

Vertical Drift Studies

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Fiducial Volume Cut

- HD takes a fiducial volume as ~ 70 cm from the boundary walls
 - `bool isFid = (fabs(vtx.X())<310 && fabs(vtx.Y())<550 && vtx.Z()>50 && vtx.Z()<1244)`
- We want to do something similar for VD so the equivalent cut becomes
 - `bool isFid = (fabs(vtx.X())<300 && fabs(vtx.Y())<680 && vtx.Z()>40 && vtx.Z()<850)`
- Also processed the horizontal drift files in the original 1D simulation, similar to what was used in the training before.
 - MCC11
- VD training/validation/test samples (Smaller Stats) : ~ 550 k, ~ 70 k, ~ 70 k
- VD training/validation/test samples (Full Stats) : 1.5 million, ~ 85 k, ~ 85 k
- My HD training/validation/test samples : ~ 530 k, 66k, 66k
- Original training had ~ 3.2 million events so almost 2x the current statistics

Recap on Results

CVN VD (Smaller Stats)

	precision	recall	f1-score	support
CC Numu	0.87	0.90	0.89	18715
CC Nue	0.86	0.88	0.87	18625
CC Nutau	0.50	0.32	0.39	4506
NC	0.87	0.90	0.88	27786
accuracy			0.85	69632
macro avg	0.78	0.75	0.76	69632
weighted avg	0.85	0.85	0.85	69632

INFO:root:flavour confusion matrix (rows = predicted classes, cols = actual classes):

```
[[16874  312  942 1227]
 [  274 16332 1121 1158]
 [  246  650 1441  532]
 [ 1321 1331 1002 24869]]
```

CVN HD

	precision	recall	f1-score	support
CC Numu	0.87	0.92	0.90	18541
CC Nue	0.88	0.88	0.88	18060
CC Nutau	0.46	0.27	0.34	4503
NC	0.86	0.89	0.87	25456
accuracy			0.85	66560
macro avg	0.77	0.74	0.75	66560
weighted avg	0.84	0.85	0.85	66560

INFO:root:flavour confusion matrix (rows = predicted classes, cols = actual classes):

```
[[17089  306 1077 1063]
 [  157 15816  964 1030]
 [  294  462 1229  679]
 [ 1001 1476 1233 22684]]
```

CVN HD Old

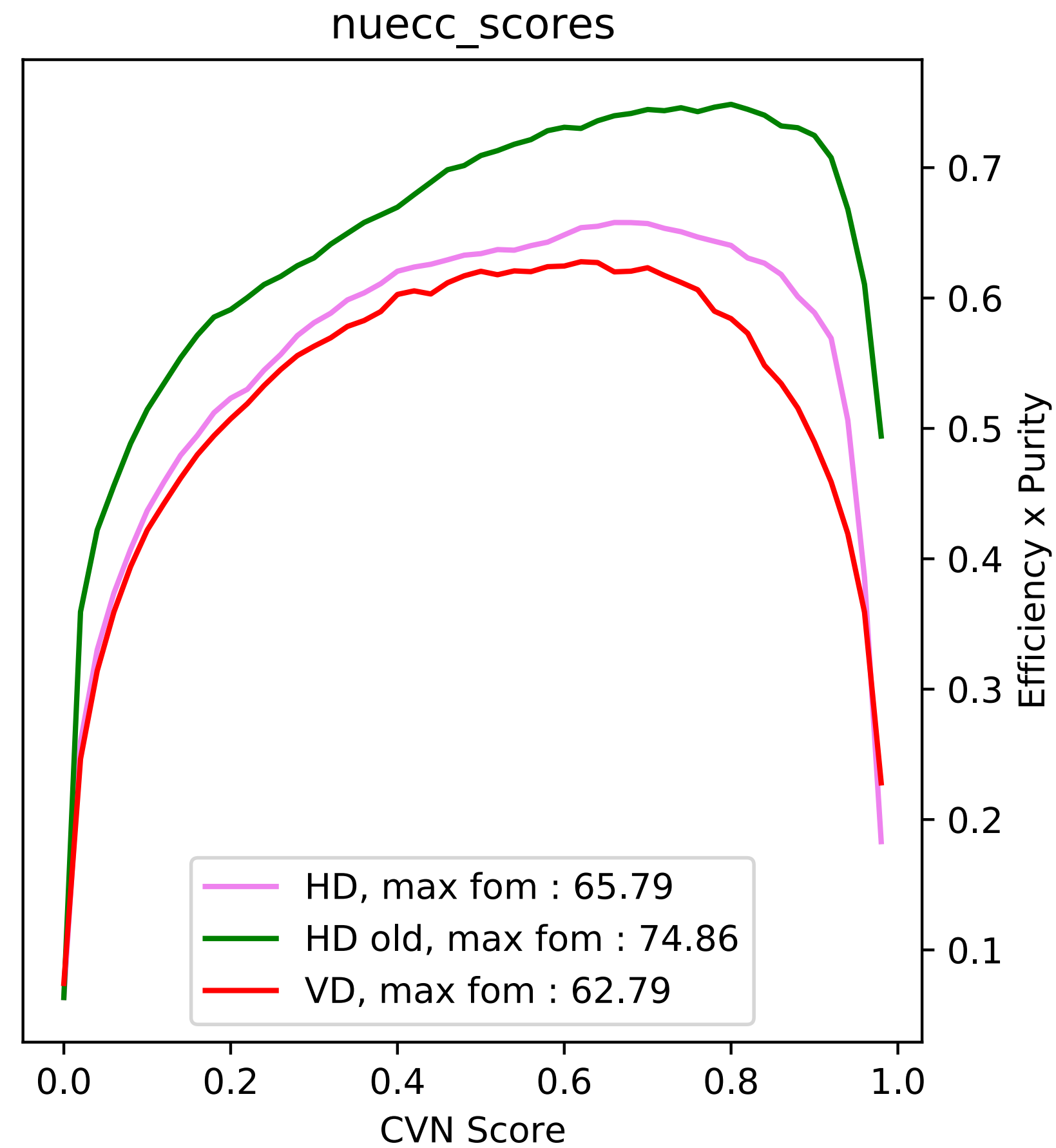
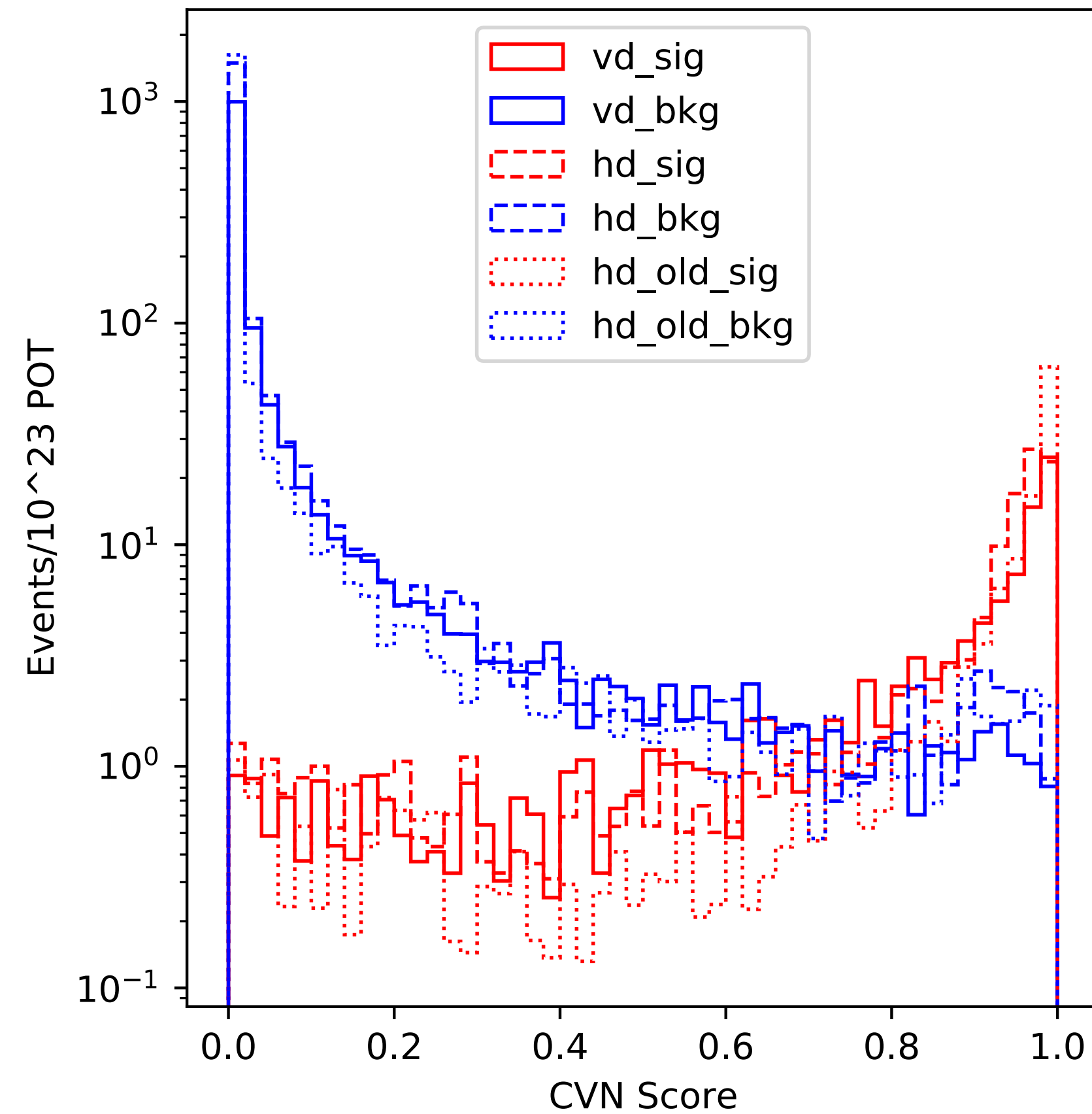
	precision	recall	f1-score	support
CC Numu	0.91	0.93	0.92	18541
CC Nue	0.90	0.92	0.91	18060
CC Nutau	0.54	0.33	0.41	4503
NC	0.88	0.91	0.90	25456
accuracy			0.88	66560
macro avg	0.81	0.77	0.78	66560
weighted avg	0.87	0.88	0.88	66560

INFO:root:flavour confusion matrix (rows = predicted classes, cols = actual classes):

```
[[17291  151  858  719]
 [  181 16667  958  700]
 [  220  272 1477  765]
 [  849  970 1210 23272]]
```

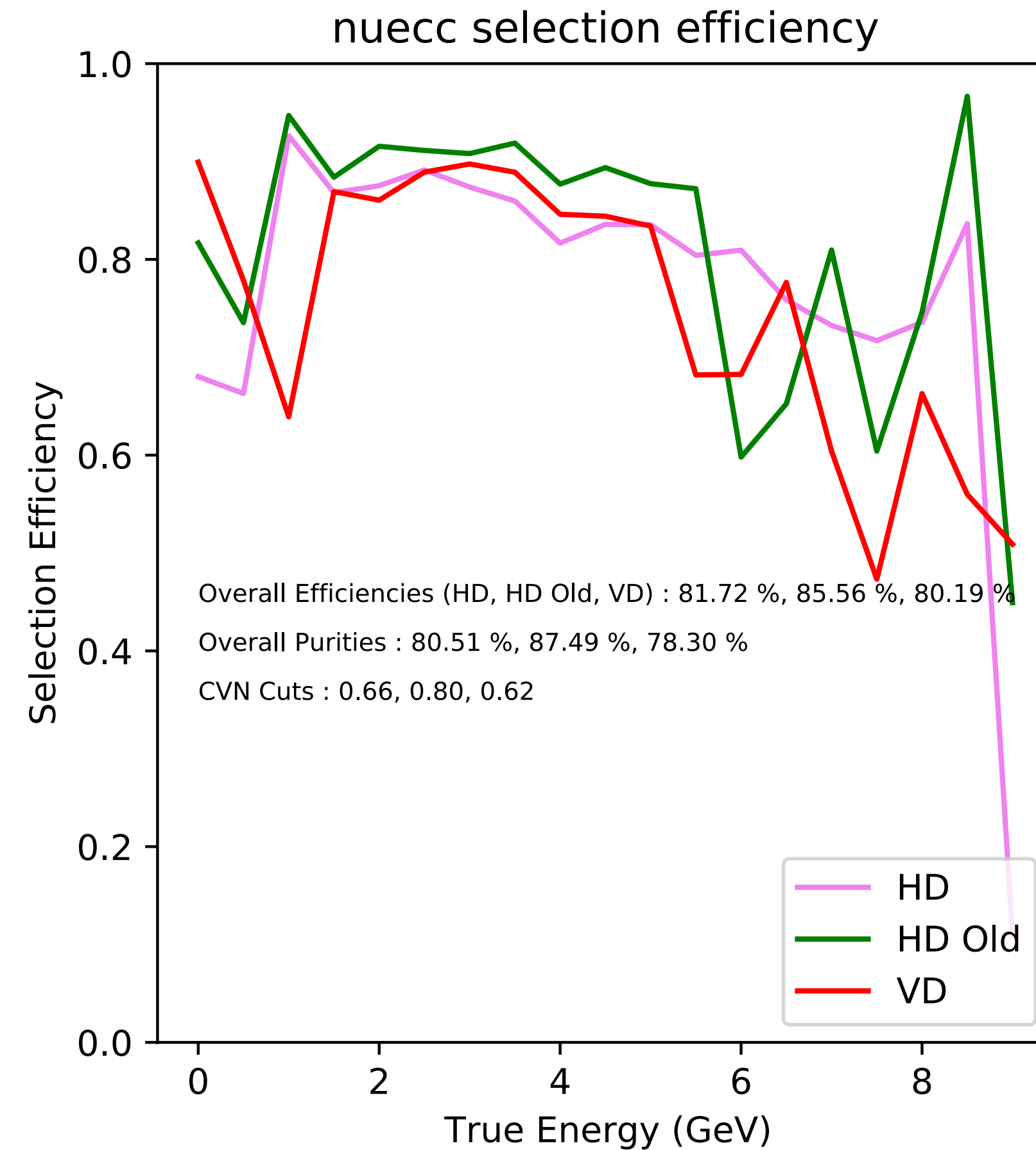
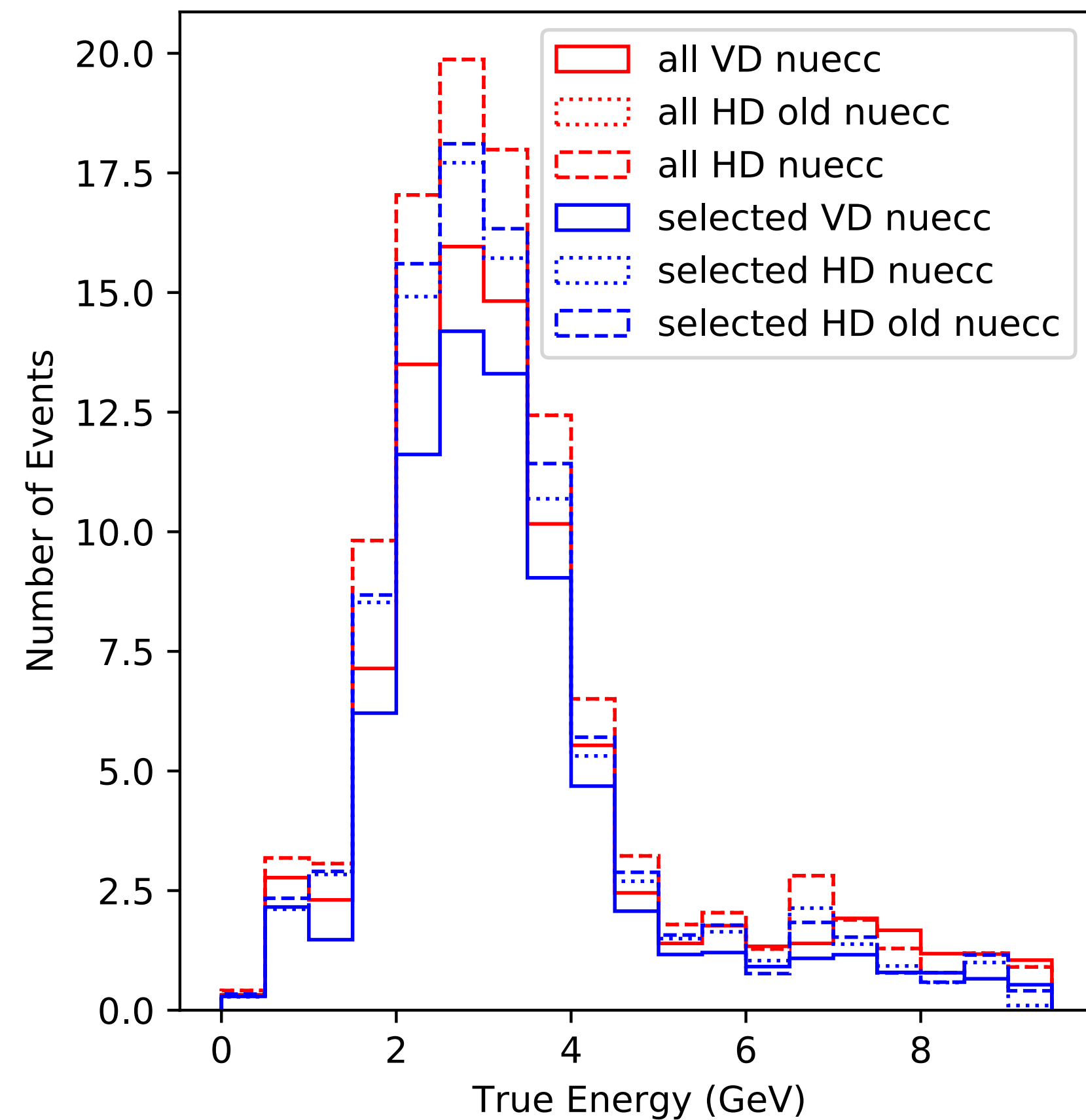
- CVN HD is my training on MCC11 files w/ similar statistics as VD
- CVN HD Old is applying the old trained model in the CVN paper from Leigh and Saul on the MCC11 HD dataset
 - Can easily imagine ~2-3 points in precision/recall coming from more training statistics

Recap on Results



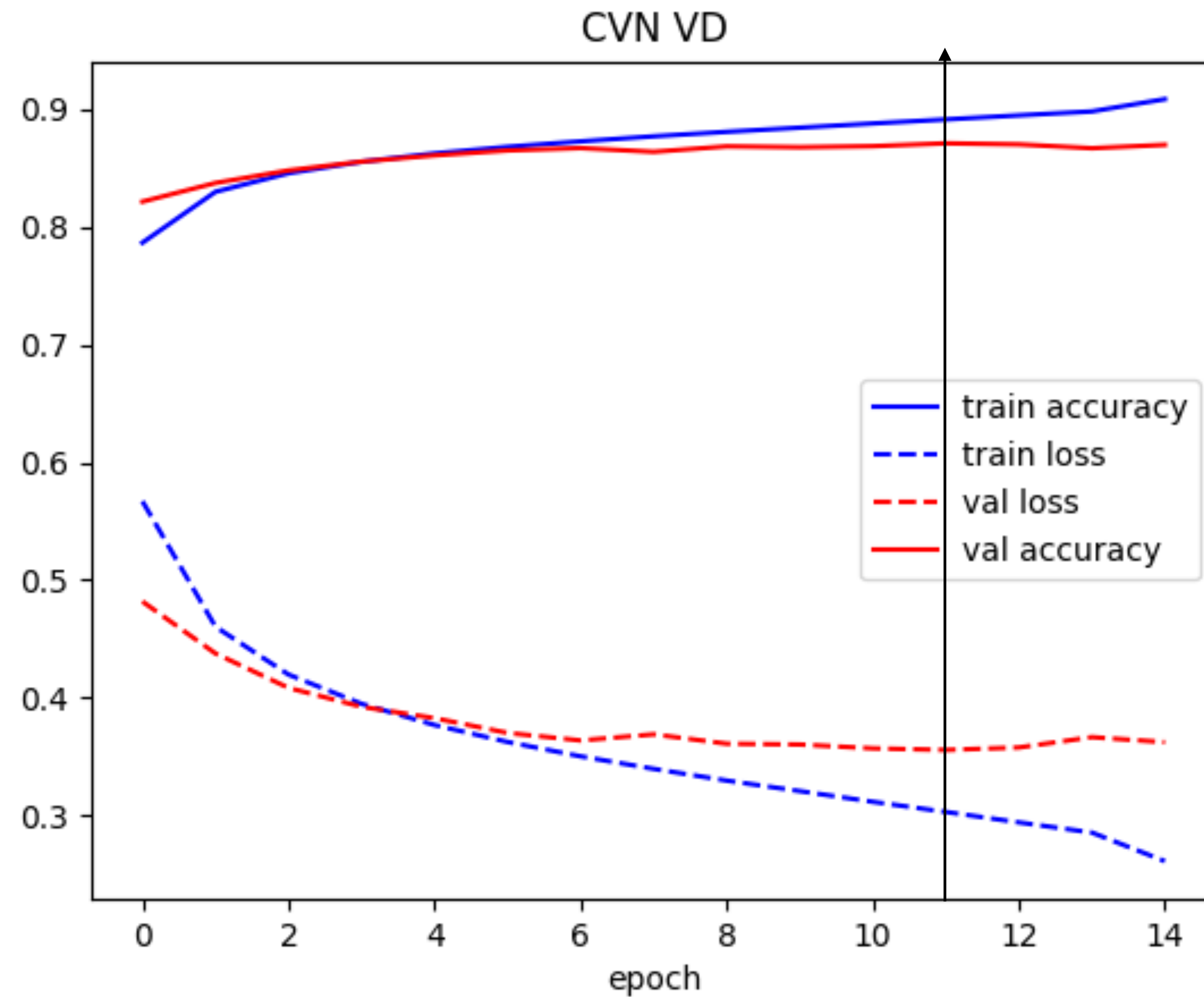
- Comparing the 3, it seems VD and HD are close — whereas “HD Old” outperforms them both
- POT Normalization approximately, ran over 2000 events for each swap (numuCC, nueCC, nutauCC) and extrapolated total POT from there — nuTauCC background approximately halved from before

Recap on Results



- For efficiency and purity numbers, after approximate POT normalisation, I get ~80% efficiency and purity for both VD and HD
- For HD old, its 86% and 88% (eff, purity) respectively — which is pretty close to what they report in the paper (NB : they report it wrt reco energy since at nueCC peak, high energy NCs dominate background, not 2-4 GeV NCs like here)

Full Stats Training



- Training still going on, showing results for trained model at epoch 11.. seems to be close to converging

CVN VD (Smaller Stats)

```

precision    recall  f1-score   support

CC Numu      0.87    0.90    0.89    18715
CC Nue       0.86    0.88    0.87    18625
CC Nutau     0.50    0.32    0.39    4506
NC           0.87    0.90    0.88    27786

accuracy                    0.85    69632
macro avg      0.78    0.75    0.76    69632
weighted avg   0.85    0.85    0.85    69632

```

INFO:root:flavour confusion matrix (rows = predicted classes, cols = actual classes):

```

[[16874  312  942 1227]
 [ 274 16332 1121 1158]
 [ 246  650 1441  532]
 [ 1321 1331 1002 24869]]

```



CVN VD (Full Stats)

INFO:root:flavour report:

```

precision    recall  f1-score   support

CC Numu      0.89    0.91    0.90    22762
CC Nue       0.86    0.91    0.89    22606
CC Nutau     0.53    0.36    0.43    5389
NC           0.90    0.89    0.89    33211

accuracy                    0.87    83968
macro avg      0.80    0.77    0.78    83968
weighted avg   0.86    0.87    0.86    83968

```

INFO:root:flavour confusion matrix (rows = predicted classes, cols = actual classes):

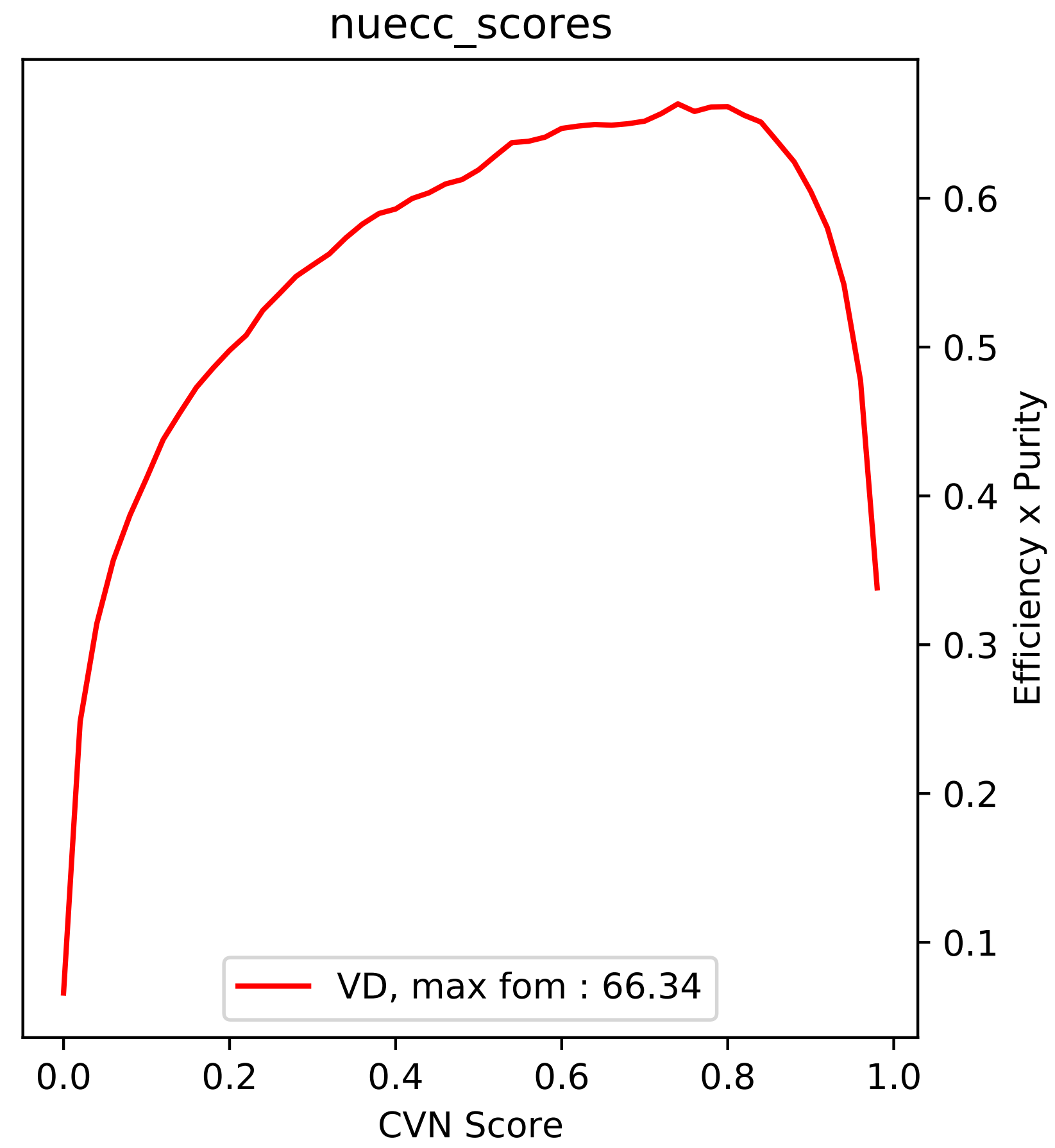
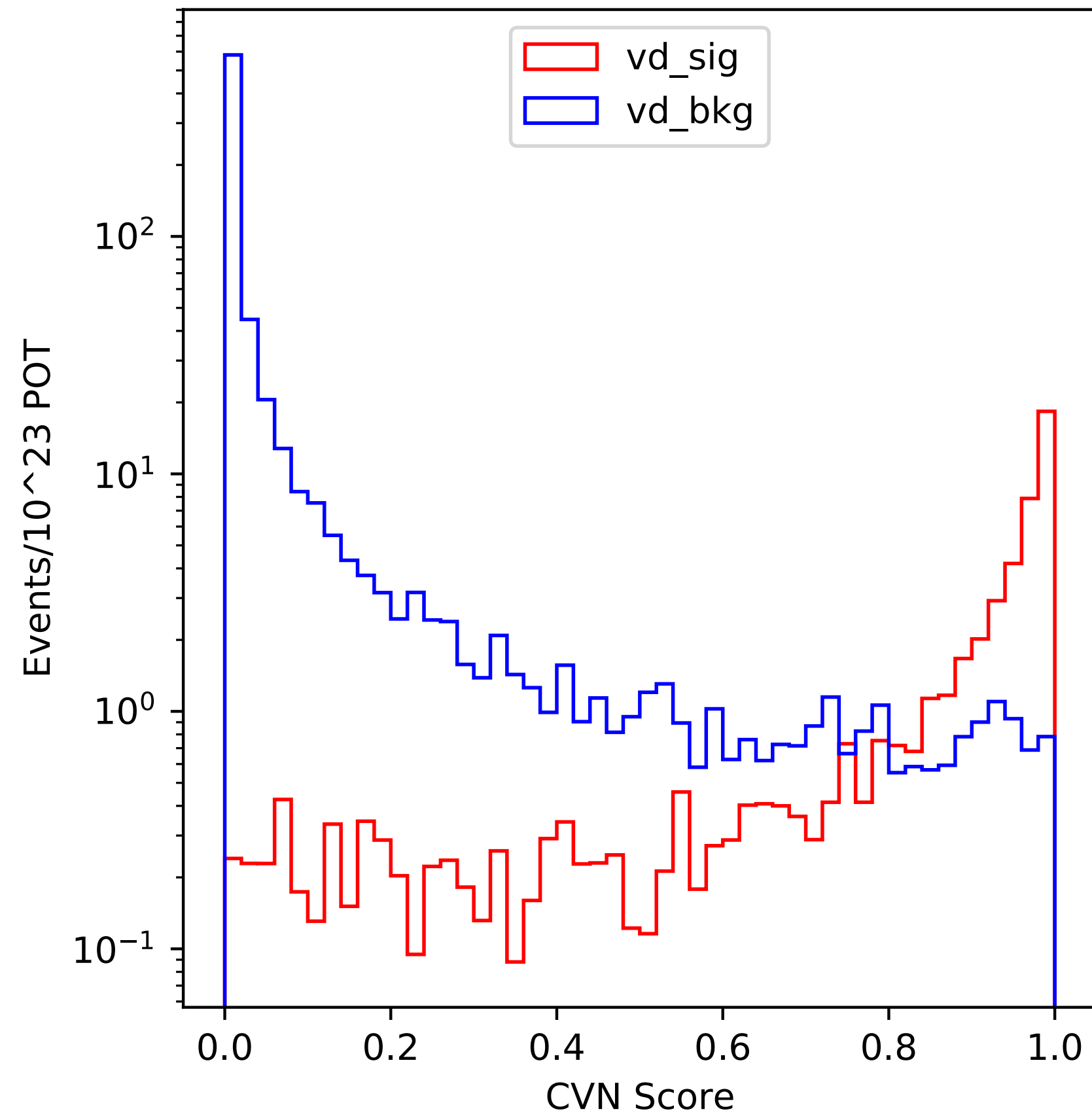
```

[[20826  273  997 1315]
 [ 347 20536 1449 1464]
 [ 330  613 1957  768]
 [ 1259 1184  986 29664]]

```

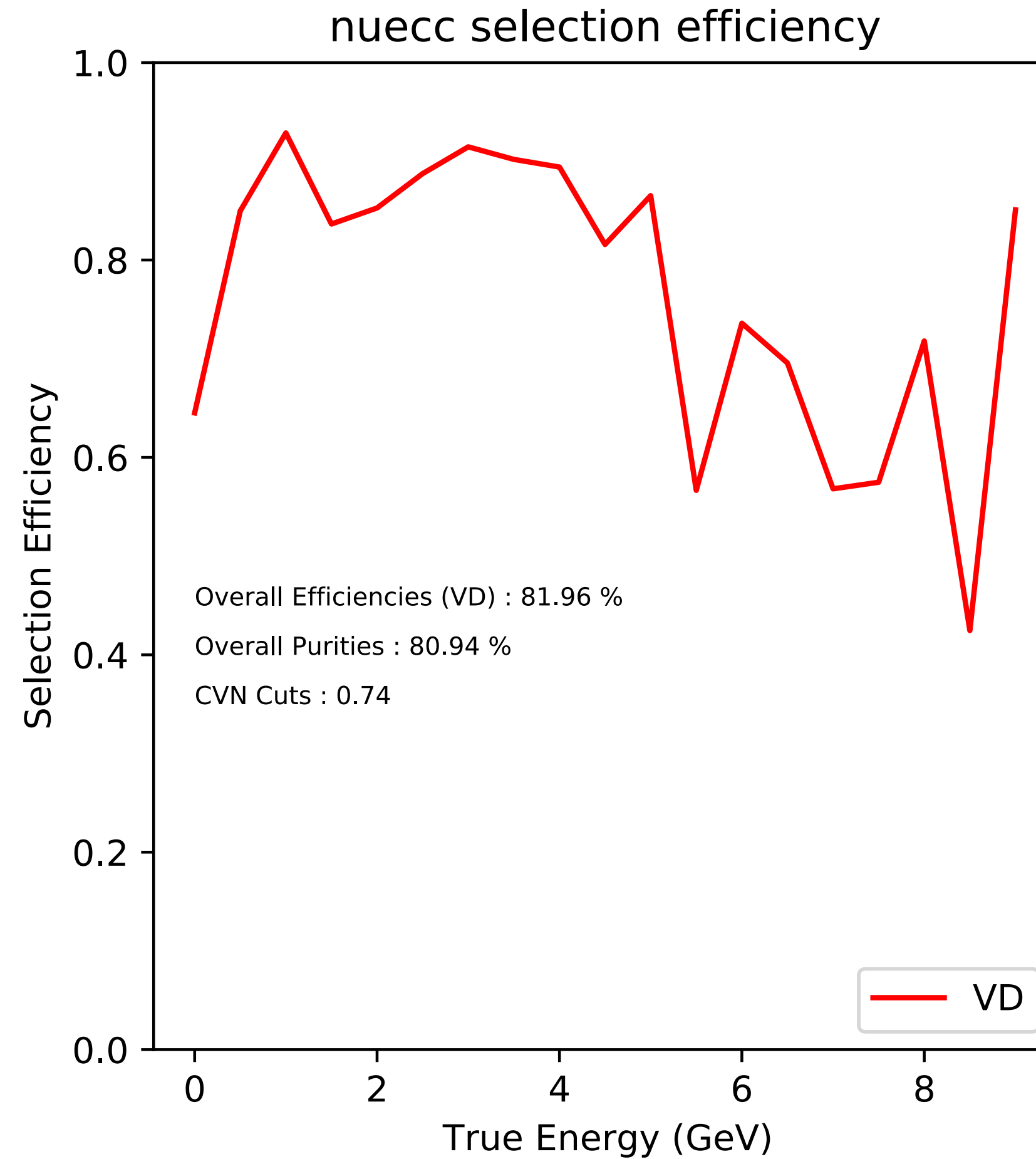
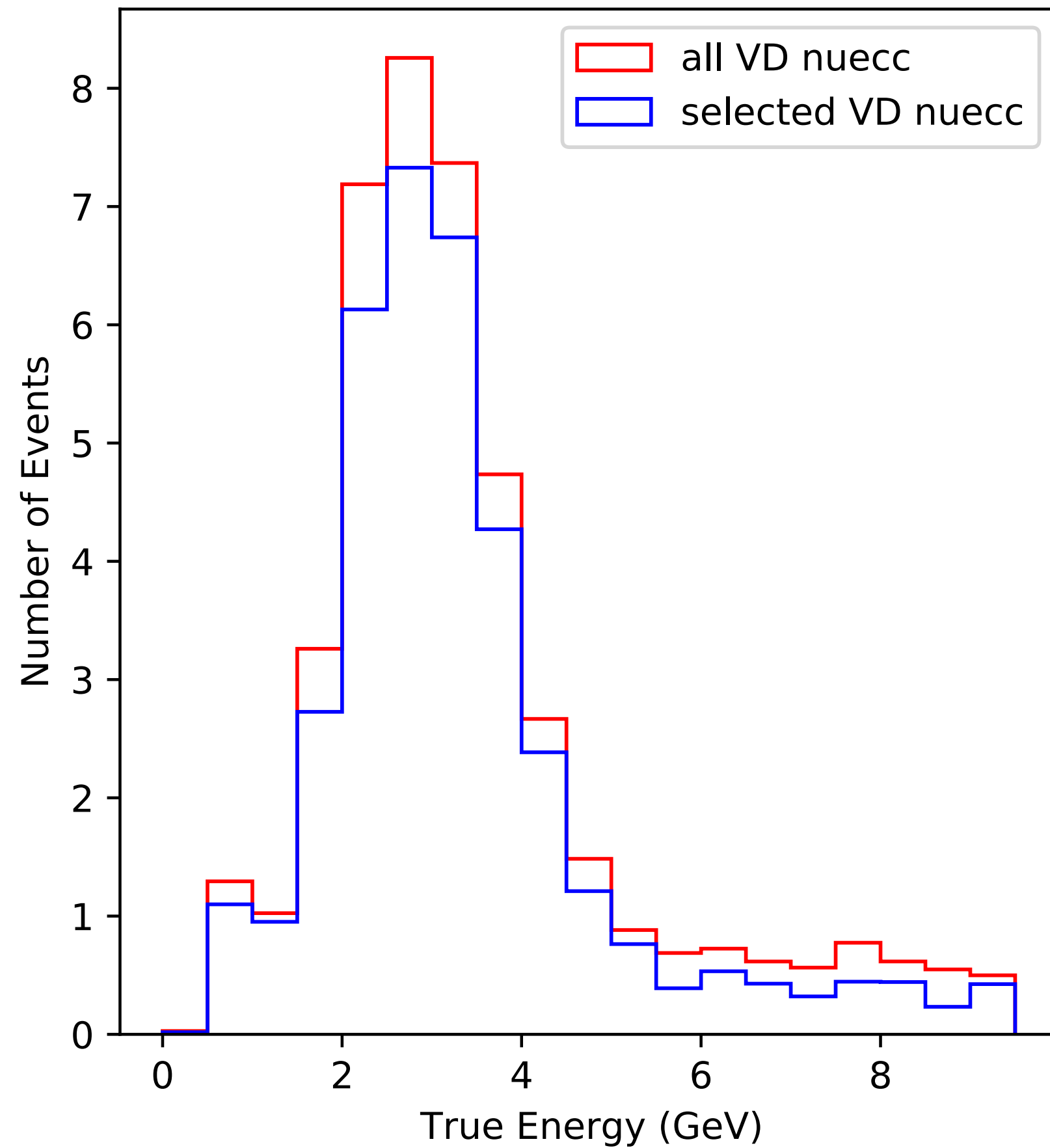
- Improvement from before, not a whole lot. Mainly in the nutauCC predictions
- Does do a little better differentiating nueCCs/numuCCs from NC

NueCC



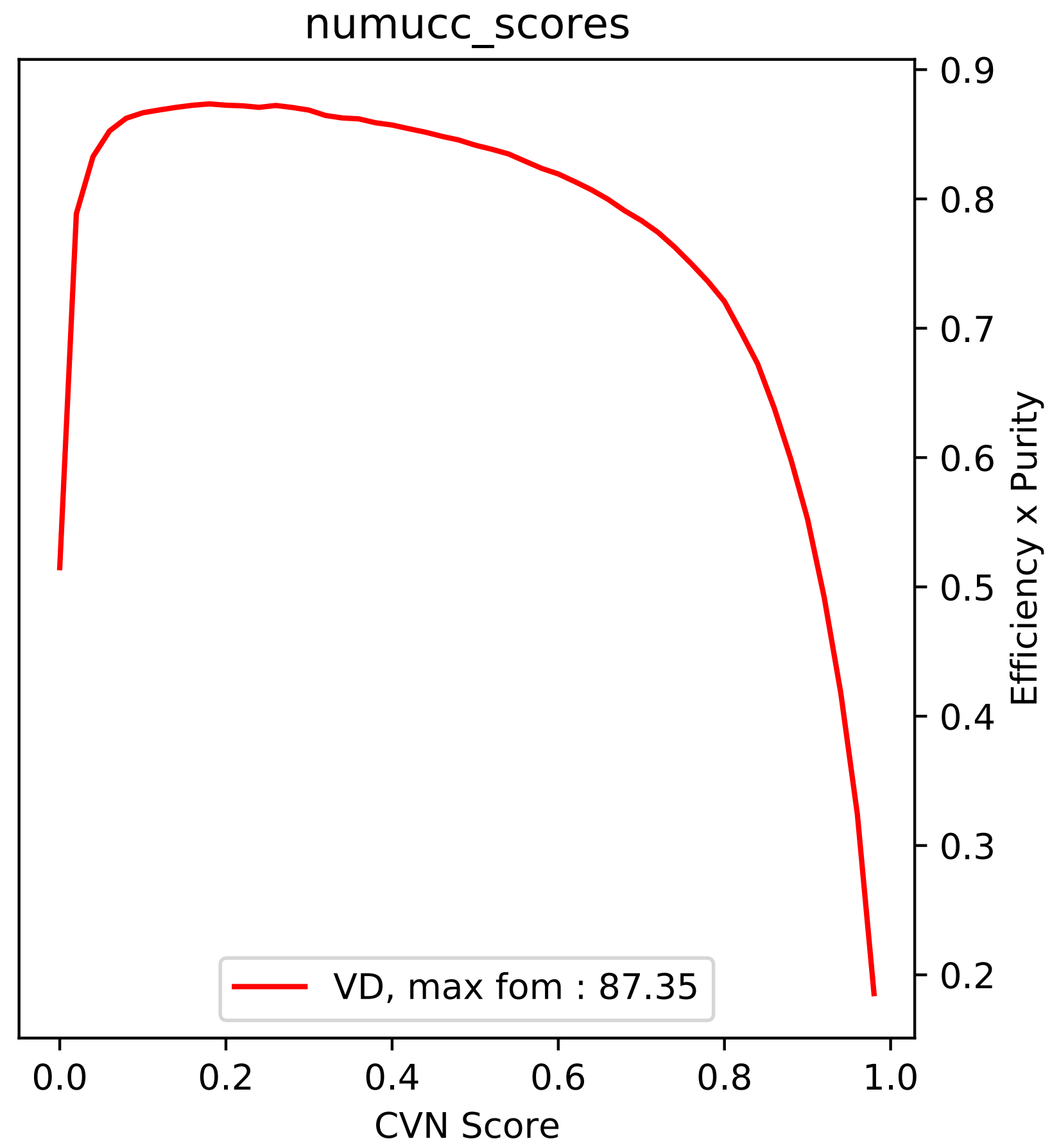
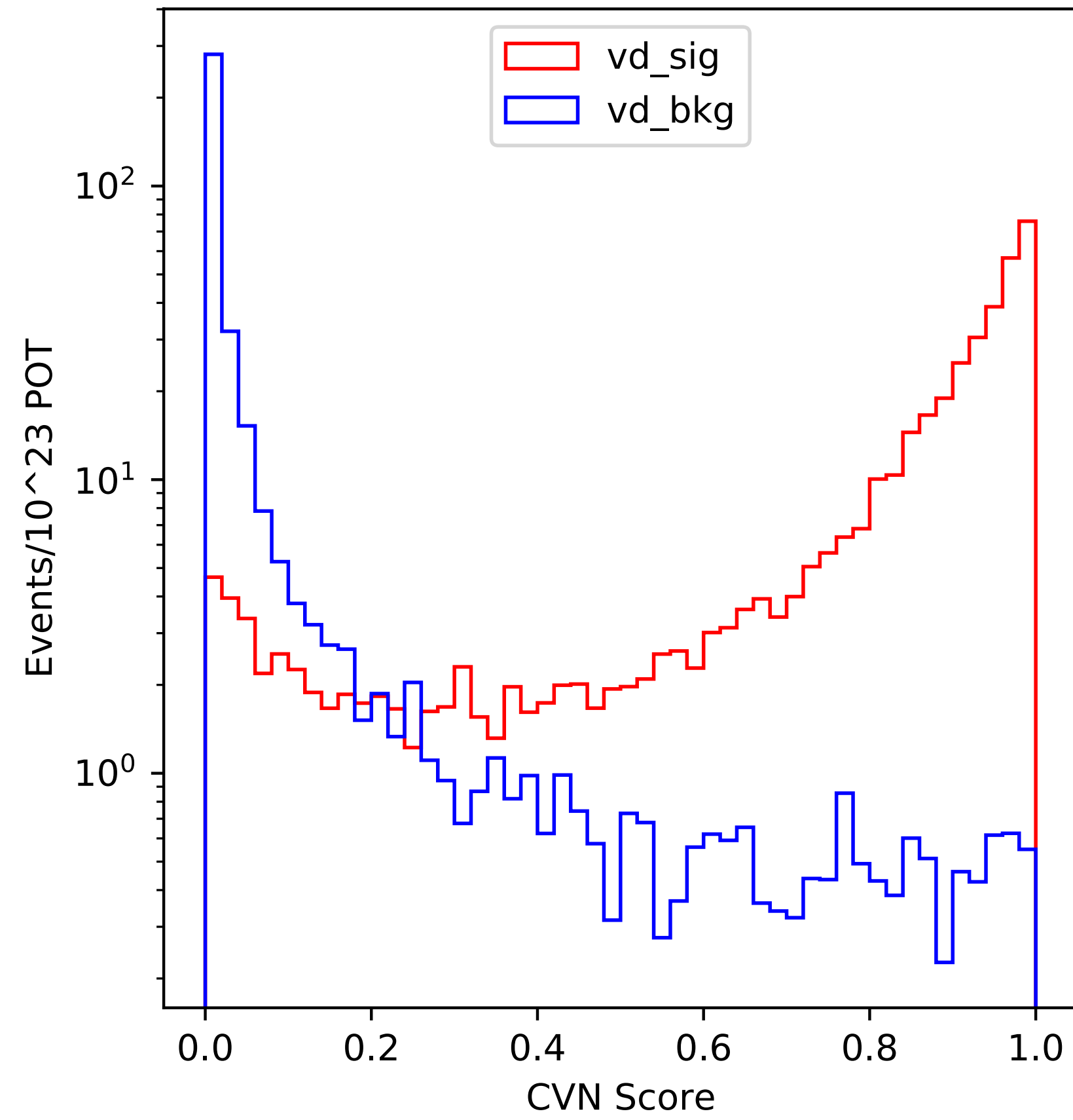
- Distributions with “dumb” oscillation weights + pot normalised

NueCC



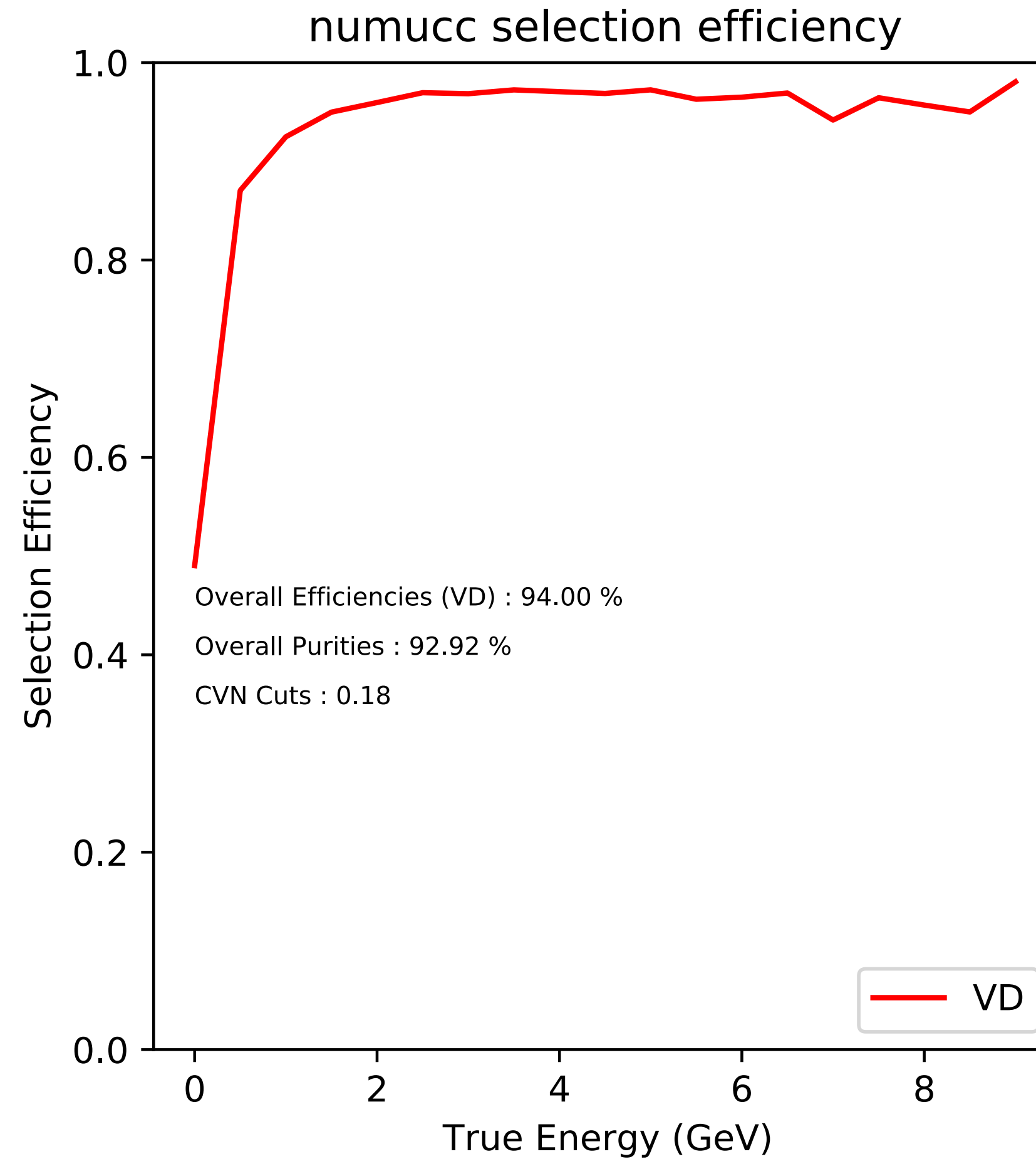
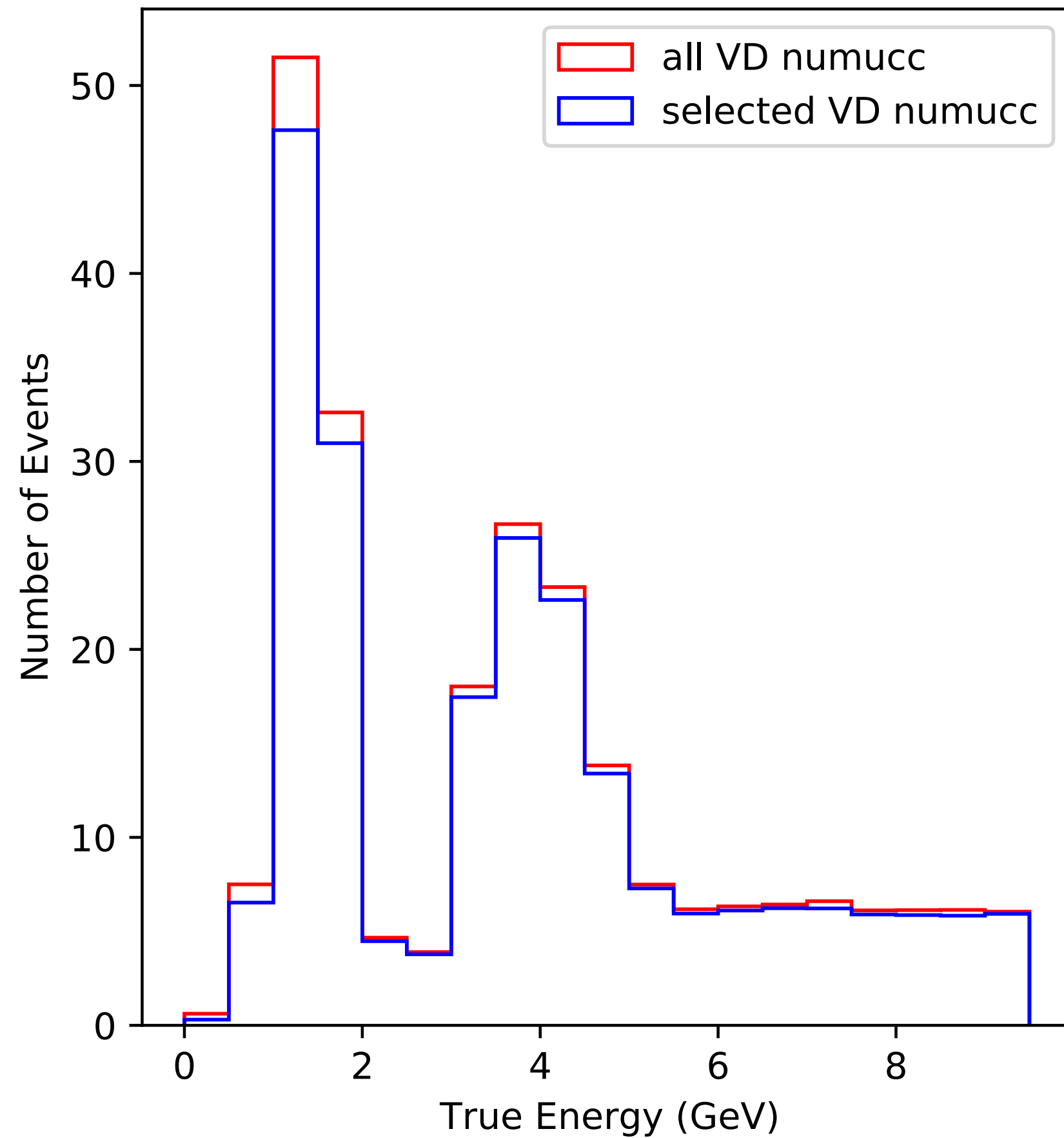
- Eff, Purity numbers after optimised CVN cut based on FOM - (82%, 81%)
- Better than before but not as good as the HD numbers

NumuCC



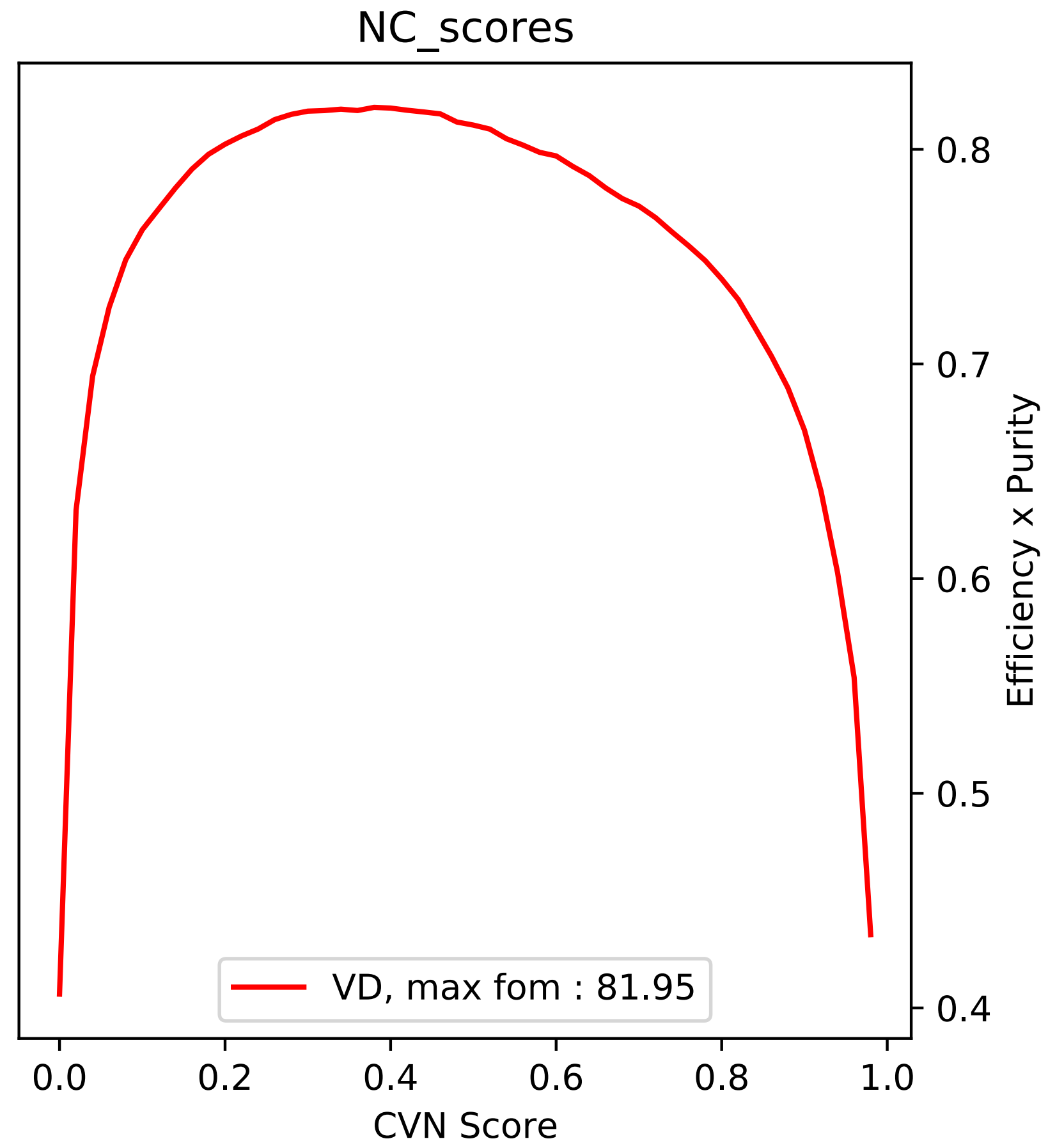
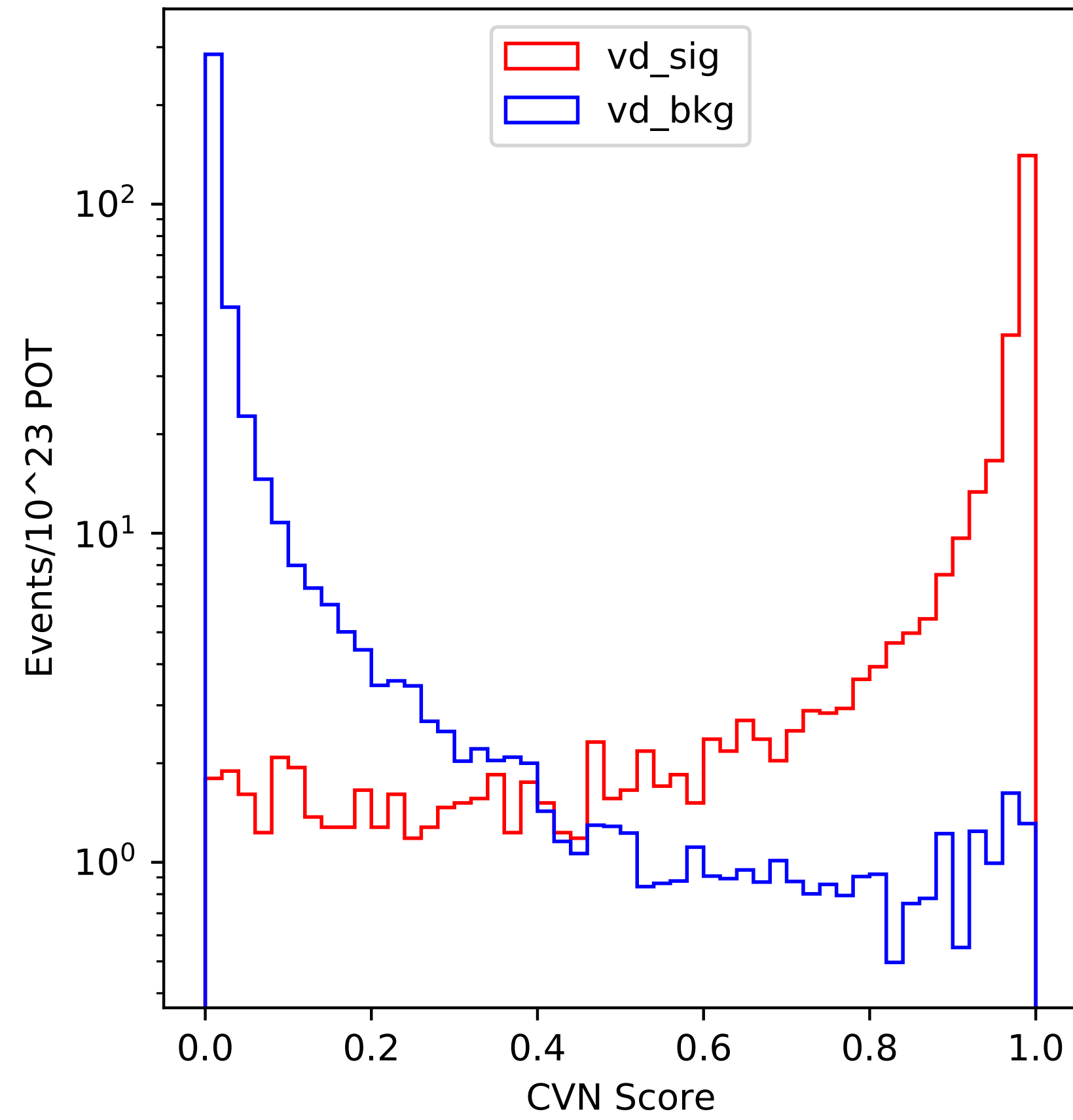
- Distributions with “dumb” oscillation weights + pot normalised

NumuCC



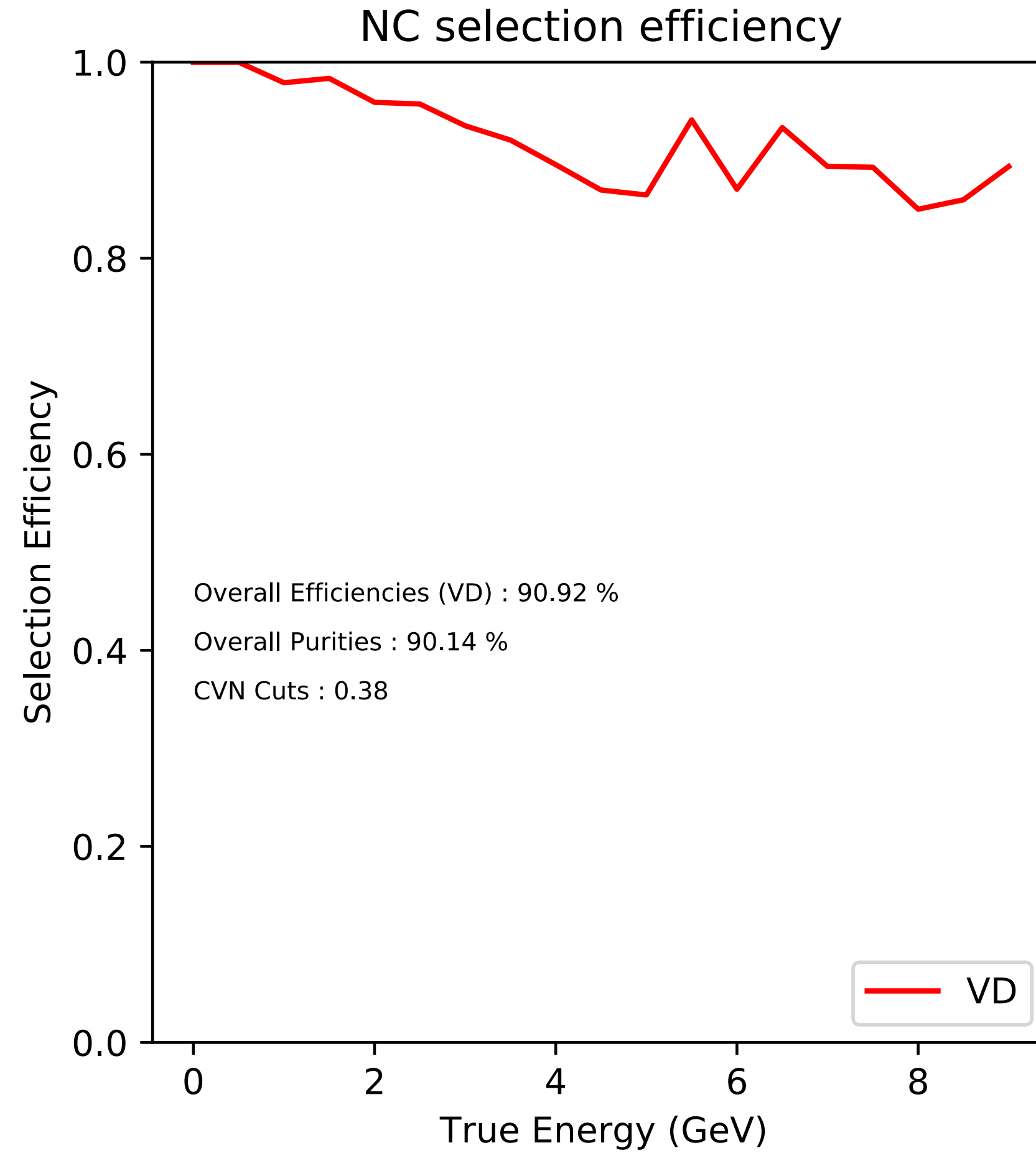
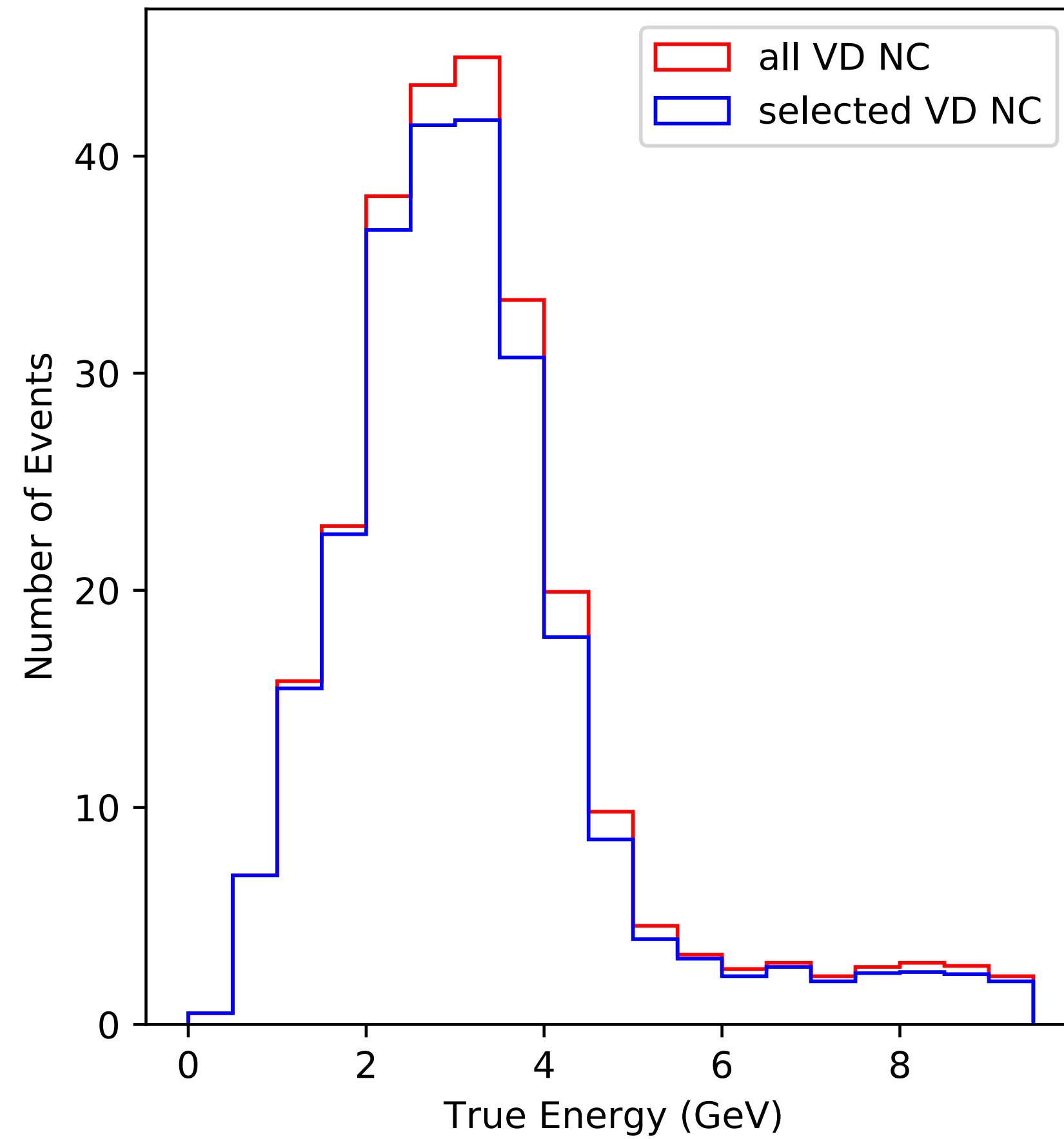
- Eff, Purity numbers after optimised CVN cut based on FOM - (94%, 93%)

NC



- Distributions with “dumb” oscillation weights + pot normalised

NC



- Eff, Purity numbers after optimised CVN cut based on FOM - (91%, 91%)

Summary

- Raw network performance for VD seems to be stabilised a little bit behind the old HD trainings
 - Biggest difference in efficiency and purity numbers for nueCCs
 - Not very far off but enough to matter?
 - Hard to say how much more we can squeeze out of it, unfortunately the biggest improvement from the large sample seems to come in the nuTauCC classification
- Can look at training using wire cell deconvolve charge inputs/sim channels to set upper bounds on these numbers
- Obviously lot is different in the new simulation. Any thoughts about this drop?