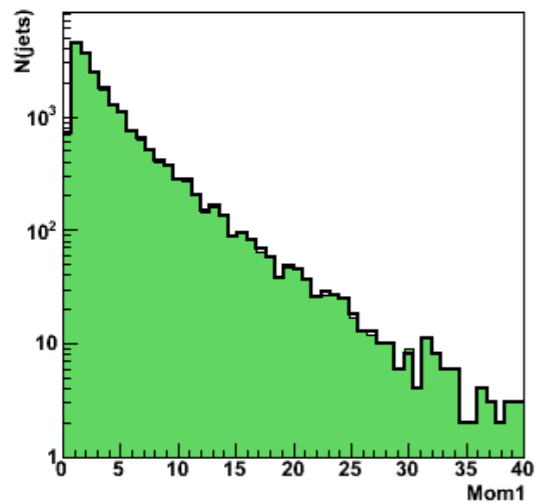
More flavour tag inputs...

LDCPrime_02Sc

cheat - line; ConvTag - histo

Momentum most significant track

Momentum second most significant track



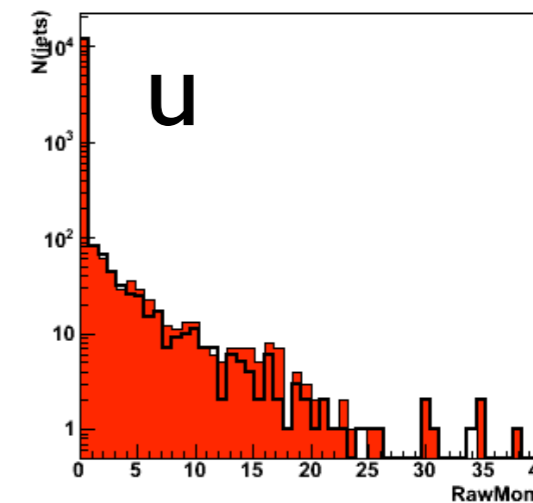
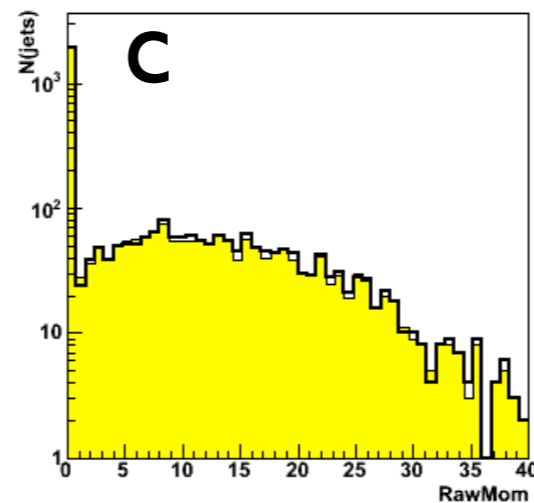
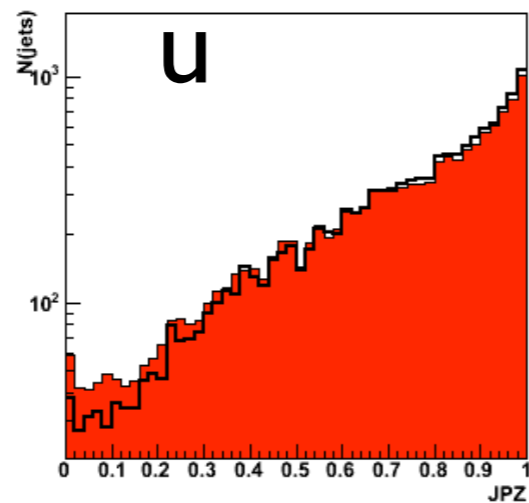
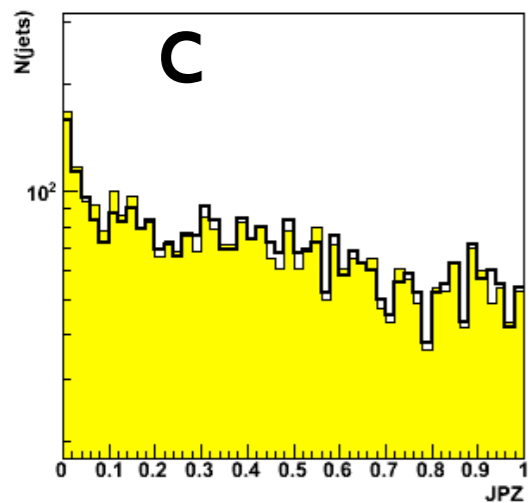
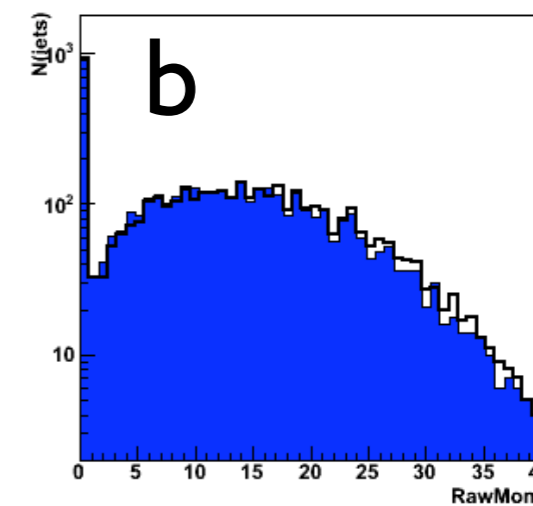
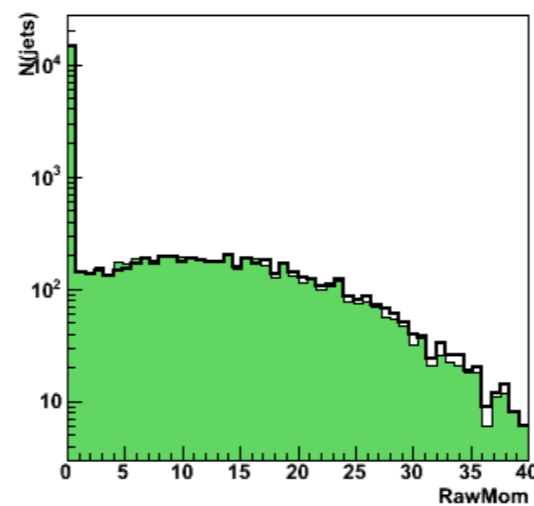
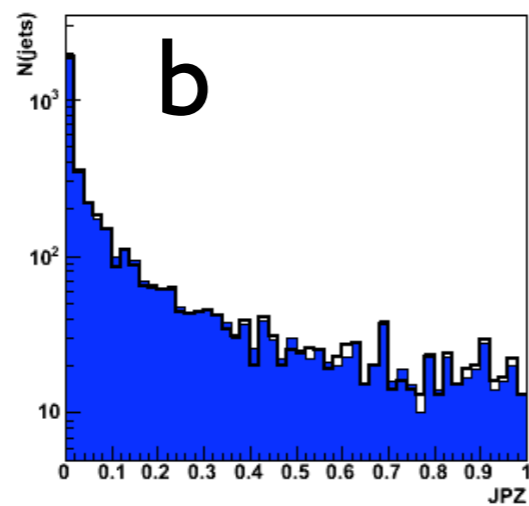
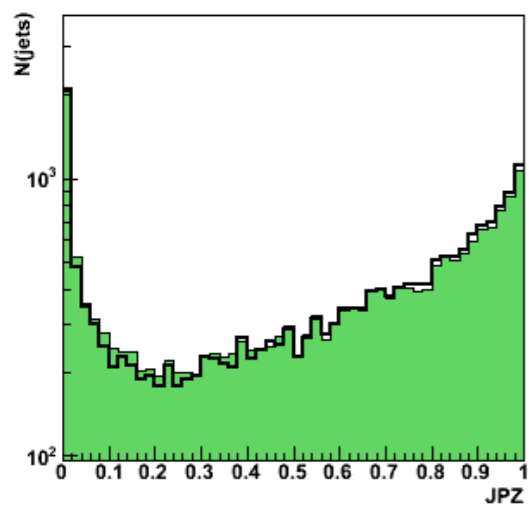
More flavour tag inputs...

LDCPrime_02Sc

cheat - line; ConvTag - histo

Joint probability Z

Raw momentum



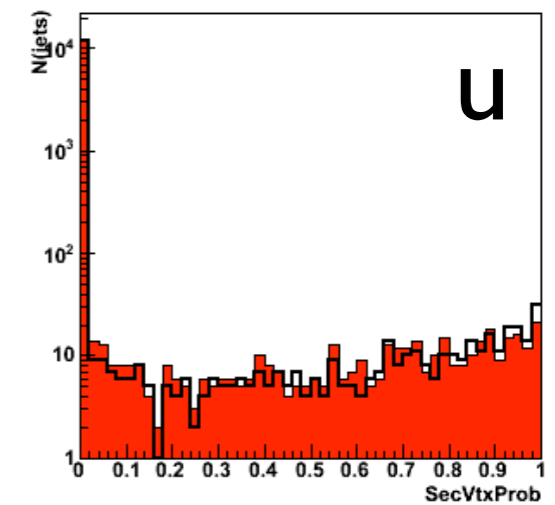
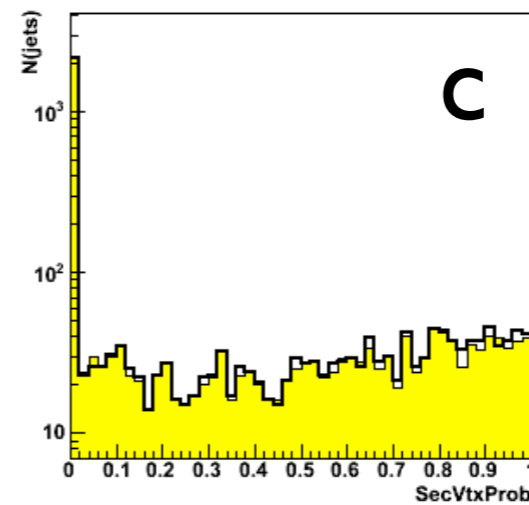
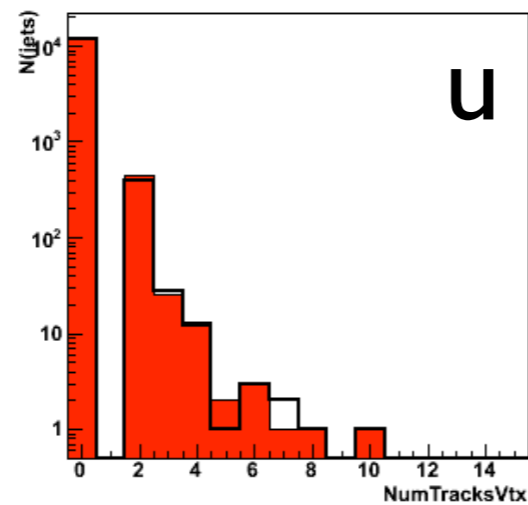
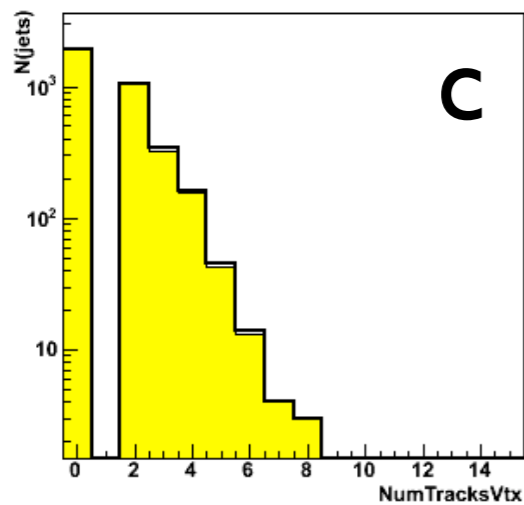
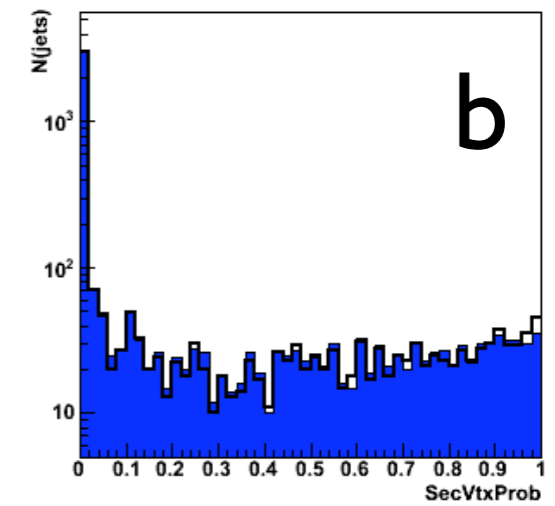
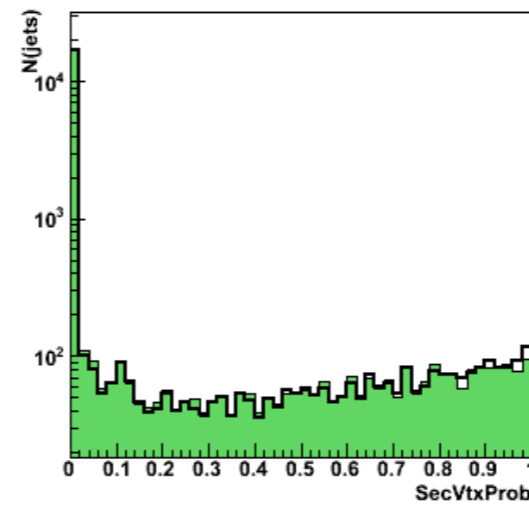
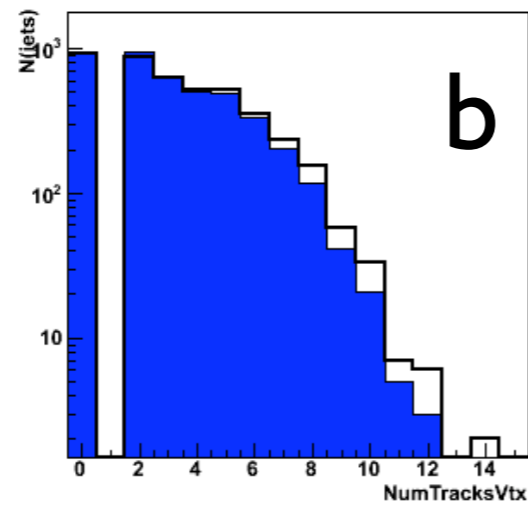
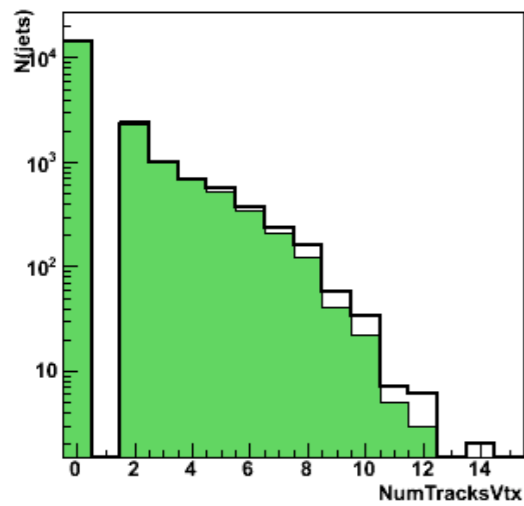
More flavour tag inputs...

LDCPrime_02Sc

cheat - line; ConvTag - histo

Num. of tracks vertex

Secondary vertex prob.



Track selection tuning

Track selection tuning

- Vary the parameters for track selection for the vertex reconstruction (IPFIT, ZVRES) and for the flavour tag inputs (FTI).
- Aim: Improve flavour tagging keeping the performance of vertex reconstruction and flavour tag inputs.

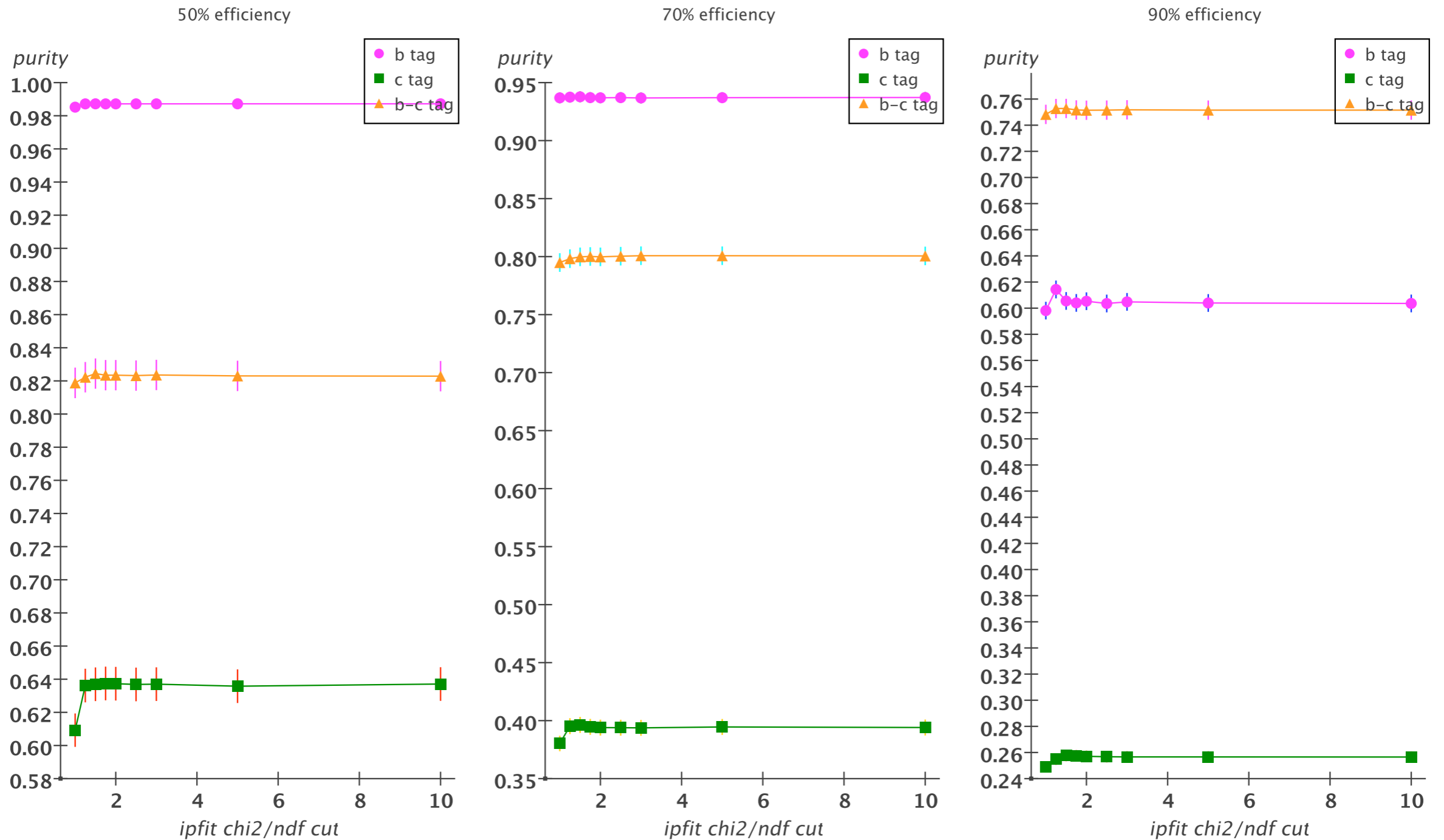
- **Monte Carlo sample** (Talini):
 - $e^+e^- \rightarrow Z \rightarrow qq$, $\sqrt{s} = 91.2$ GeV;
 - 10000 events;
 - Detector model: LDCPrime_02Sc.
- **Reconstruction:**
 - ilcsoft v01-03-06-p02; LCFIVertex HEAD;
 - tracking: FullLDCTracking;
 - clustering + particle flow: PandoraPFA;
 - jets: Satoru jet finder, durhamnjet, njet = 2;

Track selection parameters

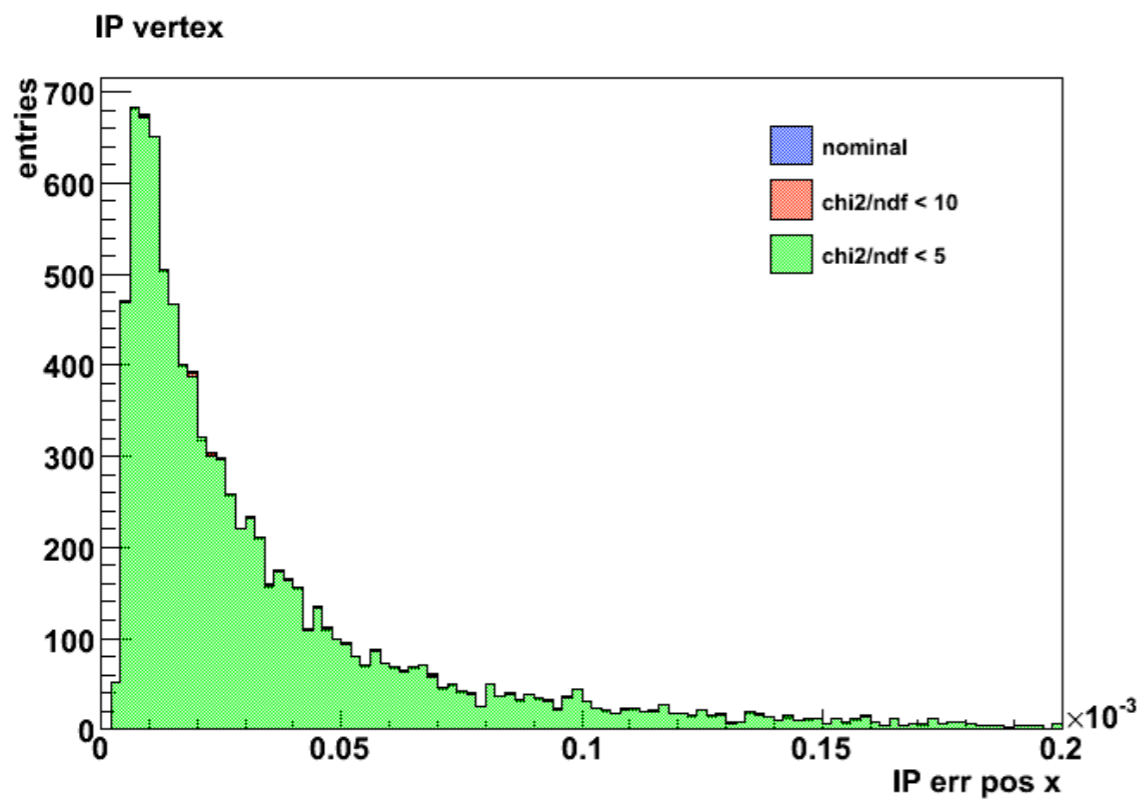
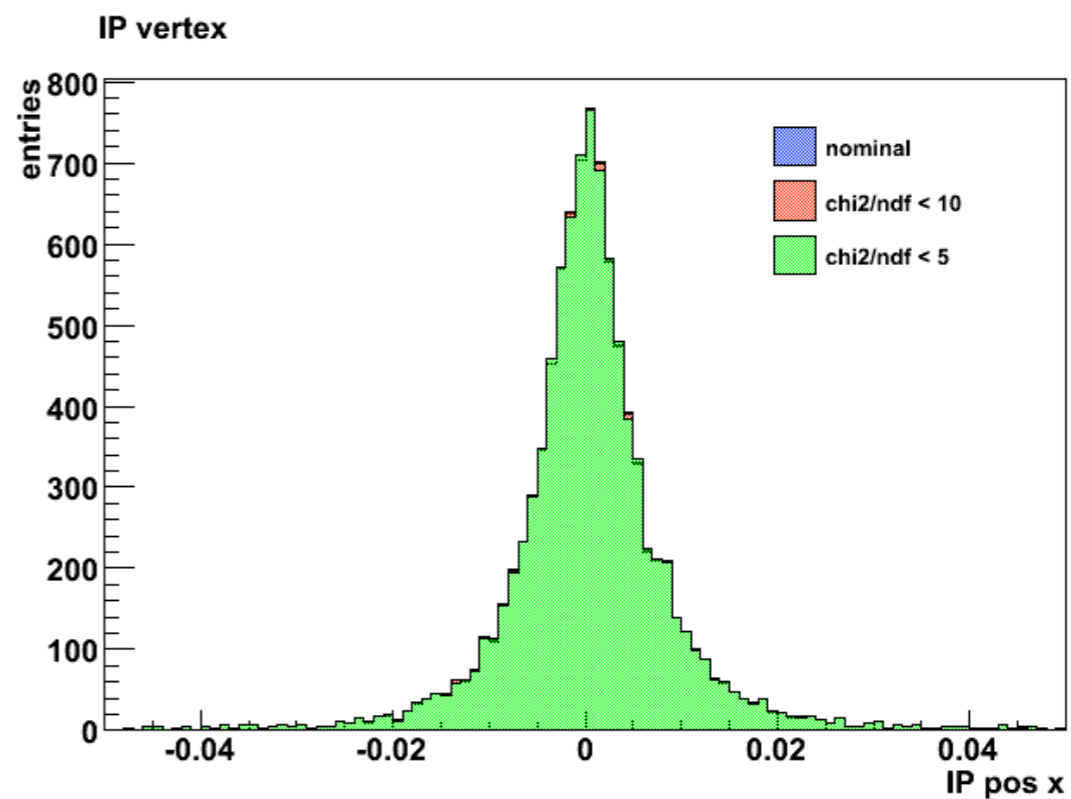
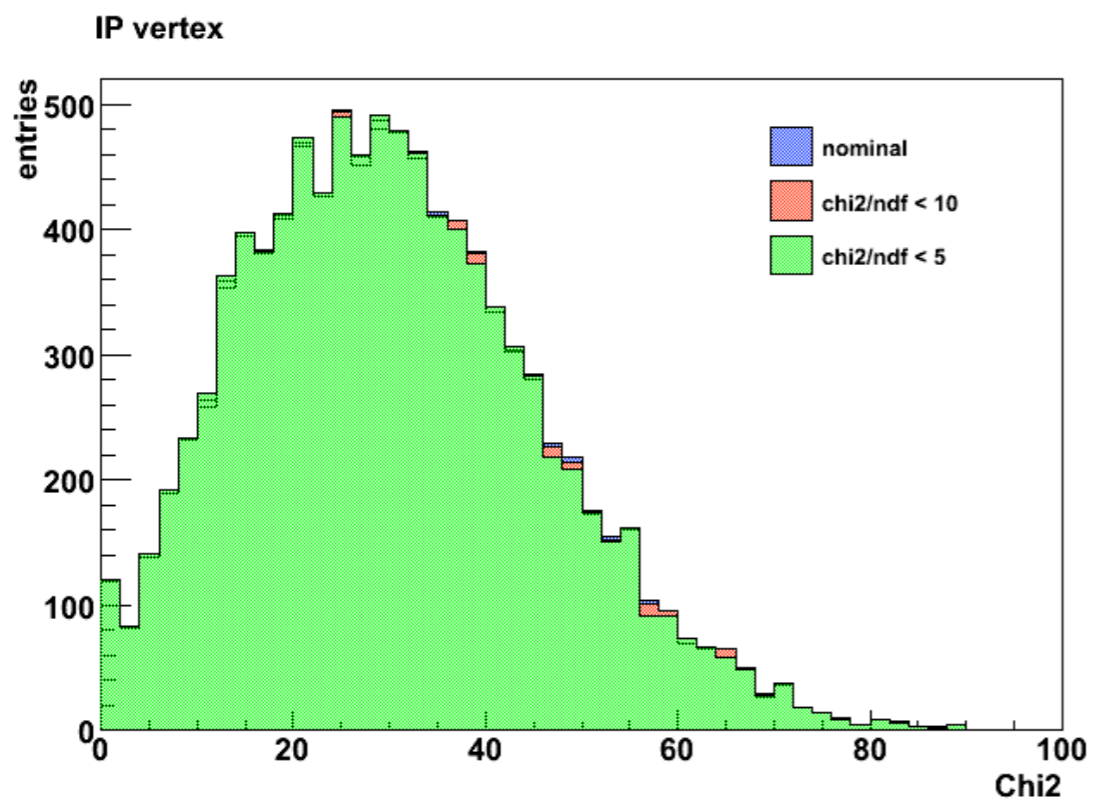
Description	xml parameter names	Code default	ipfit.xml	zvres.xml	fti.xml
Cut on χ^2 /ndf of track fit	a1_Chi2OverDOFEnable a2_Chi2OverDOFCutLowerThan a3_Chi2OverDOFCutValue	10	10 X	10 X	10 X
Cut on d0 (R ϕ impact parameter)	b1_D0Enable b2_D0CutLowerThan b3_D0CutValue	20	50 \checkmark (mm)	10 \checkmark (mm)	20 \checkmark (mm)
Cut on d0 error	c1_D0ErrEnable c2_D0ErrCutLowerThan c3_D0ErrCutValue	0.25	0.025 X (mm)	0.25 \checkmark (mm)	0.025 X (mm)
Cut on z impact parameter	d1_Z0Enable d2_Z0CutLowerThan d3_Z0CutValue	20	50 \checkmark (mm)	20 \checkmark (mm)	20 \checkmark (mm)
Cut on error on z imp param	e1_Z0ErrEnable e2_Z0ErrCutLowerThan e3_Z0ErrCutValue	0.25	0.025 X (mm)	0.025 X (mm)	0.025 X (mm)
Cut on pT of track	f1_PTEnable f2_PTCutLowerThan f3_PTCutValue	0.1	0.1 X (GeV/c)	0.1 \checkmark (GeV/c)	0.1 \checkmark (GeV/c)
cut on Ks, Λ decay tracks	h1_MCPIDEnable h2_CutPIDS h3_MonteCarloLCRelationCollection	0	X	\checkmark +- 310 +- 3122	\checkmark +- 310 +- 3122

X: disabled; \checkmark : enabled

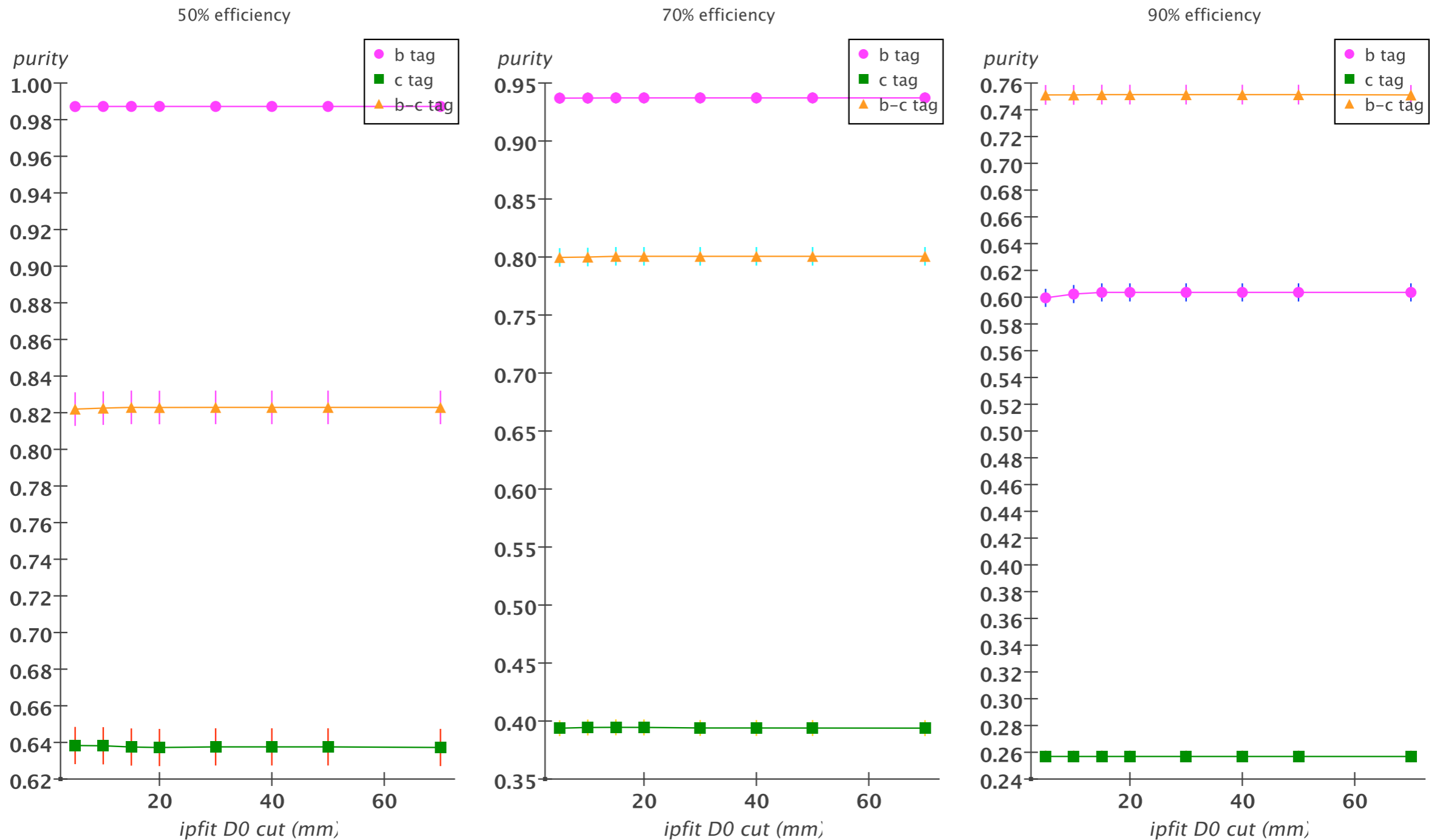
IPFIT: flavour tag purity as a function of the χ^2/ndf of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: no cut



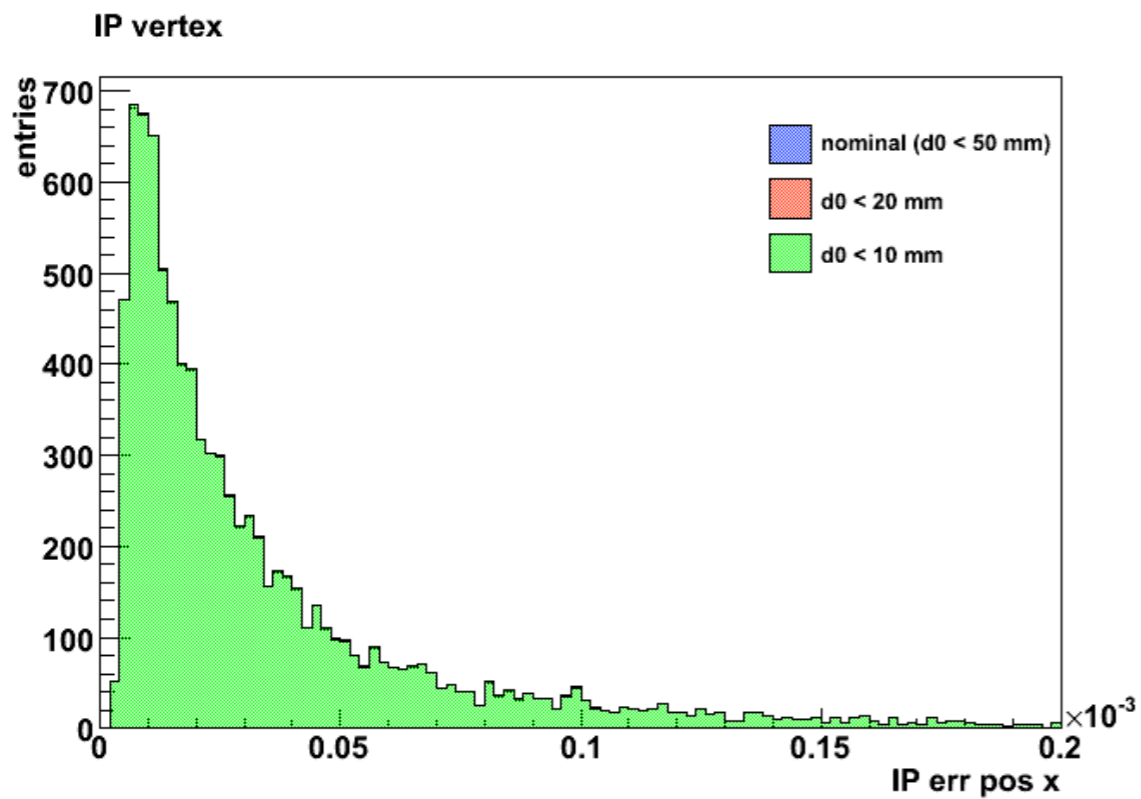
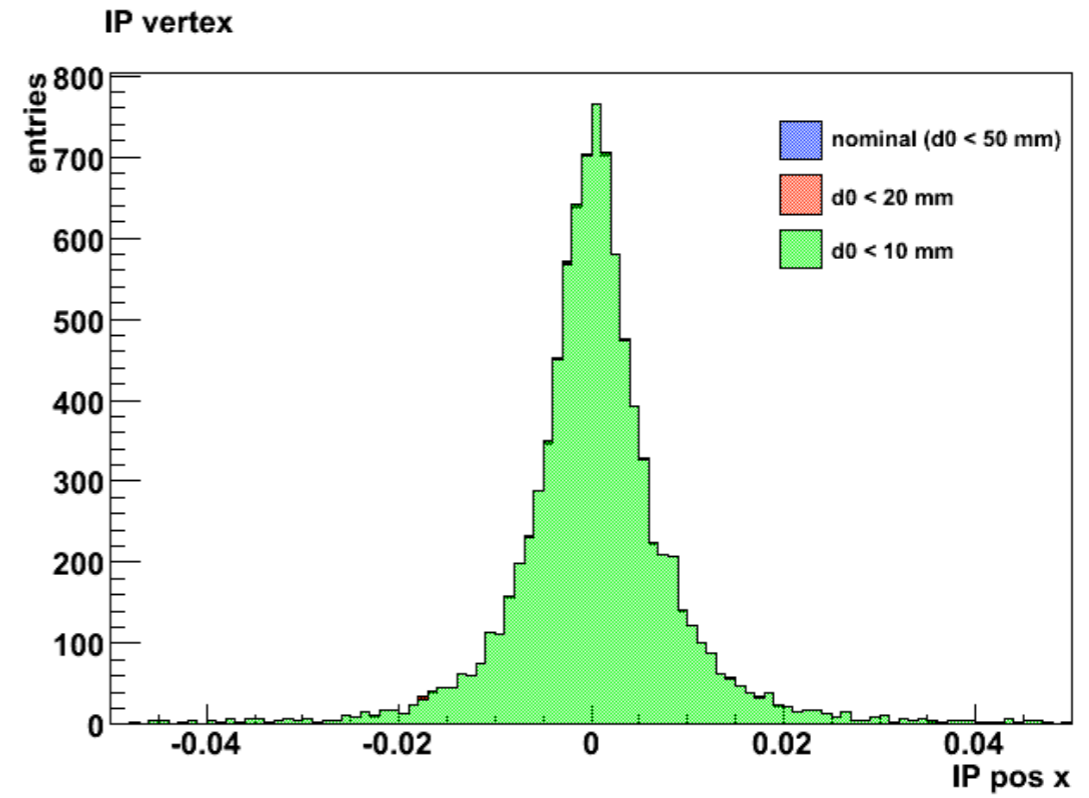
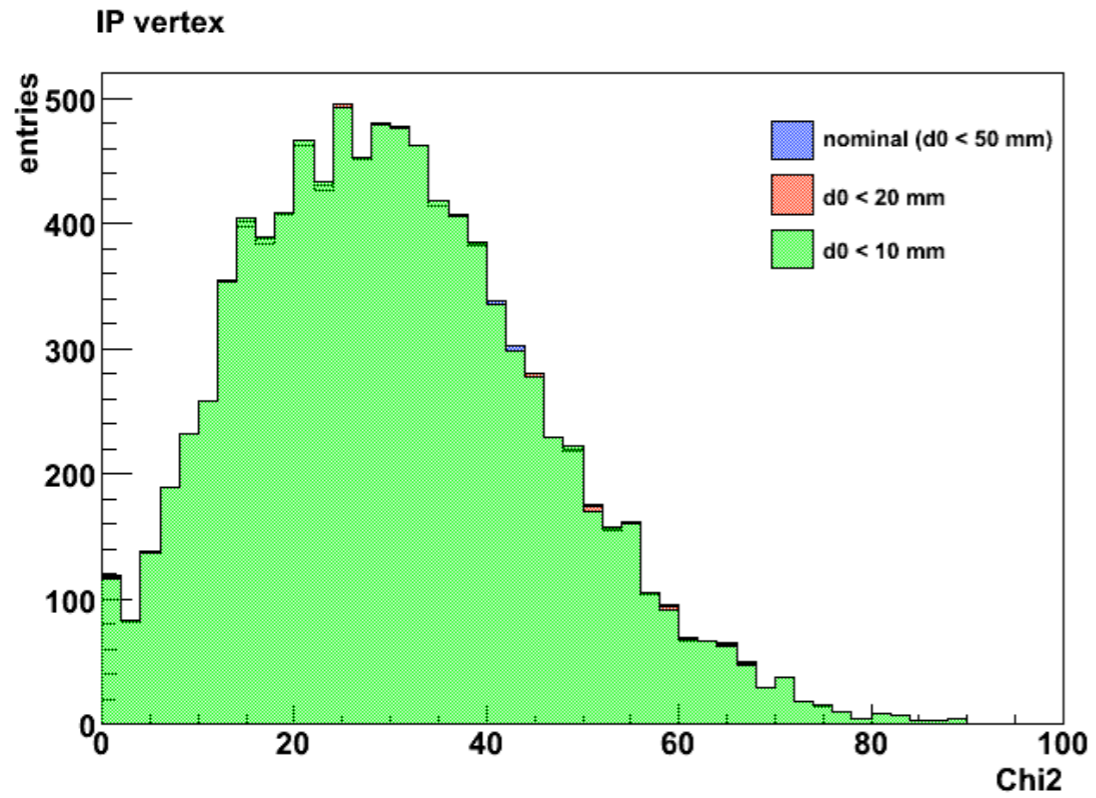
IP vertex (chi2/ndf cut)



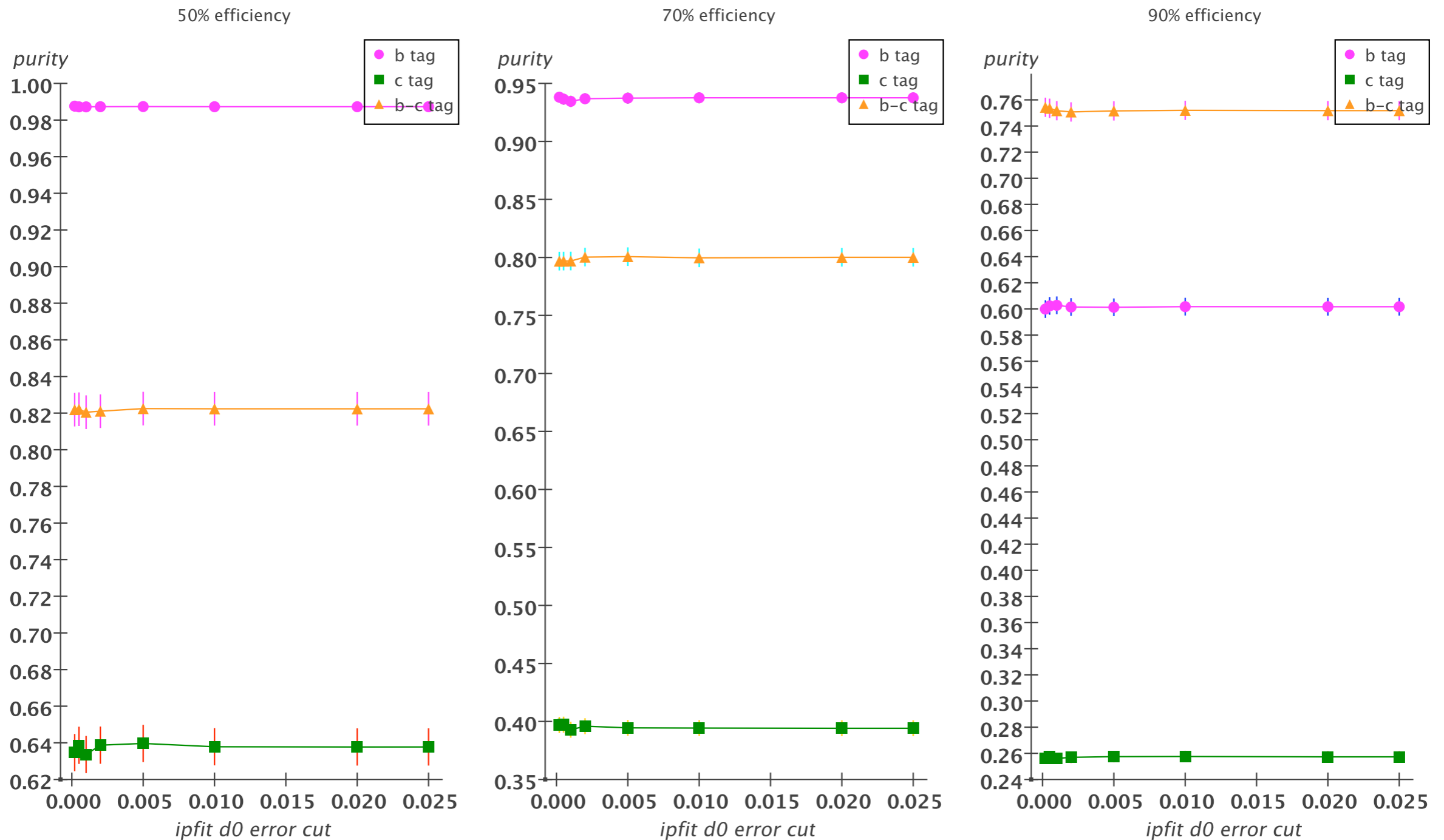
IPFIT: flavour tag purity as a function of the d_0 of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: 50mm



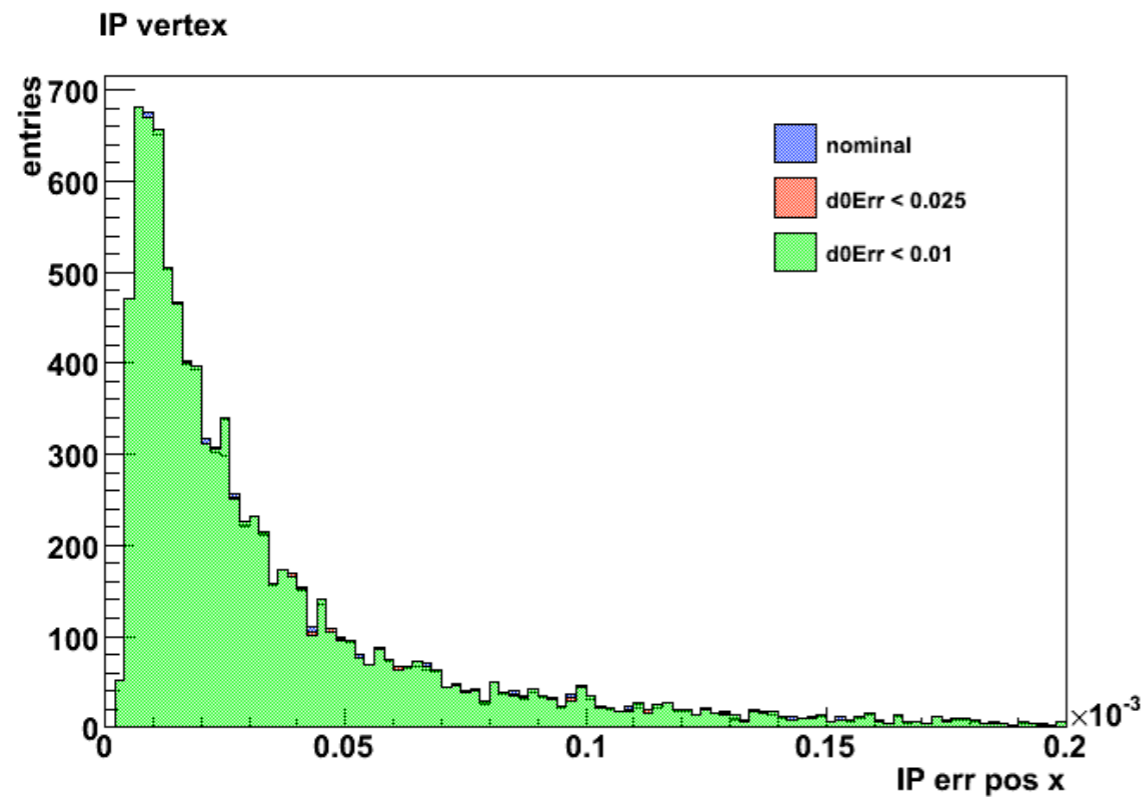
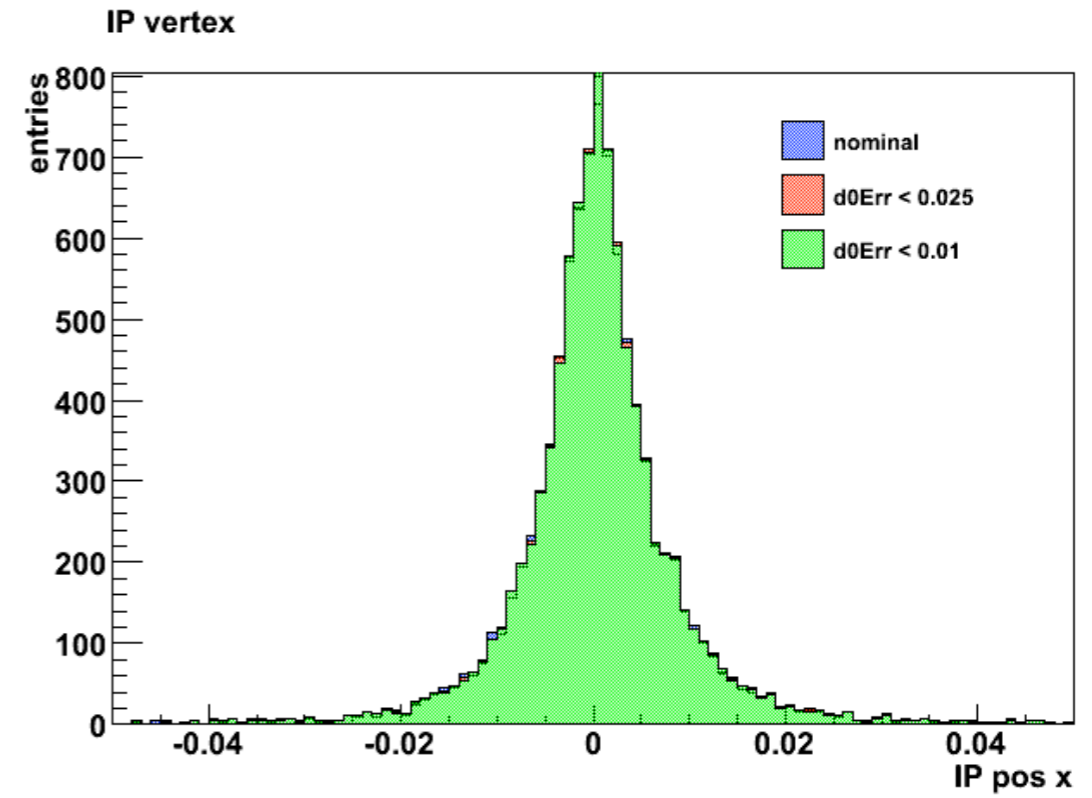
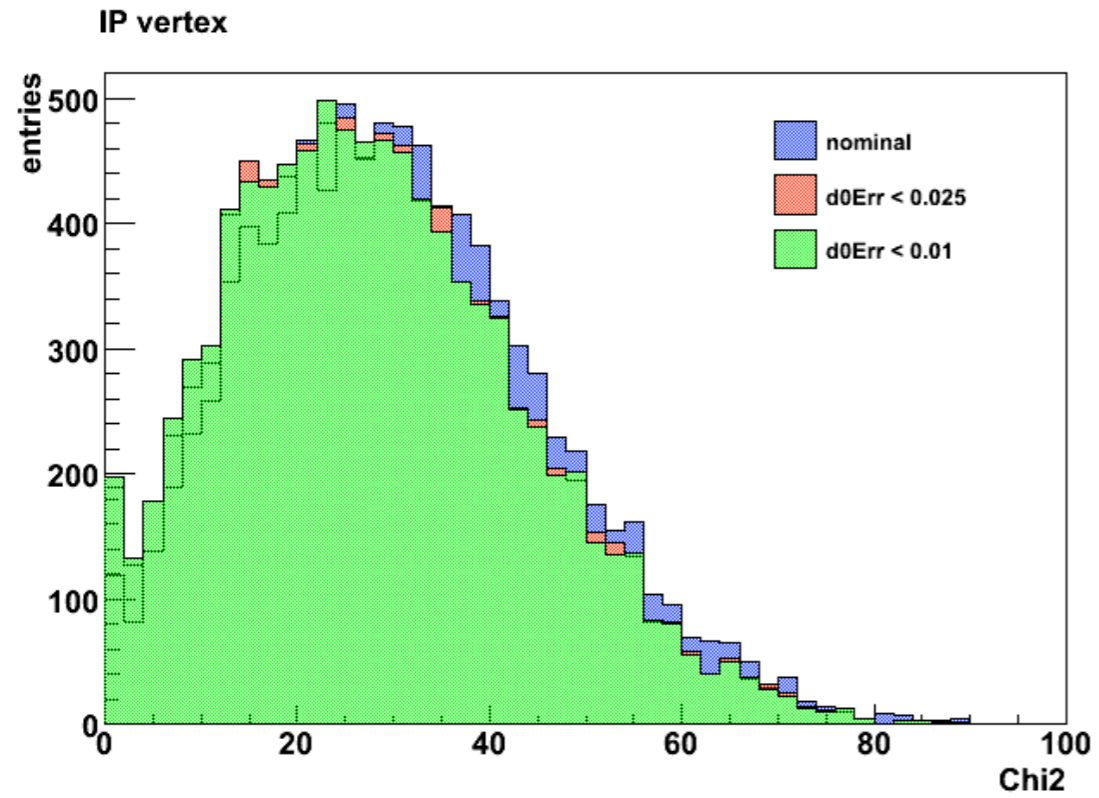
IP vertex (d0 cut)



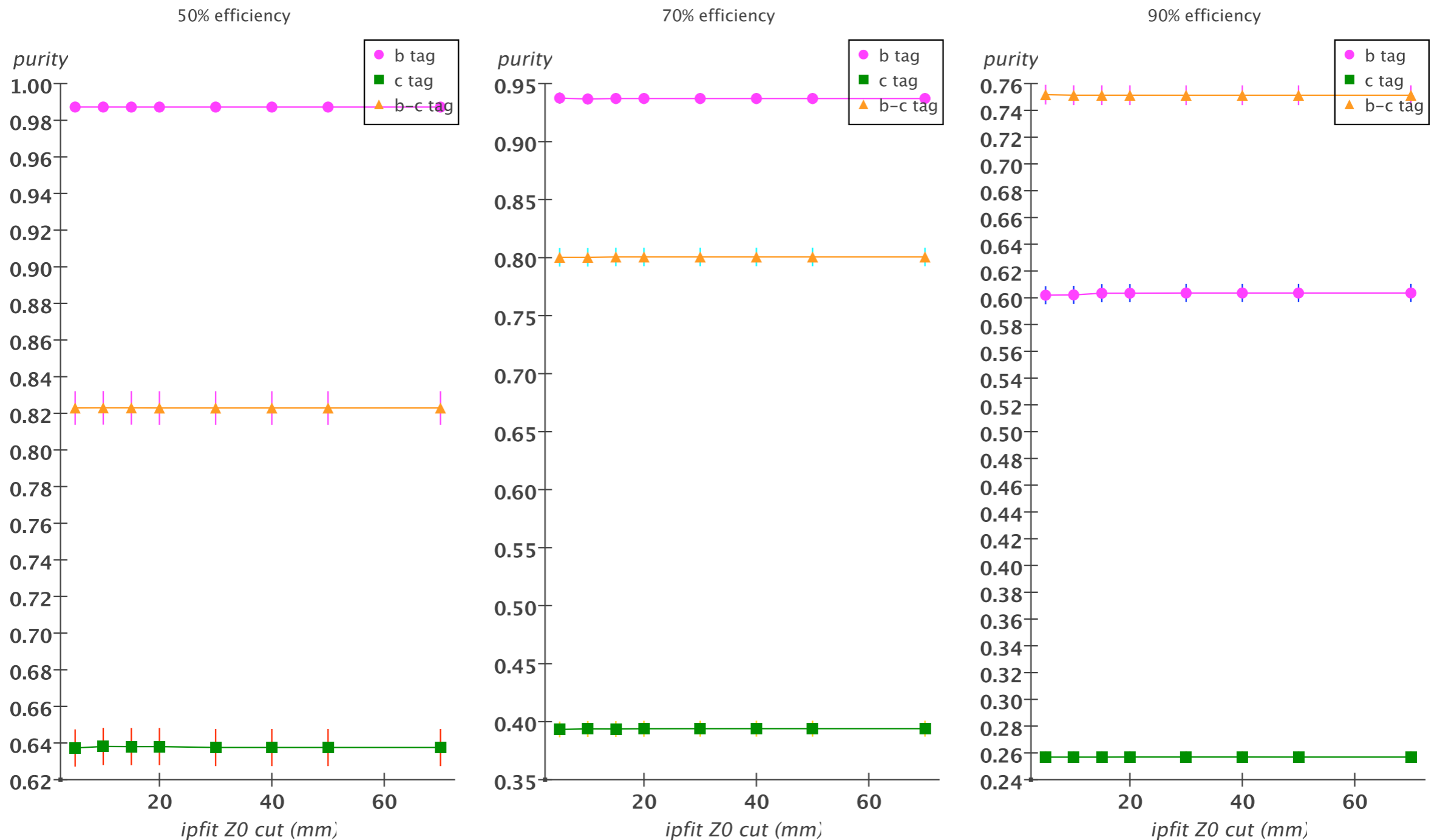
IPFIT: flavour tag purity as a function of the $d0err$ of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: no cut



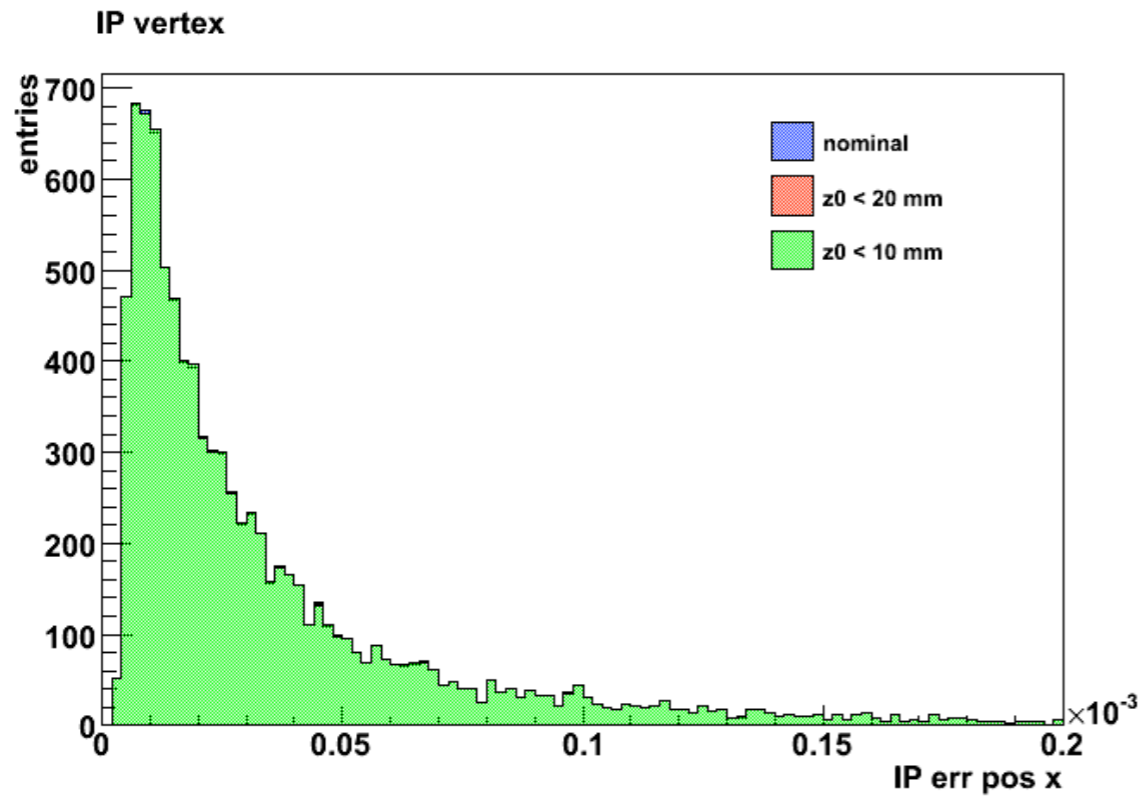
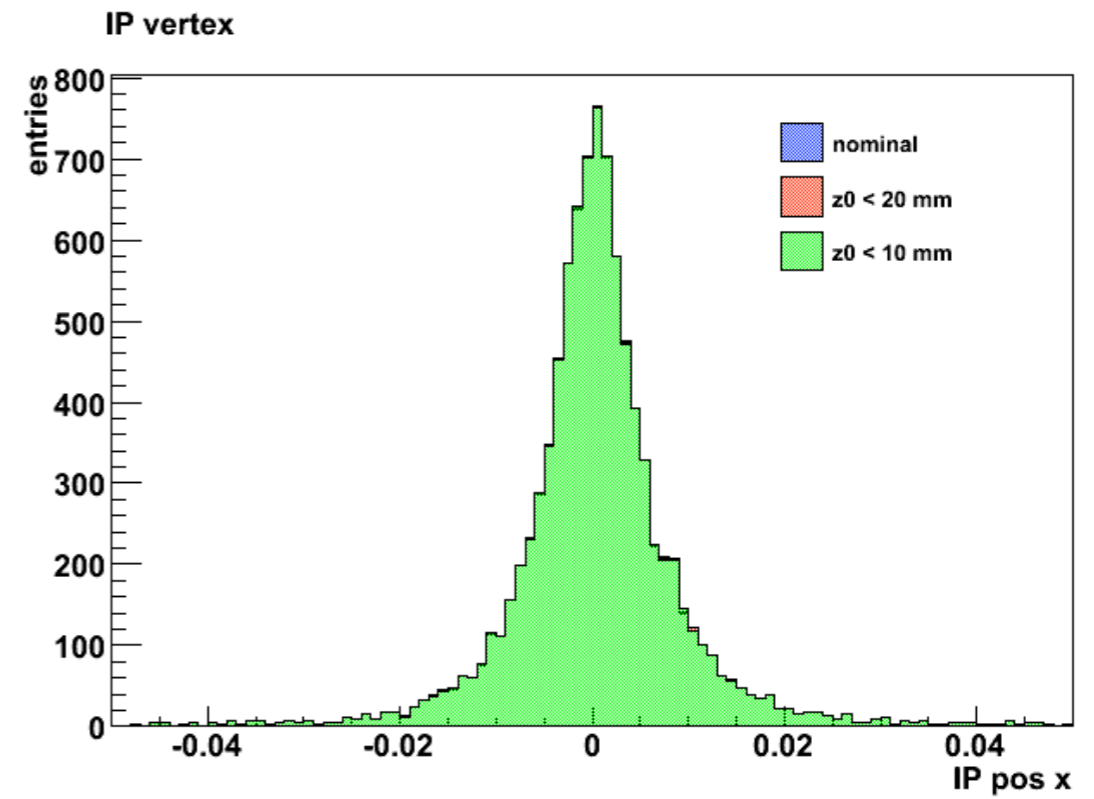
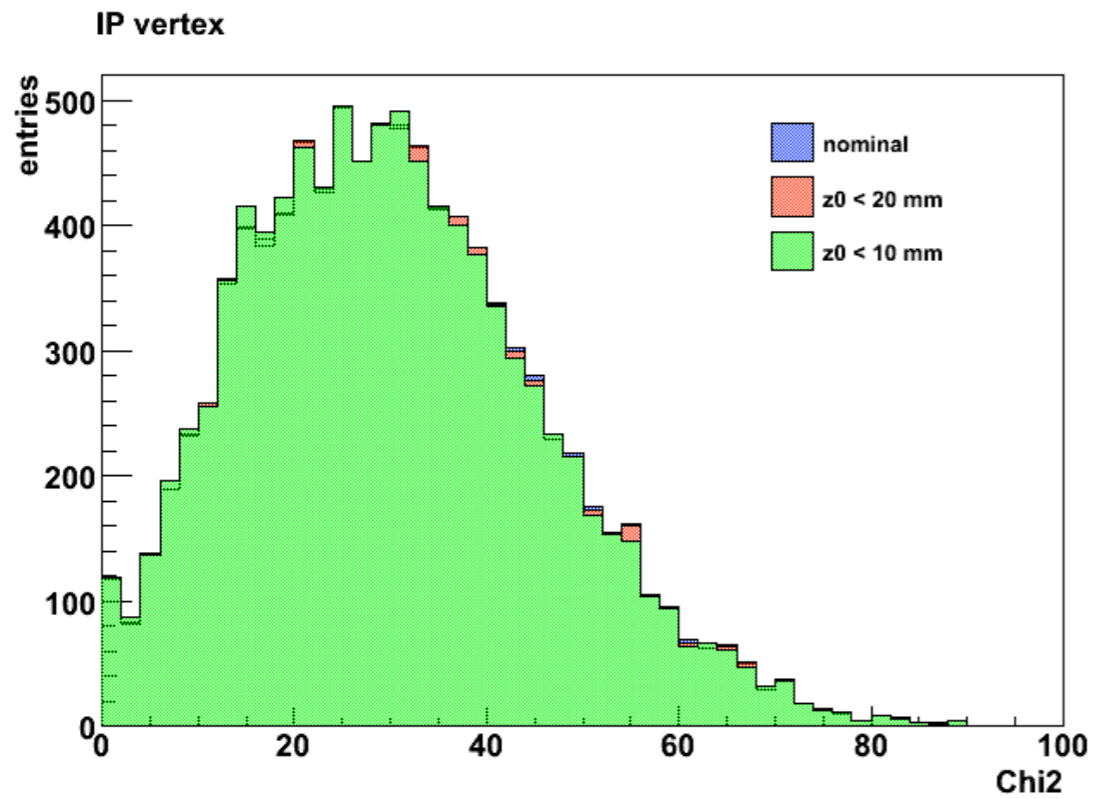
IP vertex (d0err cut)



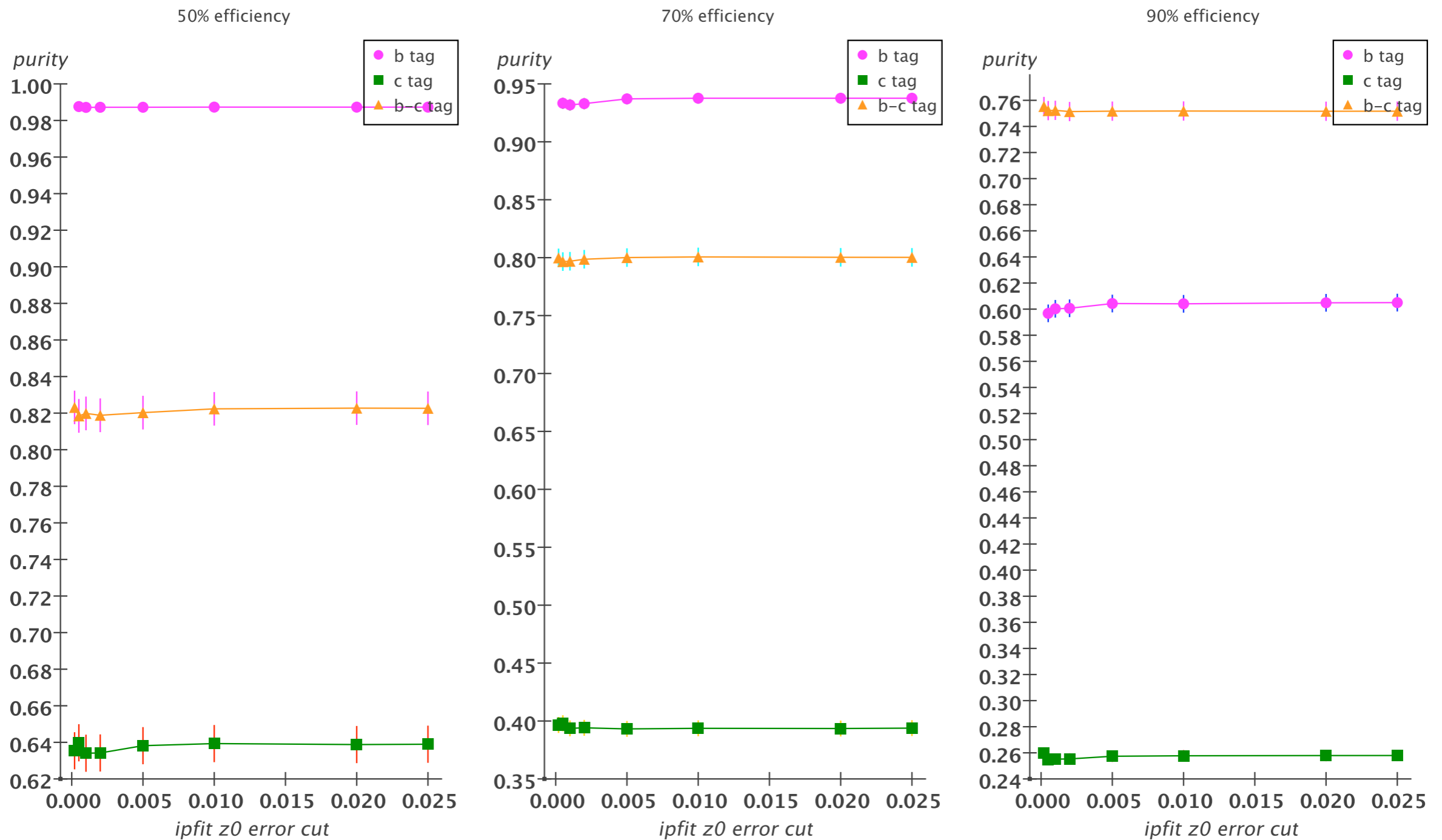
IPFIT: flavour tag purity as a function of the z_0 of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: 50 mm



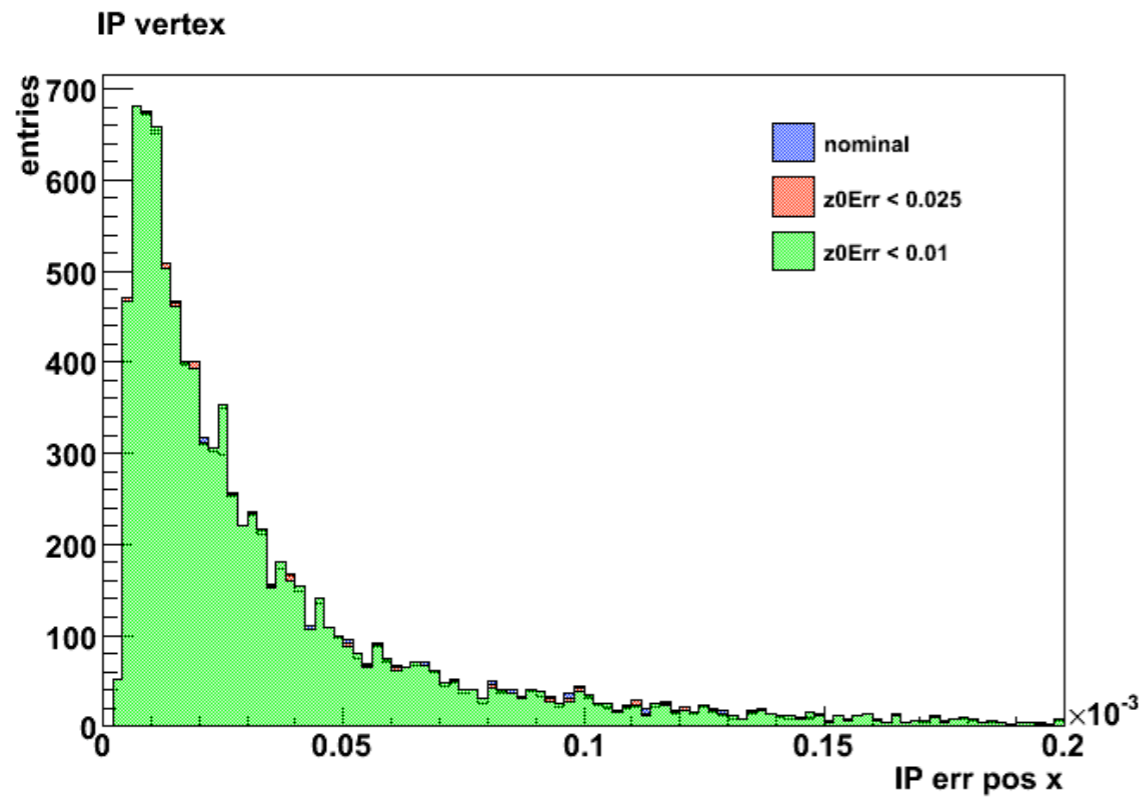
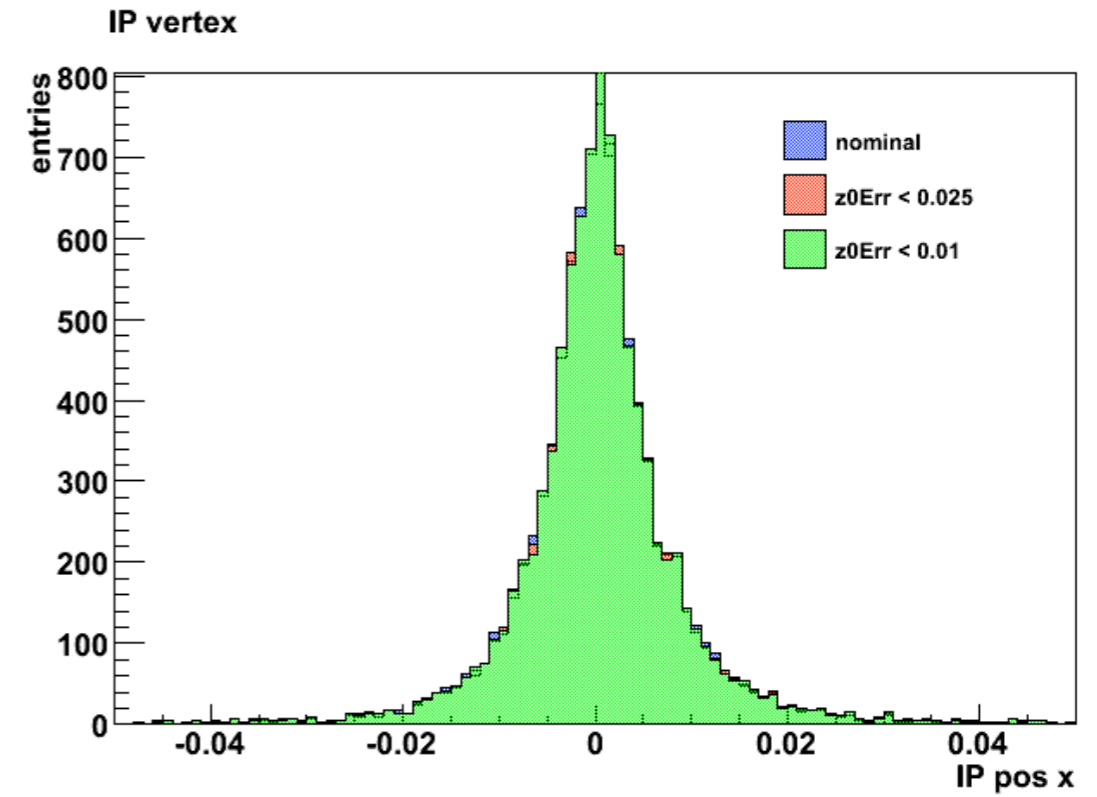
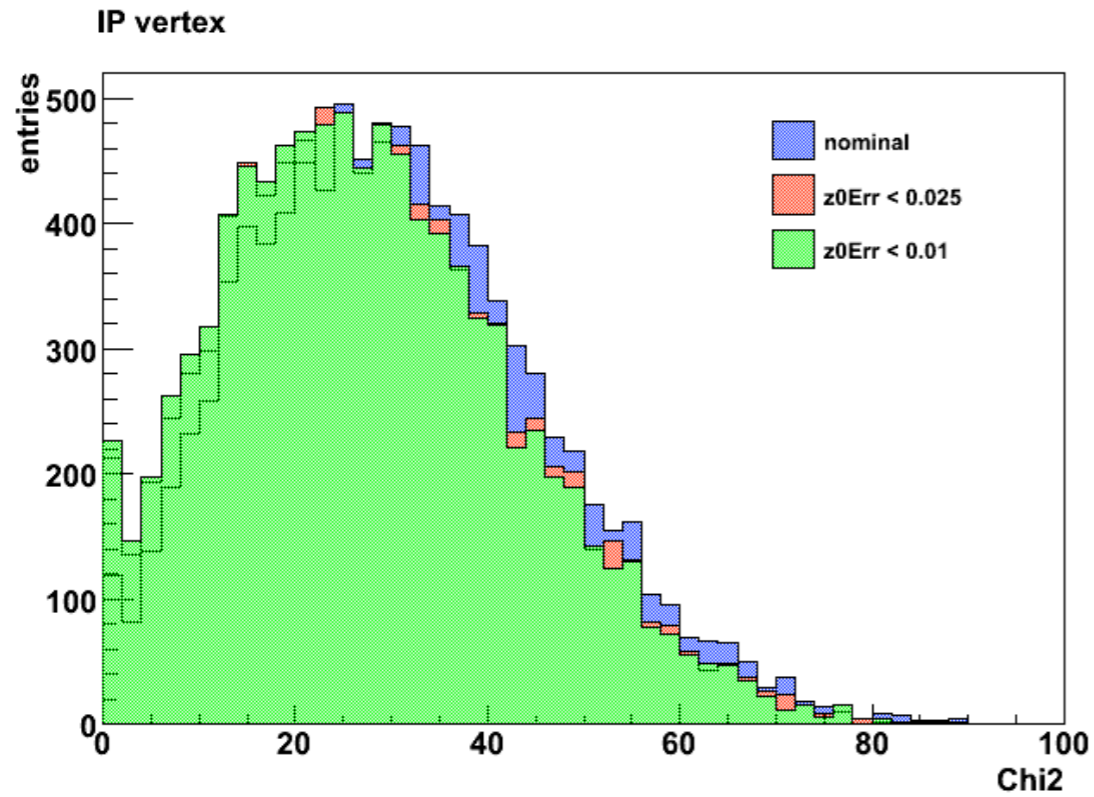
IP vertex (z0 cut)



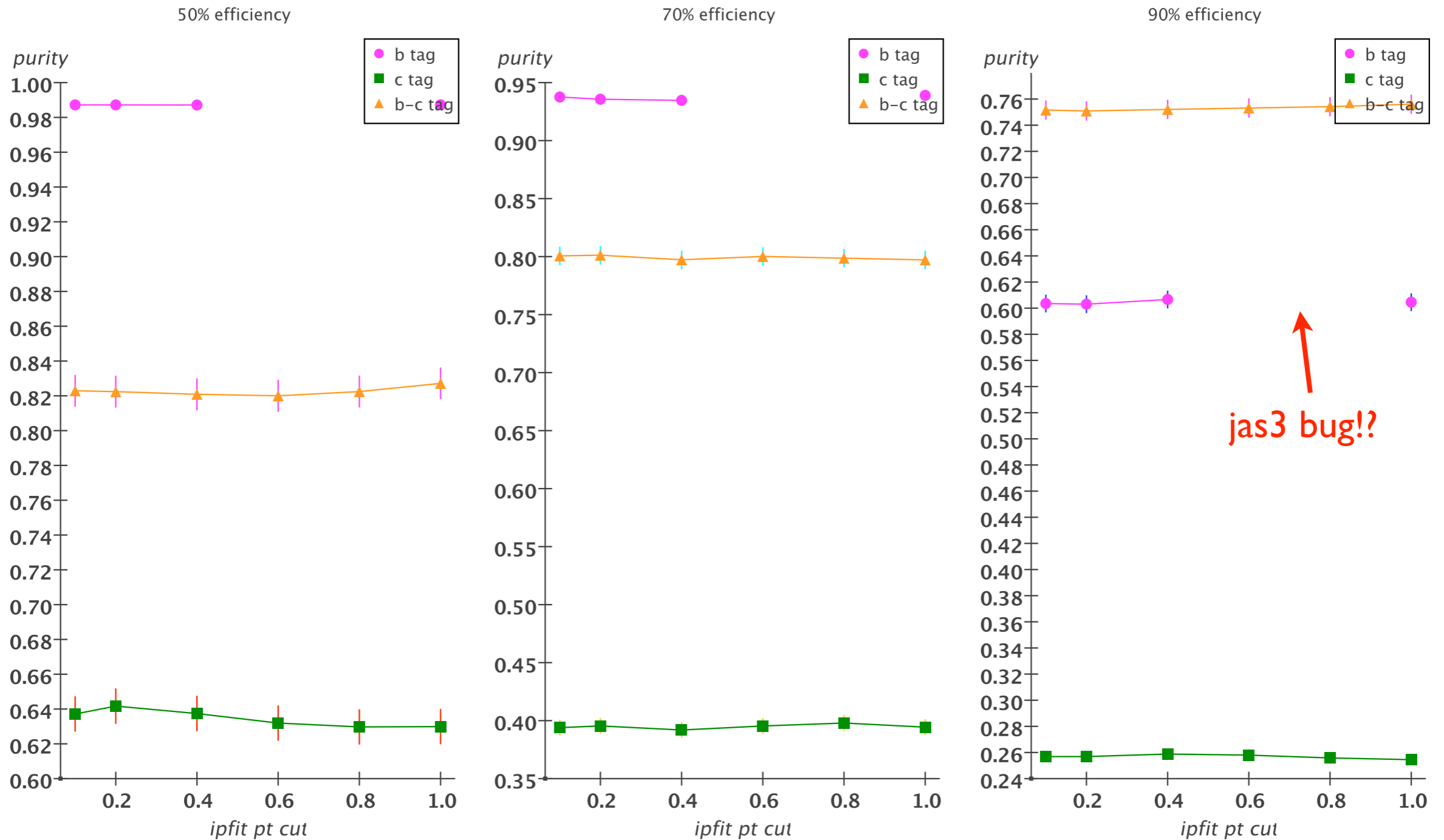
IPFIT: flavour tag purity as a function of the z_0 err of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: no cut



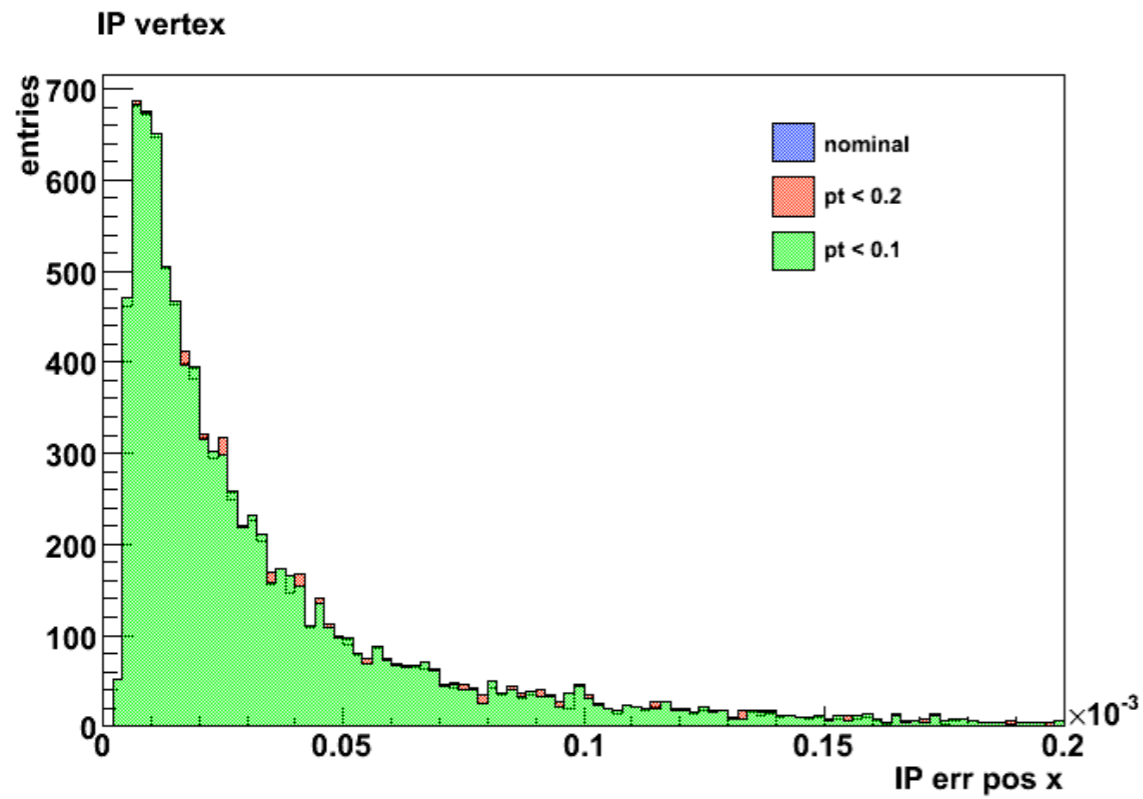
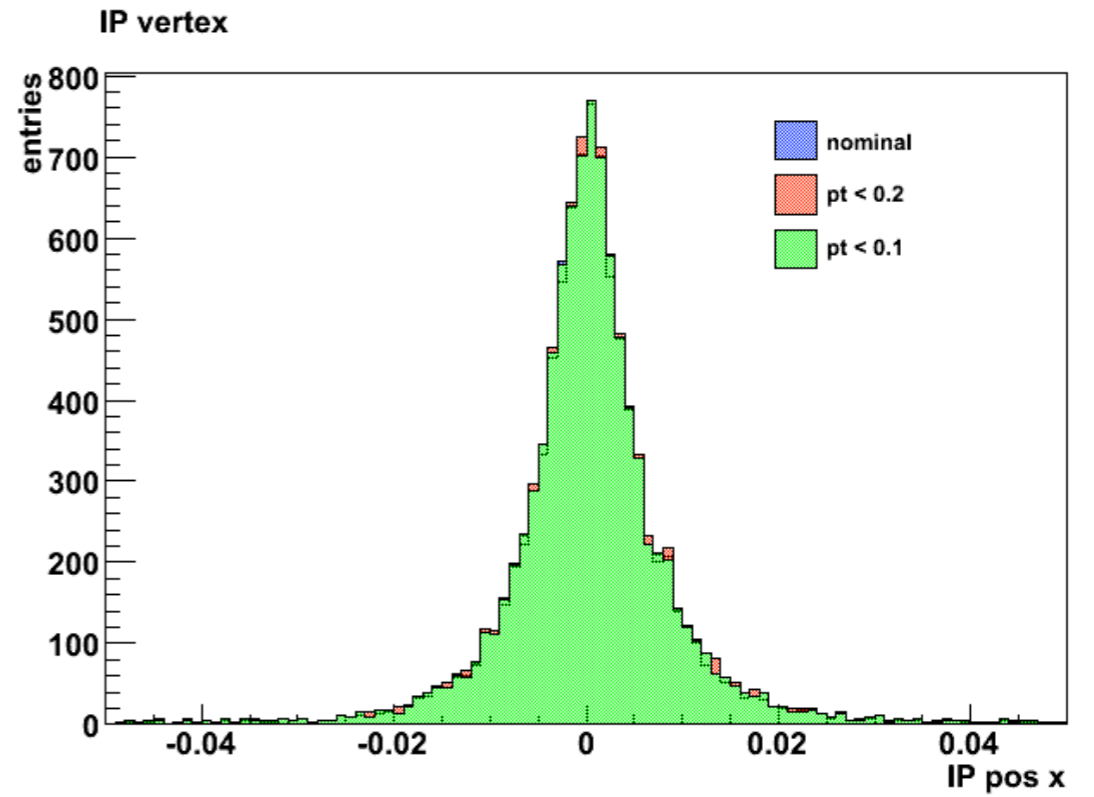
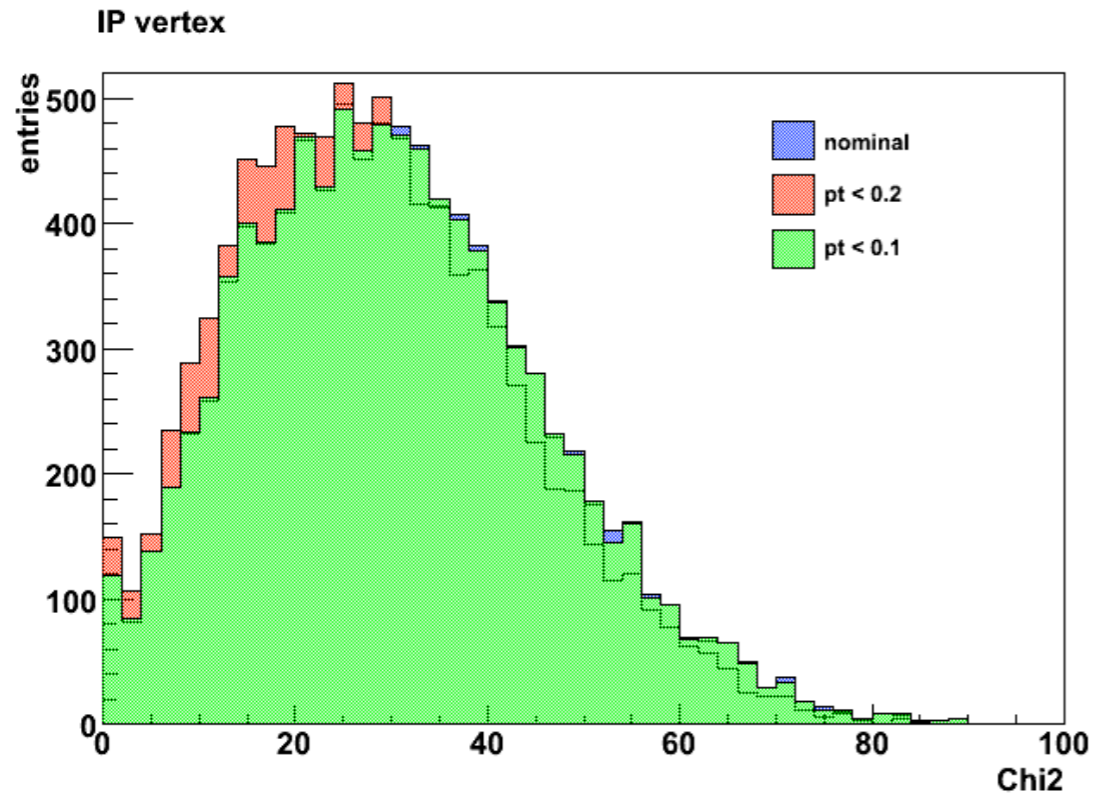
IP vertex (z0err cut)



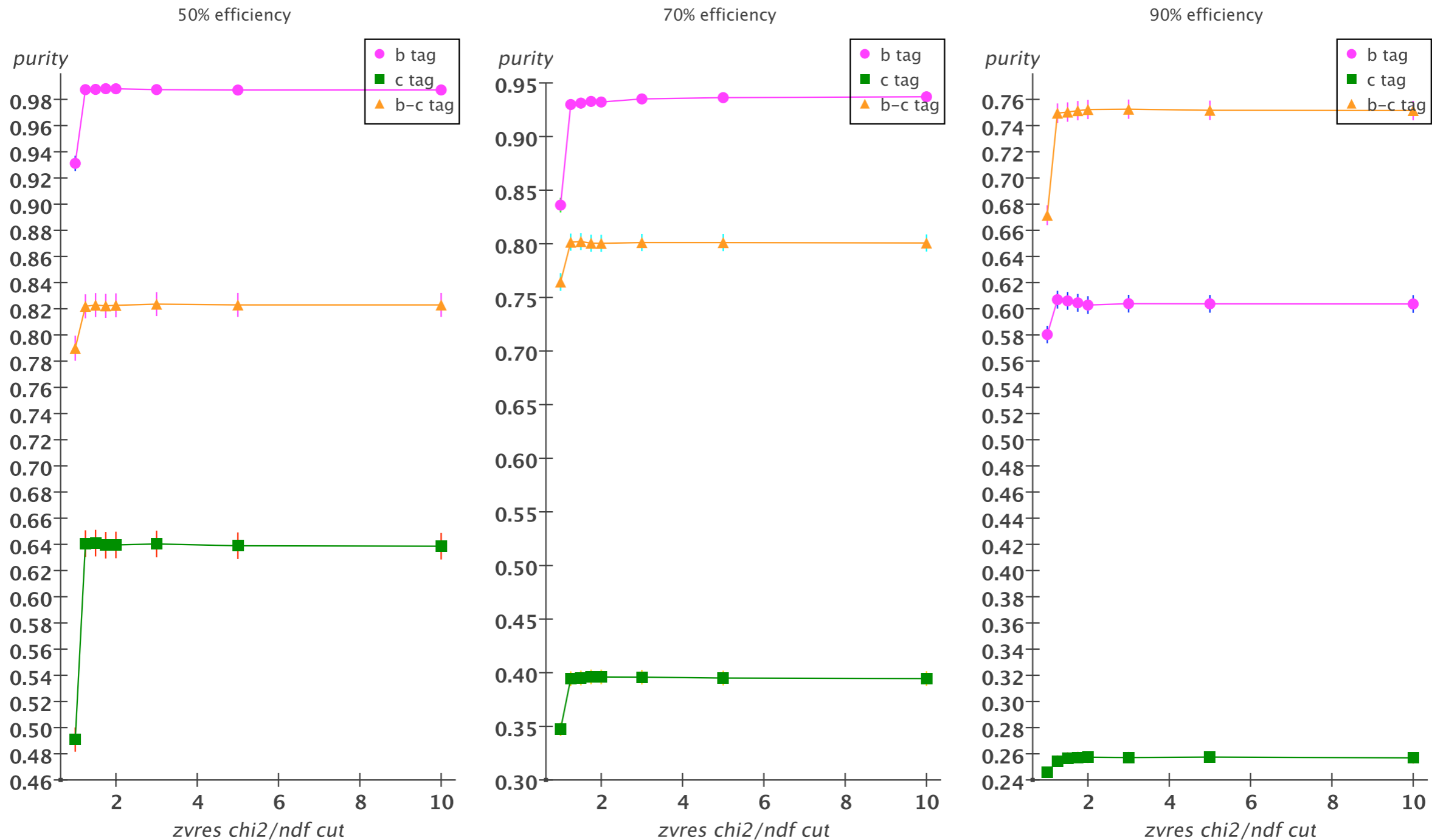
IPFIT: flavour tag purity as a function of the pt of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: no cut



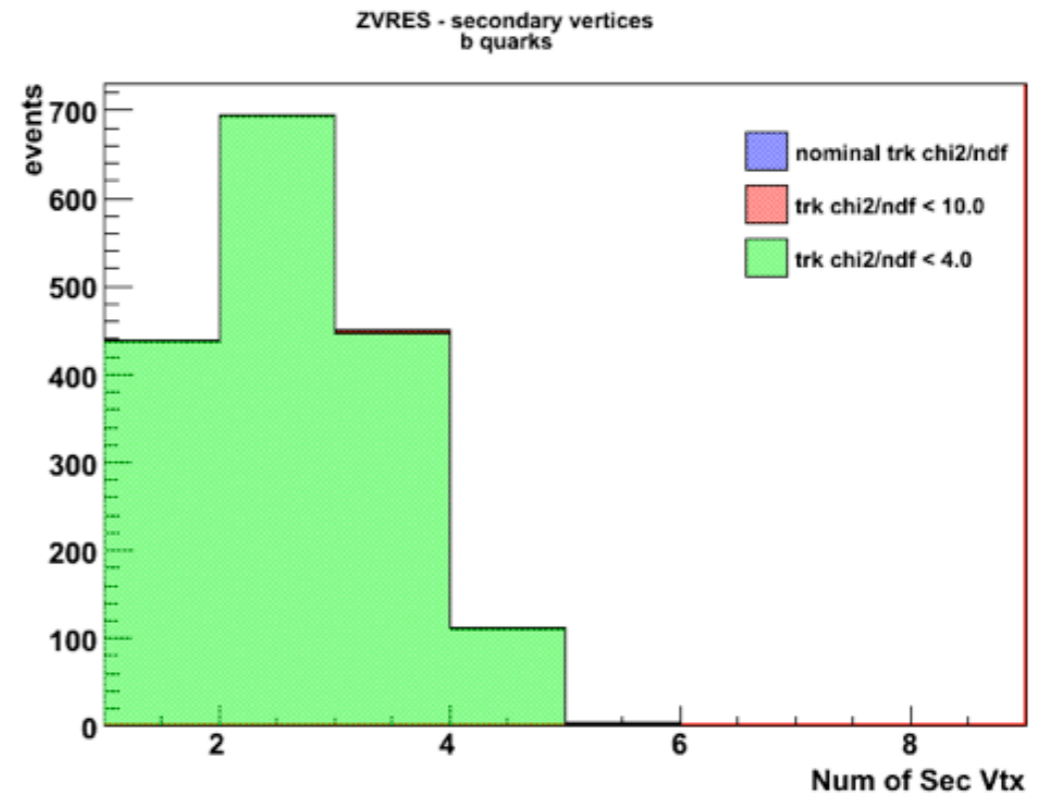
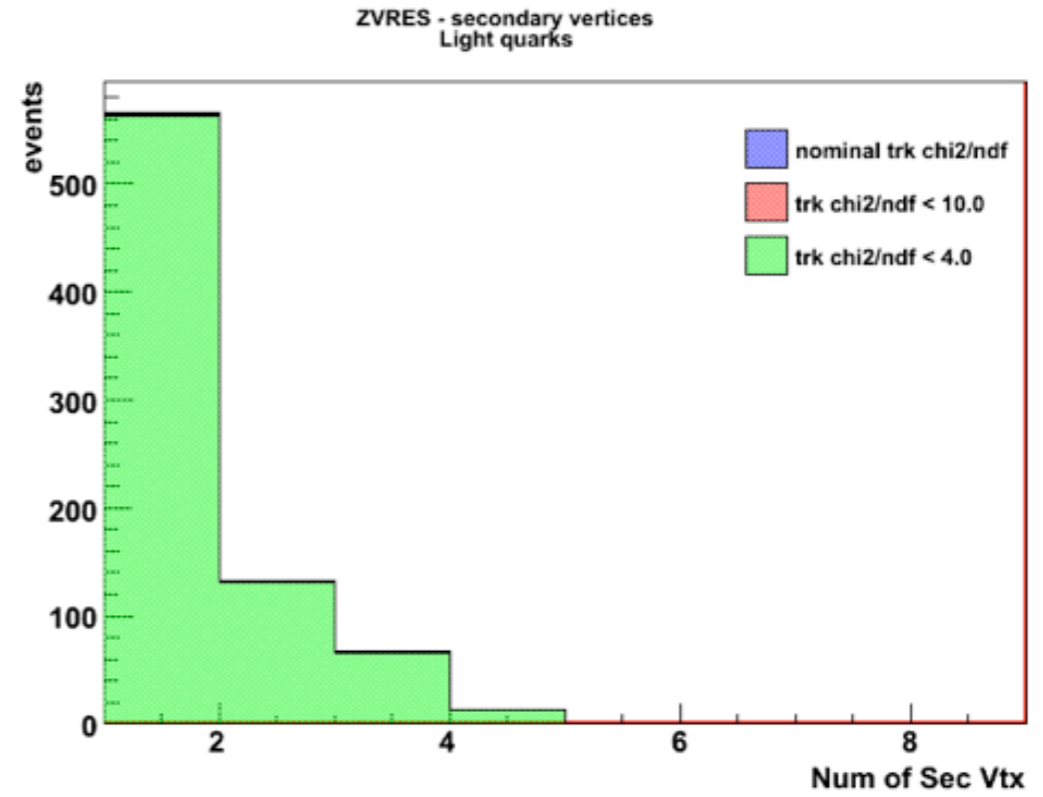
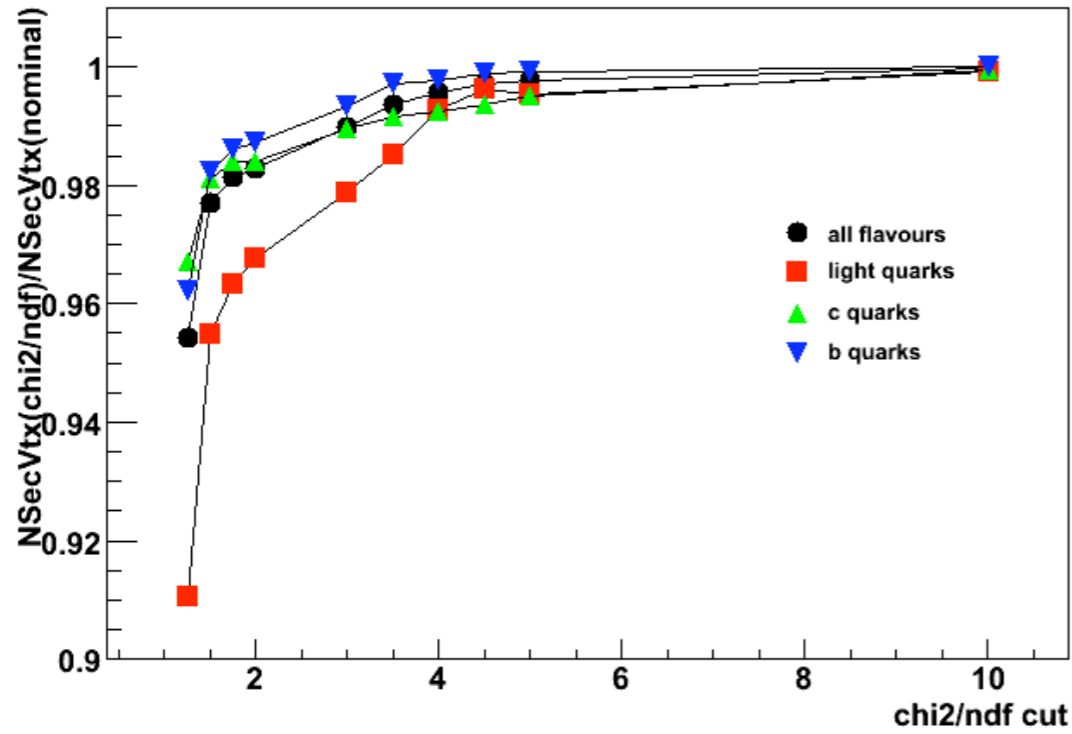
IP vertex (pt cut)



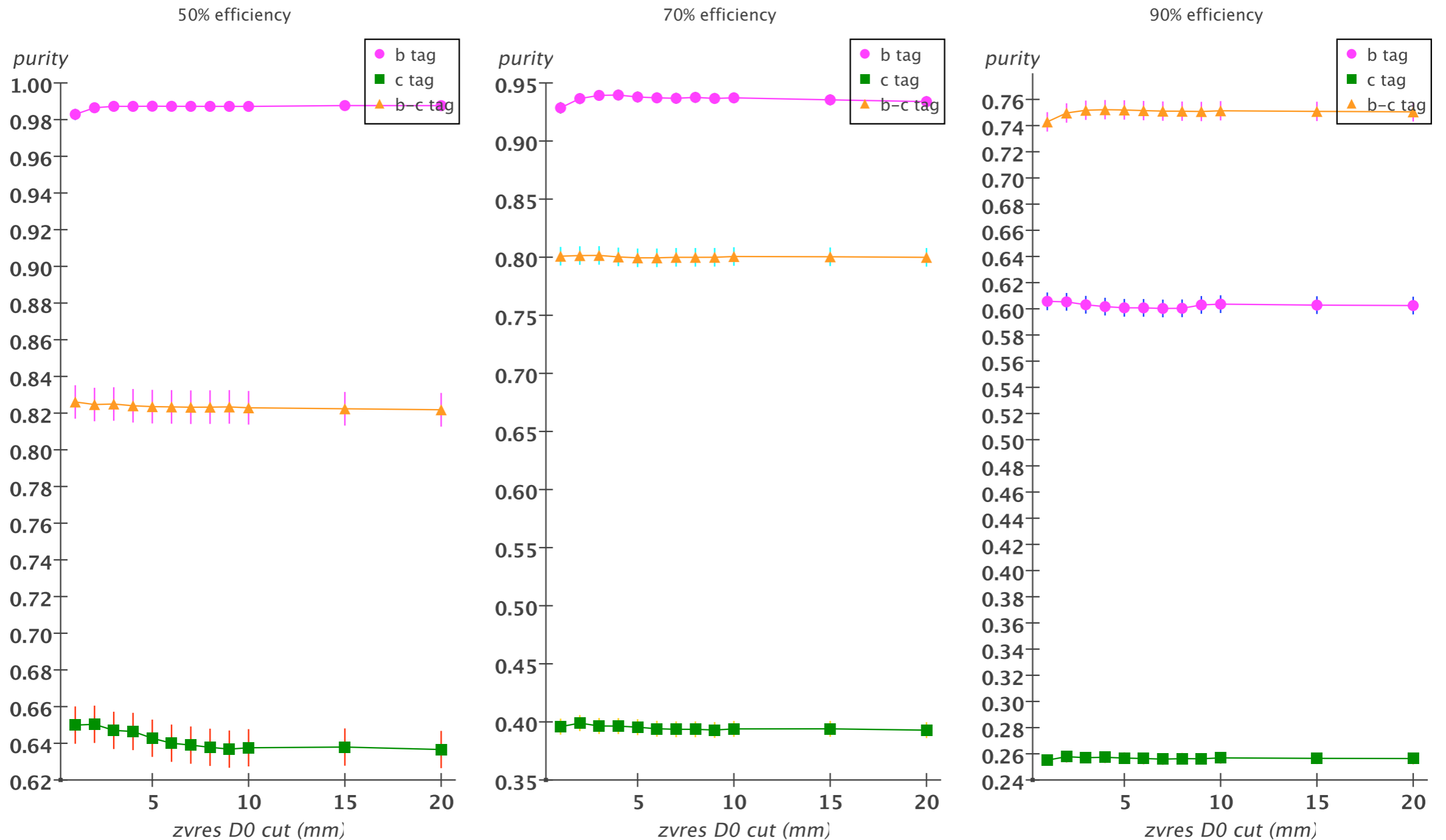
ZVRES: flavour tag purity as a function of the χ^2/ndf of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: no cut



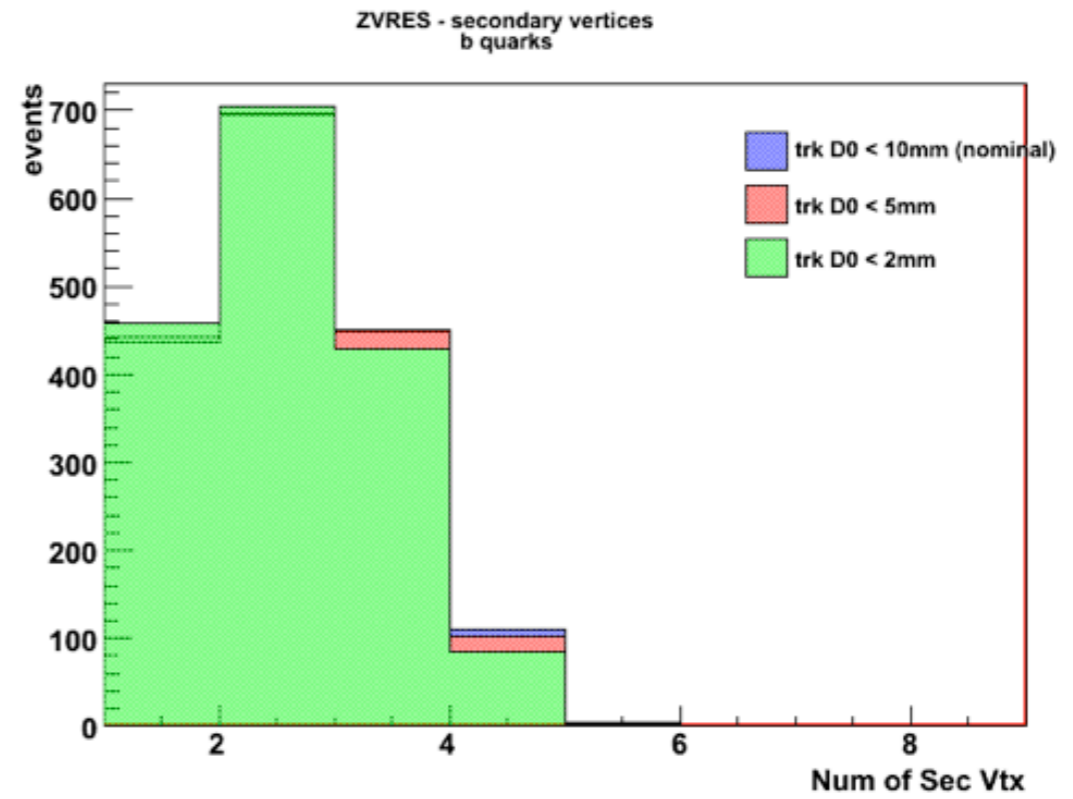
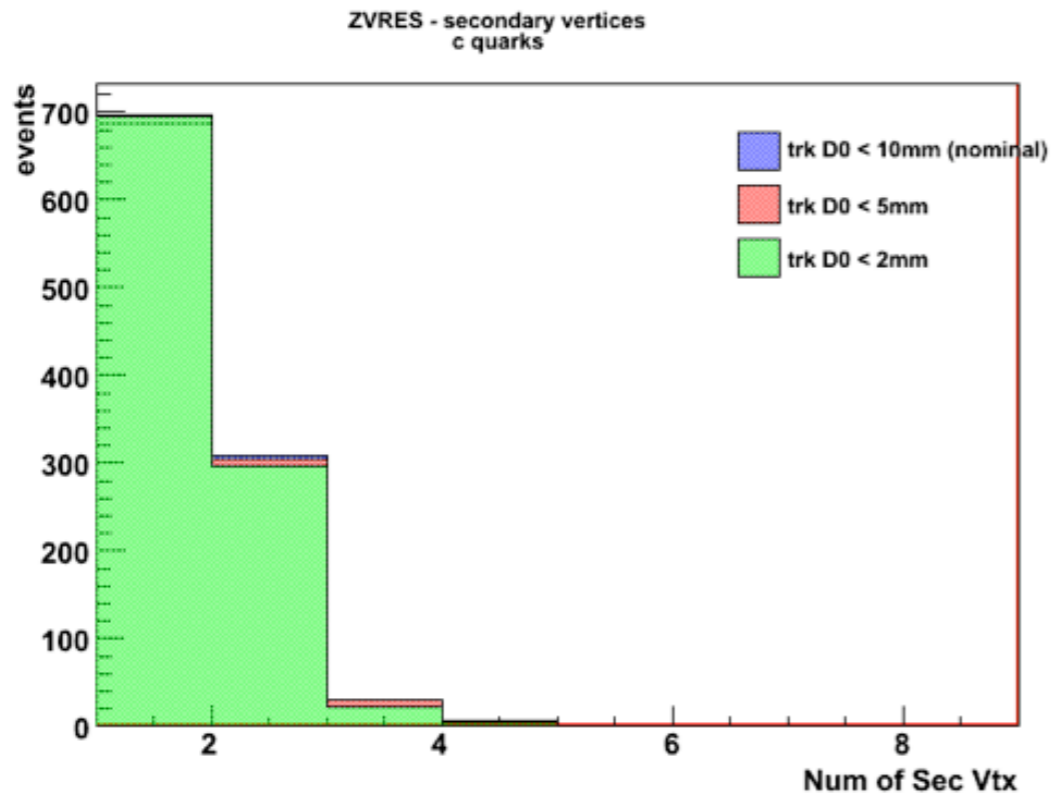
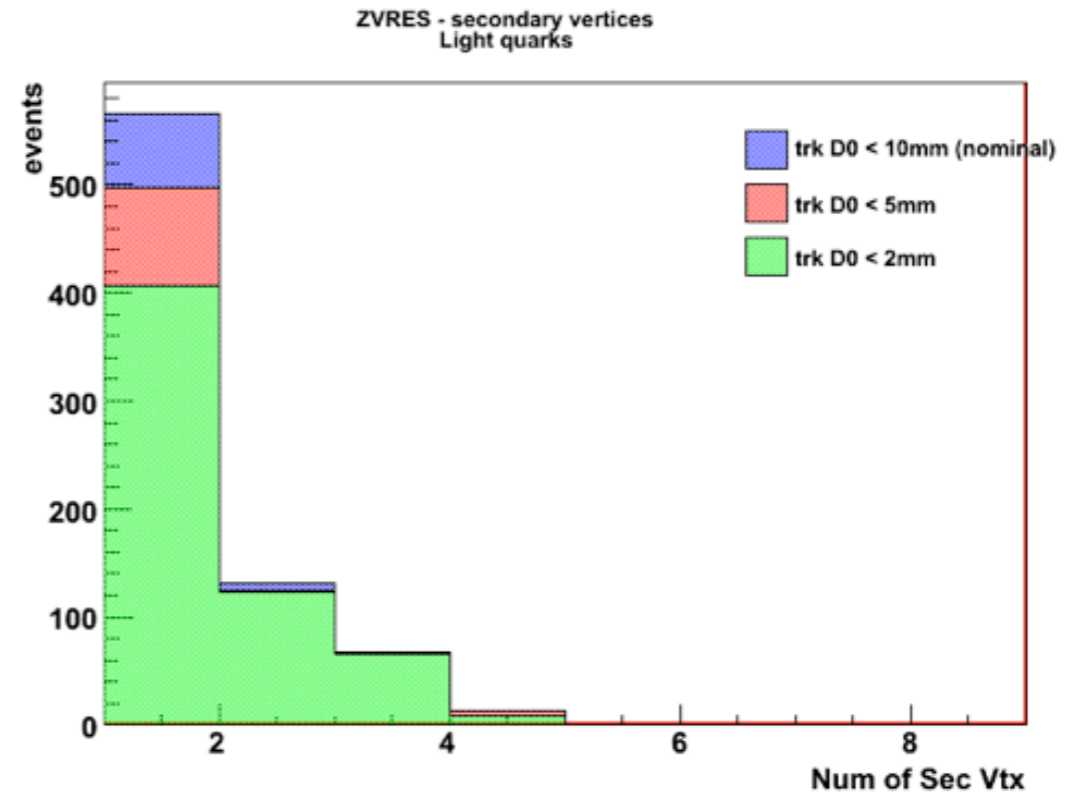
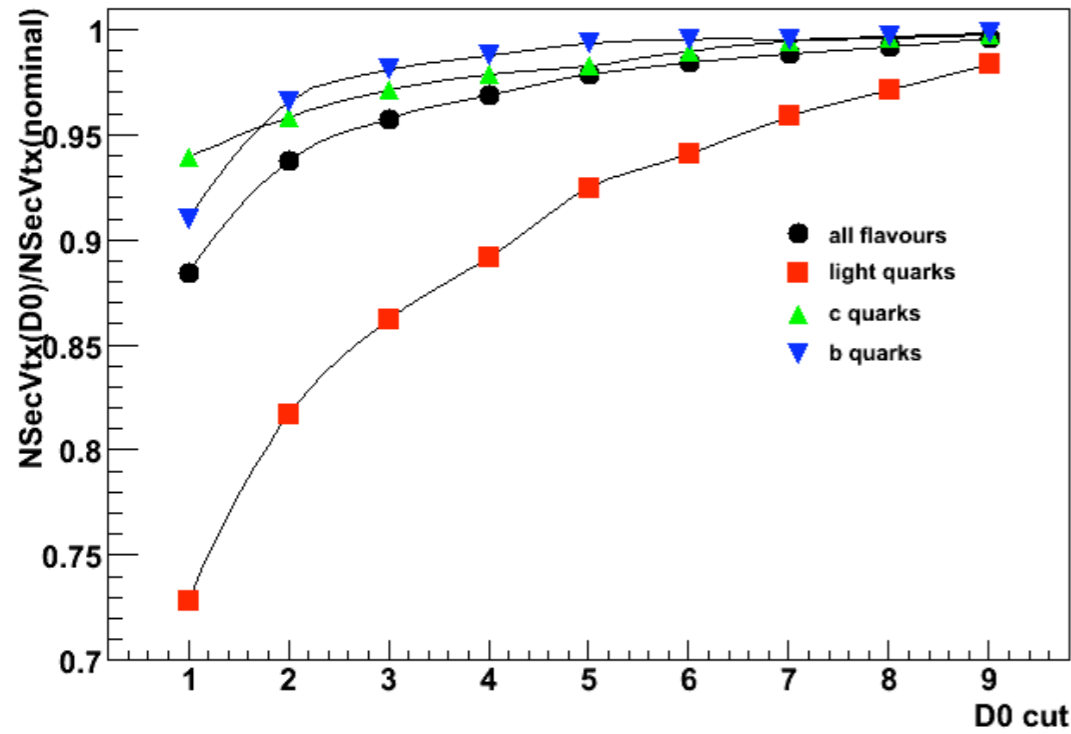
ZVRES secondary vertices (chi2/ndf cut)



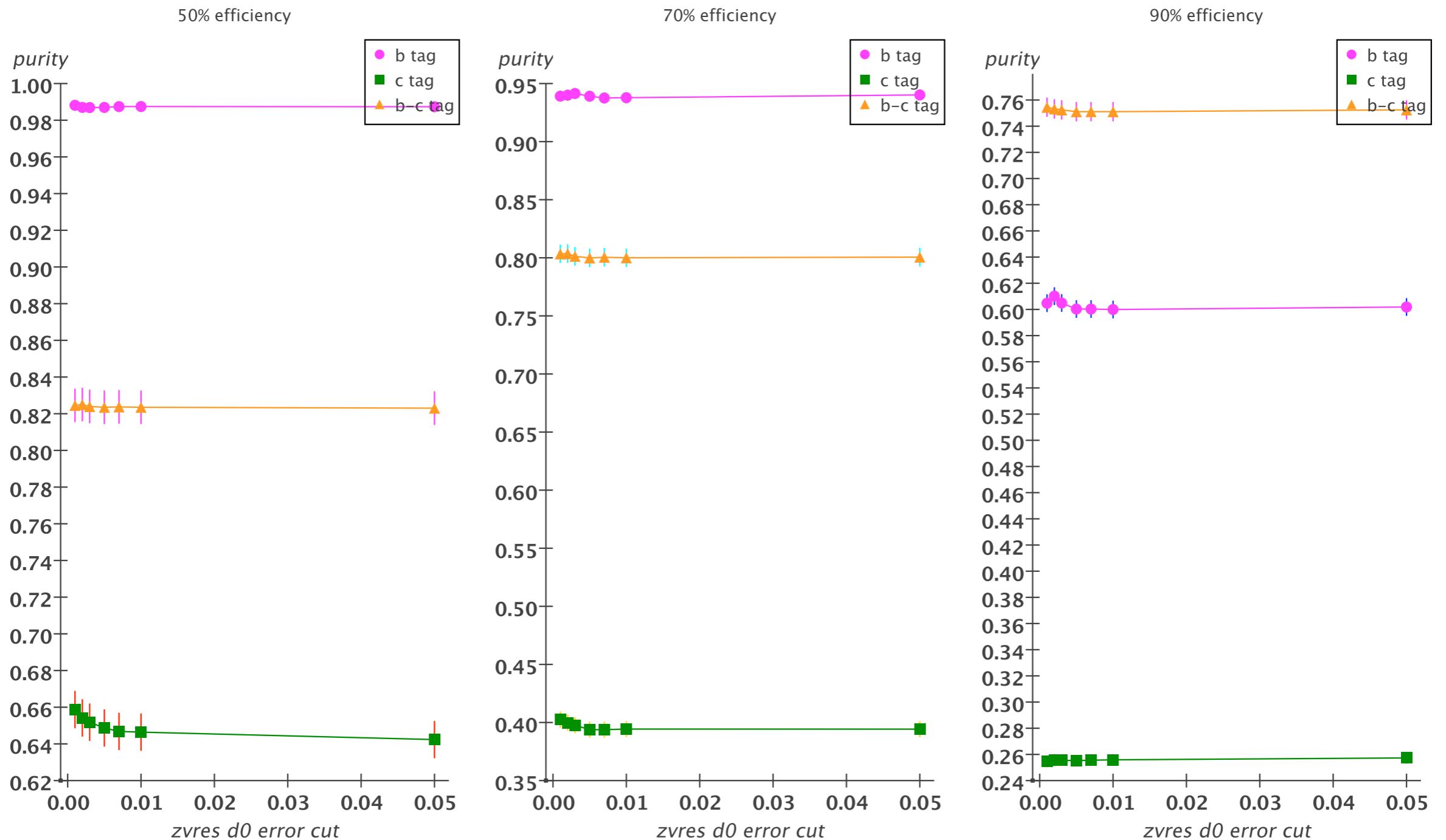
ZVRES: flavour tag purity as a function of the d_0 of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: 10mm



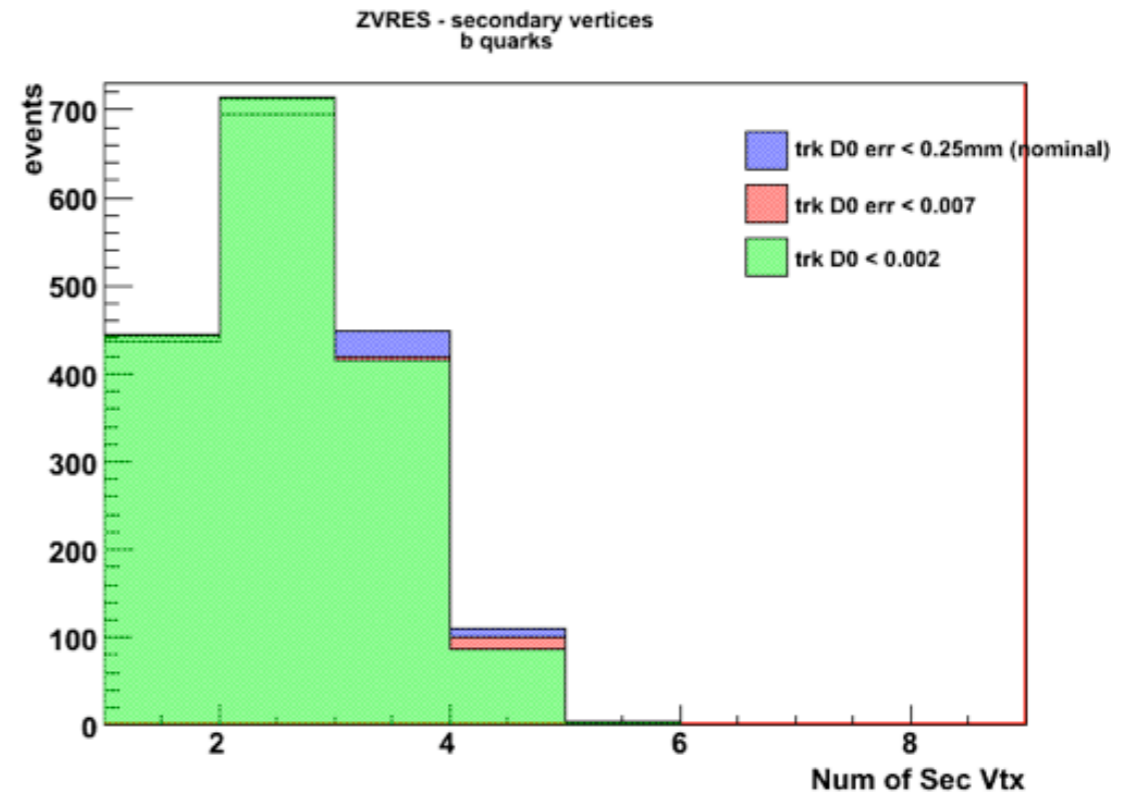
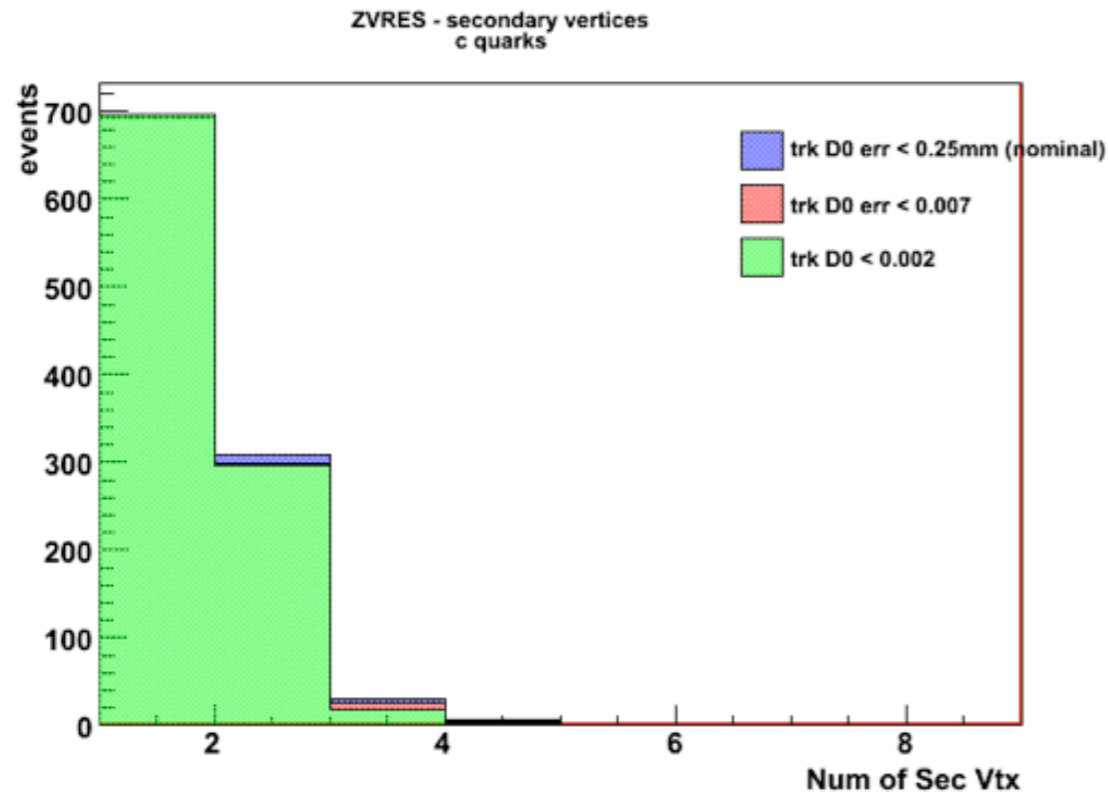
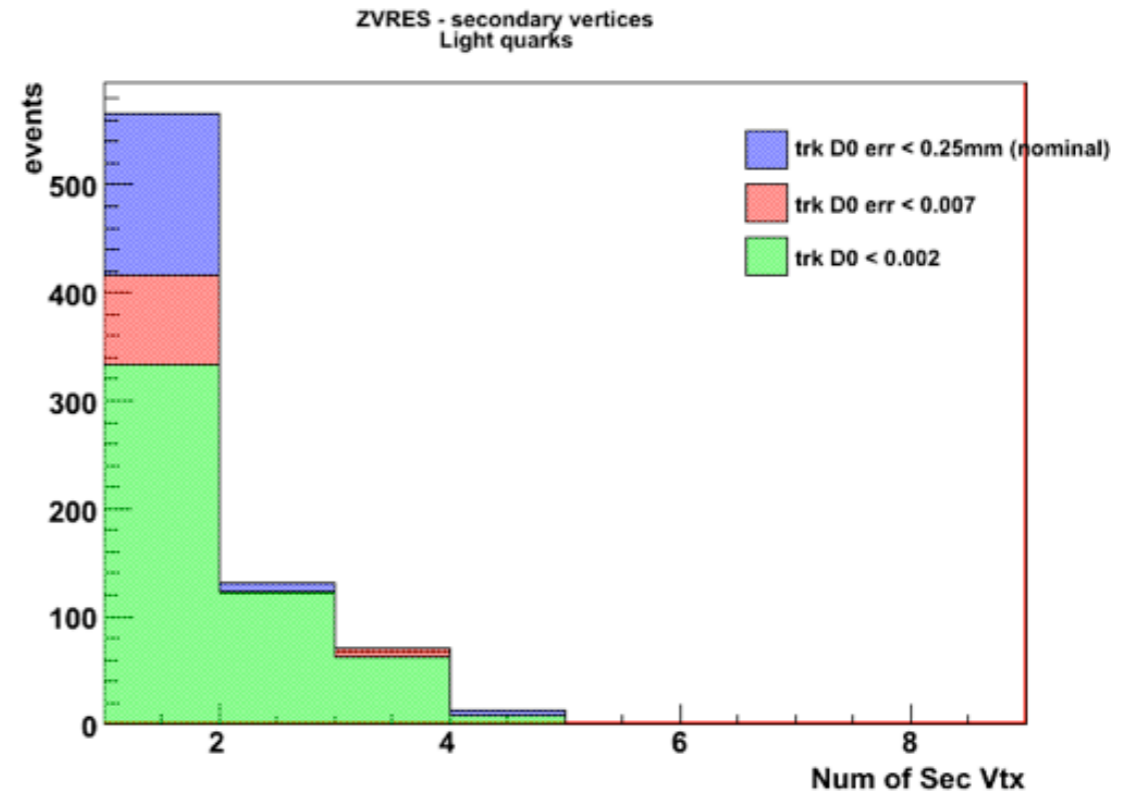
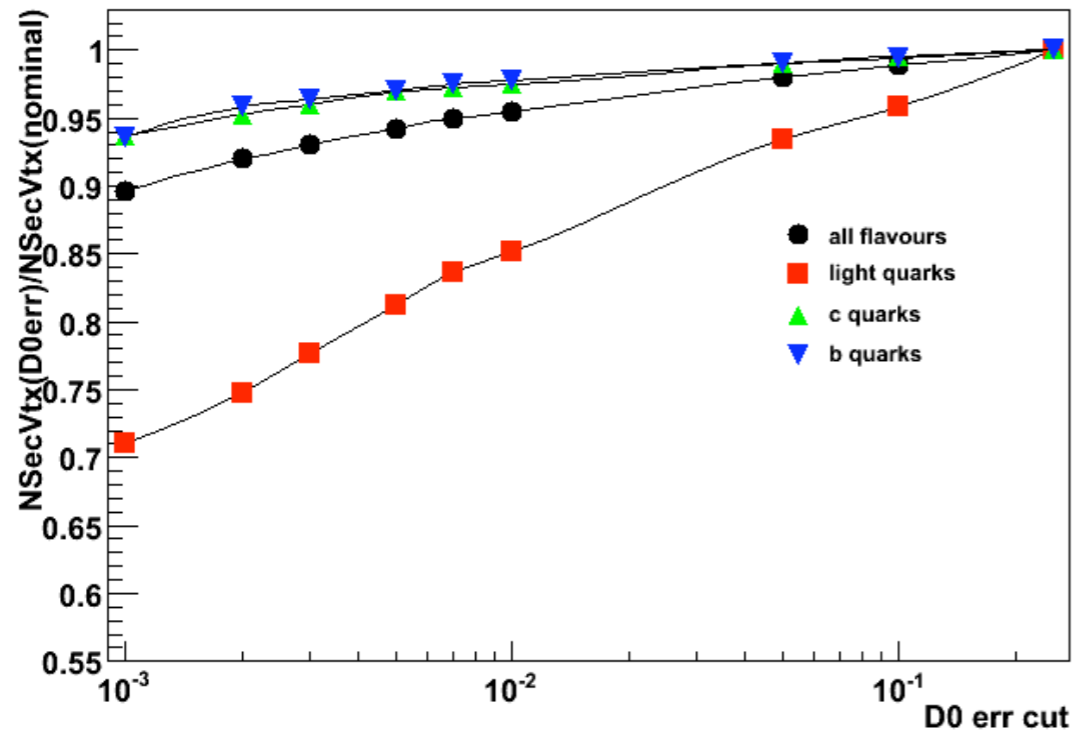
ZVRES secondary vertices (d0 cut)



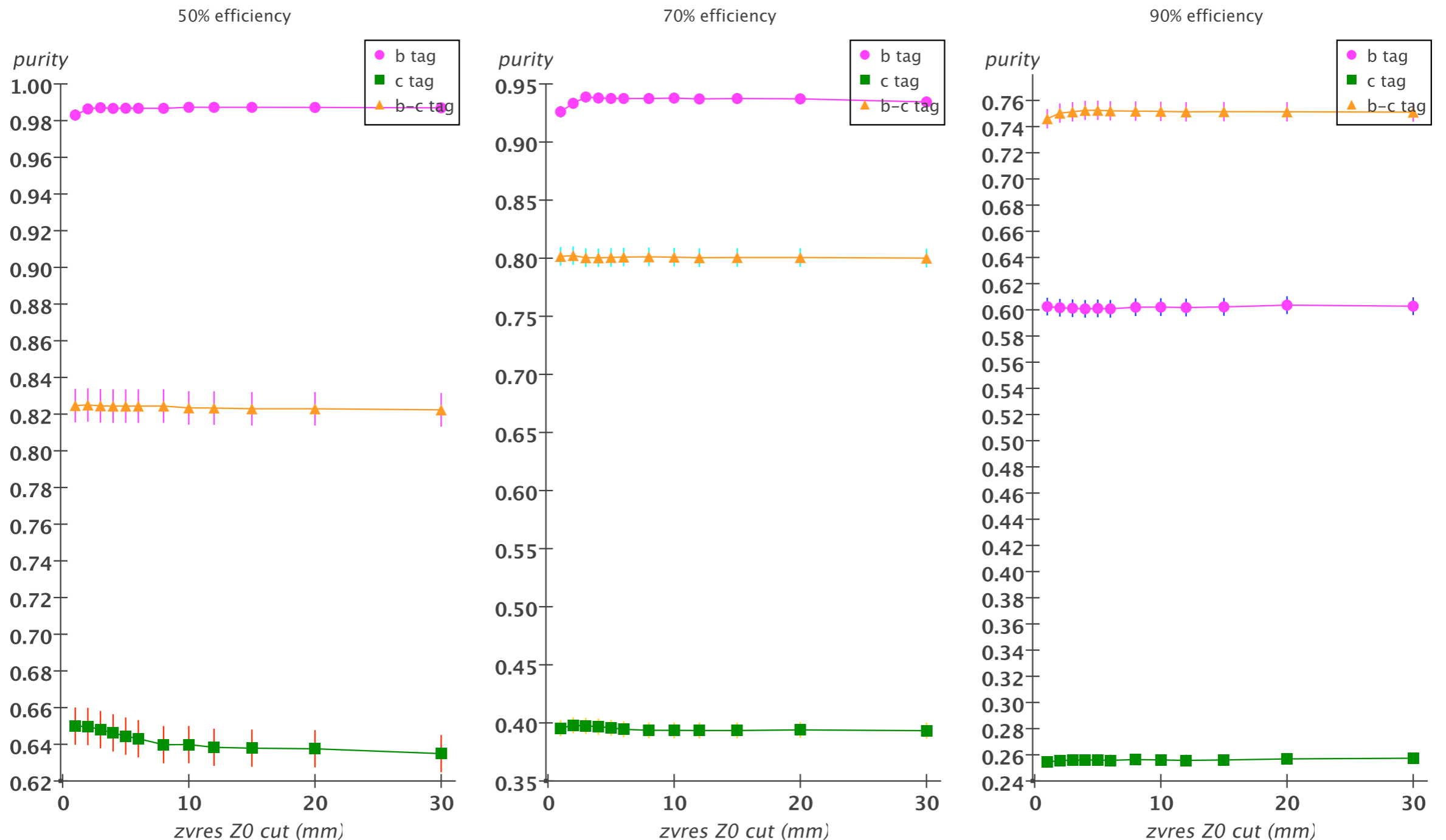
ZVRES: flavour tag purity as a function of the $d0err$ of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: 0.25mm



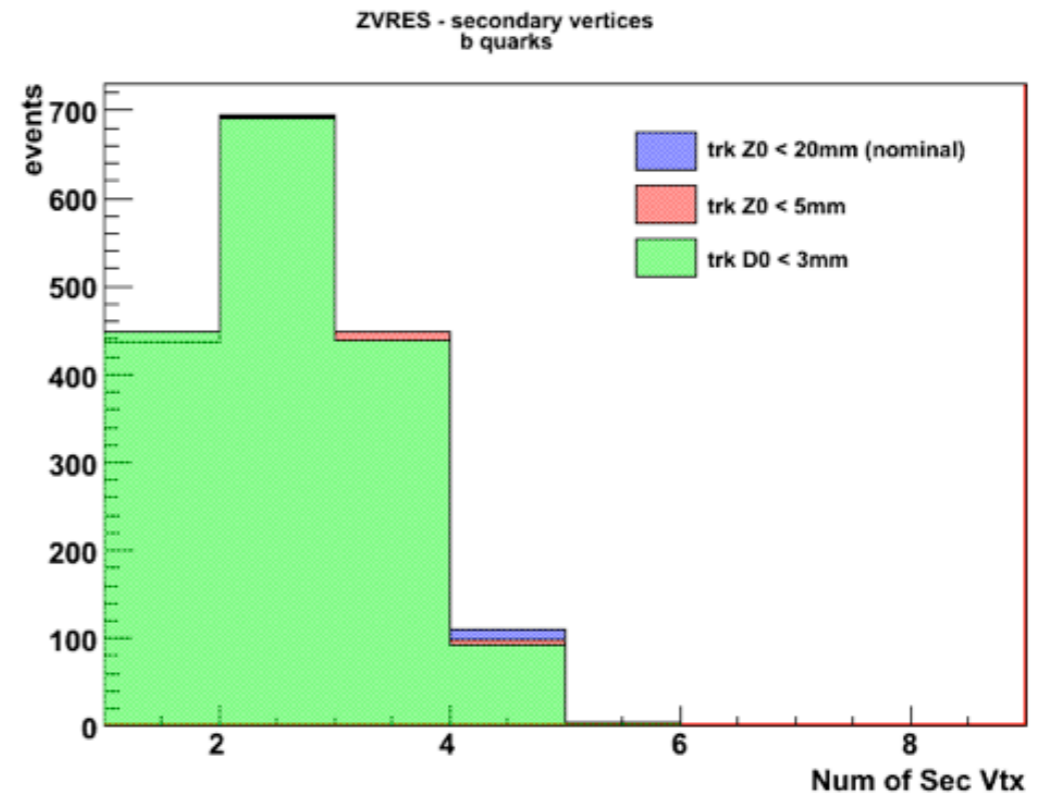
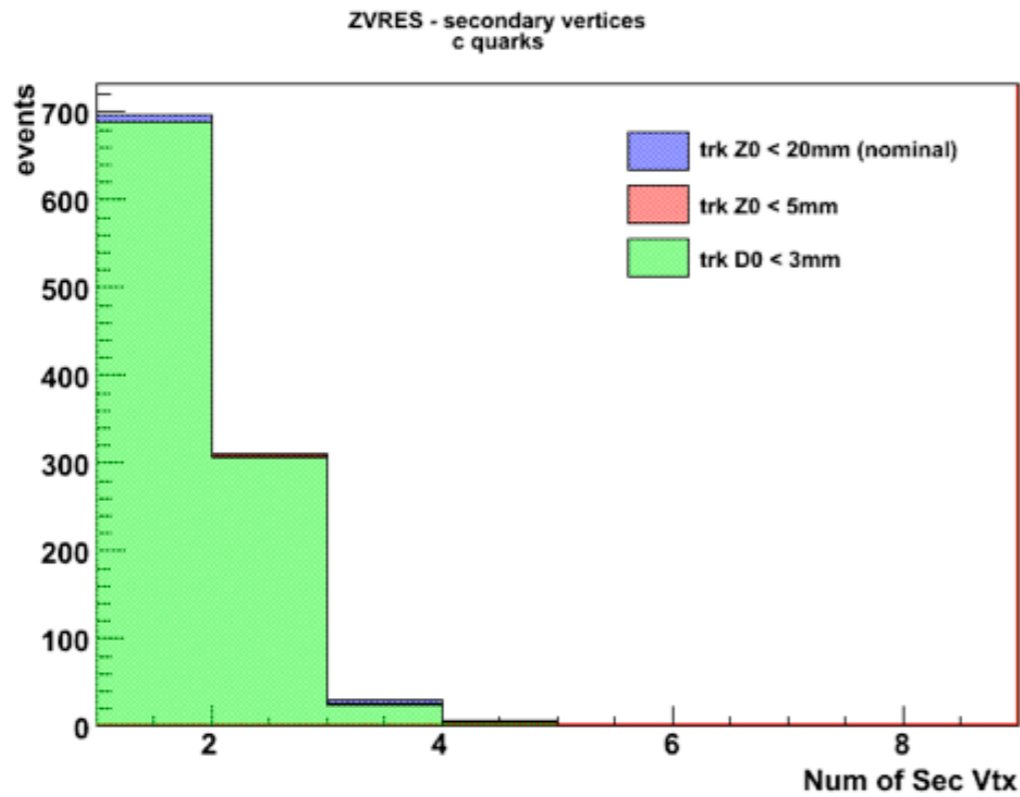
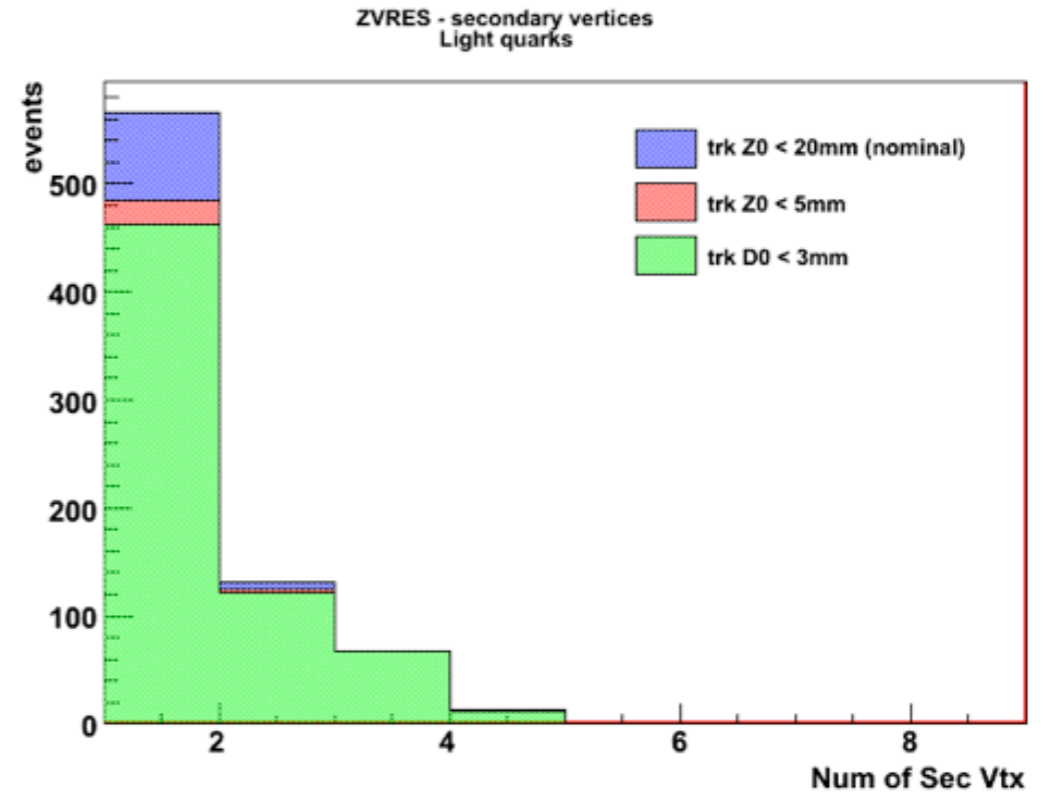
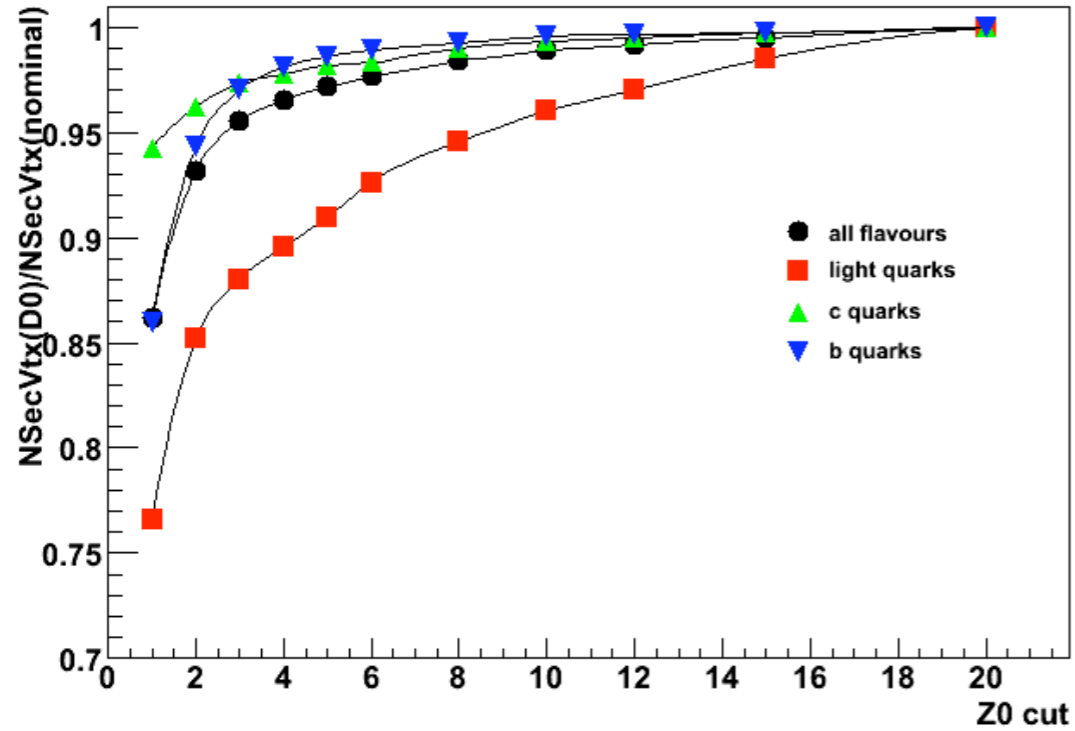
ZVRES secondary vertices (d0 error cut)



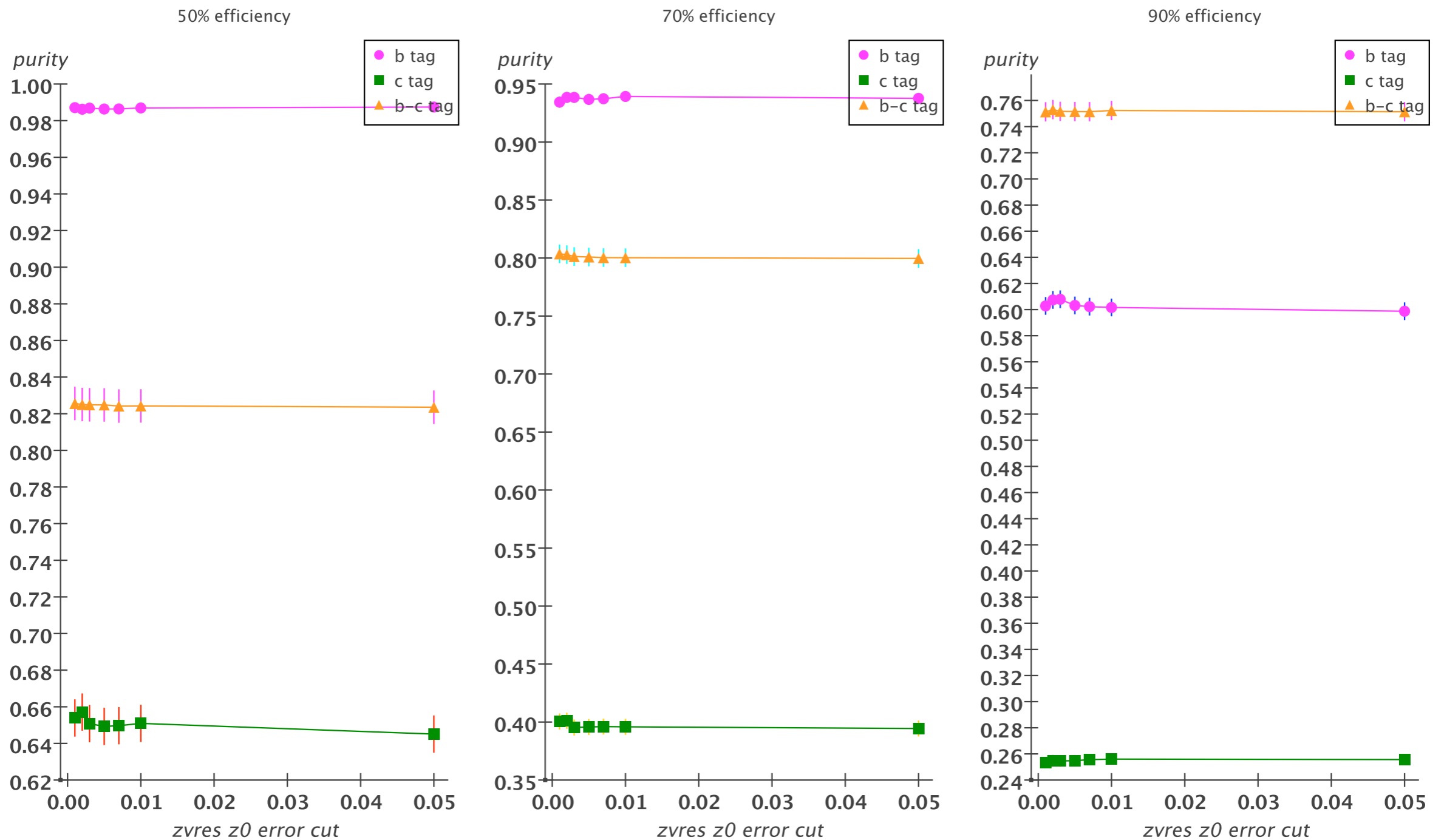
ZVRES: flavour tag purity as a function of the z_0 of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: 20 mm



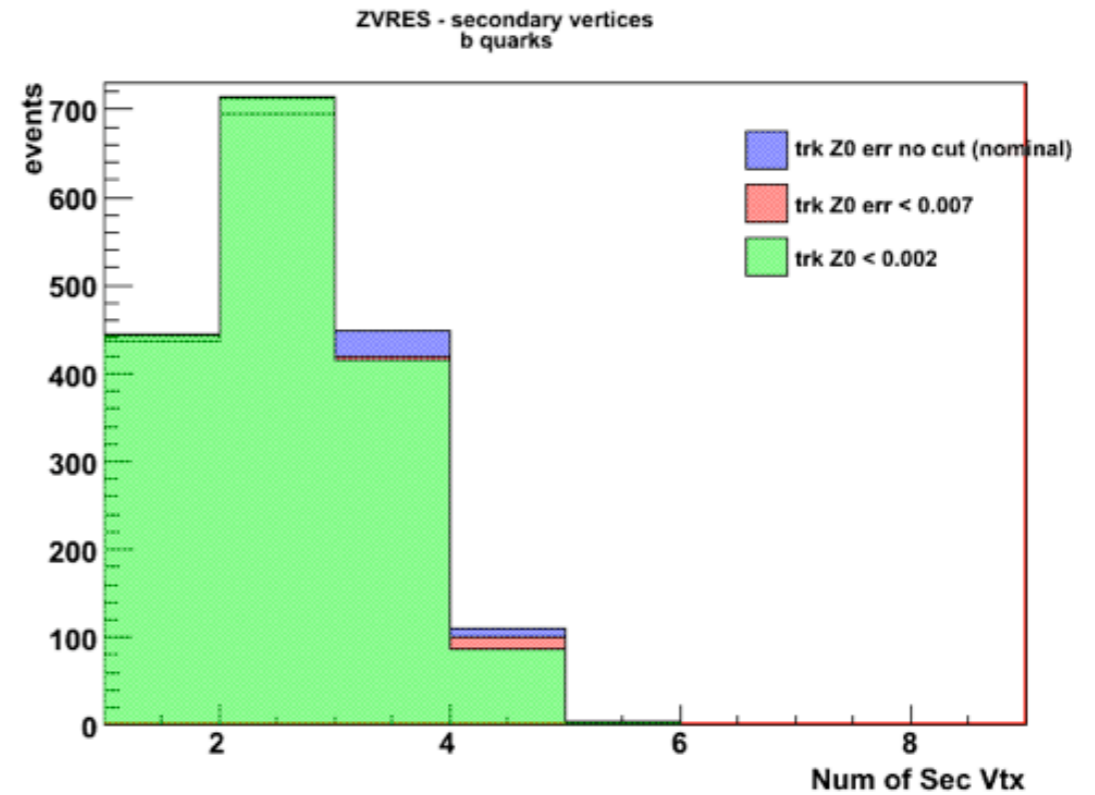
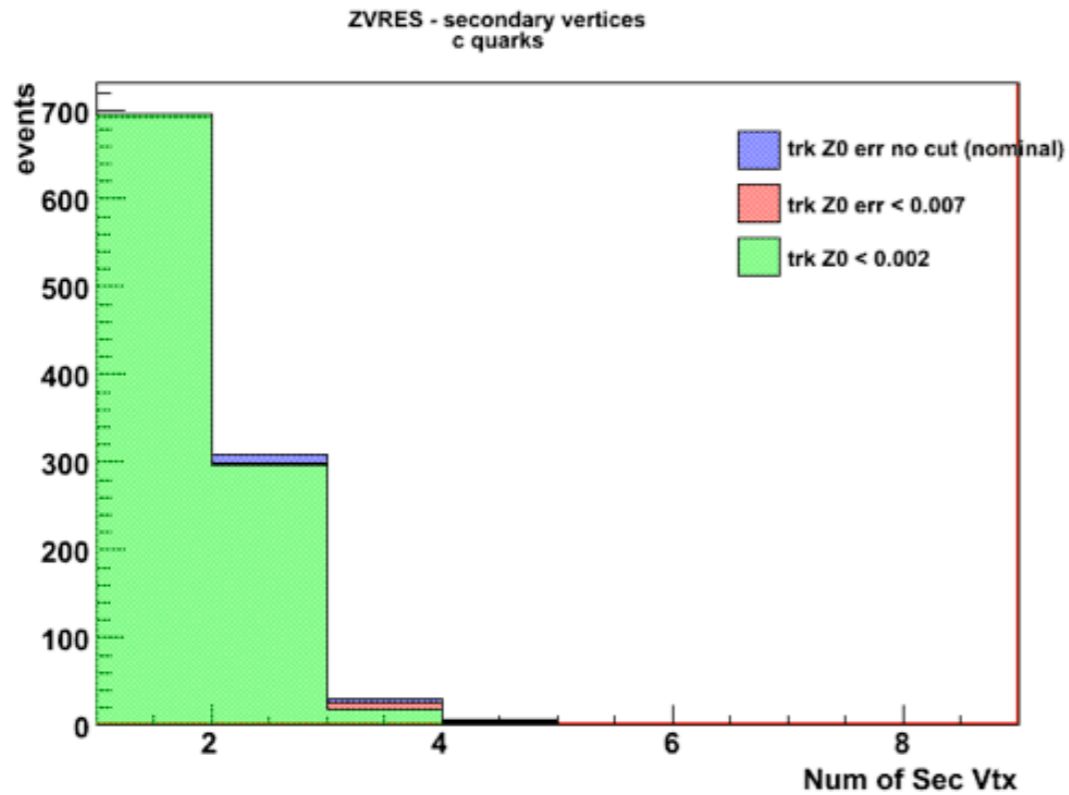
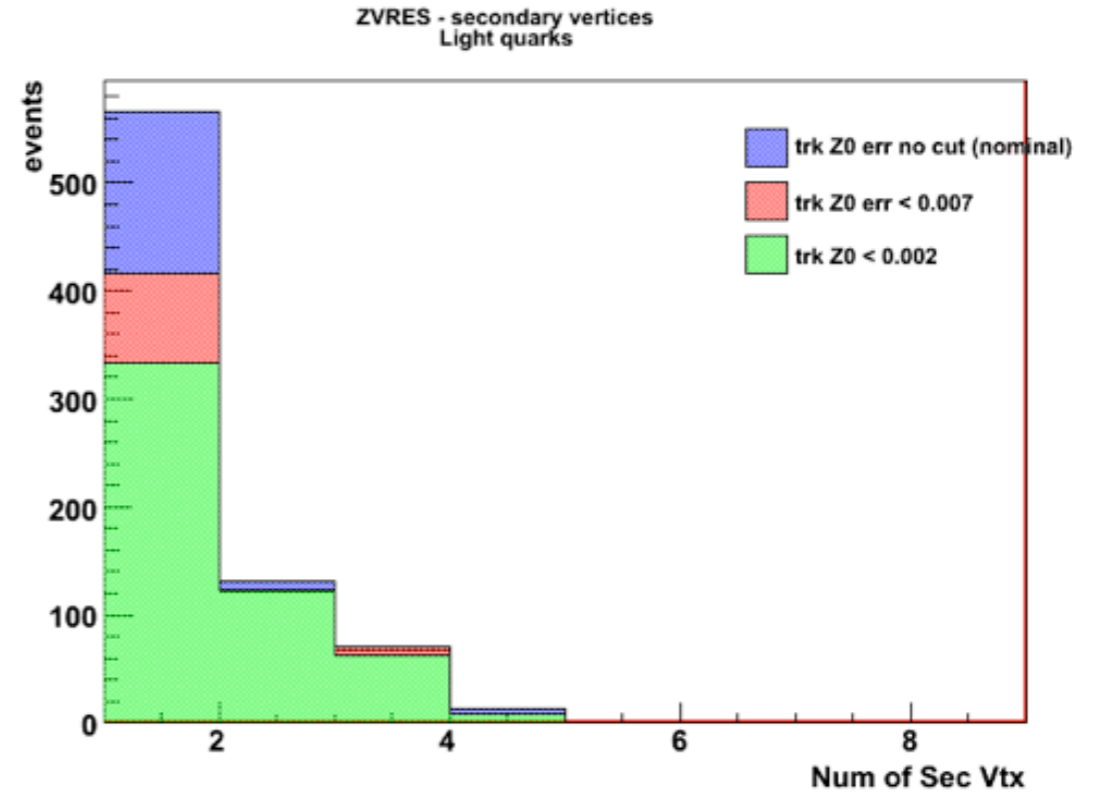
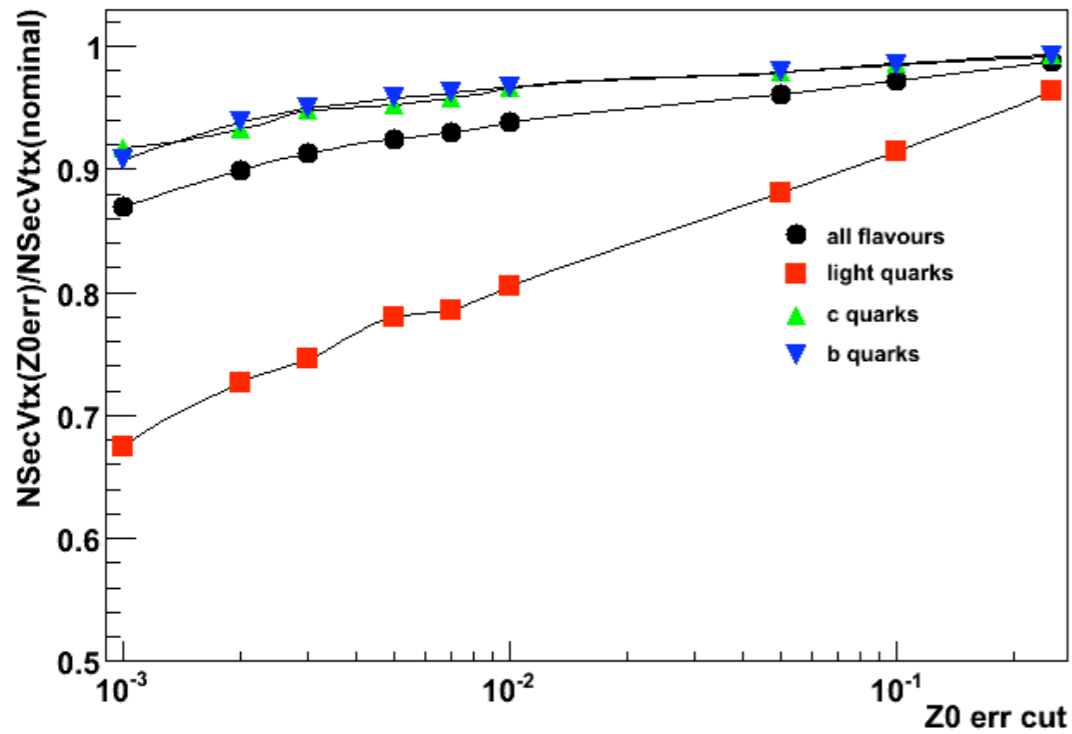
ZVRES secondary vertices (z0 cut)



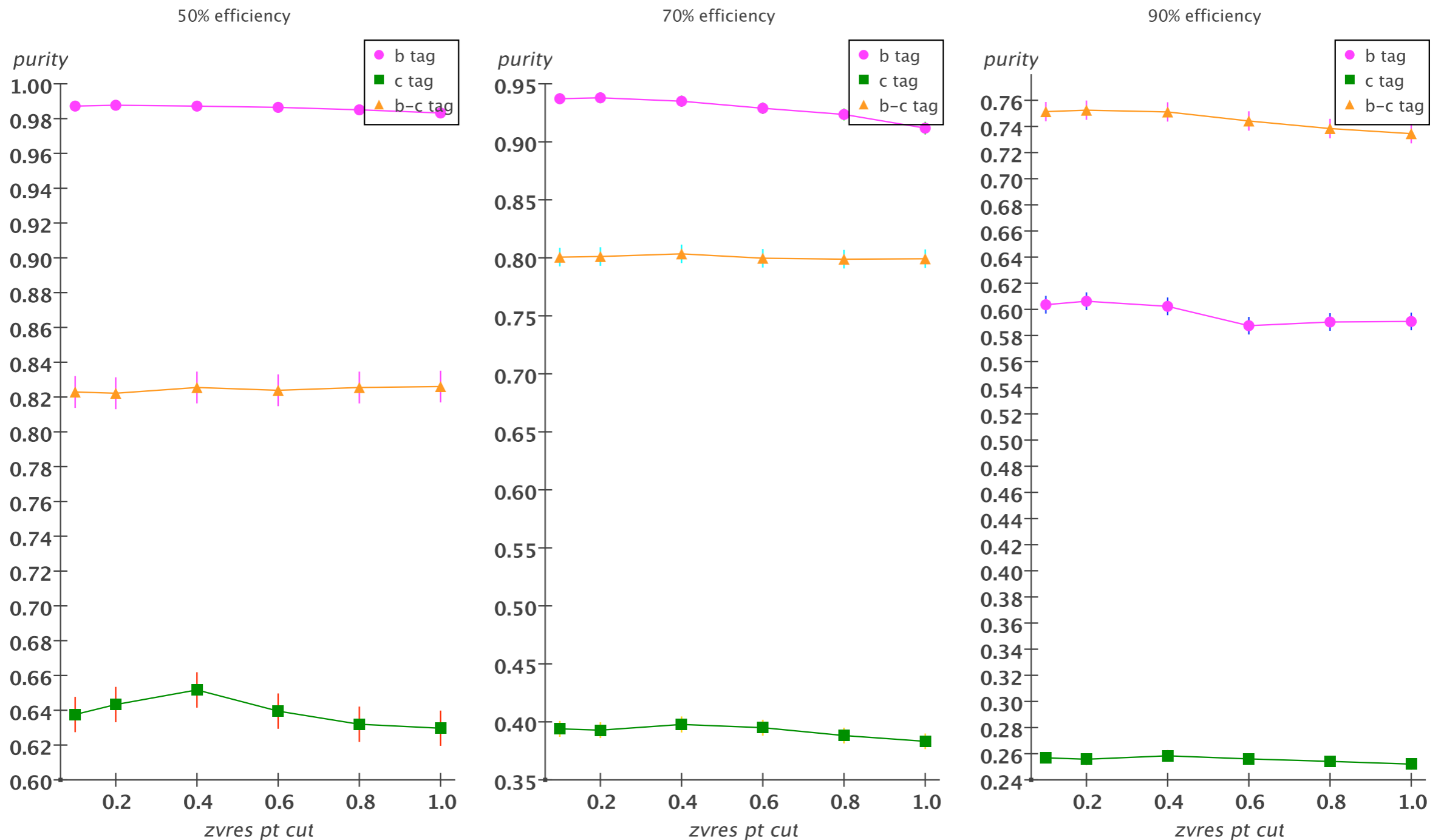
ZVRES: flavour tag purity as a function of the z_0 err of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: no cut



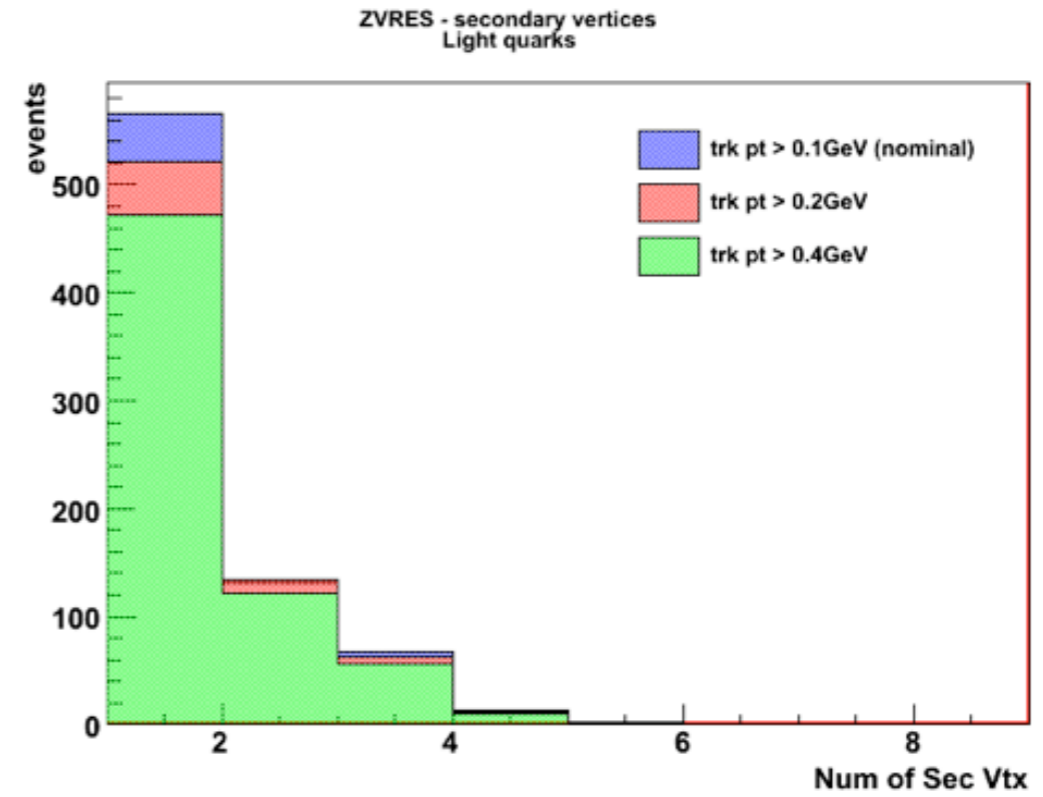
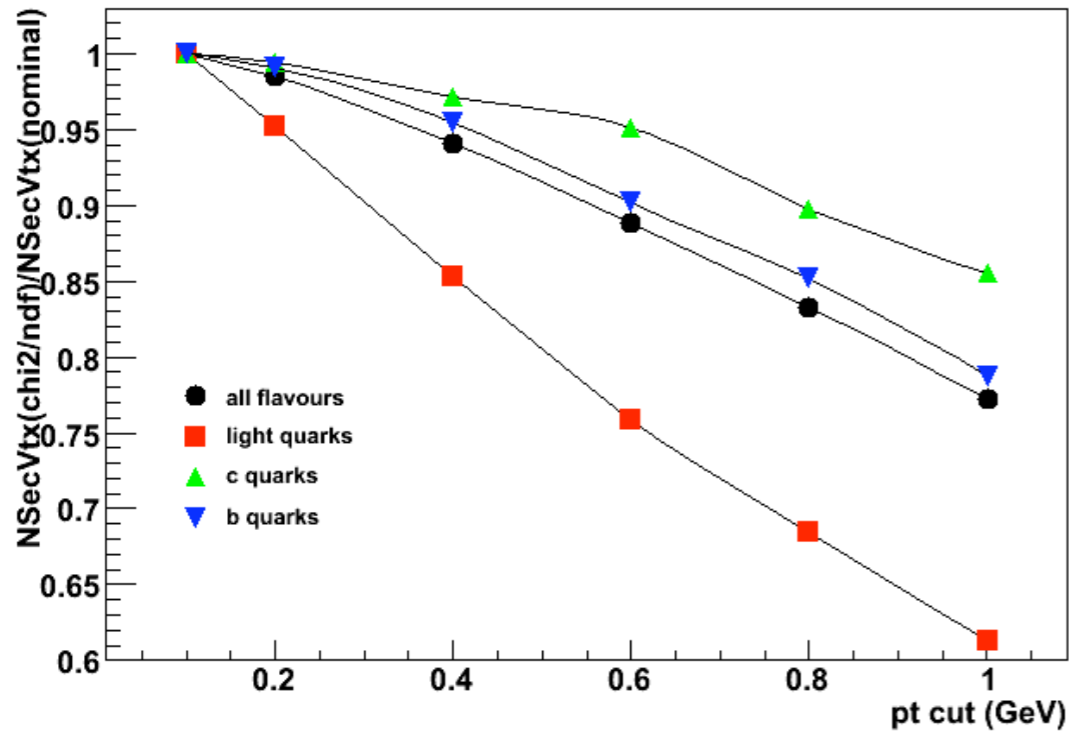
ZVRES secondary vertices (z0 error cut)



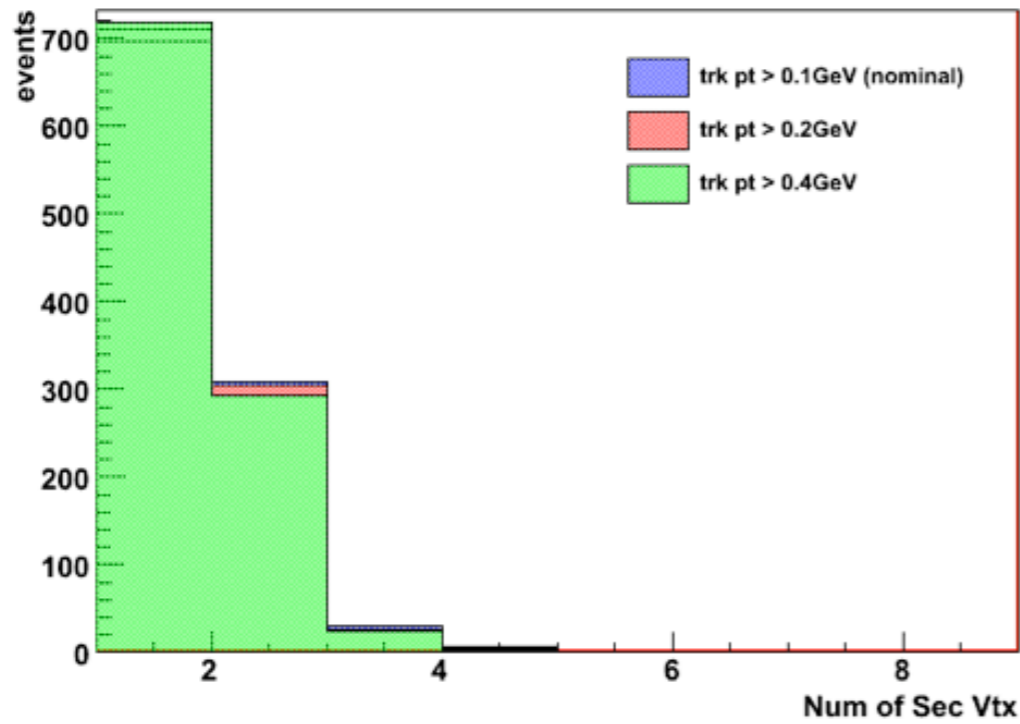
ZVRES: flavour tag purity as a function of the pt of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: 0.1 GeV



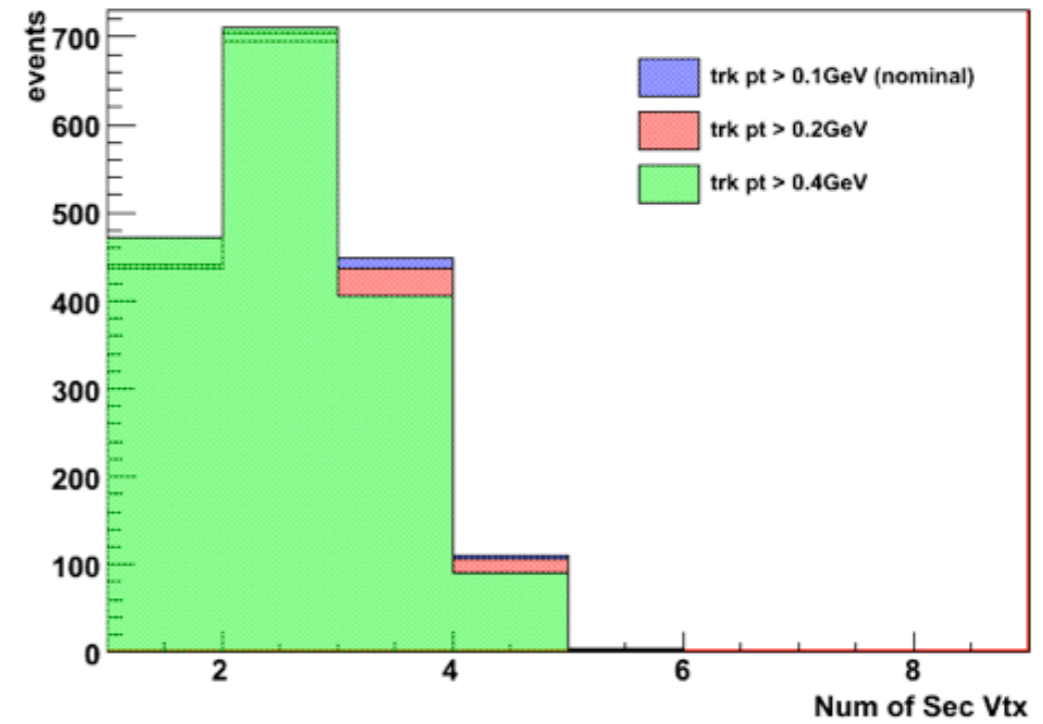
ZVRES secondary vertices (pt cut)



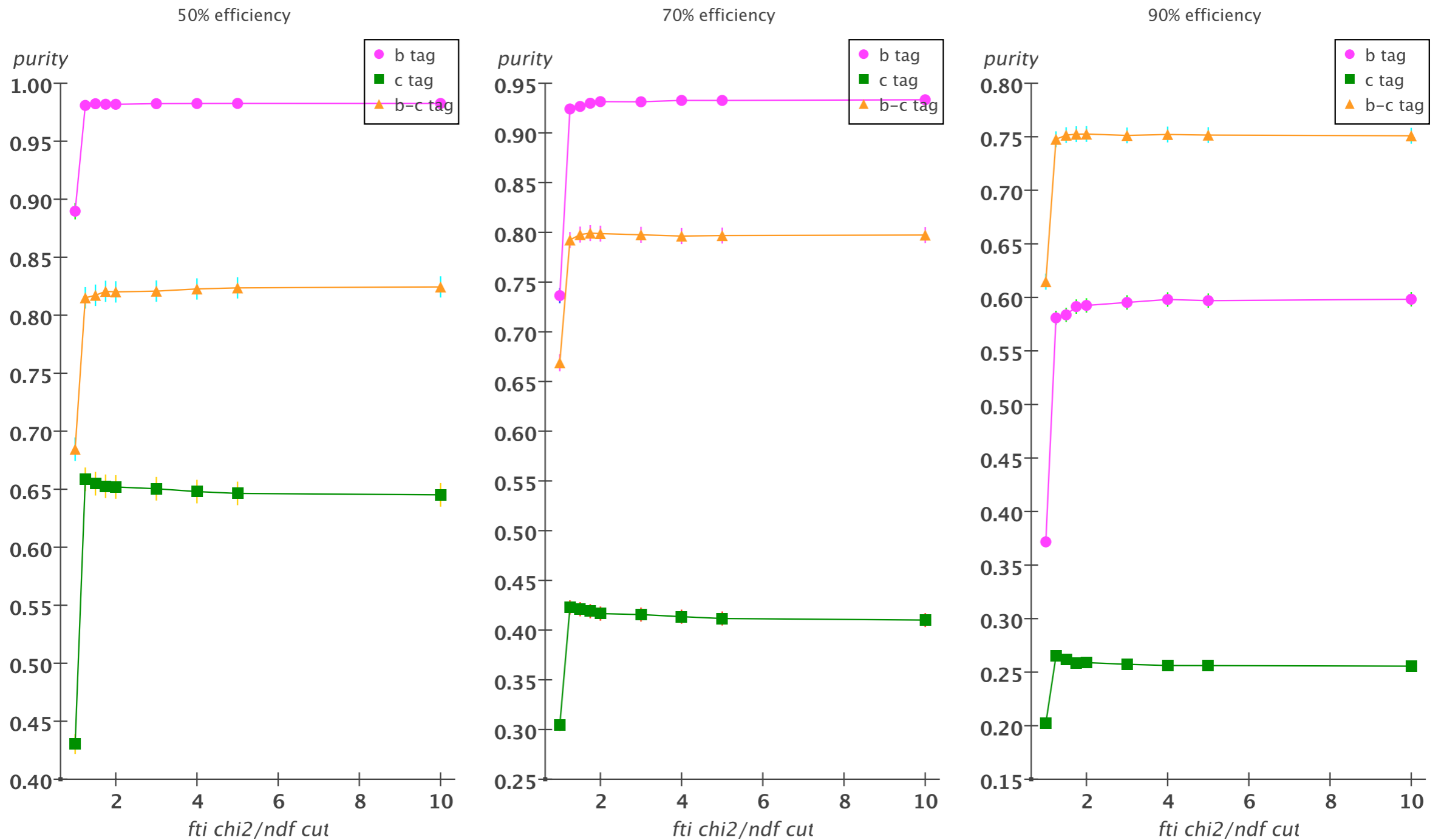
ZVRES - secondary vertices
c quarks



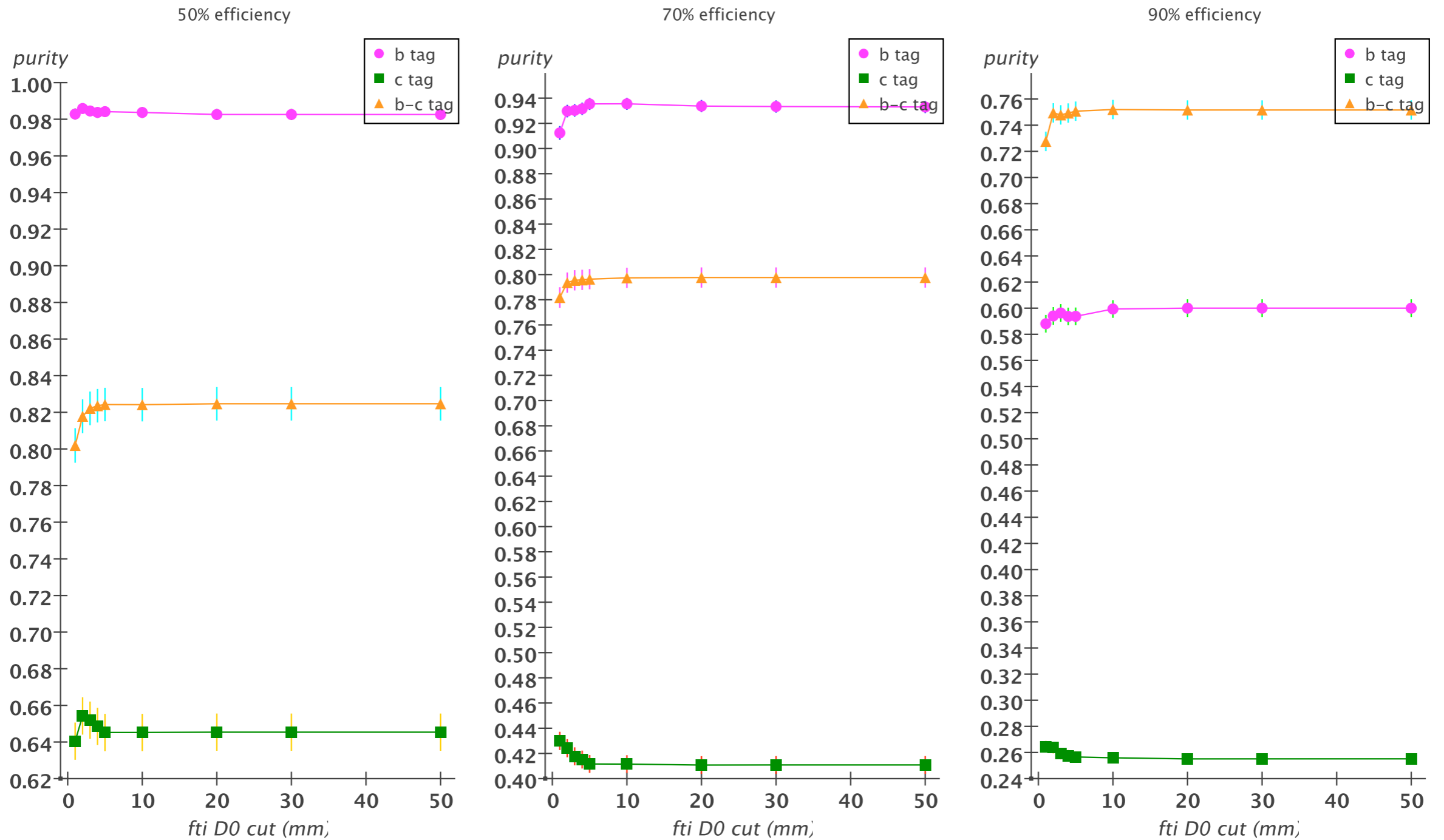
ZVRES - secondary vertices
b quarks



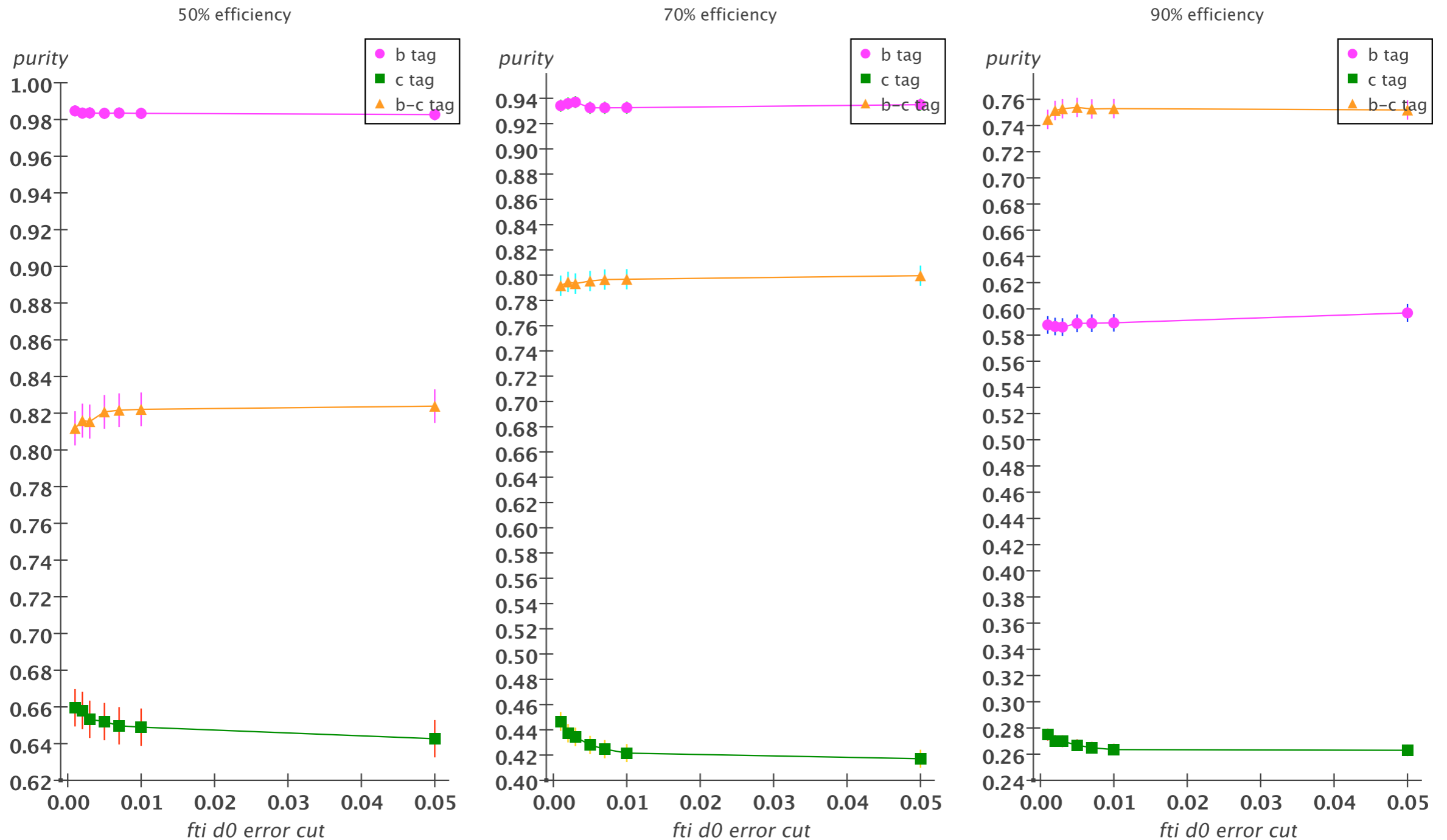
FTI: flavour tag purity as a function of the χ^2/ndf of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: no cut



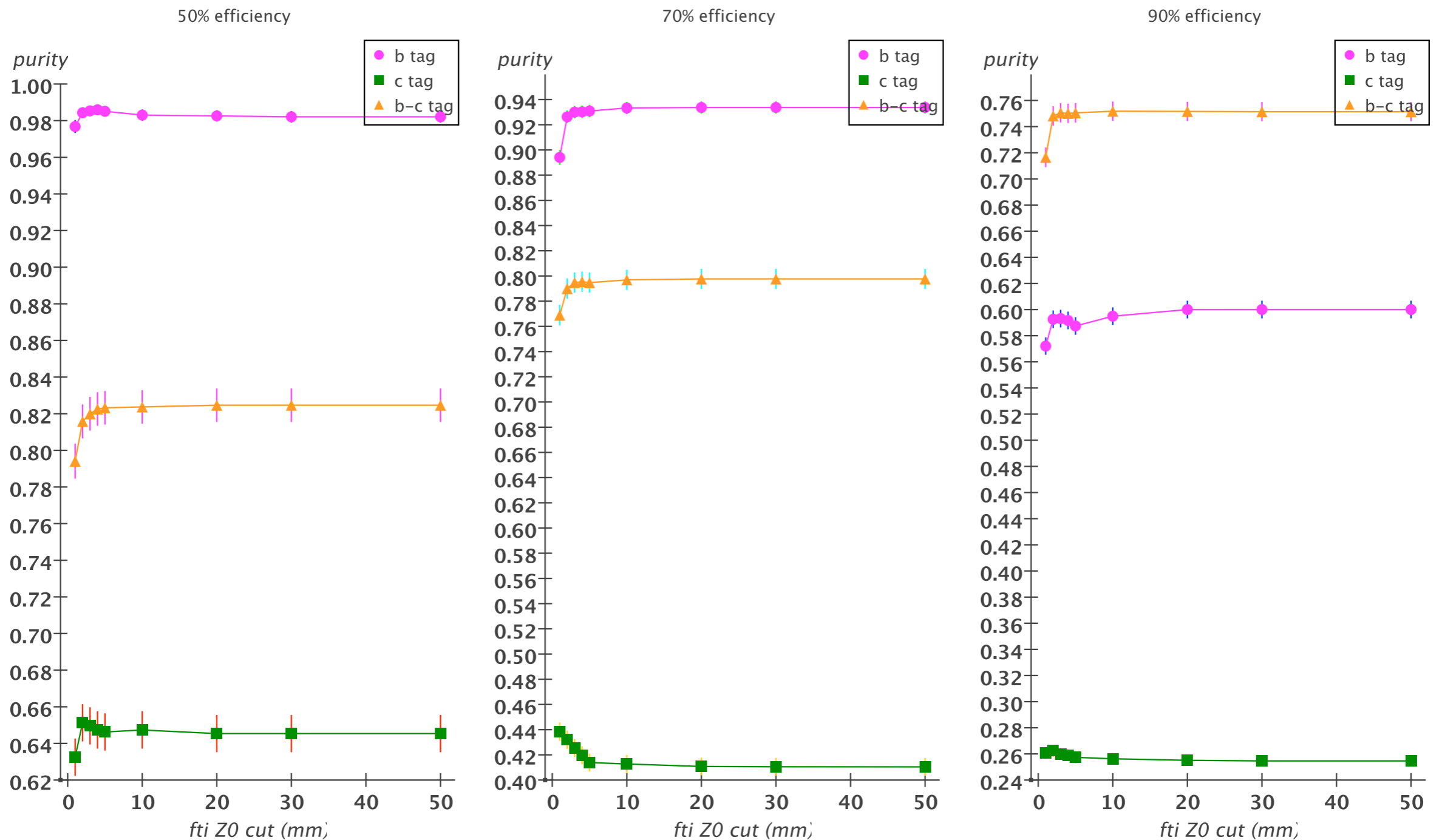
FTI: flavour tag purity as a function of the d_0 of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: 10mm



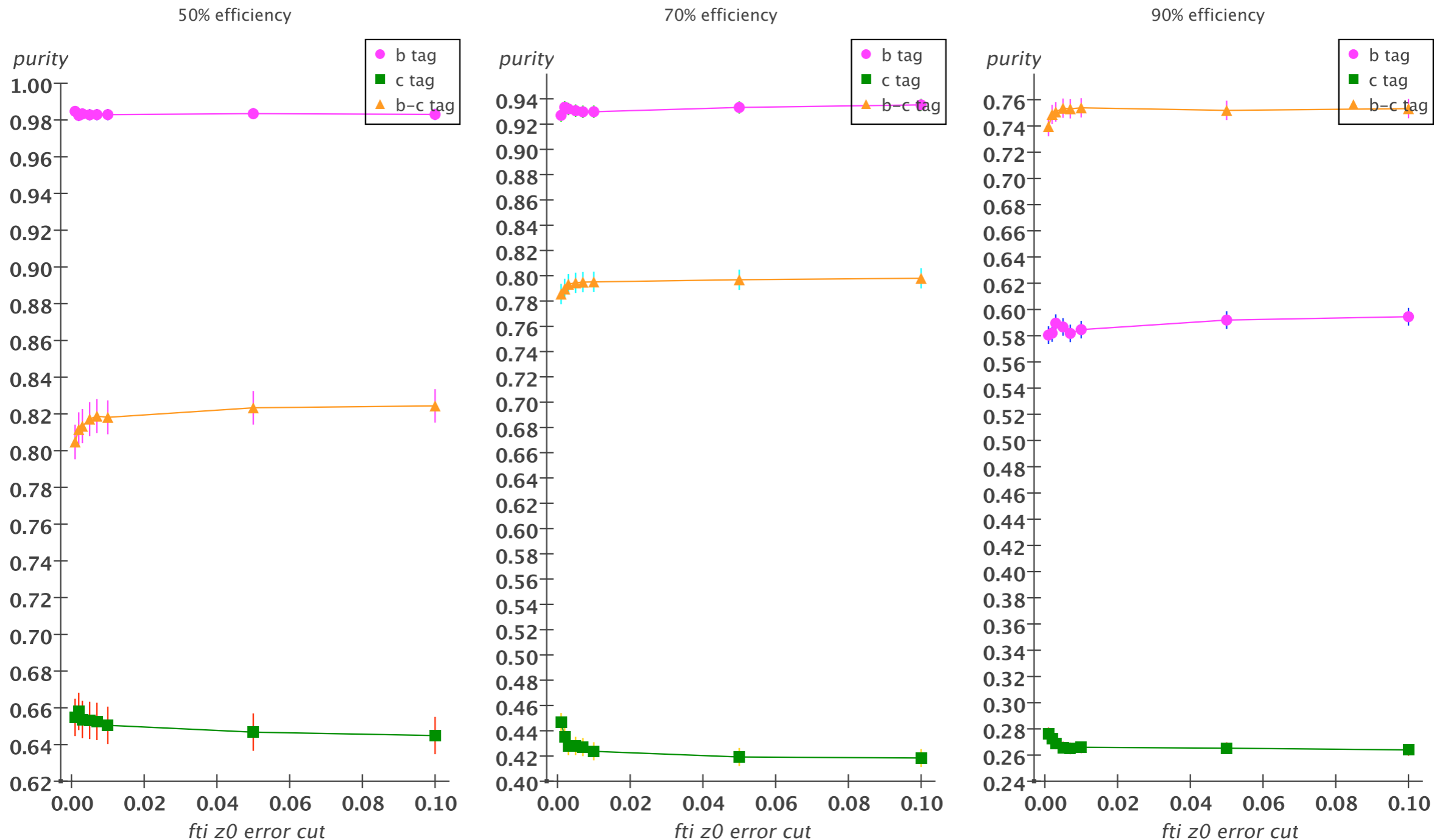
FTI: flavour tag purity as a function of the d_0 err of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: 0.25mm



FTI: flavour tag purity as a function of the z_0 of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: 20 mm



FTI: flavour tag purity as a function of the z_0 err of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: no cut



FTI: flavour tag purity as a function of the pt of the tracks at efficiencies of 50%, 70% and 90% (any number of vertices) - nominal cut: 0.1 GeV

