### Study of D<sup>0</sup> 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  $Br(D^0 \rightarrow \nu\bar{\nu}) = 1.1 \cdot 10^{-30}$ , which is beyond the reach of current collider experiments.

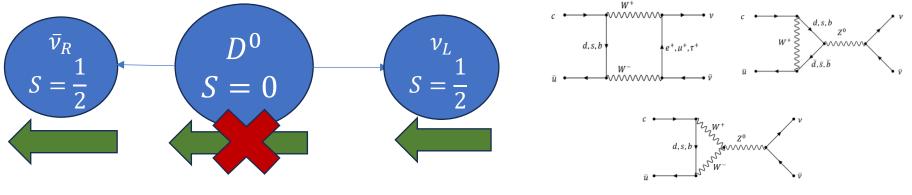


Figure1: Scheme of helicity suppressing

**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  $fb^{-1}$  data samples at 90% CL at belle [Phys. Rev. D 95, 011102(R)]

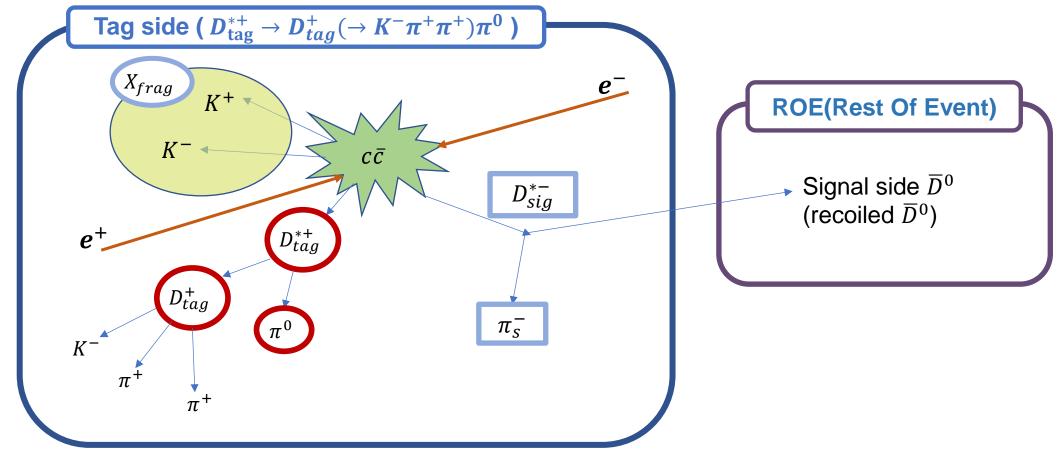
# MC simulation samples

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

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

- MC15ri generic MC( $1ab^{-1}$ ) is used as generic background MC sample
  - Will move on run dependent MC soon
- 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	$K^0_S K^+$	1.5
$K^-\pi^+\pi^+\pi^-$	8.1	$K_S^0 \pi^+$	1.5	$pK_s^0$	1.1	$K^0_S K^0_S \pi^+$	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	$K^{+}\pi^{-}\pi^{+}K^{0}_{S}$	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	$\pi^+\pi^0 \tilde{K}^0_S$	0.5
$\pi^-\pi^+\pi^0$	1.5	$K^{+}K^{0}_{S}K^{0}_{S}$	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	$p^{+}K^{0}_{S}\pi^{0}$	2.0		
$K^-K^+$	0.4			$p^{+}K^{0}_{S}\pi^{+}\pi^{-}$	1.6		
$K^-K^+\pi^0$	0.3			$\pi^+\pi^-\Sigma^+$	4.5		
$K^{-}K^{+}K^{0}_{S}$	0.4			$\pi^+\pi^-\pi^0\Sigma^+$	1.2		
$\pi^0 K^0_S$	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

$\boxed{D^{*+} or D^+}$	$D^{*0} or D^0$	$\Lambda_c^+$	$D_s^{*+} \text{ or } D_s^+$
nothing $(K^+K^-)$	$\pi^+(K^+K^-)$	$\pi^+ \bar{p}$	$K_S^0$
$\pi^{0}(K^{+}K^{-})$	$\pi^+\pi^0(K^+K^-)$	$\pi^+\pi^0\bar{p}$	$\pi^0 K_S^0$
$\pi^{+}\pi^{-}(K^{+}K^{-})$	$\pi^{+}\pi^{+}\pi^{-}(K^{+}K^{-})$	$\mid \pi^+\pi^-\pi^+\bar{p}\mid$	$\pi^+ K^-$
$ \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<sub>frag</sub>* 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  $\pi^0(\to \gamma\gamma)$ ,  $K^0_S(\to \pi^+\pi^-)$ ,  $\Lambda^0(\to p^+\pi^-)$ ,  $\Sigma^+(\to p^+\pi^0)$ ,

$$\frac{E_{d_1}-E_{d_2}}{E_{d_1}+E_{d_2}} | \text{ of } \pi^0(\to\gamma\gamma), K_S^0(\to\pi^+\pi^-), \Lambda^0(\to p^+\pi^-), \Sigma^+(\to p^+\pi^0) \text{ etc...}$$

• For  $D_{tag}^*$  training

 $\Delta M (= M_{D_{tag}^*} - M_{D_{tag}}), \text{ momentum of } \pi_s^{\pm}, \gamma, \pi^0,$ 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

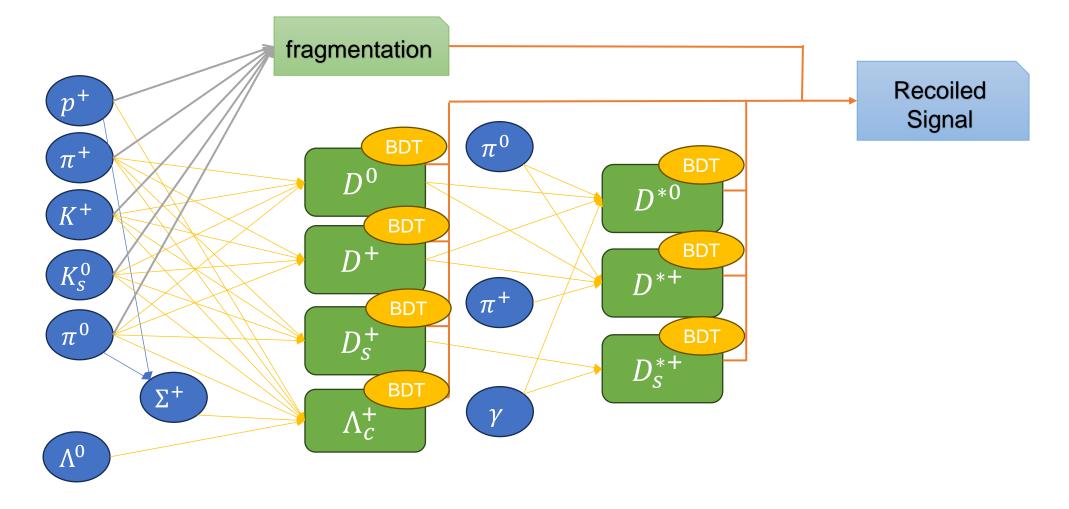
- For tracks : dr < 1.0 , |dz| < 3.0 and InCDCAcceptance
- $\pi^{\pm}$ : 15 candidates with Highest pionID after pionID > 0.01
- $K^{\pm}$ : 10 candidates with Highest kaonID after kaonID > 0.1
- $p^{\pm}$ : 10 candidates with highest protonID after protonID > 0.1
- $\gamma$  : beamBackgroundSuppression > 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

•  $K_S^0, \Lambda^0$ :

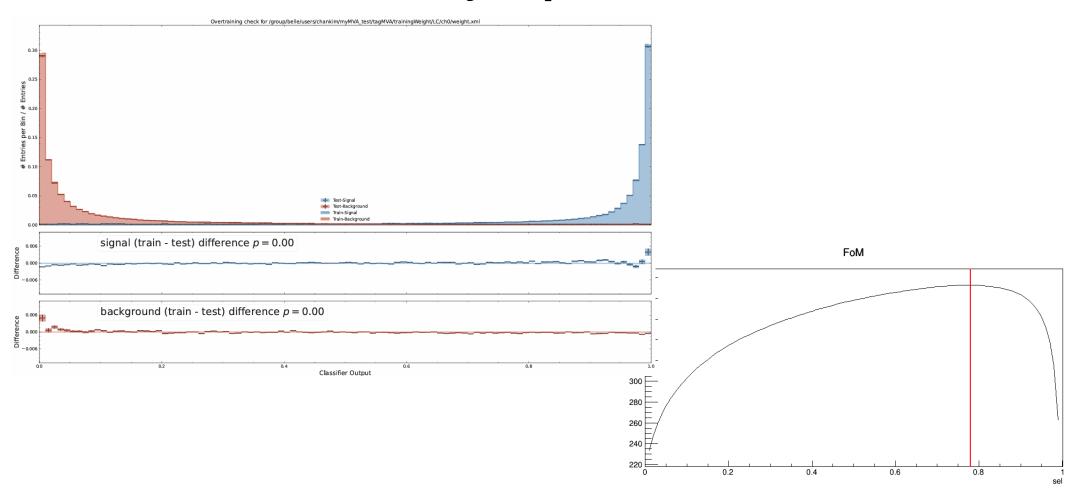
- mass and dr and  $\chi^2$  and angle between Momentum and Vertex Vector selection on  $\Lambda^0$
- goodBelleKshort for  $K_S^0$  (similar selection to  $\Lambda^0$ )

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• \Sigma^+ : reconstructed from \Sigma^+ \rightarrow p^+ \pi^0 and mass cut (1.08 < M < 1.28)
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# Flow of Charm Tagger

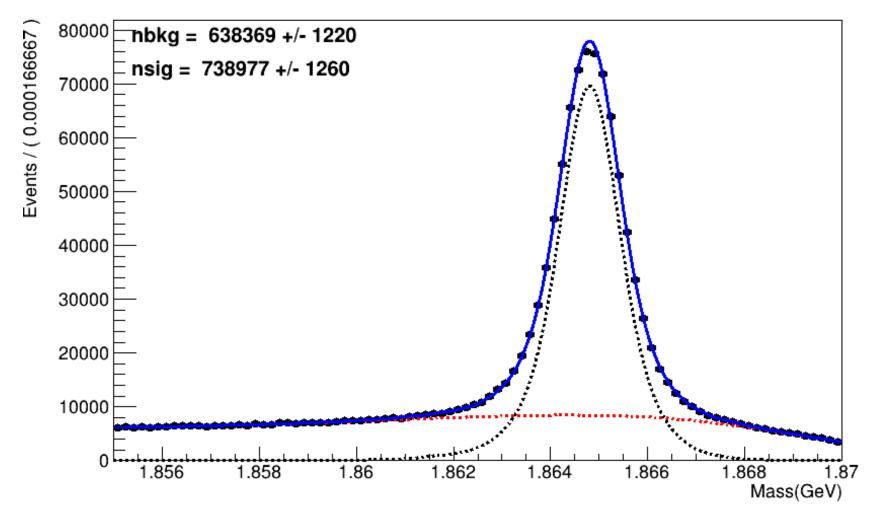


# One example about training : $\Lambda_c^+ \rightarrow p^+ K^- \pi^+$



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

Inclusvie recoil D<sup>0</sup> Mass



### Variables for extracting signal side D

- Recoil mass  $(M_{miss}(D_{tag}^*X_{frag}\pi_s^{\pm}) \text{ or } M_{recoil}(D^0))$ 
  - $e^+e^- \rightarrow D^*_{tag}X_{frag}\pi^+_sD^0$
  - $p^{\mu}(e^{+}) + p^{\mu}(e^{-}) \left(p^{\mu}(D^{*}_{tag}) + p^{\mu}(X_{frag}) + p^{\mu}(\pi^{+}_{s})\right) = p^{\mu}(D^{0}_{sig})$

• 
$$M_{recoil}(D^0) = \sqrt{p^{\mu}(D^0) * p_{\mu}(D^0)}$$

- 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  $(M_{D^0}, E_{ECL})$

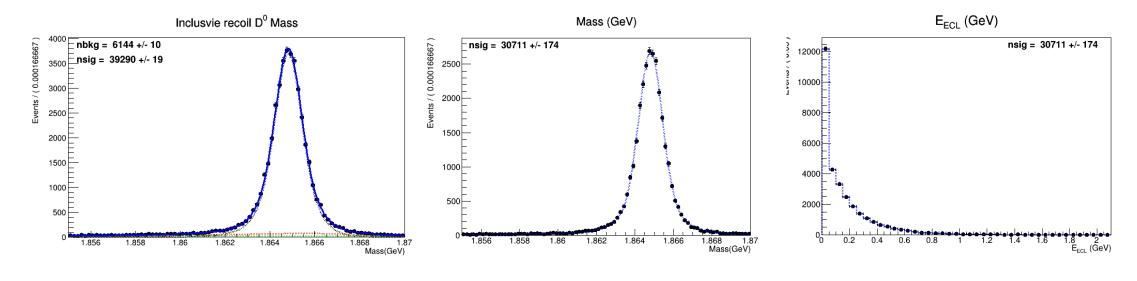
### Fitting strategy

- Inclusive D fitting
  - 1D fitting :  $M_{D^0}$
  - Signal pdf: 2 gaussians + 1 bifurcated gaussian
    - Shape fixed on inclusive D fit result of  $D^0 \rightarrow \nu \bar{\nu}$  signal MC
  - Background pdf : argus + linear
- Exclusive D fitting
  - 2D fitting :  $(M_{D^0}, E_{ECL})$
  - Signal pdf :
    - For  $M_{D^0}$ , signal pdf from exclusive D fitting
    - For  $E_{ECL}$ , 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)

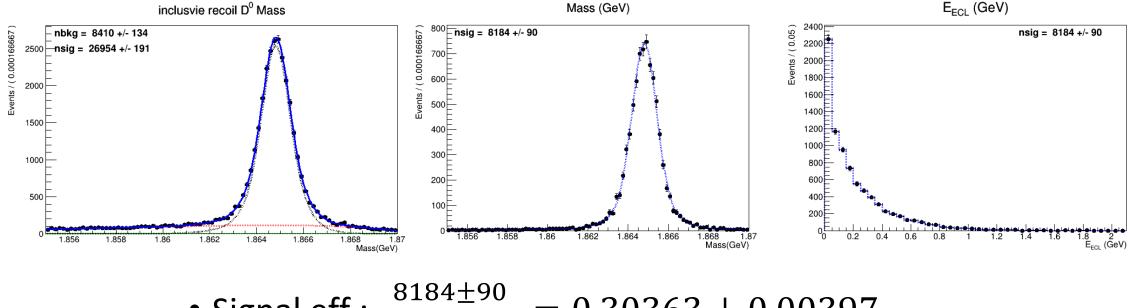
- Exclusive :  $D^0$  with selection on signal side fit on the 1.855 GeV <  $M_{D^0}$  < 1.870 GeV &  $E_{ECL}$  < 2.1 GeV
  - Selection for exclusive  $D^0$  on Signal MC  $(D^0 \rightarrow \nu \bar{\nu})$ 
    - no remaining tracks,  $\pi^0$ ,  $K_L^0$ ,  $K_S^0$ ,  $\Lambda^0$
  - Selection for exclusive  $D^0$  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



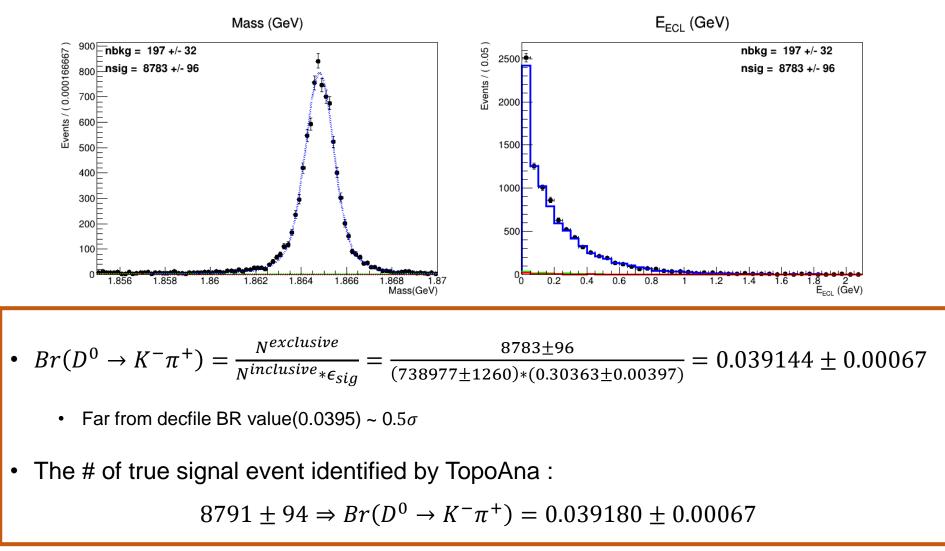
• Signal eff :  $\frac{30711 \pm 174}{39290 \pm 19} = 0.78165 \pm 0.00444$ 

# Signal efficiency on control sample

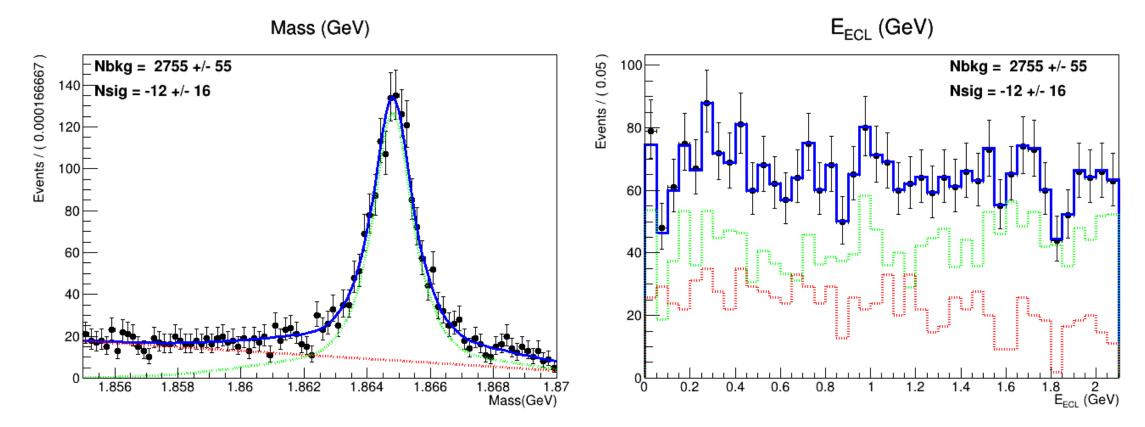


• Signal eff :  $\frac{8184 \pm 90}{26954 \pm 191} = 0.30363 \pm 0.00397$ 

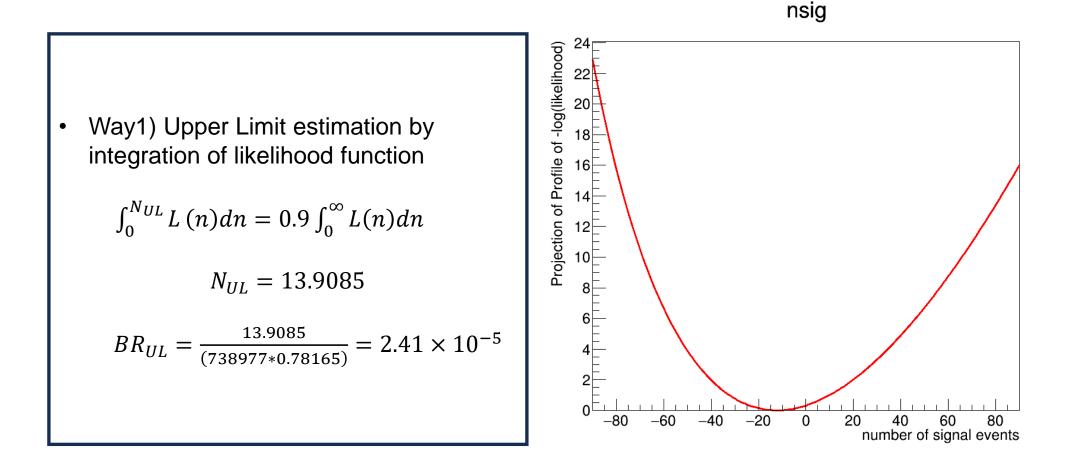
### BR measurement on generic MC(uds/mixed/charged)



# 2D fit on generic MC for $D^0 \rightarrow$ invisibles (no signals)



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

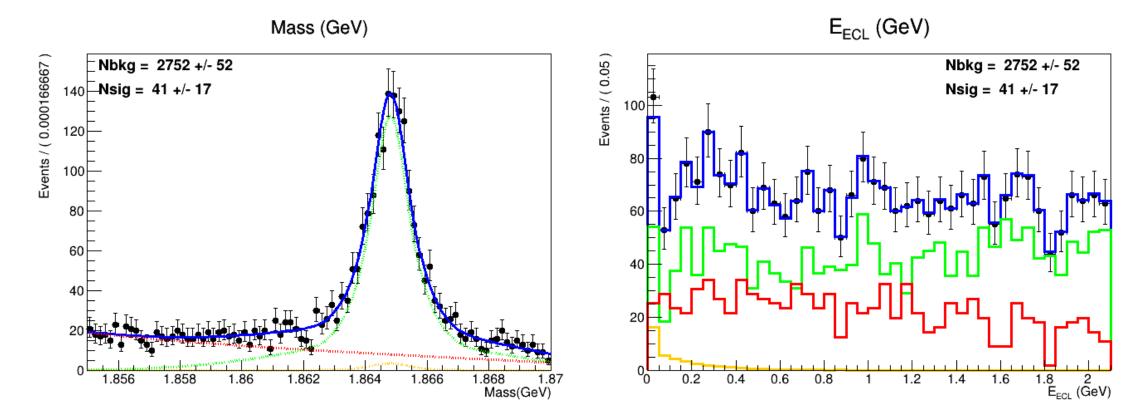


# Upper limit estimation of $D^0 \rightarrow invisibles$

HypoTest Scan Result

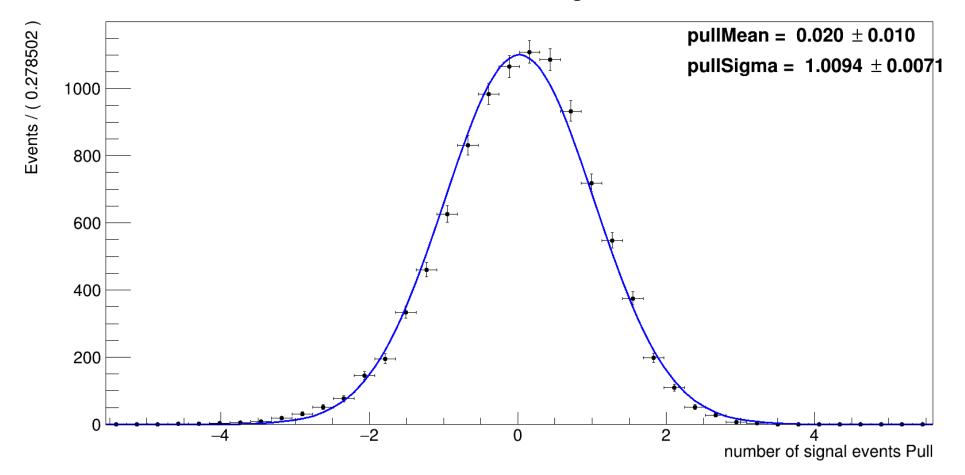
p value Observed CLs+b Observed CLb Way2) Upper limit estimation Expected CLs - Median 0.8 Expected CLs  $\pm$  1  $\sigma$ by CLs method Expected CLs  $\pm 2 \sigma$  $N_{UL} = 16.4465$ 0.6  $BR_{UL} = \frac{16.4465}{(738977*0.78165)} = 2.85 \times 10^{-5}$ 0.4 0<sup>L</sup> 20 30 10 50 60 70 90 40 80 Nsig

### 2D fit on generic MC for $D^0 \rightarrow \text{invisibles}$ (with signal embedding $\#(D^0 \rightarrow \text{invisibles}) = 50$ )

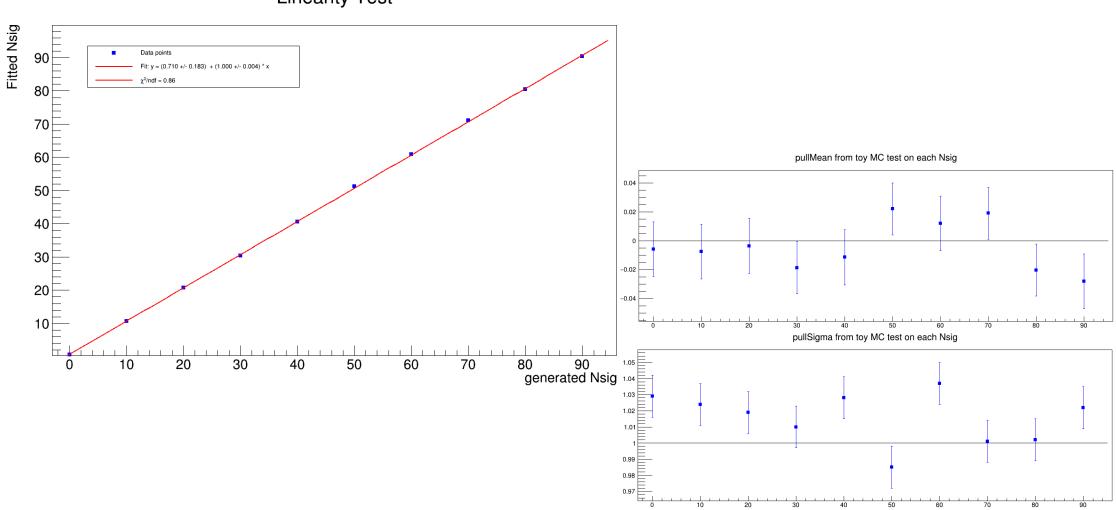


### ToyMC test of 2D fit about exclusive $D^0 \rightarrow$ invisible

A RooPlot of "number of signal events Pull"

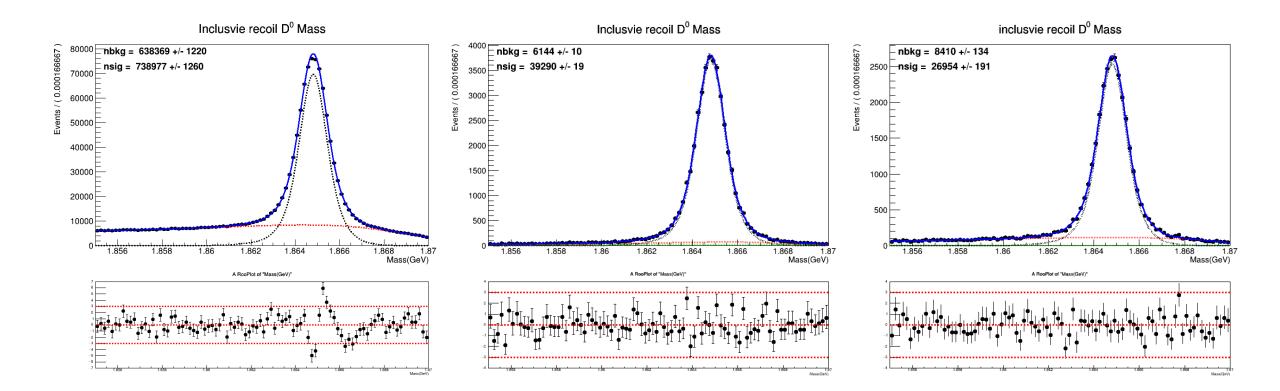


#### Linearity test of 2D fit about exclusive $D^0 \rightarrow$ invisible



Linearity Test

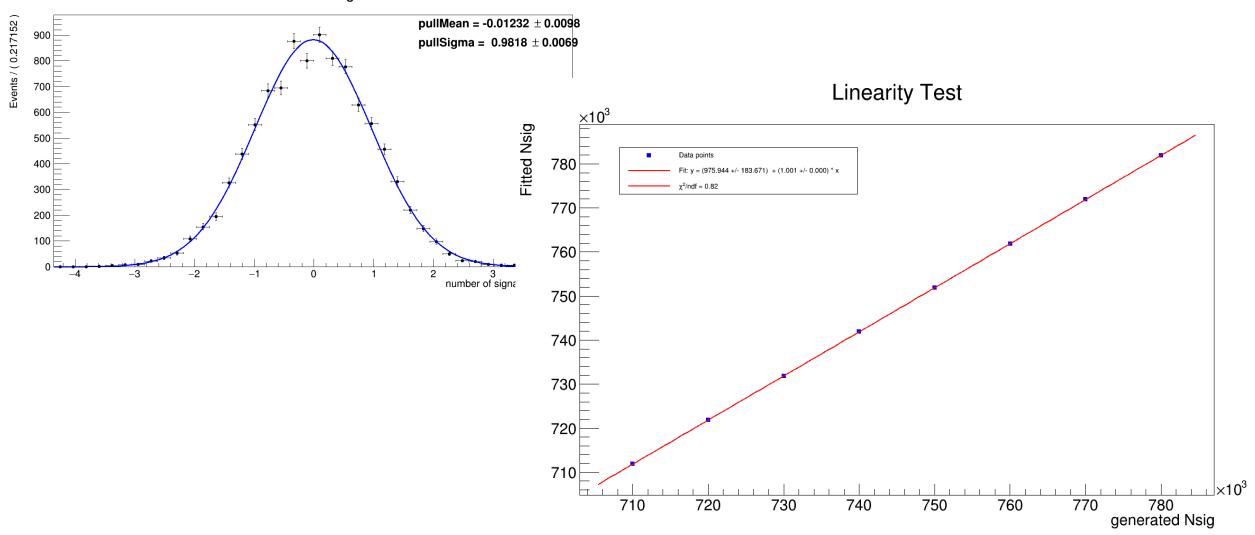
### Check fit result with pull distribution for inclusive D



Left : generic MC, middle : signal MC, right : control sample

### Inclusive D fit result check with ToyMC on generic MC

A RooPlot of "number of signal events Pull"



### Next plan

#### • Study systematic uncertainty

- Naively thinking, the 1<sup>st</sup> priority of systematics is systematic uncertainty from charm tagger
  - $\Rightarrow$  so, I would like to ask whether it is okay to see data for getting systematics

Decay	BF [%]
$D^0 \to K^- \pi^+$	3.948 %
$D^0 \rightarrow K^- \pi^+ \pi^0$	14.4 %
$D^0 \to K_S^0 \pi^+ \pi^-$	2.80 %
$D^0 \to K^- \pi^+ \pi^- \pi^+$	8.22 %
Total BF	29.368 %

Table: Decay channels to get systematic uncertainty about charm tagger

- Rough procedure:
  - Measure BF of each decays on Data(run-dependent) and MC
  - Check the Data/MC difference on measured BF of each decays
- May study semi-leptonic mode such as K I nu in future
- Systematics from fit strategy, model and gamma selection and tracking efficiency and  $K_S^0$ ,  $K_L^0$  and  $\Lambda^0$  selection efficiency in exclusive veto conditions

#### Backup : check on correlation between previous BDT variables

	d2Asy	d3Asy	d0pro	٩	chiPr	d1kao	d1pro	ц,	d1pio	d0ka.o	mpac	cosAn	d0pio	cosTo	Σ	dx	d2 d0 1	d3d01	d2 Asy	d3Asy	d0 pro	٩	chiPr	d1kao	d1pro	Ъ	d1 pio	d0ka.o	mpac	cosAn	d0pio	cosTo	Σ	dx	d2 d0 1	d3d01
d3d01	43	8	2	-26	0	7	6	-0	1	-6	-0	1	7	-201	2	-12	-@	100	-23	-1	5	-243	0	6	6	-0	-23	-8	-2	2	6	-243	-@	-6	- -	100
d2d01	1	2	æ	-7	0	5	5	쏍	5	-43	-2	1	4	T	2	-14	100	-8	-7	2	6	-0	0	5	6	-0	2	æ	-2	2	3	11	ন্ম	-20	100	-8 -
хр	- 2	-2	3	77	0	2	4	-0	6	-5	-2	7	7	-L	7	100	-114	-12	2	-8	5	69	1	Ø	7	-0	1	-8	-2	7	1	-2	8	100	-10	-8
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Signal

Background

#### Backup: Check on correlation between BDT variables

	qz	d2Asy	d3Asy	chiPr	dIpio	þ	cosAn	dOkao	cosTo	Σ	dx	d2d01	10PEp	dz	d2Asy	daAsy	chiPr	dIpio	þ	cosAn	d0kao	cosTo	Σ	dx.	d2d01	109Eb
d3d01	-0	43	1	0	1	-@	2	-6	-201	2	-12	-8	100		-2	-a		-2	-@	2	48	-28	-@	-5	-8	100
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d0kao -	0	1	1	-@	4	0	-@	100	Ø	-0	æ	æ	-3	- 0	3	1	1	20	ංනු	8	100	13	0	-8	æß	-8
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Signa

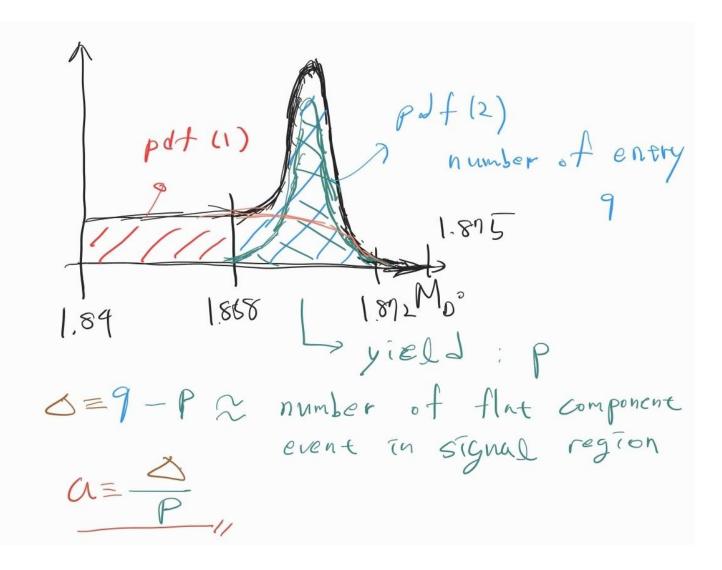
Background

### Backup : details of fit procedure

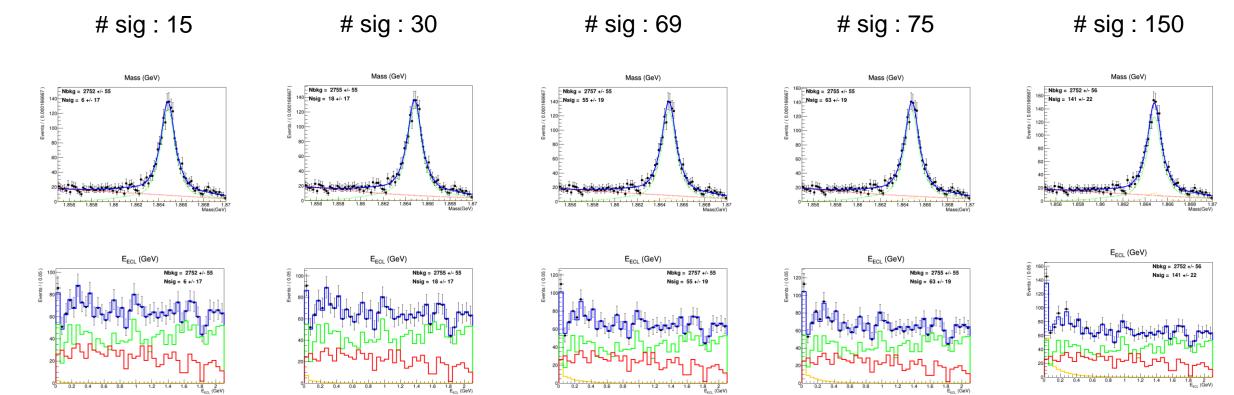
- Fit inclusive D0 on signal MC
- Fixed the signal PDF
- $E_{ECL}$  histogram PDF
  - Signal PDF from signal MC
  - Background PDF is from  $E_{ECL}$  histogram on background events
    - Flat background PDF is from  $E_{ECL}$  on  $1.855 < M_{D^0} < 1.860 - (1)$
    - Peak background PDF : histogram PDF from  $E_{ECL}$  on  $1.860 < M_{D^0} < 1.870 - - (2)$ => peak background PDF :  $(2)^*(1 + a) - (1)^*a$

#### Backup : variable a

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

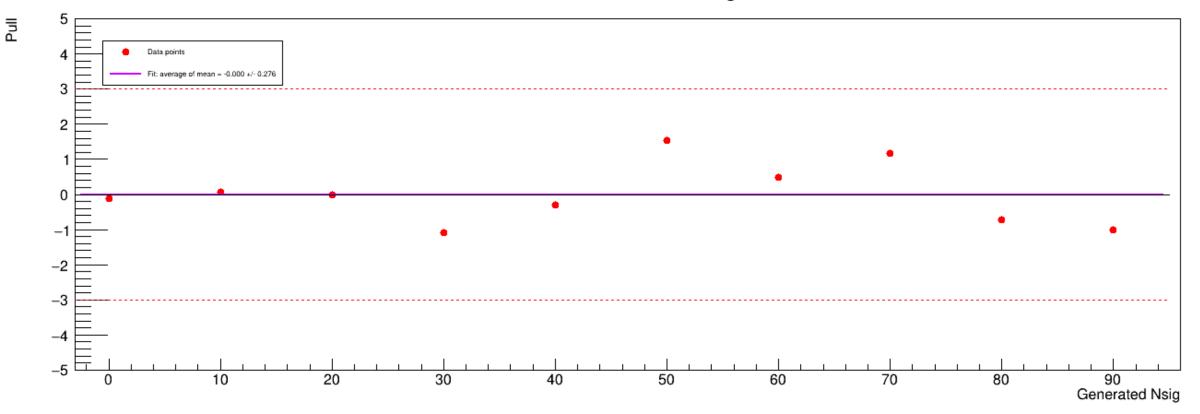


# Backup : signal embedded fit result (15,30,69,75,150)



#### Backup: checking linearity test result about exclusive D fit with $Z(=\frac{N_{fitted}-N_{expected}}{\sigma})$

Pull vs Generated Nsig



#### Backup: checking linearity test result about inclusive D fit with $Z(=\frac{N_{fitted}-N_{expected}}{\sigma})$

Pull vs Generated Nsig

