

# Study of $D^0$ decays to the invisible final states at Belle II

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# Introduction to analysis

- In SM, heavy (B or D) decays to  $\nu\bar{\nu}$  is helicity suppressed with an expected branching fraction of  $\text{Br}(D^0 \rightarrow \nu\bar{\nu}) = 1.1 \cdot 10^{-30}$ , which is beyond the reach of current collider experiments.

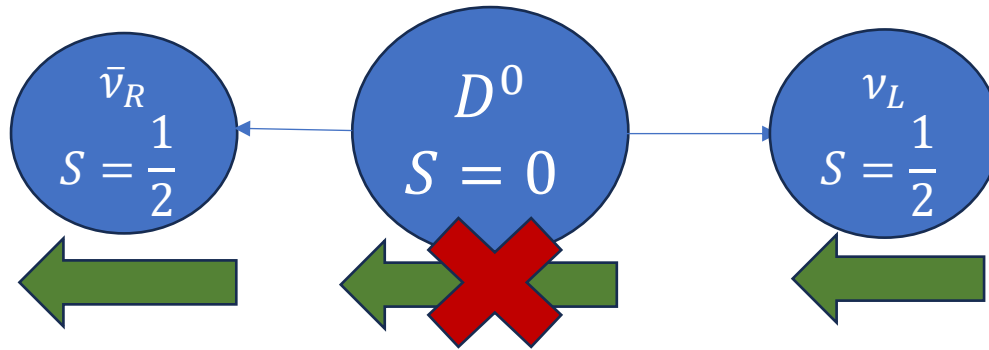


Figure1: Scheme of helicity suppressing

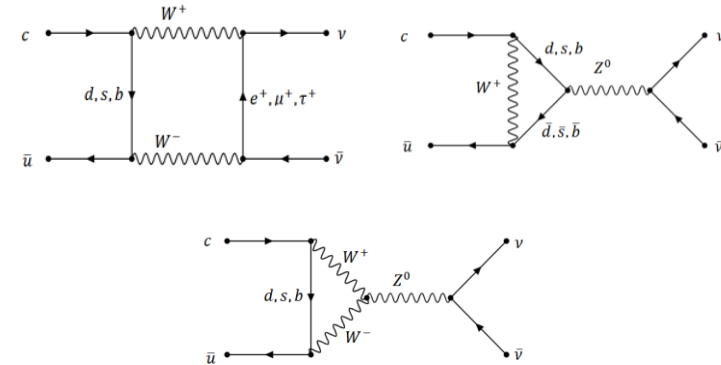


Figure2: Feynman diagram for  $D^0 \rightarrow \nu\bar{\nu}$

- Therefore, search for  $D^0 \rightarrow$  invisible final states is sensitive to new physics
- The previous result is  $BR_{UL} = 9.4 \times 10^{-5}$  on  $924 \text{ fb}^{-1}$  data samples at 90% CL at belle [Phys. Rev. D 95, 011102(R)]

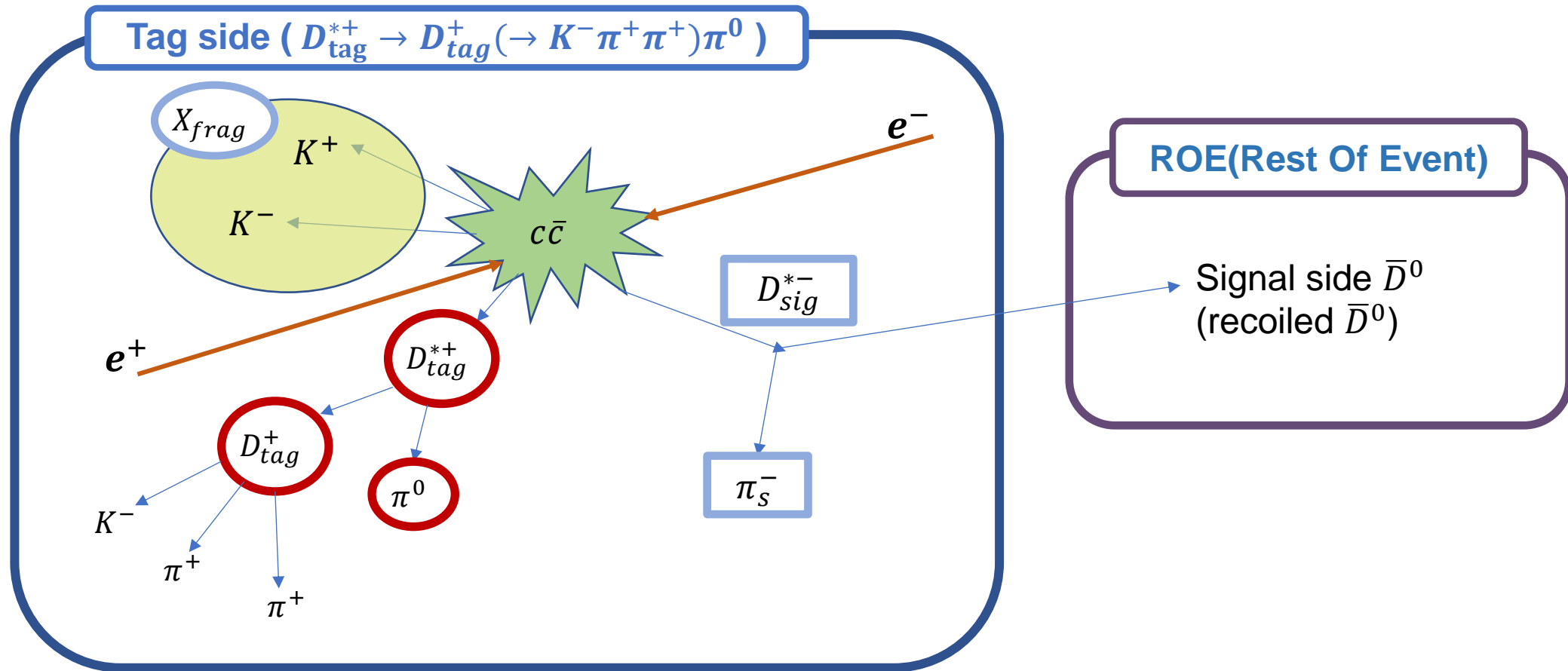
# MC simulation samples

- 20M signal MC samples were used
- Signal Event used for simulation :

$$e^+e^- \rightarrow c\bar{c} \rightarrow D_{tag}X_{frag}D_{sig}^{*+}$$
$$D_{sig}^{*+} \rightarrow D_{sig}^0\pi^+$$
$$D_{sig}^0 \rightarrow \nu\bar{\nu}$$

- MC15ri generic MC( $1ab^{-1}$ ) is used as generic background MC sample
- 20M Control sample ( $D^0 \rightarrow K^-\pi^+$ )
  - $D^0 \rightarrow K^+K^-, K^+\pi^-, K^-\pi^+\pi^0$  MC is also used as background for control sample study

# Analysis Method : Charm Tagger



**Figure3:** schematics of signal event with tag side decay  $D_{tag}^{*+} \rightarrow D_{tag}^+ (\rightarrow K^- \pi^+ \pi^+) \pi^0$

# Description of Charm tagging Procedure

## Reconstruction $D_{tag}, D_{tag}^*$

1. Reconstruct  $D_{tag}$  using the pre-chosen decay channels
2. Reconstruct  $D_{tag}^*$

## Recoil part 1 ( $D^{*+}$ )

1. Calculate  $M_{miss}(D_{tag}^{(*)}X_{frag})$  which is regarded as mass of  $D_{sig}^{*+}$
2. Apply kinematic mass constrained Fit on  $M_{miss}(D_{tag}^{(*)}X_{frag})$  to  $m_{D^{*+}}$
3. BCS of  $D_{sig}^{*+}$  by using chiProb from step 2

## Recoil part 2 ( $D^0$ )

5. Using slow pion, calculate  $M_{miss}(D_{tag}^*X_{frag}\pi_s^+)$  which is regarded as mass of signal side  $D^0$
6. BCS of  $D^0$  by using angle between  $D_{sig}^0$  and tag side hadron in cm frame

**Table1.** Tag reconstruction channels

$D^0$ decay	$Br(\%)$	$D^+$ decay	$Br(\%)$	$\Lambda_c^+$ decay	$Br(\%)$	$D_s^+$ decay	$Br(\%)$
$K^- \pi^+$	3.9	$K^- \pi^+ \pi^+$	9.4	$pK^- \pi^+$	5.0	$K^+ K^- \pi^+$	5.5
$K^- \pi^+ \pi^0$	13.9	$K^- \pi^+ \pi^+ \pi^0$	6.1	$pK^- \pi^+ \pi^0$	3.4	$K_S^0 K^+$	1.5
$K^- \pi^+ \pi^+ \pi^-$	8.1	$K_S^0 \pi^+$	1.5	$pK_S^0$	1.1	$K_S^0 K_S^0 \pi^+$	5.4
$K^- \pi^+ \pi^+ \pi^- \pi^0$	4.2	$K_S^0 \pi^+ \pi^0$	6.9	$\Lambda^0 \pi^+$	1.1	$K^+ K^- \pi^+ \pi^0$	5.6
$K_S^0 \pi^+ \pi^-$	2.9	$K_S^0 \pi^+ \pi^+ \pi^-$	3.1	$\Lambda^0 \pi^+ \pi^0$	3.6	$K_S^0 K^- \pi^+ \pi^+$	1.5
$K_S^0 \pi^+ \pi^- \pi^0$	5.4	$K^+ K^- \pi^+$	1.0	$\Lambda^0 \pi^+ \pi^+ \pi^-$	2.6	$K^+ \pi^- \pi^+ K_S^0$	1.0
$K^- \pi^+ \pi^0 \pi^0$	8.9	$K^- K^+ \pi^+ \pi^0$	0.7	$p^+ \pi^- \pi^+$	0.5	$\pi^+ \pi^- \pi^+$	1.0
$\pi^- \pi^+$	0.1	$\pi^- \pi^+ \pi^+$	0.3	$p^+ K^- K^+$	0.1	$\pi^+ K_S^0$	0.1
$\pi^- \pi^+ \pi^- \pi^+$	0.8	$\pi^- \pi^+ \pi^+ \pi^0$	1.2	$p^+ K^- \pi^+ \pi^0 \pi^0$	0.1	$\pi^+ \pi^0 K_S^0$	0.5
$\pi^- \pi^+ \pi^0$	1.5	$K^+ K_S^0 K_S^0$	0.3	$p^+ \pi^- \pi^+ \pi^- \pi^+$	0.2	$K^- K^+ \pi^+ \pi^- \pi^+$	0.7
$\pi^- \pi^+ \pi^0 \pi^0$	1.0	$\pi^+ \pi^0$	0.1	$p^+ K_S^0 \pi^0$	2.0		
$K^- K^+$	0.4			$p^+ K_S^0 \pi^+ \pi^-$	1.6		
$K^- K^+ \pi^0$	0.3			$\pi^+ \pi^- \Sigma^+$	4.5		
$K^- K^+ K_S^0$	0.4			$\pi^+ \pi^- \pi^0 \Sigma^+$	1.2		
$\pi^0 K_S^0$	1.2			$\pi^0 \Sigma^+$	1.2		
sum	53.1	sum	30.5	sum	28.2	sum	22.8

$D_{tag}^*$  reconstruction channels  
and fragmentations for each tag particle

$D^{*+}$ decay	Br(%)	$D^{*0}$ decay	Br(%)	$D_s^{*+}$ decay	Br(%)
$D^0 \pi^+$	67.7	$D^0 \pi^0$	61.9	$D_s^+ \gamma$	93.5
$D^+ \pi^0$	30.7	$D^0 \gamma$	38.1		
sum	98.4	sum	100.0	sum	93.5

**Table3:**  $D_{tag}^*$  channel

$D^{*+}$ or $D^+$	$D^{*0}$ or $D^0$	$\Lambda_c^+$	$D_s^{*+}$ or $D_s^+$
nothing( $K^+ K^-$ )	$\pi^+(K^+ K^-)$	$\pi^+ \bar{p}$	$K_S^0$
$\pi^0(K^+ K^-)$	$\pi^+ \pi^0(K^+ K^-)$	$\pi^+ \pi^0 \bar{p}$	$\pi^0 K_S^0$
$\pi^+ \pi^-(K^+ K^-)$	$\pi^+ \pi^+ \pi^-(K^+ K^-)$	$\pi^+ \pi^- \pi^+ \bar{p}$	$\pi^+ K^-$
$\pi^+ \pi^- \pi^0(K^+ K^-)$			$\pi^+ \pi^- \pi^0 K_S^0$
			$\pi^+ K^-$
			$\pi^+ \pi^0 K^-$
			$\pi^+ \pi^- \pi^+ K^-$

**Table4:**  $X_{frag}$  channel (total 24 channels)

# fastBDT training for Charm Tagging

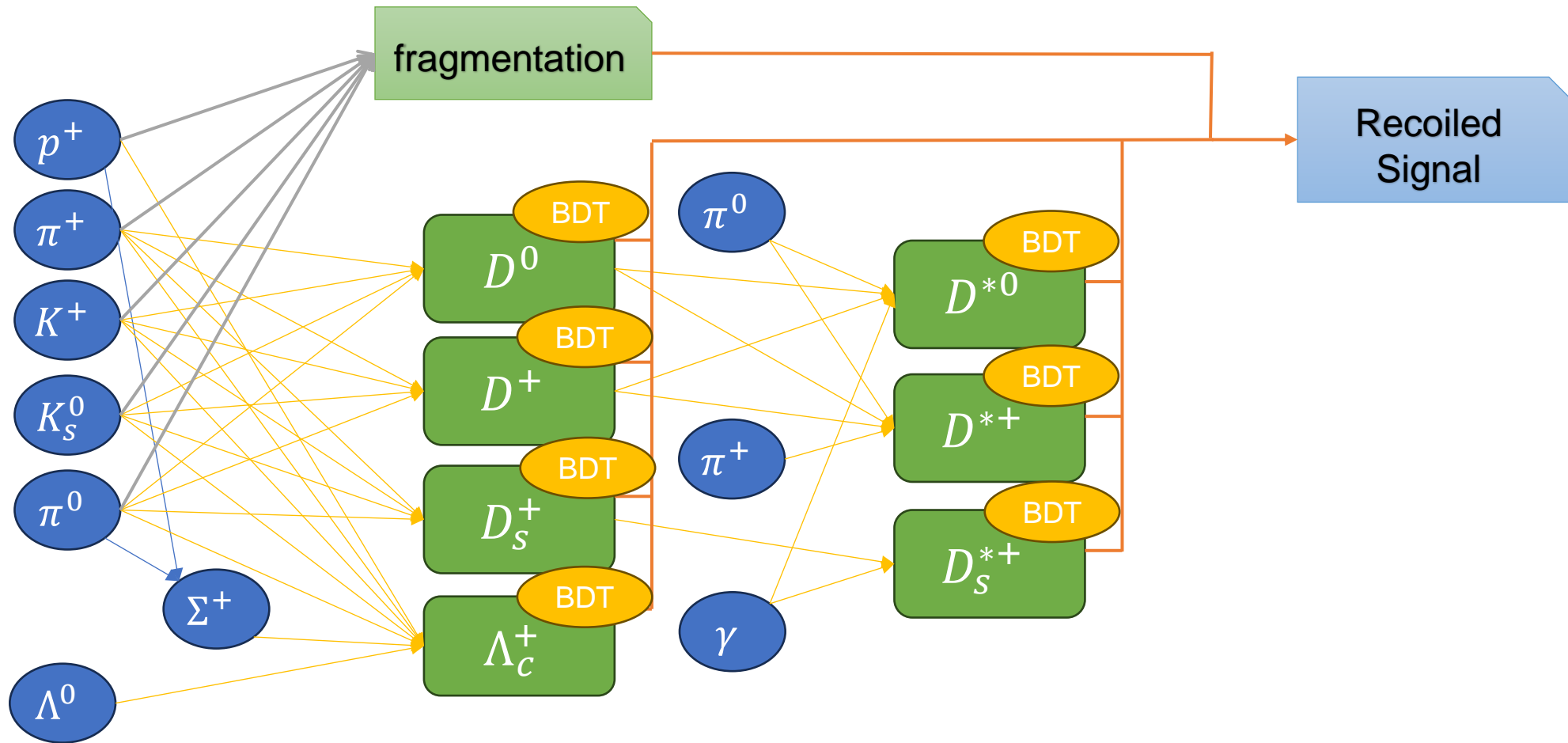
- Input Variables of fastBDT  
(reduce # of input variables according to high correlation and low importance)
  - For  $D_{tag}$  training  
M, xp, dr(flight length), dz, chiProb, cosToThrustOfEvent,  
cosAngleBetweenMomentumAndVertexVectorInXYPlane,  
PID of daughters, cosHelicityAngle(2 body or 3 body decays),  
angle between 2 daughters of  $\pi^0(\rightarrow \gamma\gamma)$ ,  $K_S^0(\rightarrow \pi^+\pi^-)$ ,  $\Lambda^0(\rightarrow p^+\pi^-)$ ,  $\Sigma^+(\rightarrow p^+\pi^0)$ ,  
 $|\frac{E_{d1}-E_{d2}}{E_{d1}+E_{d2}}|$  of  $\pi^0(\rightarrow \gamma\gamma)$ ,  $K_S^0(\rightarrow \pi^+\pi^-)$ ,  $\Lambda^0(\rightarrow p^+\pi^-)$ ,  $\Sigma^+(\rightarrow p^+\pi^0)$  etc...
  - For  $D_{tag}^*$  training  
 $\Delta M(= M_{D_{tag}^*} - M_{D_{tag}})$ , momentum of  $\pi_S^\pm$ ,  $\gamma$ ,  $\pi^0$ ,  
angle between  $D_{tag}$  and  $\pi_S^\pm$ ,  $\gamma$ ,  $\pi^0$  etc...
- Hyper Parameters of BDT was optimized by applying grid search for each tag training



# Preselection of Charm Tagger

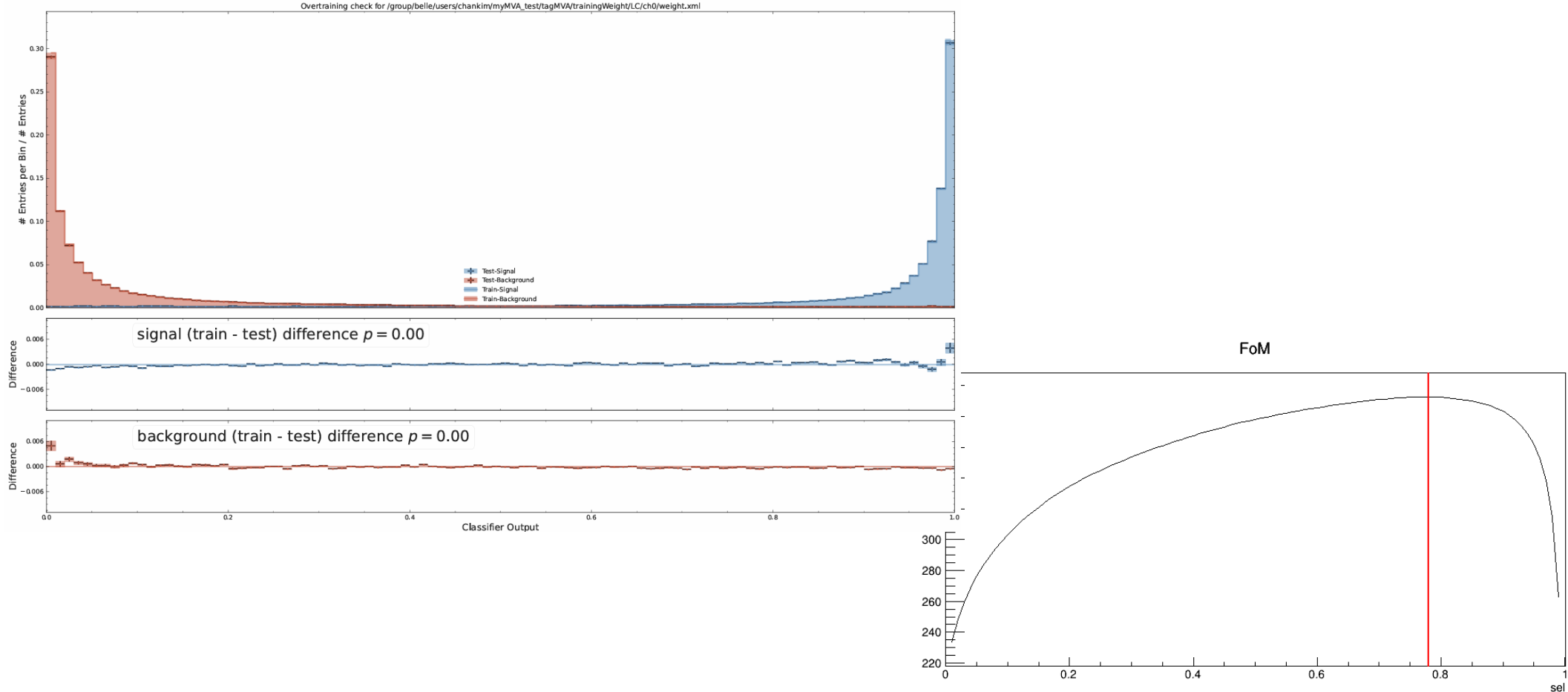
- For tracks :  $dr < 1.0$  ,  $|dz| < 3.0$  and InCDCAcceptance
- $\pi^\pm$  : 15 candidates with Highest pionID after pionID  $> 0.01$
- $K^\pm$  : 10 candidates with Highest kaonID after kaonID  $> 0.1$
- $p^\pm$  : 10 candidates with highest protonID after protonID  $> 0.1$
- $\gamma$  : beamBackgroundSuppression  $> 0.5$  & fakePhotonSuppression  $> 0.1$   
( $E > 0.1$  for  $\gamma$  in  $D_s^{*+} \rightarrow D_s^+ \gamma, D^{*0} \rightarrow D^0 \gamma$ )
- for fragmentations, PID selection of  $\pi^\pm, K^\pm, p^\pm$  is on 0.1, 0.9, 0.9 and additionally require  $p > 0.1$  GeV
  
- $K_S^0, \Lambda^0$  :
  - mass and  $dr$  and  $\chi^2$  and angle between Momentum and Vertex Vector selection on  $\Lambda^0$
  - goodBelleKshort for  $K_S^0$  (similar selection to  $\Lambda^0$ )
- $\Sigma^+$  : reconstructed from  $\Sigma^+ \rightarrow p^+ \pi^0$  and mass cut ( $1.08 < M < 1.28$ )

# Flow of Charm Tagger



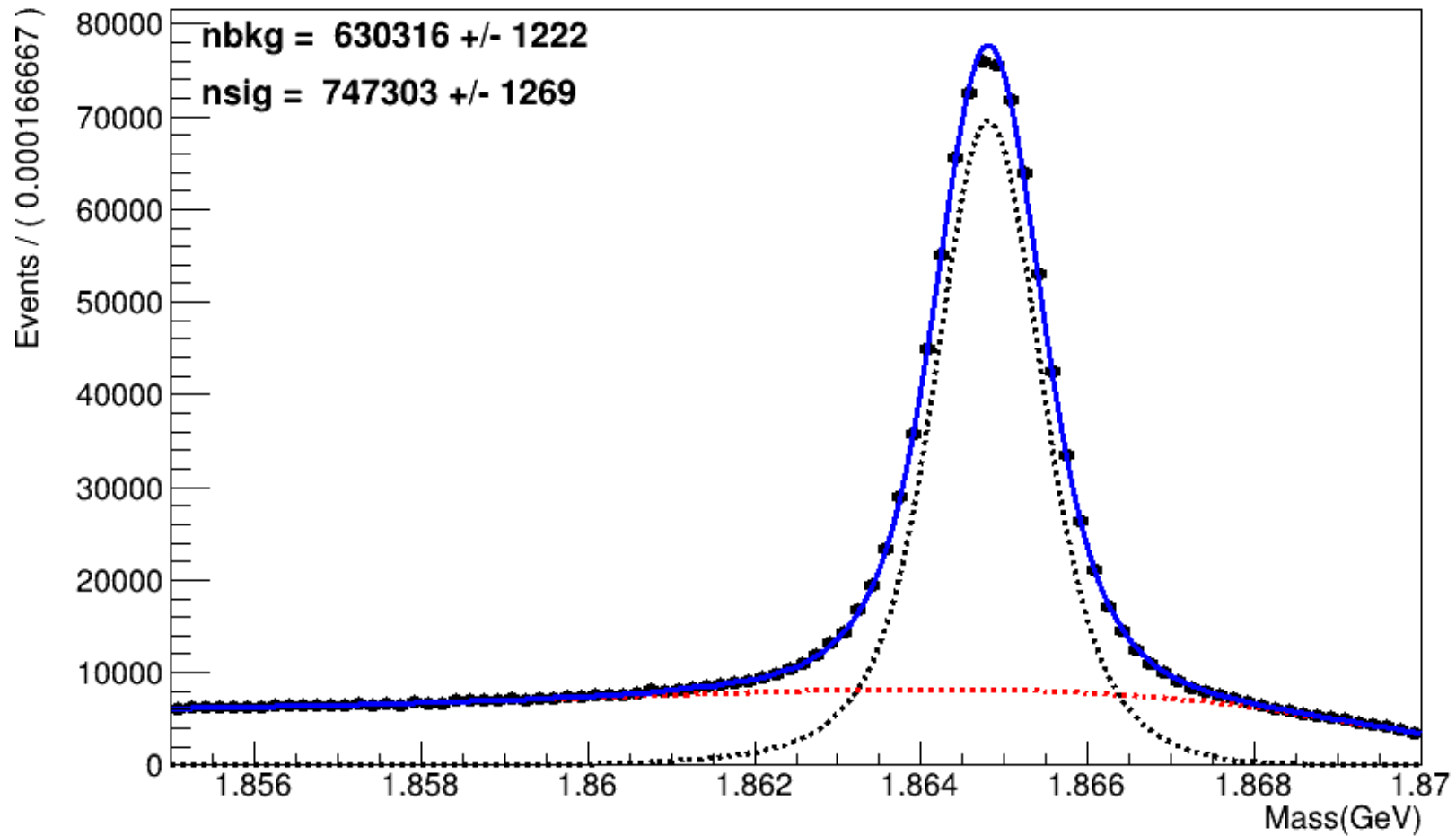
# One example about training :

$$\Lambda_c^+ \rightarrow p^+ K^- \pi^+$$



# Reconstructed $D^0$ from charm tagger on generic MC

Inclusive recoil  $D^0$  Mass



# Variables for extracting signal side D

- Recoil mass ( $M_{miss}(D_{tag}^* X_{frag} \pi_s^\pm)$  or  $M_{recoil}(D^0)$ )
  - $e^+ e^- \rightarrow D_{tag}^* X_{frag} \pi_s^\pm D^0$
  - $p^\mu(e^+) + p^\mu(e^-) - (p^\mu(D_{tag}^*) + p^\mu(X_{frag}) + p^\mu(\pi_s^\pm)) = p^\mu(D_{sig}^0)$
  - $M_{recoil}(D^0) = \sqrt{p^\mu(D^0) * p_\mu(D^0)}$
  - **Inclusive  $D^0$  : recoiled  $D^0$  (no requirement on signal side)**  
=> 1D fit on signal side recoil  $M_{D^0}$
- $E_{ECL}$  : sum of energies from roe of tag side remained in electromagnetic calorimeter(ECL) cluster
  - **Exclusive  $D^0$  : recoiled  $D^0$  (requirement on signal side)**  
=> 2D fit on signal side ( $M_{D^0}, E_{ECL}$ )

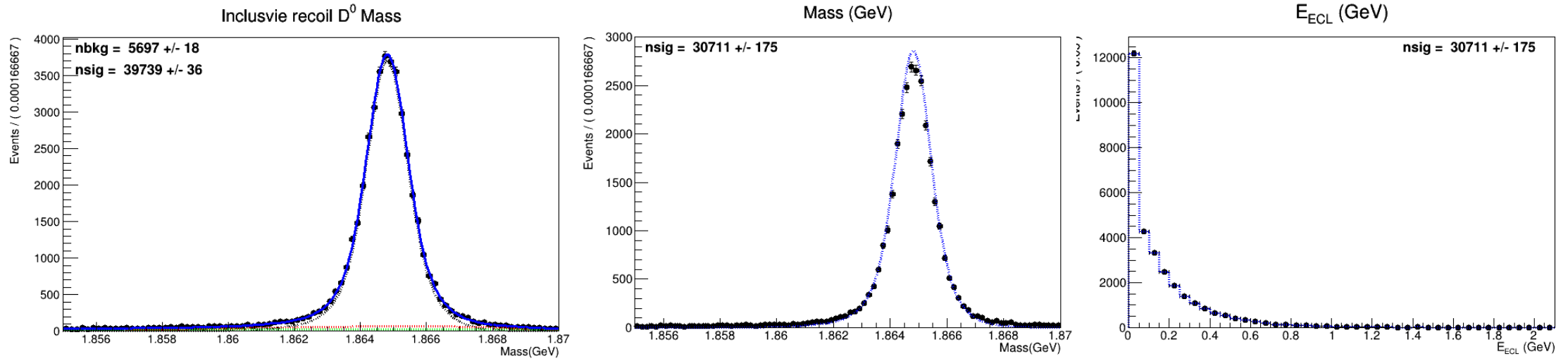
# Fitting strategy

- Inclusive D fitting
  - 1D fitting :  $M_{D^0}$
  - Signal pdf : 2 gaussians + 1 bifurcated gaussian
  - Background pdf : argus + linear
- Exclusive D fitting
  - 2D fitting :  $(M_{D^0}, E_{ECL})$
  - Signal pdf : signal pdf from inclusive D fitting & histogram pdf from signal MC study
  - Background pdf :
    - Flat: Argus + linear & histogram PDF from MC study
    - Peak: 3 gaussians & histogram PDF from MC study

## Exclusive D requirement (signal extraction)

- Exclusive :  $D^0$  with selection on signal side  
fit on the  $1.855 \text{ GeV} < M_{D^0} < 1.870 \text{ GeV}$  &  $E_{ECL} < 2.1 \text{ GeV}$ 
  - Selection for exclusive  $D^0$  on **Signal MC** ( $D^0 \rightarrow \nu\bar{\nu}$ )
    - no remaining tracks,  $\pi^0, K_L^0, K_S^0, \Lambda^0$
  - Selection for exclusive  $D^0$  on **Control sample** ( $D^0 \rightarrow K^-\pi^+$ )
    - 2 remaining tracks and 1 reconstructed  $D^0(K^-\pi^+)$
    - no  $\pi^0, K_L^0, K_S^0, \Lambda^0$
    - $|\Delta E| < 0.1 \text{ GeV}$  ( $\Delta E \equiv E(\text{recoil } D^0) - E_{K\pi}$ )

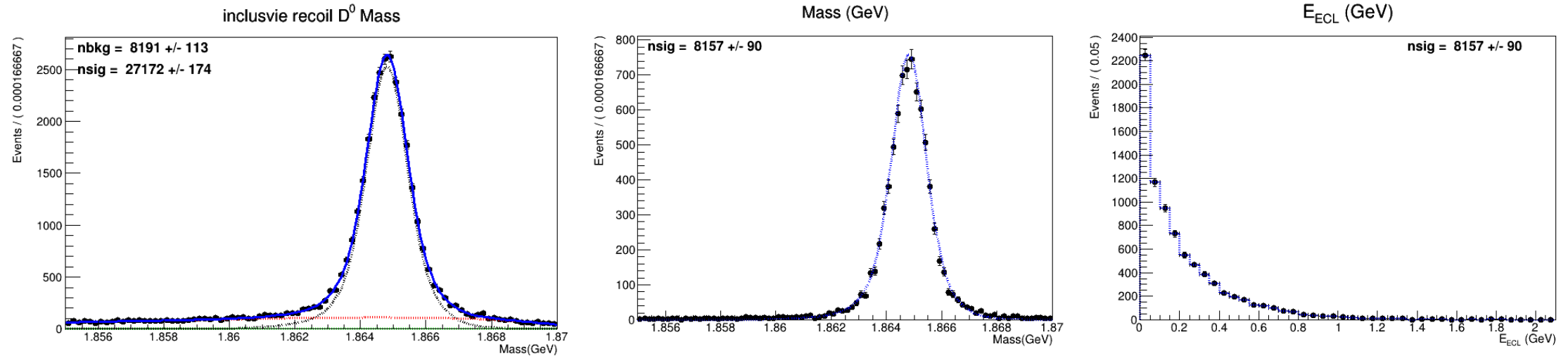
# Signal efficiency on signal MC



- Signal eff :  $\frac{30711 \pm 175}{39739 \pm 36} = 0.77282 \pm 0.00446$

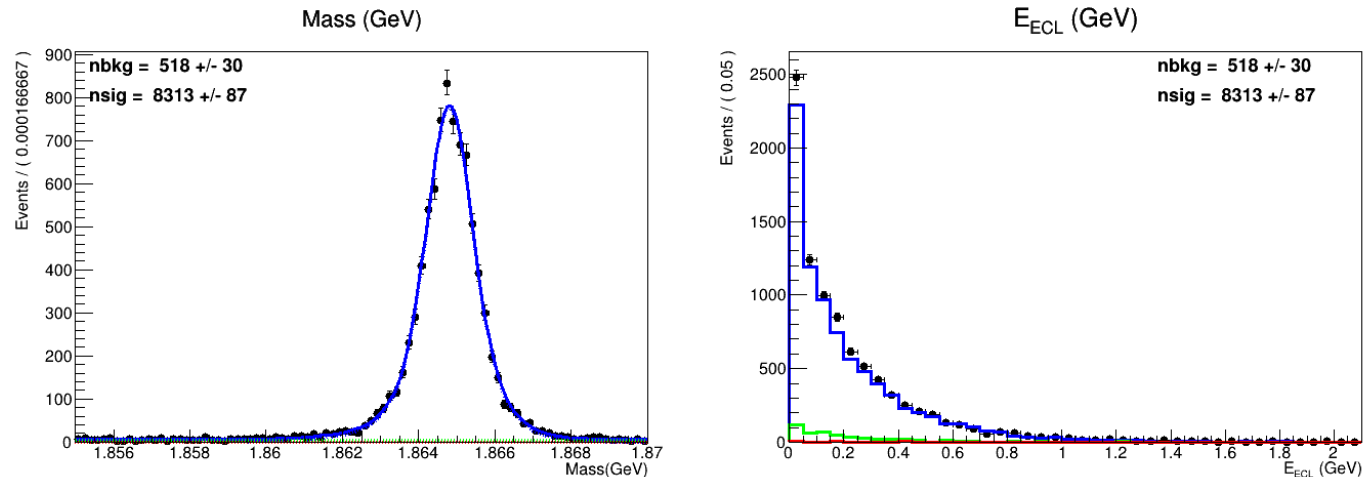


# Signal efficiency on control sample



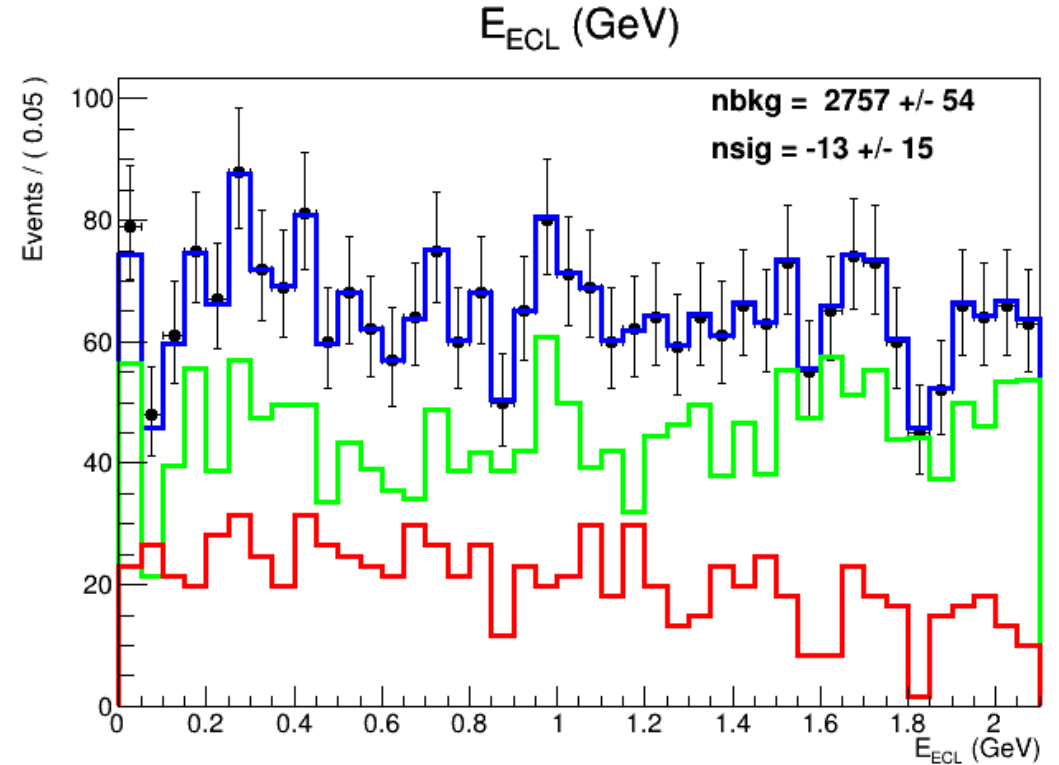
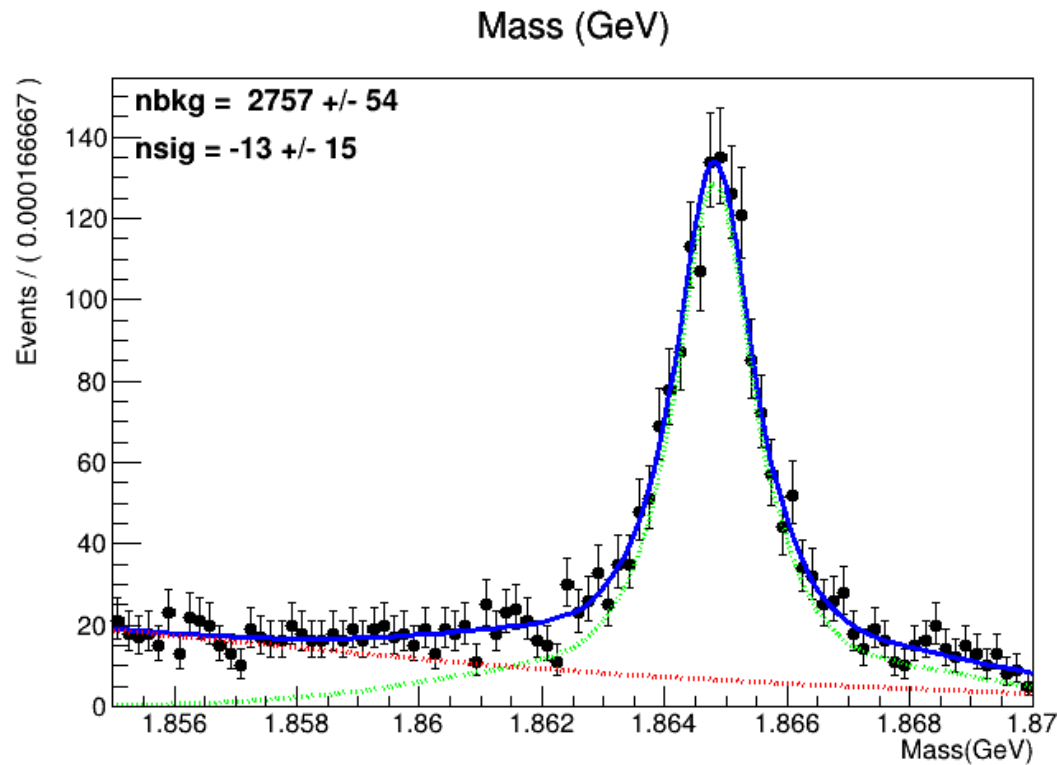
- Signal eff :  $\frac{8157 \pm 90}{27172 \pm 174} = 0.30020 \pm 0.00383$

# BR measurement on generic MC(uds/mixed/charged)



- $Br(D^0 \rightarrow K^- \pi^+) = \frac{N^{exclusive}}{N^{inclusive} * \epsilon_{sig}} = \frac{8313 \pm 87}{(747303 \pm 1269) * (0.30020 \pm 0.00383)} = 0.03706 \pm 0.00061$ 
  - Far from decile BR value(0.0395)  $\sim 4\sigma$
- The # of true signal event identified by TopoAna :
 
$$8757 \pm 94 \Rightarrow Br(D^0 \rightarrow K^- \pi^+) = 0.03903 \pm 0.00065$$
- This difference seems to be from similarity of  $E_{ECL}$  shape for the peaking background component
  - Main source is  $D^0 \rightarrow K^+ \pi^-$ ,  $K^+ K^-$ ,  $K^- \pi^+ \pi^0$  consists of 80% of background events
  - It seems to be irreducible error with current fitting strategy...
  - For validation of charm tagger, it is also possible to apply other fit strategy only for signal extraction about this control sample study

# 2D fit on generic MC for $D^0 \rightarrow$ invisibles



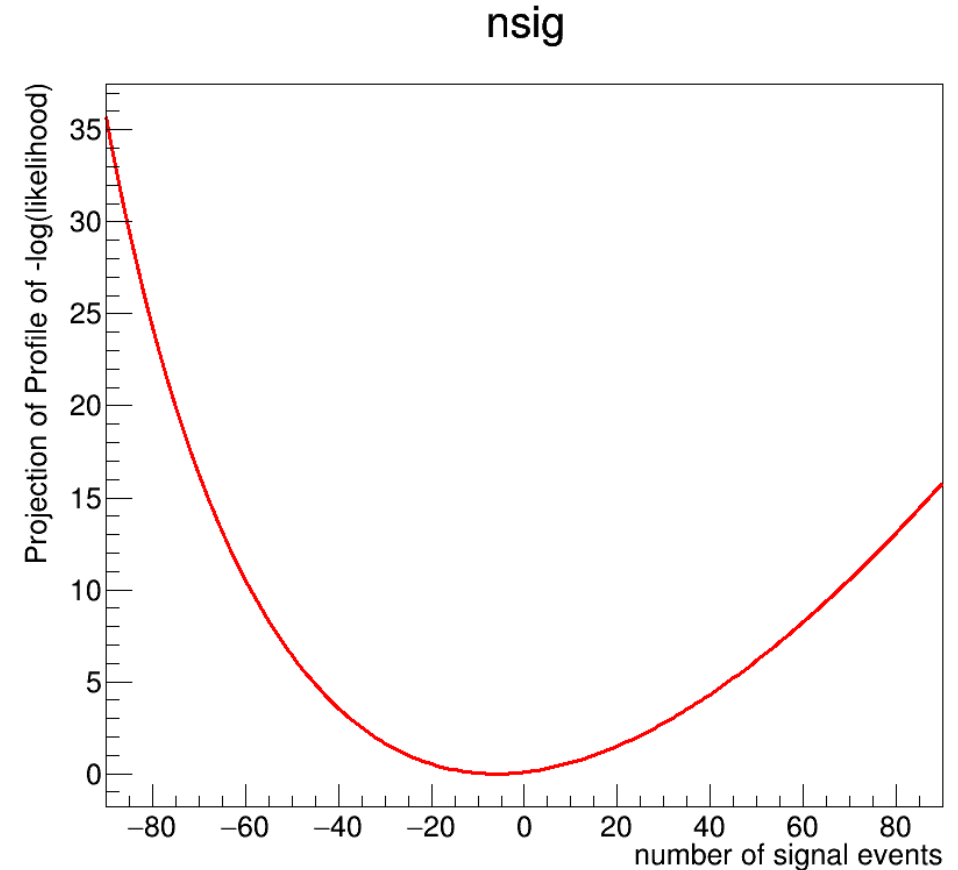
# Upper limit estimation of $D^0 \rightarrow$ invisibles

- Way1) Upper Limit estimation by integration of likelihood function

$$\int_0^{N_{UL}} L(n) dn = 0.9 \int_0^{\infty} L(n) dn$$

$$N_{UL} = 11.4984$$

$$BR_{UL} = \frac{11.4984}{(747303 \cdot 0.77282)} = 2.0 \times 10^{-5}$$



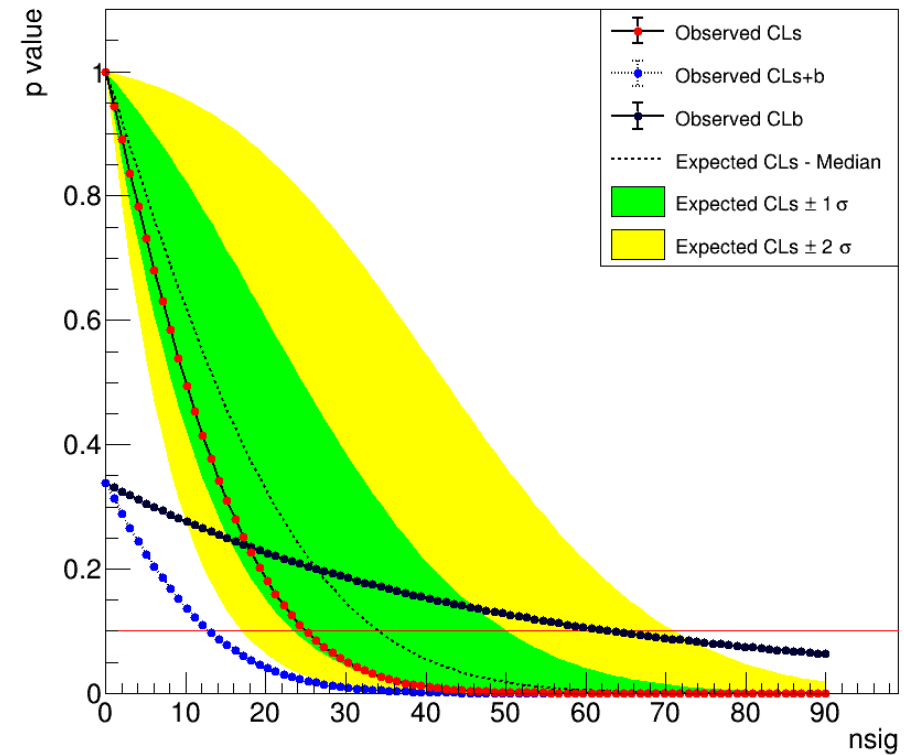
# Upper limit estimation of $D^0 \rightarrow$ invisibles

- Way2) Upper limit estimation by CLs method

$$N_{UL} = 22.0664$$

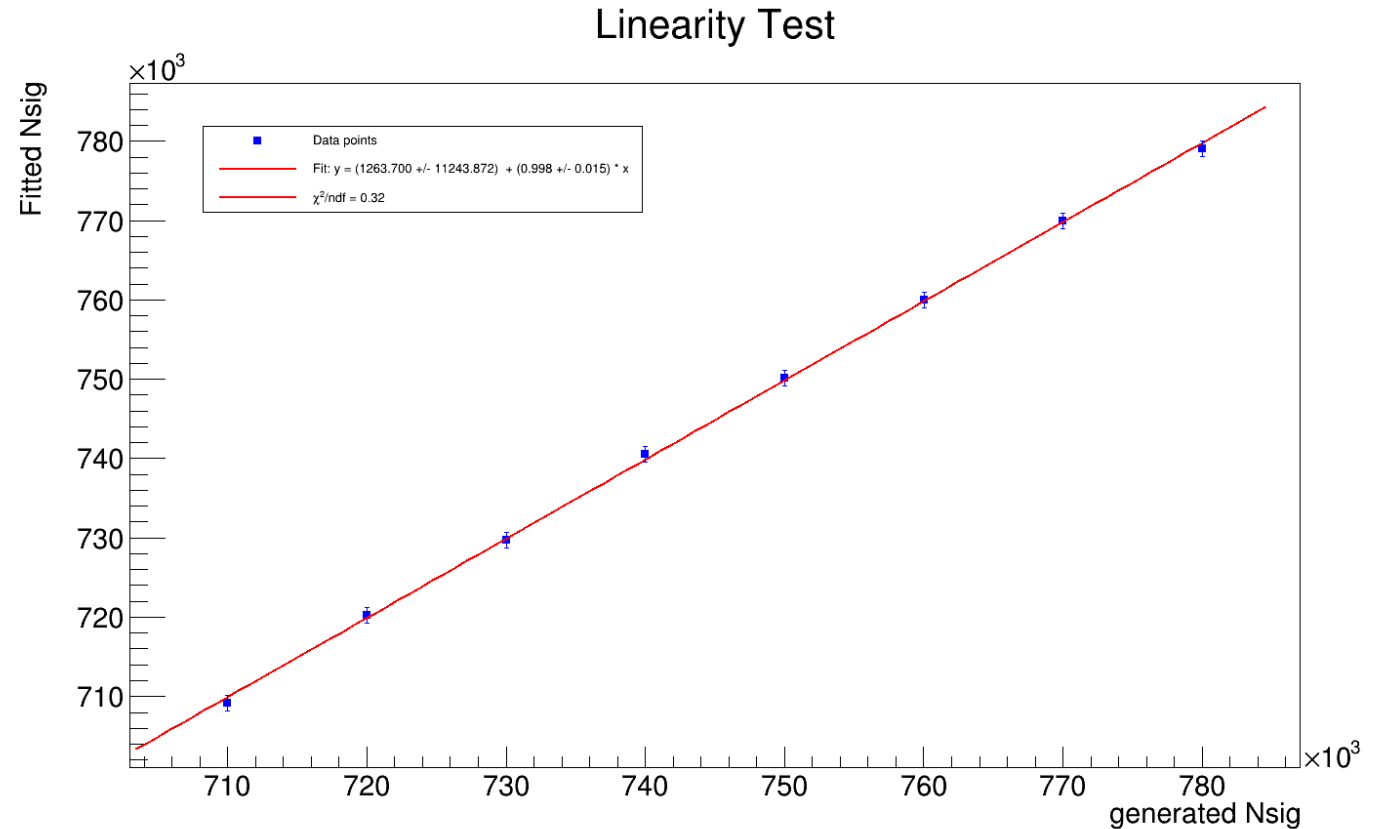
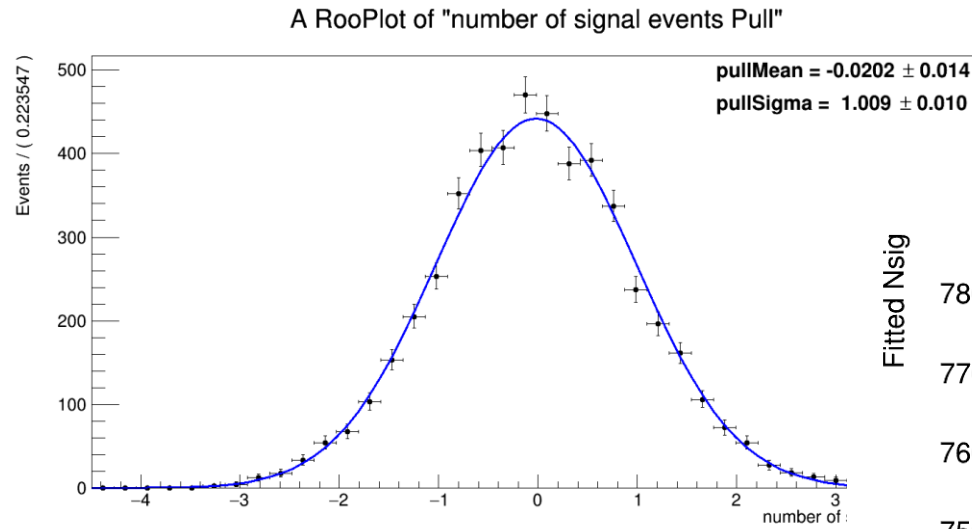
$$BR_{UL} = \frac{22.0664}{(747303 * 0.77282)} = 3.8 \times 10^{-5}$$

HypoTest Scan Result





# Inclusive D fit result check with ToyMC on generic MC



# As next step

- Check control sample BR measurement result with other fit strategy using reconstructed  $D^0$  instead of recoil mass
- Move on the run-dependent MC sample
- Study systematics
  - => 1<sup>st</sup> priority : systematic uncertainty from charm tagger
  - => PID, tracking, fitting strategy etc...

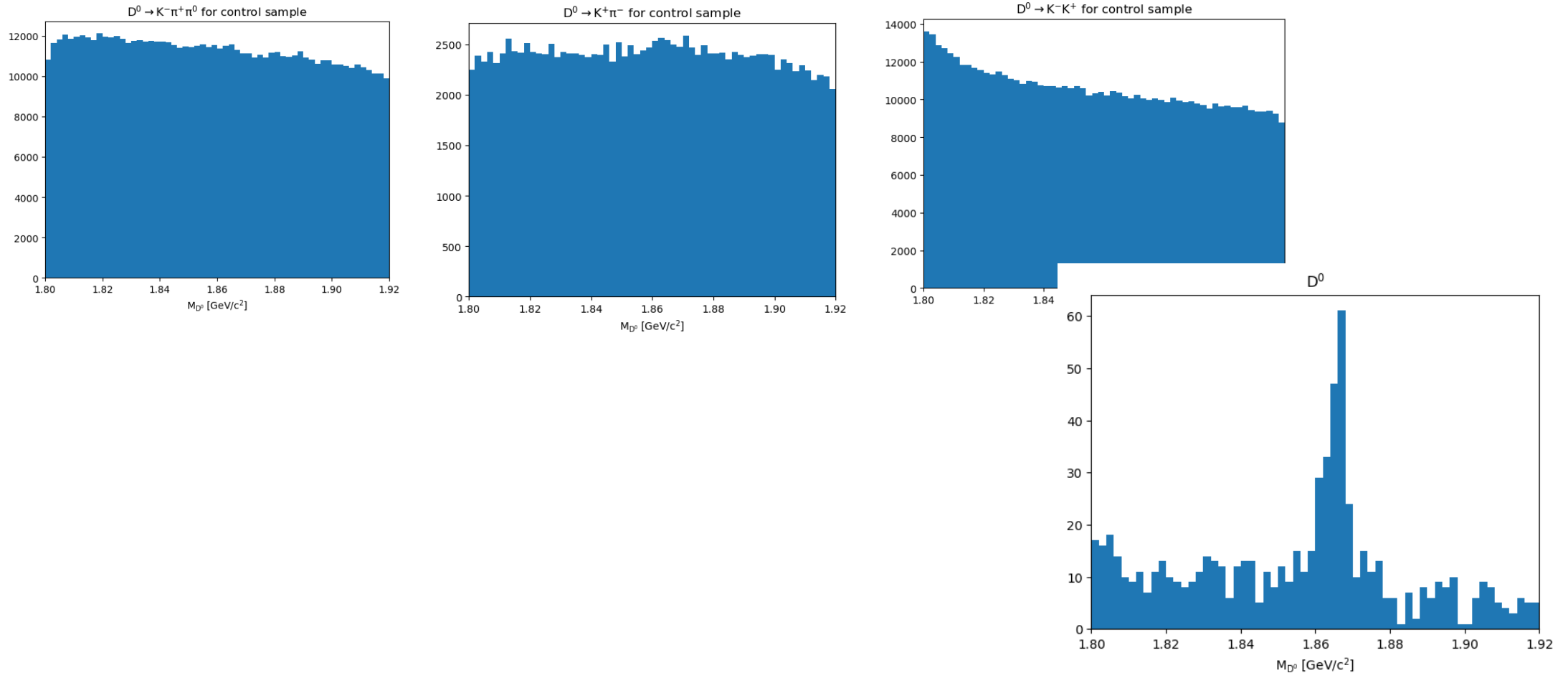
FYI) currently communicating with Kristof who is developing ccbareFI (comparison and some variables in BDT)



# Question & comment from WG report

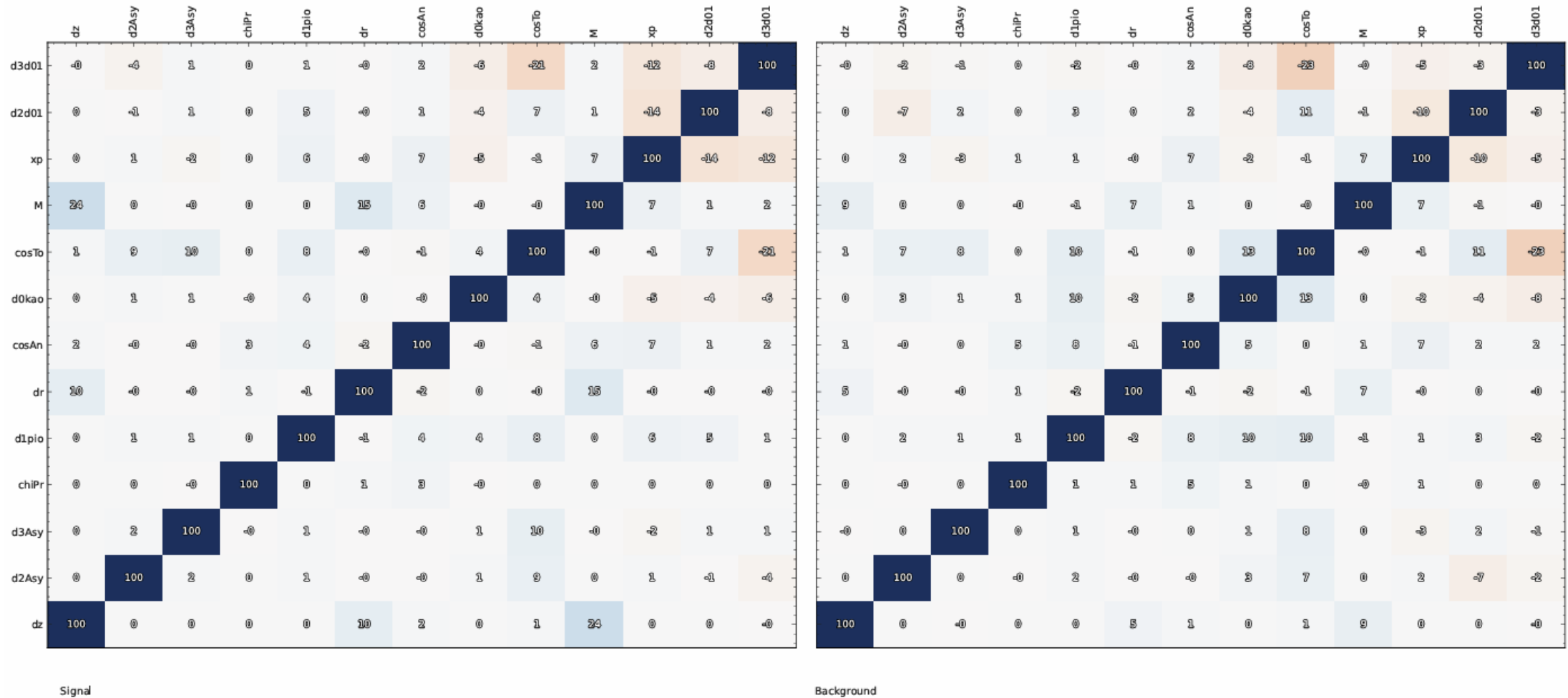
- Any plan to develop common tool that everybody can use in future?
  - Now we just started to compare the result and have discussion, but I hope we can do it
- What kind of dataset do you plan to use
  - I would like to concentrate on only belle II data set for publication
- Did you train every single tag reconstruction channels and optimization for every single channels
  - Yes, totally the 56 BDTs and optimized with FoM for every single channels
- Have you considered to measure the BR of  $D^0 \rightarrow K_L^0 K_L^0$  with this method?
  - I may possibly do it as next topic, but I would like to focus on current topic at now
- How do you get the signal efficiency
  - exclusive D : subset of inclusive D and calculated from  $D^0 \rightarrow K^- \pi^+$  MC
- From Kristof, he said he will upgrade ccbarFEI with my BDT setup and optimization
  - Currently, training your modes is quite memory intensive and I am also doing some bug fixing on the FEI
  - Charm Tagger has almost same yield as ccbarFEI, but with less backgrounds (seems it may be from optimization)

# Reconstructed $M_{K^-\pi^+}$ distribution on $D^0 \rightarrow K^-\pi^+\pi^0, K^+\pi^-, K^+K^-$





# Backup: Check on correlation between BDT variables

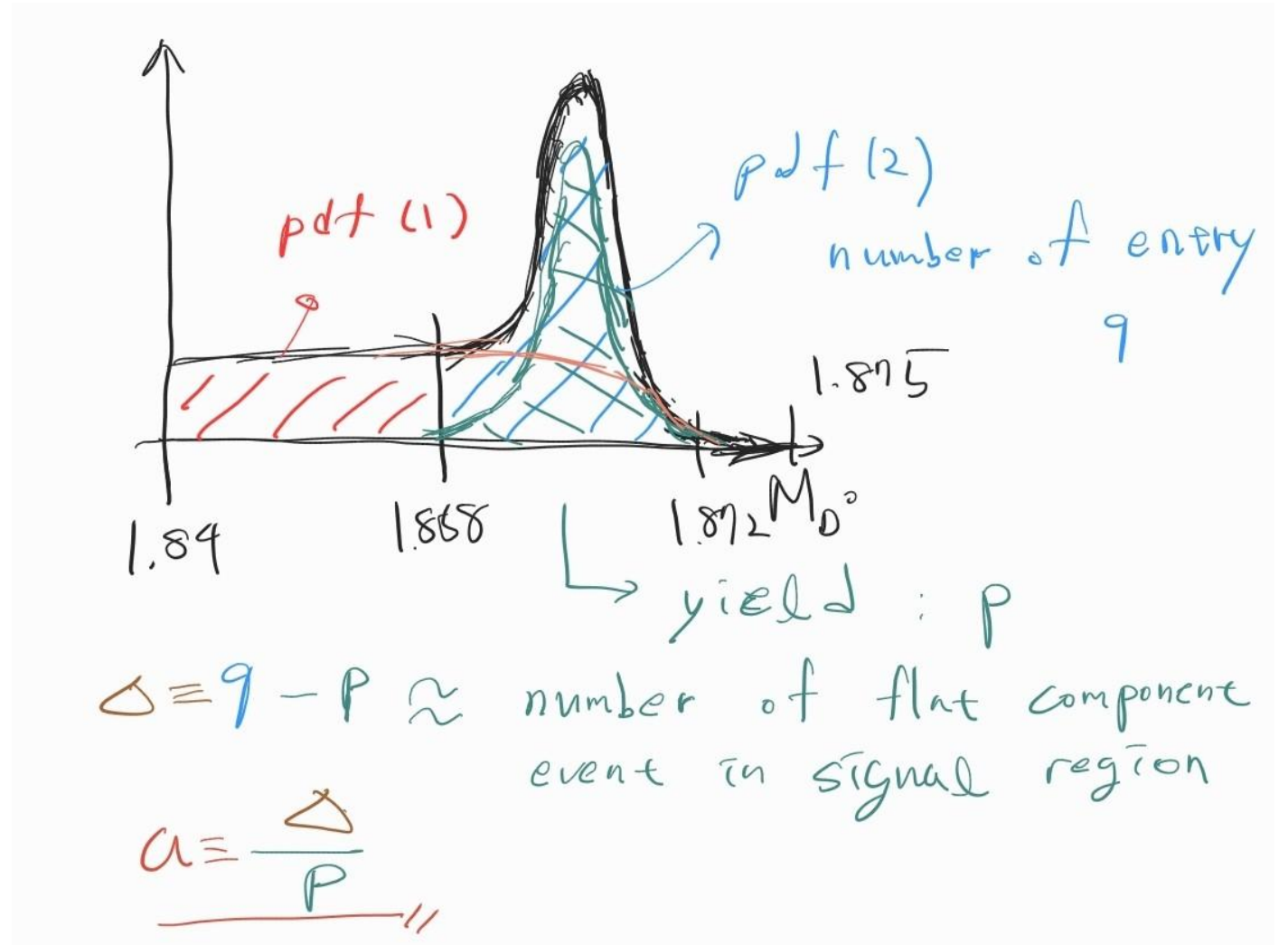


# Backup : details of fit procedure

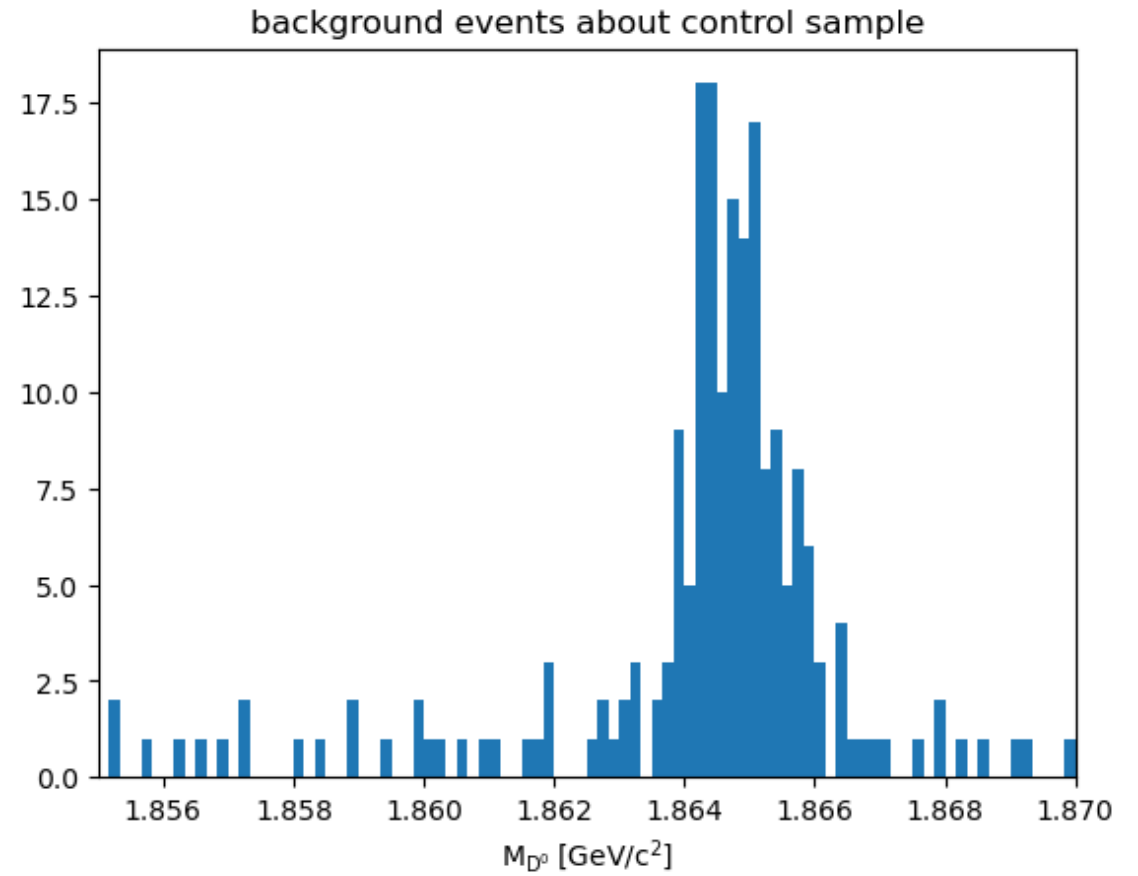
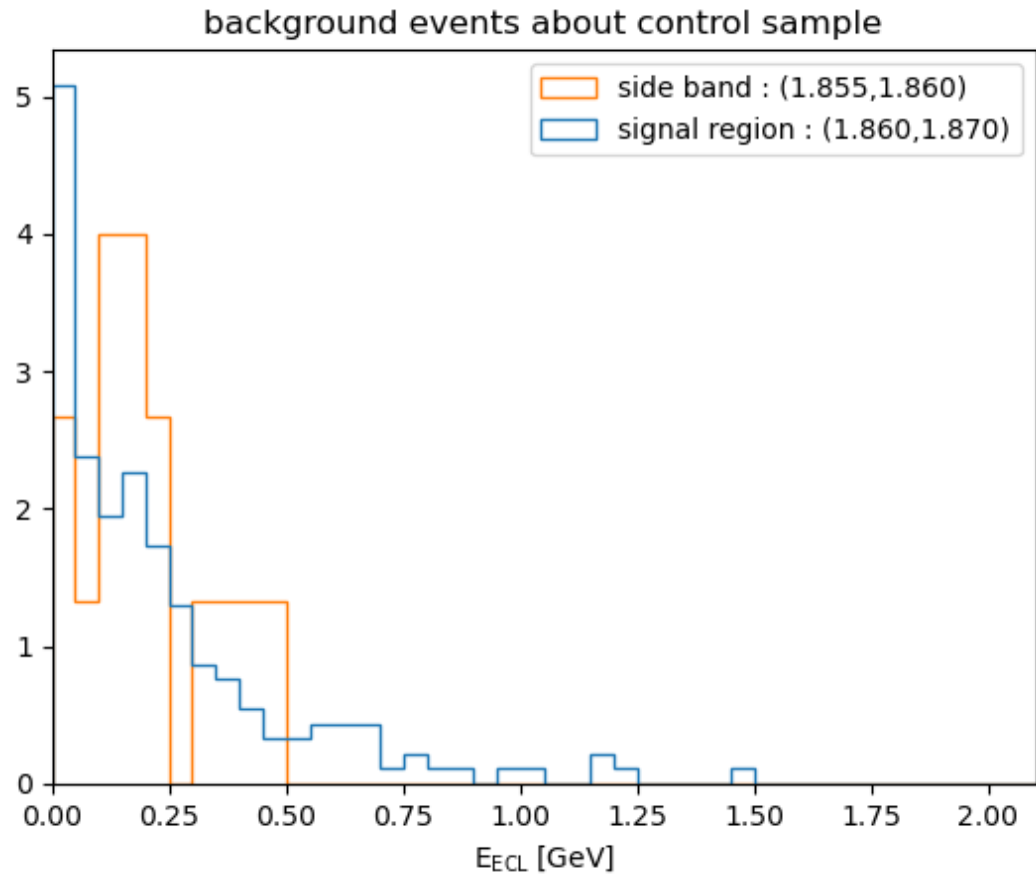
- Fit inclusive  $D^0$  on signal MC
- Fixed the signal PDF
- $E_{ECL}$  histogram PDF
  - Signal PDF from signal MC
  - Background PDF is from  $E_{ECL}$  histogram on background events
    - Flat background PDF is from  $E_{ECL}$  on  $1.855 < M_{D^0} < 1.860$  - - - (1)
    - Peak background PDF :  
histogram PDF from  $E_{ECL}$  on  $1.860 < M_{D^0} < 1.870$  - - - (2)  
=> peak background PDF :  $(2) \cdot (1 + a) - (1) \cdot a$

## Backup : variable a

- The value of a can be roughly estimated
- a is floating number with small range around the estimated value



# Backup : Check on background events ( $D^0 \rightarrow K^+ \pi^-, K^+ K^-, K^- \pi^+ \pi^0$ ) about control sample study



Backup : Check ( $M_{D^0}, E_{ECL}$ ) on background events ( $D^0 \rightarrow K^+\pi^-, K^+K^-, K^-\pi^+\pi^0$ )  
about control sample study

