# $\Omega_c^0$ lifetime measurement

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#### Outline

- 1. Data and simulation
- 2. Prompt yield extraction
  - Strategy to extract prompt yields
  - Fit to MC samples of signal, normalization and control modes
  - Sanity checks
  - Fit to D<sup>0</sup> data
- 3. Efficiency estimation
  - Comparison of Dalitz distributions for MC and data of  $D^0$  mode
  - Comparison of kinematic distributions for MC and data of  $D^0$  mode
  - Comparison of topological distributions for MC and data of  $D^0$  mode
  - Comparison of samples with different L0 trigger
  - Comparison of event multiplicity
  - Corrections to MC by weighting
  - Decay time distribution with MC corrections

### Section 1

### Data and simulation

#### 2016 data and simulation samples available for this analysis

Modes	Data	MC	
signal	$\Omega_c^0 \rightarrow p K^- K^- \pi^+$	Turbo	ReDecay $ au = 250  \mathrm{fs}, 500  \mathrm{fs}$
normalization	$\equiv^0_c \rightarrow pK^-K^-\pi^+$	Turbo	ReDecay
control	$\Omega_b^- \to \Omega_c^0 (\to p K^- K^- \pi^+) \pi^-$	Stripping	Full Sim.
control	$\Omega_{b}^{-} \rightarrow \Omega_{c}^{0} (\rightarrow p K^{-} K^{-} \pi^{+}) \mu^{-} \overline{\nu}_{\mu}$	Stripping	Full Sim. $ au = 250  \mathrm{fs}$
control	$\Xi_b^- \to \Xi_c^0 (\to p K^- K^- \pi^+) \mu^- \overline{\nu}_\mu$	Stripping	Full Sim.
control	$D^{*+} \rightarrow D^0 (\rightarrow K^- K^+ \pi^- \pi^+) \pi^+$	Turbo	ReDecay
control	$B^+  ightarrow \overline{D}{}^0 ( ightarrow K^+ K^- \pi^+ \pi^-) \pi^+$	Stripping	Full Sim.

### Section 2

#### Prompt yield extraction

#### Strategy to extract prompt yields

- Contamination from *b*-hadron decays is evident in large decay time bins
- Use  $\log_{10}(\chi^2_{IP})$  as discriminating variable
- Model the prompt and secondary components with the Bukin function

$$\mathcal{P}(x;\mu,\sigma,\xi,\rho_{1},\rho_{2}) = \begin{cases} \exp\left\{\frac{(x-x_{1})\xi\sqrt{\xi^{2}+1\sqrt{2\ln 2}}}{\sigma\left(\sqrt{\xi^{2}+1}-\xi\right)^{2}\ln\left(\sqrt{\xi^{2}+1}+\xi\right)} + \rho_{1}\left(\frac{x-x_{1}}{\mu-x_{1}}\right)^{2} - \ln 2\right\} & x \leq x_{1}, \\ \exp\left\{-\left[\frac{\ln\left(1+2\xi\sqrt{\xi^{2}+1}\frac{x-\mu}{\sigma\sqrt{2\ln 2}}\right)}{\ln\left(1+2\xi^{2}-2\xi\sqrt{\xi^{2}+1}\right)}\right]^{2} \times \ln 2\right\} & x_{1} < x < x_{2}, \\ \exp\left\{\frac{(x-x_{2})\xi\sqrt{\xi^{2}+1}\sqrt{2\ln 2}}{\sigma\left(\sqrt{\xi^{2}+1}-\xi\right)^{2}\ln\left(\sqrt{\xi^{2}+1}+\xi\right)} + \rho_{2}\left(\frac{x-x_{2}}{\mu-x_{2}}\right)^{2} - \ln 2\right\} & x \geq x_{2}. \end{cases}$$

where

$$x_1 = \mu + \sigma \sqrt{2 \ln 2} \left( \frac{\xi}{\sqrt{\xi^2 + 1}} - 1 \right)$$

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#### Illustration of the Bukin functions

Bukin functions with various asymetry and tail parameters with  $\mu=0,\sigma=1,\rho_1=0$ 



#### Prompt and secondary MC samples

- Components after all selections



Validation needed with *b*-decay MC samples

Decay time t is defined as

$$t \equiv \frac{\vec{p} \cdot \vec{r}}{p^2} \times m$$

where  $\vec{p}$  is the momentum vector,  $\vec{r}$  the vector pointing from PV to decay vertex, and *m* the invariant mass of the charm hadron

- All fits converge with accurate error matrix
- $\mu$ : dependent on *t* with clear pattern for prompt and secondary
- $\sigma$ : vary with t for prompt and secondary
- $\xi$ : vary with t for prompt and secondary
- $\rho_1, \rho_2$ : vary with t for prompt and secondary

#### Fit results in decay-time bins: $\mu$



#### Fit results in decay-time bins: $\sigma$



#### Fit results in decay-time bins: $\xi$



### Fit results in decay-time bins: $\rho_1$



#### Fit results in decay-time bins: $\rho_2$



 $\blacksquare$  Fit with all parameters free in equally-sized 20 decay time bins between 0.3  $\rm ps$  and 3  $\rm ps$ 

## $D^0$ results in even decay-time bins: $\mu$



#### $D^0$ results in even decay-time bins: $\sigma$



#### $D^0$ results in even decay-time bins: $\xi$



#### $D^0$ results in even decay-time bins: $ho_1$



#### $D^0$ results in even decay-time bins: $\rho_2$



- Check whether the fit can re-produce the input prompt fraction with MC samples
- Fit the combined (prompt+secondary) MC with free  $\mu_{\text{prompt}}$  and  $\mu_{\text{secondary}}$ , while fixing other parameters to values of seperate fits



# Comparison of input and extracted yield: $\Omega_c^0$



### Likelihood scan of $\mu_{\mathrm{prompt}}$ and $\mu_{\mathrm{secondary}}$

•  $\Omega_c^0$  bin 0





# Comparison of input and extracted yield: $\Xi_c^0$



#### Likelihood scan of $\mu_{\mathrm{prompt}}$ and $\mu_{\mathrm{secondary}}$

■ Ξ<sup>0</sup><sub>c</sub> bin 4



# Fit results of combined MC samples: $D^0$



# Comparison of input and extracted yield: $D^0$



## Sanity check: fit background-injected combined MC

- Check whether the fit can re-produce the input prompt fraction
- Fit the combination of prompt MC, secondary MC and background from data
  - Background sample is generated from the RooKeysPdf of mass-sideband data
  - Background sample size is determined from the signal/background ratio in data
  - logIPCHI2 of MC sample is shifted by 0.15 to match the data
- Fit configuration
  - $\mu_{\mathrm{prompt}}$  and  $\mu_{\mathrm{secondary}}$  free to float
  - Other parameters fixed to separately-fit results
  - Number of background fixed to the input value

#### Fit results of background-injected combined MC



# Comparison of input and extracted yield: $D^0$



#### Mass fit

- In  $1865 \pm 45 \,\mathrm{MeV}/c^2$  mass region
- Gaussian  $+ 2^{nd}$ -order Chebychev
- IPCHI2 fit
  - In  $1865 \pm 2.5 imes 5.65 \, \mathrm{MeV}/c^2$  signal region
  - Prompt and secondary signal components:  $\mu_{prompt}$  and  $\mu_{secondary}$  free and other parameters fixed to MC
  - Background: kernel estimation with mass-sideband data Lower sideband: [1820, 1830]  $MeV/c^2$ Upper sideband: [1900, 1910]  $MeV/c^2$
  - The total number of backgrounds fixed to values from mass fit
  - Binning scheme: the same as the signal mode

#### Comparison of background logIPCHI2

- MC cannot provide a large combinatorial background sample
- Real data is necessary for background studies



### Physics backgrounds in $D^{*+} \rightarrow \pi^+ D^0 (\rightarrow K^- K^+ \pi^- \pi^+)$

- $\Lambda_c^+$  background: p mis-identified as  $K^-$
- Suppressed by tight PID cuts


### Physics backgrounds in $D^{*+} \rightarrow \pi^+ D^0 (\rightarrow K^- K^+ \pi^- \pi^+)$

•  $\pi$  swap:  $D^{*+} \rightarrow \pi^+ D^0 (\rightarrow K^- K^+ \pi^- \pi^+)$ 

• Not significant due to tight  $\chi^2_{IP}$  cuts



#### Physics backgrounds in $D^{*+} \rightarrow \pi^+ D^0 (\rightarrow K^- K^+ \pi^- \pi^+)$

- $K^0_{s}$  background:  $D^0 \rightarrow K^- K^+ K^0_{s} (\rightarrow \pi^+ \pi^-)$
- Exist but not significant in the signal region



## Fit results of $D^0$ data



#### Fit results of yields





#### Comparison of data and MC: prompt fraction



# Comparison of data and MC: $\mu_{\mathrm{prompt}} - \mu_{\mathrm{secondary}}$



#### Comparison of data and MC: prompt yield



#### Likelihood scan of $\mu_{\mathrm{prompt}}$ and $\mu_{\mathrm{secondary}}$



#### Effect of initial values

Repeat the fit (bin 0) with random initial values in parameter space (µ<sub>prompt</sub>, µ<sub>secondary</sub>, N<sub>prompt</sub>, N<sub>secondary</sub>)



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#### Lifetime measurement

- Study the effect of unequal binning scheme
- Fit data in equally-sized 20 decay time bins between 0.3 ps and 3 ps







#### Comparison of prompt yield in even decay-time bins



#### Section 3

## Efficiency estimation

# Comparison of Dalitz distributions for MC and data of $D^0$ mode

- Five variables are needed to describe the  $D^0 o K^+ K^- \pi^+ \pi^-$  decay
- Choose Cabibbo-Maksymowicz (CM) variables
  - m(K<sup>+</sup>K<sup>-</sup>)
  - $m(\pi^+\pi^-)$
  - cos(θ<sup>K+K-</sup><sub>K+</sub>): the cosine of the angle between the direction of the D<sup>0</sup> and that of one of the kaons in the rest frame of the two kaons
  - cos(θ<sup>π+π-</sup><sub>π+</sub>): the cosine of the angle between the direction of the D<sup>0</sup> and that of one of the pions in the rest frame of the two pions
  - cos(φ): the cosine of the angle in the D<sup>0</sup> rest frame between the plane defined by the directions of the two kaons and the plane defined by the directions of the two pions
- Data sWeights are calcualted from mass fit
- Decay time binning scheme: equally-sized 20 decay time bins between 0.3 ps and 3 ps

#### Comparison of $m(K^+K^-)$ in Bin 0

- No  $\phi$  resonance in phase-space MC
- Prompt and secondary MC agree well



## Comparison of $m(\pi^+\pi^-)$ in Bin 0

- No  $\rho$  resonance in phase-space MC
- Prompt and secondary MC agree well



# Comparison of $\cos(\theta_{K^+}^{K^+K^-})$ in Bin 0

- Data-MC discrepency is small
- Data is not symmetric w.r.t. 0
- Prompt and secondary MC agree well



## Comparison of $\cos(\theta_{\pi^+}^{\pi^+\pi^-})$ in Bin 0

- Data is not symmetric w.r.t. 0
- Prompt and secondary MC agree well



#### Comparison of $cos(\phi)$ in Bin 0

- Data is not symmetric w.r.t. 0
- Prompt and secondary MC agree well



Only the decay-time-dependent discrepency is relevant in this measurement

## Decay time dependence of $m(K^+K^-)$

#### Decay-time dependent



## Decay time dependence of $m(\pi^+\pi^-)$

#### Weakly decay-time dependent



# Decay time dependence of $\cos(\theta_{K^+}^{K^+K^-})$

#### Decay-time independent



# Decay time dependence of $\cos( heta_{\pi^+}^{\pi^+\pi^-})$

#### Decay-time independent



#### Decay time dependence of $\cos(\phi)$

#### Decay-time independent



# Comparison of $p_{\rm T}$ ( $D^0$ )



# Comparison of $p_{\mathrm{T}}$ $(\pi^+)$



# Comparison of $p_{\rm T}$ (K<sup>-</sup>)







#### Comparison of samples with different L0 trigger

- LOHadronDecision\_TOS v.s. LOGlobal\_TIS && !LOHadronDecision\_TOS
- $D^0 p_{\rm T}$  is quite different



## Comparison of samples with different L0 trigger (cont.)

- LOHadronDecision\_TOS v.s. LOGlobal\_TIS && !LOHadronDecision\_TOS
- Daughters'  $p_{\rm T}$  is quite different



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## Comparison of nTracks


- Weights are calculated sequentially by comparing the decay-time-integrated MC prompt sample and sWeighted data
  - PID weights from PIDCalib
  - Based on PID weights, calculate  $(m(K^+K^-), m(\pi^+\pi^-))$  weights
  - Based on PID and  $(m(K^+K^-), m(\pi^+\pi^-))$  weights, calculate  $p_{\rm T}(D^0)$  weights
- cos(θ<sup>K+K<sup>-</sup></sup><sub>K<sup>+</sup></sub>), cos(θ<sup>π+π<sup>-</sup></sup><sub>π<sup>+</sup></sub>), cos(φ) not weighted due to weak dependence on decay time
- Here we assume the relevant distributions of the prompt and secondary components in data are the same

#### Comparison before and after corrections to MC

#### • Consider PID and $(m(K^+K^-), m(\pi^+\pi^-))$ weights



#### Comparison before and after corrections to MC (cont.)

• Consider PID and  $(m(K^+K^-), m(\pi^+\pi^-))$  weights



#### Comparison before and after corrections to MC

• Consider PID,  $(m(K^+K^-), m(\pi^+\pi^-))$  and  $p_T(D^0)$  weights



#### Comparison before and after corrections to MC (cont.)

• Consider PID,  $(m(K^+K^-), m(\pi^+\pi^-))$  and  $p_T(D^0)$  weights



#### PID corrections with PIDCalib

- PID cuts applied: Kaon  $DLL_{K\pi} > 10$  and Pion  $DLL_{K\pi} < 0$
- Use PIDCalib to get the correct PID efficiency
- Use default binning scheme
  - *P*[GeV/c]:
    - $[ \ 3.0, \ 9.3, \ 15.6, \ 19.0, \ 24.4, \ 29.8, \ 35.2, \ 40.6, \ 46.0, \ 51.4 \ ]$
    - [ 51.4, 56.8, 62.2, 67.6, 73.0, 78.4, 83.8, 89.2, 94.6, 100.0 ]
  - η: [ 1.5, 2.375, 3.25, 4.125, 5.0 ]

#### Performance histograms: (left) Magdown and (right) MagUp



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Lifetime measurement

#### Reweighting with Boosted Decision Trees

- Try to reweight MC with GBReweighter from the hep\_ml package, a reweighter algorithm based on ensemble of regression trees
  - PID weights from PIDCalib
  - Based on PID weights, calculate

 $w(m(K^+K^-), m(\pi^+\pi^-), \cos(\theta_{K^+}^{K^+K^-}), \cos(\theta_{\pi^+}^{\pi^+\pi^-}), \cos(\phi))$ 

- Build unbiased predictions with folding algorithm
  - training data is splitted into n equal parts
  - train n reweighters, each one is trained using n-1 folds
  - predict each event with the reweighter that did not use it during training
- Here we assume the relevant distributions of the prompt and secondary components in data are the same, i.e. subtract the background in real data by fitting the mass spectrum

#### Comparison before and after corrections to MC

#### Reweighting with GBReweighter



# Comparison before and after corrections to MC (cont.)

#### Reweighting with GBReweighter



#### Decay time distribution with MC corrections



# BACKUP









































# Fit mass of $D^0$ data





