

BESIII Charm Meeting

Phase-space binned analysis of strong-phase parameters in $D^0 \rightarrow K^+ K^- \pi^+ \pi^-$

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Outline

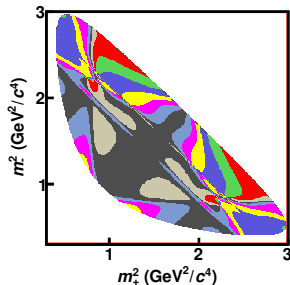
- 1 Introduction
- 2 Binning scheme
- 3 Formalism and measurement strategy
- 4 Selection
- 5 Fit of double tag yields
- 6 Likelihood fit
- 7 Input-output checks
- 8 Fit results
- 9 Bonus measurements
- 10 Summary and conclusion

Introduction

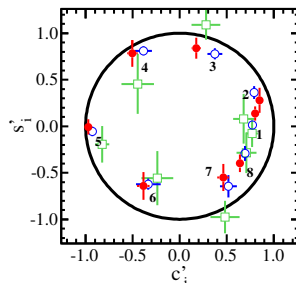
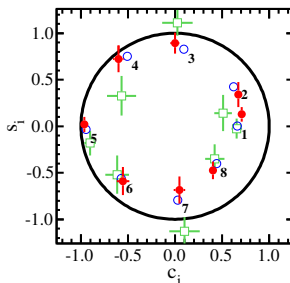
- Aim of analysis: Measure the strong-phase difference between D^0 and $\bar{D}^0 \rightarrow K^+ K^- \pi^+ \pi^-$ using the $8 \text{ fb}^{-1} \psi(3770)$ dataset
- Important input to:
 - ① Measurement of the CKM angle γ
 - ② Studies of charm mixing and CPV
 - ③ Strong-phase measurements of other decay modes
- Phase-space integrated measurement: [Phys. Rev. D **107** 032009](#)
- Strong phases are diluted when integrated over the 5D phase space
 \implies Perform analysis in bins
- Binned model-dependent analysis of γ has already been performed at LHCb: [arXiv:2301.10328](#)

Introduction - Previous strong-phase analyses

- Golden mode for strong-phase analysis: $D^0 \rightarrow K_{S,L}^0 \pi^+ \pi^-$
- Measurements of c_i and s_i performed in 2×8 bins:
 - CLEO: [Phys. Rev. D **82** 112006](#)
 - BESIII: [Phys. Rev. D **101** 112002](#)



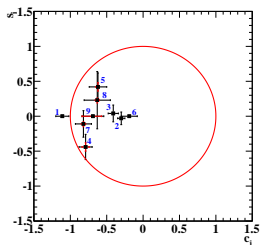
(a) Equal δ binning scheme



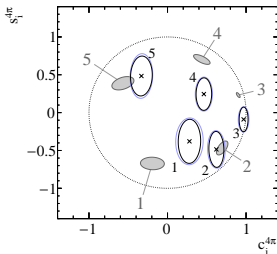
(b) c_i and s_i for (left) $K_S \pi \pi$ and (right) $K_L \pi \pi$

Introduction - Previous strong-phase analyses

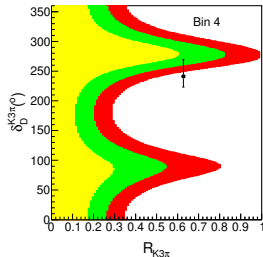
- Four-body decays have a 5D phase space, but their binning scheme may be defined analogously
- Many different strategies for 5D binning schemes:
 - $D \rightarrow K_S^0 \pi^+ \pi^- \pi^0$: [JHEP 2018 82 \(2018\)](#)
 - $D \rightarrow \pi^+ \pi^- \pi^+ \pi^-$: [JHEP 2018 144 \(2018\)](#)
 - $D \rightarrow K^- \pi^+ \pi^- \pi^+$: [JHEP 2021 164 \(2021\)](#)



(a) $K_S \pi \pi \pi^0$



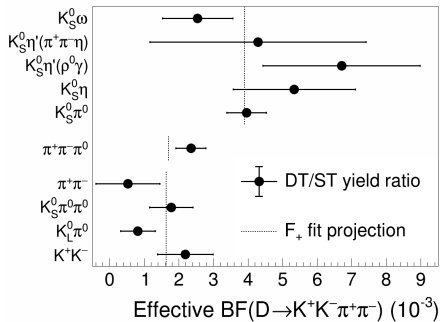
(b) 4π



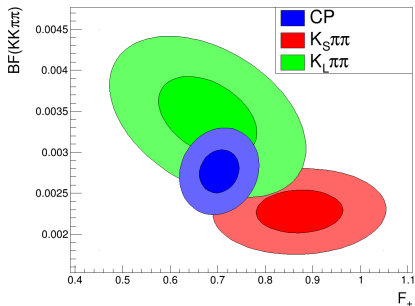
(c) $K \pi \pi \pi$

Introduction - Previous strong-phase analyses

- The phase-space integrated strong phase of $D \rightarrow K^+ K^- \pi^+ \pi^-$ was previously measured: [Phys. Rev. D **107** 032009](#)
- The asymmetry in the branching fraction measured using CP even and CP odd tags is sensitive to the CP-even fraction F_+
 - $c_i = 2F_+ - 1$ is the amplitude-average cosine of the strong phase



(a) BF asymmetry



(b) F_+ combination

Strong phases are determined from yield asymmetries

- When integrating over a phase-space region, variations in strong phases dilute the asymmetries
- We require a binning scheme to minimise the dilution and enhance asymmetry effects

How to bin a 5-dimensional phase space?

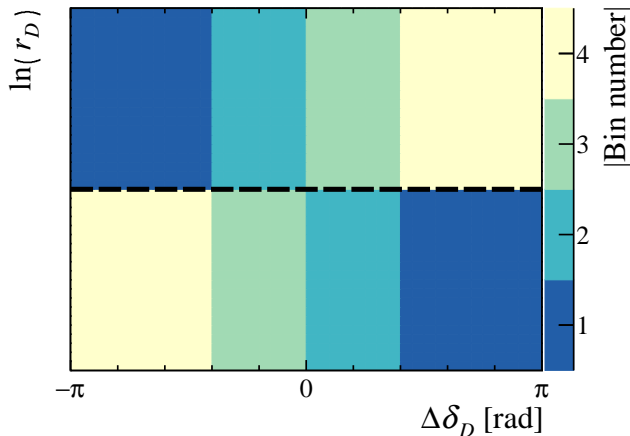
- Generate C++ code for LHCb amplitude model using AmpGen¹
- For each D event, calculate

$$\frac{\mathcal{A}(D^0)}{\mathcal{A}(\bar{D}^0)} = r_D e^{i\delta_D}$$

- Bin along δ_D and r_D
 - Bin boundaries in δ_D are moved to maximise sensitivity to γ

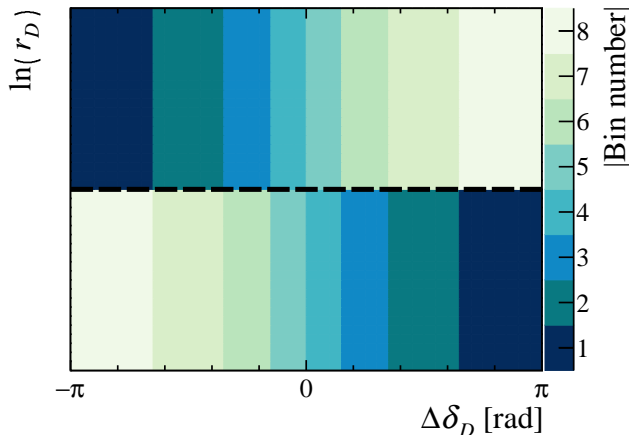
¹AmpGen by Tim Evans

Binning scheme



Bins $i < 0$ on top, $i > 0$ below
Binning scheme used in this analysis

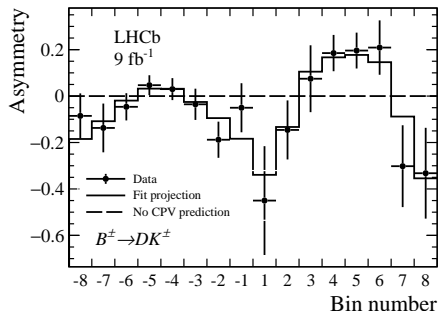
Binning scheme



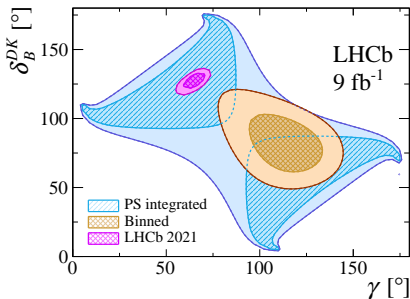
Bins $i < 0$ on top, $i > 0$ below

Possible binning scheme for future analysis with 20 fb^{-1} dataset

Model-dependent LHCb measurement



(a) Bin asymmetries in $B^\pm \rightarrow DK^\pm$



(b) Measurement of γ at LHCb

$$\gamma = (116^{+12}_{-14})^\circ$$

This result is currently model independent, and needs external inputs from BESIII to become a proper model-independent measurement!

$D\bar{D}$ pair from $\psi(3770)$ is prepared in a $\mathcal{C} = -1$ state

- D mesons are “quantum correlated”
- The decay of $D \rightarrow K^+ K^- \pi^+ \pi^-$ is correlated by the CP content of the tag mode

$$|D\bar{D}\rangle = \frac{1}{\sqrt{2}}(|D^0\rangle|\bar{D}^0\rangle - |\bar{D}^0\rangle|D^0\rangle)$$

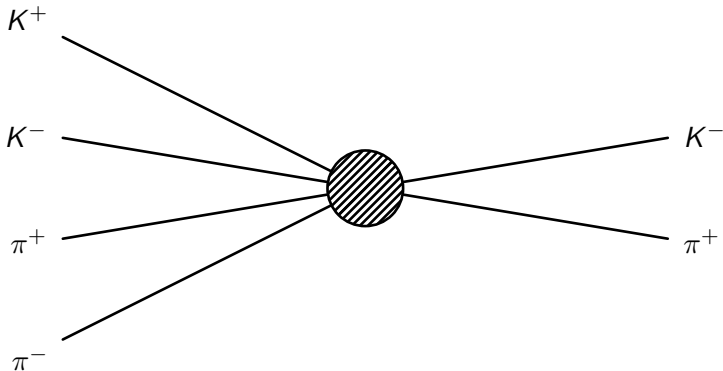
- Equivalently, in terms of CP eigenstates:

$$|D\bar{D}\rangle = \frac{1}{\sqrt{2}}(|D_-\rangle|D_+\rangle - |D_+\rangle|D_-\rangle)$$

Formalism and measurement strategy

- Tag mode can be a flavour tag

- $K\pi$, $K\pi\pi^0$, $K\pi\pi\pi$, $Ke\nu$

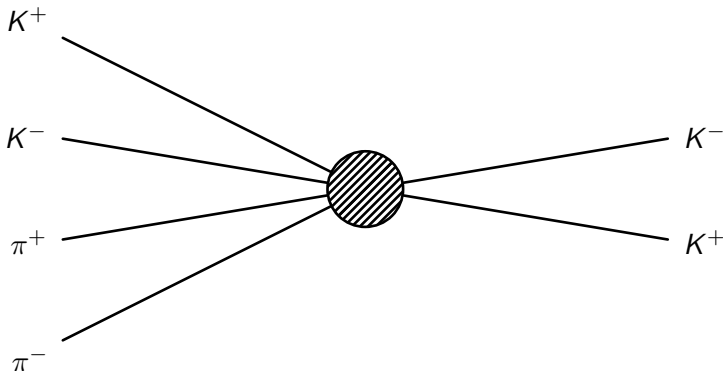


Use flavour tags to measure fraction of $D^0 \rightarrow KK\pi\pi$ decays in bin i :

$$\frac{N_i^{\text{DT}}}{N^{\text{ST}}} = \mathcal{B} \times K_i$$

Formalism and measurement strategy

- Tag mode can be a CP even tag
 - KK (fully and part. reco $KK\pi\pi$), $\pi\pi$, $\pi\pi\pi^0$, $K_S\pi^0\pi^0$, $K_L\pi^0$

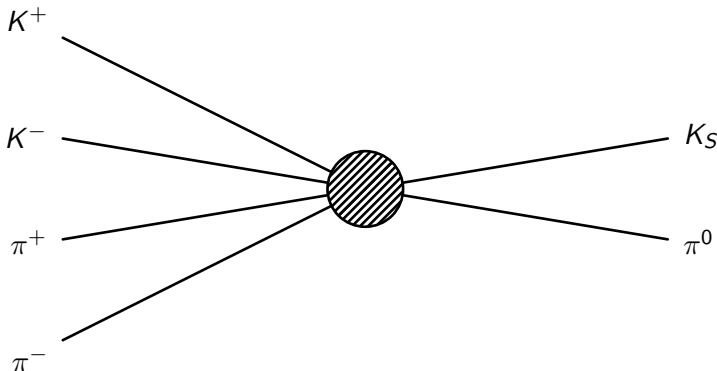


$D \rightarrow K^+K^-$, which is CP even, forces $D \rightarrow K^+K^-\pi^+\pi^-$ to be CP odd:

$$\frac{N_i^{\text{DT}}}{N^{\text{ST}}} = \mathcal{B} \times (K_i + K_{-i} - 2\sqrt{K_i K_{-i}} c_i)$$

Formalism and measurement strategy

- Tag mode can be a CP odd tag
 - $K_S\pi^0$ (fully and part.reco $KK\pi\pi$), $K_S\eta$, $K_S\eta'(\pi\pi\eta, \pi\pi\gamma)$, $K_S\pi\pi\pi^0$

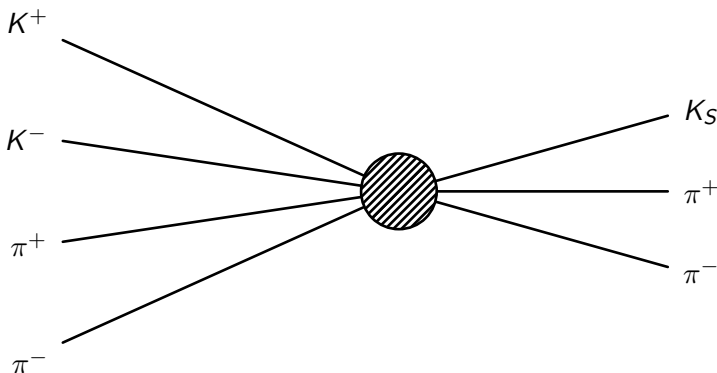


$D \rightarrow K_S^0\pi^0$, which is CP odd, forces $D \rightarrow K^+K^-\pi^+\pi^-$ to be CP even:

$$\frac{N_i^{\text{DT}}}{N^{\text{ST}}} = \mathcal{B} \times (K_i + K_{-i} + 2\sqrt{K_i K_{-i}} c_i)$$

Formalism and measurement strategy

- Tag mode can be a self-conjugate multi-body tag
 - $K_S\pi\pi$ (fully and part.reco $KK\pi\pi$), $K_L\pi\pi$

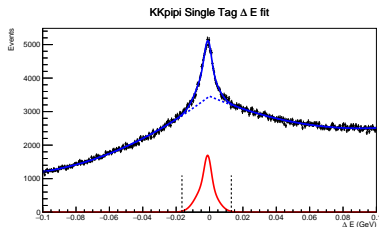


$D \rightarrow K_S^0 \pi^+ \pi^-$ has different strong phases in different bins of phase space:

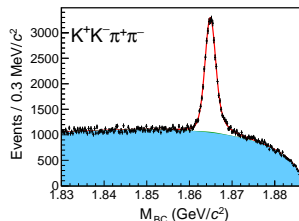
$$\frac{N_{ij}^{\text{DT}}}{N^{\text{ST}}} = \mathcal{B} \times (K_i K'_{-j} + K_{-i} K'_j - 2\sqrt{K_i K_{-i} K'_j K'_{-j}}(c_i c'_j + s_i s'_j))$$

Selection

- Double-tag analysis: Select $D \rightarrow KK\pi\pi$ events tagged with flavour, CP and multi-body tags
- Selection more or less identical to previous double-tag analyses
- $D \rightarrow KK\pi\pi$ selection:
 - 4 good charged tracks
 - 3σ window around signal peak in ΔE
 - Asymmetric $K_S KK$ veto for $m(\pi^+\pi^-) \in [477, 507] \text{ MeV}$
 - Flight significance cut at 2



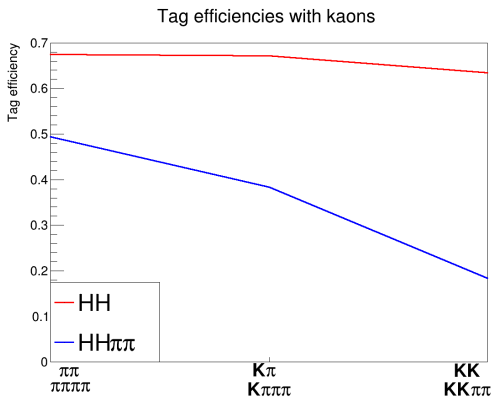
(a) ΔE window



(b) Single tag yield: $29\,227 \pm 268$

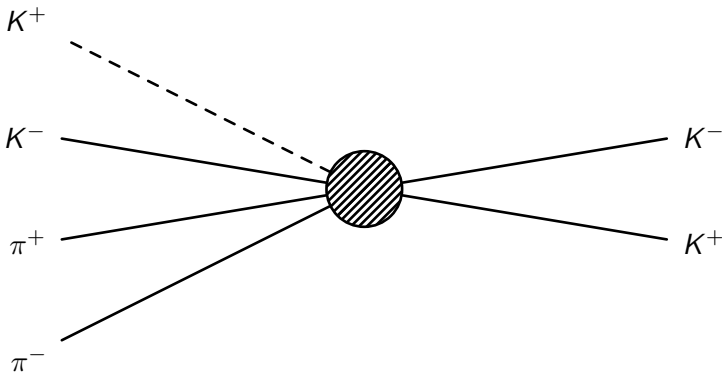
Partially reconstructed $D^0 \rightarrow K^+ K^- \pi^+ \pi^-$

- The reconstruction efficiency of $D \rightarrow KK\pi\pi$ is less than 20%
- For comparison, the efficiency of $D \rightarrow \pi\pi\pi\pi$ is around 50%!
- Poor kaon tracking efficiency at low momentum



Partially reconstructed $D^0 \rightarrow K^+ K^- \pi^+ \pi^-$

- Solution: Only reconstruct 3 of the charged D daughters
 - Presence of missing kaon is inferred from the missing momentum
 - Yields are similar to fully reconstructed sample
 - Large, but non-peaking background from $D \rightarrow K \pi \pi \pi \pi^0$

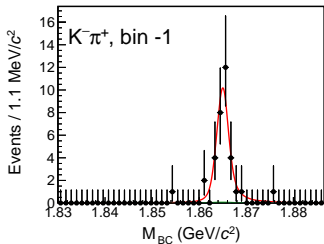


Use this technique with the $K^+ K^-$, $K_S^0 \pi^0$ and $K_S^0 \pi^+ \pi^-$ tags

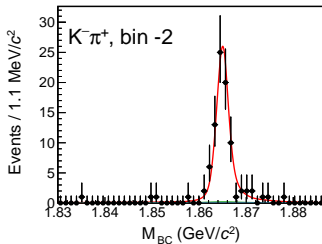
Double tag fits

- Fit strategy: Only fit signal side M_{BC} because of low statistics
- Fit model:
 - Signal: PDF from signal MC, convolved with a Gaussian
 - Flat background: Argus PDF
 - Peaking background: Shape and efficiency from MC, correct for quantum correlation
 - Simple sideband subtraction for correct signal but wrong tag event
- Fit all bins simultaneously
 - Shape is floated and shared across all bins
 - Yield of signal and combinatorial background is floated in each bin

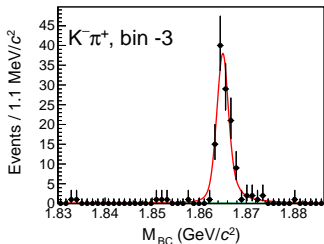
Double tag fit of $KK\pi\pi$ vs $K\pi$



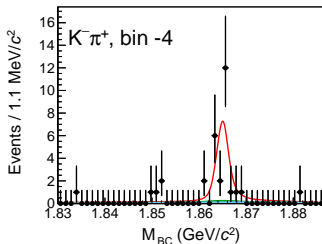
(a) Bin -1 yield: $32.4^{+6.2}_{-5.5}$



(b) Bin -2 yield: $82.7^{+9.7}_{-9.0}$

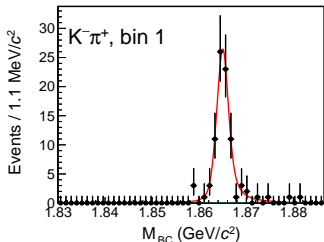


(c) Bin -3 yield: $120.3^{+11.6}_{-10.9}$

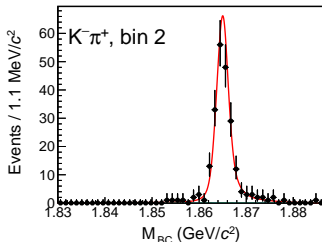


(d) Bin -4 yield: $22.4^{+5.2}_{-4.6}$

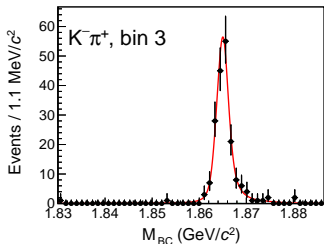
Double tag fit of $KK\pi\pi$ vs $K\pi$



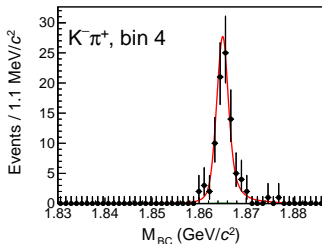
(a) Bin 1 yield: $84.5^{+9.8}_{-9.1}$



(b) Bin 2 yield: $211.2^{+15.4}_{-14.8}$

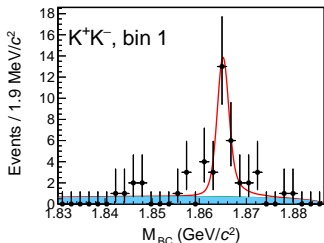


(c) Bin 3 yield: $181.0^{+14.0}_{-13.3}$

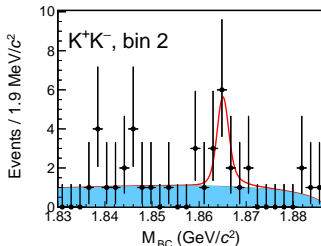


(d) Bin 4 yield: $88.6^{+9.7}_{-9.0}$

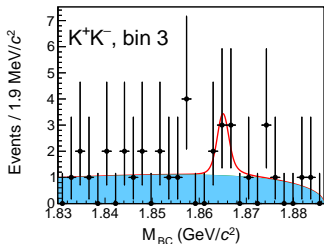
Double tag fit of $KK\pi\pi$ vs KK



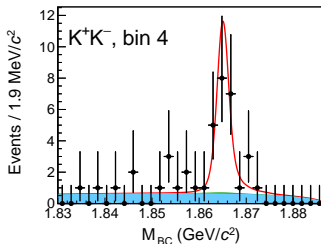
(a) Bin 1 yield: $25.3^{+6.2}_{-5.5}$



(b) Bin 2 yield: $8.8^{+4.0}_{-3.3}$

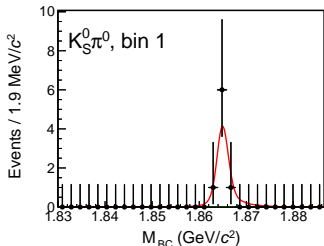


(c) Bin 3 yield: $4.5^{+3.3}_{-2.6}$

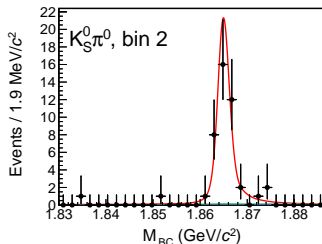


(d) Bin 4 yield: $21.1^{+5.5}_{-4.8}$

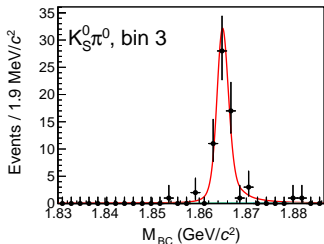
Double tag fit of $KK\pi\pi$ vs $K_S\pi^0$



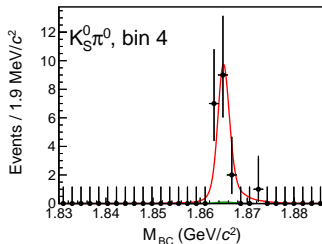
(a) Bin 1 yield: $7.9^{+3.1}_{-2.5}$



(b) Bin 2 yield: $40.4^{+6.8}_{-6.3}$

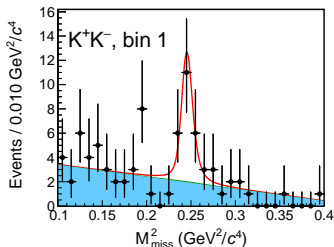


(c) Bin 3 yield: $61.1^{+8.3}_{-7.8}$

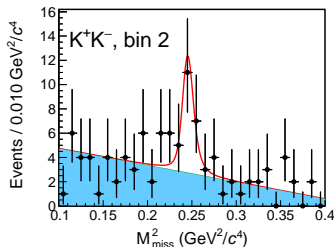


(d) Bin 4 yield: $18.3^{+4.5}_{-3.9}$

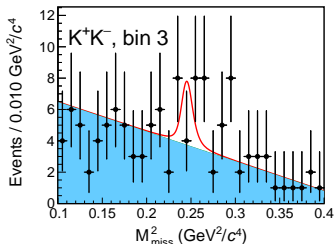
Double tag fit of partially reconstructed $KK\pi\pi$ vs KK



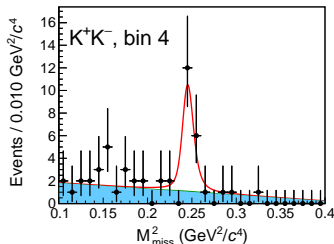
(a) Bin 1 yield: $19.6^{+5.8}_{-5.1}$



(b) Bin 2 yield: $17.7^{+6.1}_{-5.3}$

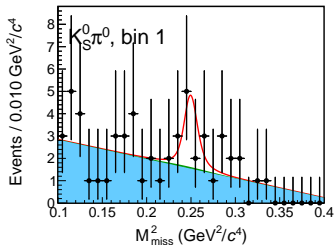


(c) Bin 3 yield: $7.5^{+6.6}_{-5.7}$

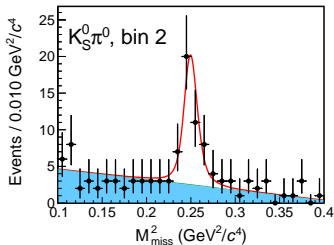


(d) Bin 4 yield: $17.3^{+5.1}_{-4.4}$

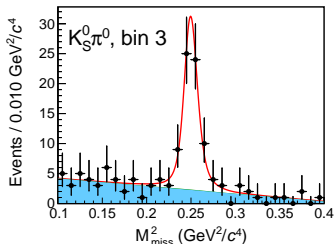
Double tag fit of partially reconstructed $KK\pi\pi$ vs $K_S\pi^0$



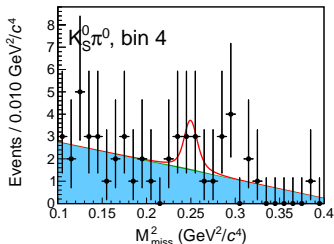
(a) Bin 1 yield: $7.2^{+4.3}_{-3.5}$



(b) Bin 2 yield: $39.3^{+8.3}_{-7.3}$



(c) Bin 3 yield: $64.4^{+9.6}_{-9.0}$



(d) Bin 4 yield: $4.8^{+3.8}_{-3.0}$

How to put this all together to determine c_i and s_i ?

- 1 In the past, the strategy has been to measure normalised K_i first, using flavour tags:

$$\frac{N_i^{\text{DT}}}{N^{\text{ST}}} \propto K_i, \quad \sum_i K_i = 1$$

- 2 However, the K_i must be corrected for efficiencies and bin migrations, in addition to Doubly Cabbibo Suppressed decays:

$$f_i = \left(1 + r_D^2 \frac{K_{-i}}{K_i} - 2\sqrt{\frac{K_{-i}}{K_i}} (c_i \cos(\delta_D) + s_i \sin(\delta_D)) \right)^{-1}$$

- 3 Finally, with the K_i fixed, c_i and s_i are determined from CP and multi-body tags:

$$\begin{aligned} \frac{N_i^{\text{DT}}}{N^{\text{ST}}} &\propto K_i + K_{-i} \mp 2\sqrt{K_i K_{-i}} c_i \\ \frac{N_{ij}^{\text{DT}}}{N^{\text{ST}}} &\propto K_i K'_{-j} + K_{-i} K'_j - 2\sqrt{K_i K_{-i} K'_j K'_{-j}} (c_i c'_j + s_i s'_j) \end{aligned}$$

We propose a simpler, but more powerful strategy:

- 1 Treat all K_i , c_i and s_i as free parameters
- 2 Include the $D \rightarrow KK\pi\pi$ branching fraction \mathcal{B} as a free parameter
- 3 Fit flavour, CP and multi-body tags simultaneously

Master equations:

$$\hat{N}_i^{\text{DT}} = N^{\text{ST}} \mathcal{B} \epsilon_{ij} \left(K_{-j} + r_D^2 K_j - 2r_D \sqrt{K_j K_{-j}} (c_j \cos(\delta_D) + s_j \sin(\delta_D)) \right)$$

$$\hat{N}_i^{\text{DT}} = N^{\text{ST}} \mathcal{B} \epsilon_{ij} (K_j + K_{-j} \mp 2\sqrt{K_j K_{-j}} c_j)$$

$$\hat{N}_{ij}^{\text{DT}} = N^{\text{ST}} \mathcal{B} \epsilon_{ijkl} (K_k K'_{-l} + K_{-k} K'_l - 2\sqrt{K_k K_{-k} K'_l K'_{-l}} (c_k c'_l + s_k s'_l))$$

ϵ_{ij} are combined efficiency and bin migration matrices

The master equations use the free parameters \mathcal{B} , K_i , c_i and s_i to make a prediction \hat{N}^{DT} to the measured double tag yields N^{DT}

Master equations

$$\hat{N}_i^{\text{DT}} = N^{\text{ST}} \mathcal{B}_{\epsilon ij} \left(K_{-j} + r_D^2 K_j - 2r_D \sqrt{K_j K_{-j}} (c_j \cos(\delta_D) + s_j \sin(\delta_D)) \right)$$

$$\hat{N}_i^{\text{DT}} = N^{\text{ST}} \mathcal{B}_{\epsilon ij} (K_j + K_{-j} \mp 2\sqrt{K_j K_{-j}} c_j)$$

$$\hat{N}_{ij}^{\text{DT}} = N^{\text{ST}} \mathcal{B}_{\epsilon ijkl} (K_k K'_{-l} + K_{-k} K'_l - 2\sqrt{K_k K_{-k} K'_l K'_{-l}} (c_k c'_l + s_k s'_l))$$

Ordinarily, we would construct a Gaussian (log)likelihood function \Rightarrow
Obtain \mathcal{B} , K_i , c_i and s_i by minimising the following function:

$$-\ln(\mathcal{L}) = \frac{1}{2} \sum_{\text{Tag}} \sum_{ij} (V^{-1})_{ij} (N_i^{\text{DT}} - \hat{N}_i^{\text{DT}}) (N_j^{\text{DT}} - \hat{N}_j^{\text{DT}})$$

$$V_{ij} = \rho_{ij} \sigma_i \sigma_j$$

¹ ρ are correlation coefficients

Master equations

$$\hat{N}_i^{\text{DT}} = N^{\text{ST}} \mathcal{B}_{\epsilon ij} \left(K_{-j} + r_D^2 K_j - 2r_D \sqrt{K_j K_{-j}} (c_j \cos(\delta_D) + s_j \sin(\delta_D)) \right)$$

$$\hat{N}_i^{\text{DT}} = N^{\text{ST}} \mathcal{B}_{\epsilon ij} (K_j + K_{-j} \mp 2\sqrt{K_j K_{-j}} c_j)$$

$$\hat{N}_{ij}^{\text{DT}} = N^{\text{ST}} \mathcal{B}_{\epsilon ijkl} (K_k K'_{-l} + K_{-k} K'_l - 2\sqrt{K_k K_{-k} K'_l K'_{-l}} (c_k c'_l + s_k s'_l))$$

Our DT yields are very small, so their uncertainties are asymmetric \implies
 Approximate covariance matrix from the asymmetric uncertainties²:

$$-\ln(\mathcal{L}) = \frac{1}{2} \sum_{\text{Tag}} \sum_{ij} (V^{-1})_{ij} (N_i^{\text{DT}} - \hat{N}_i^{\text{DT}}) (N_j^{\text{DT}} - \hat{N}_j^{\text{DT}})$$

$$V_{ij} = \rho_{ij} \sigma_i \sigma_j, \quad \sigma = \sqrt{\sigma_- \sigma_+ - (\sigma_+ - \sigma_-)(N^{\text{DT}} - \hat{N}^{\text{DT}})}$$

²[arXiv:physics/0406120](https://arxiv.org/abs/physics/0406120)

$$-\ln(\mathcal{L}) = \frac{1}{2} \sum_{\text{Tag}} \sum_{ij} (V^{-1})_{ij} (N_i^{\text{DT}} - \hat{N}_i^{\text{DT}})(N_j^{\text{DT}} - \hat{N}_j^{\text{DT}})$$
$$V_{ij} = \rho_{ij} \sigma_i \sigma_j, \quad \sigma = \sqrt{\sigma_- \sigma_+ - (\sigma_+ - \sigma_-)(N^{\text{DT}} - \hat{N}^{\text{DT}})}$$

- The above likelihood has good coverage for flavour and CP tags...
- ... but not for multi-body decays
 - Bins with $\sigma_- \approx 0$ make the fit unstable
 - Fit convergence was found to be less than 60%
- In multi-body decays, use the full unbinned likelihood directly
 - Fit convergence improves to over 95%
 - Much slower, but much more accurate

One last technical detail...

- 8 K_i parameters, but $\sum_i K_i = 1$, so only 7 are independent
- We could set $K_4 = 1 - \sum_{i \neq 4} K_i$, but such a parameterisation is unstable due to strong correlations
- Use a recursive fraction parameterisation ([JHEP 2021 169 \(2021\)](#))
- Recursive fractions R_i are defined as:

$$R_i \equiv \begin{cases} K_i, & i = -4 \\ K_i / \sum_{j \geq i} K_j, & -4 < i < +4 \\ 1, & i = +4. \end{cases}$$

Parameterisation of K_i

Visualisation of this parameterisation:

1. Definition of a'_1

$$\underbrace{a_1}_{\equiv a'_1} + \underbrace{a_2 + a_3 + \dots}_{\equiv 1 - a'_1} = 1$$

2. Eliminate a_1

$$a_2 + a_3 + a_4 + \dots = 1 - a'_1$$

3. Definition of a'_2

$$\underbrace{\frac{a_2}{1 - a'_1}}_{\equiv a'_2} + \underbrace{\frac{a_3 + a_4 + \dots}{1 - a'_1}}_{\equiv 1 - a'_2} = 1$$

Repeat this procedure until the last coefficient, which is 1

Input-output checks

Input-output checks are performed by generating toy datasets

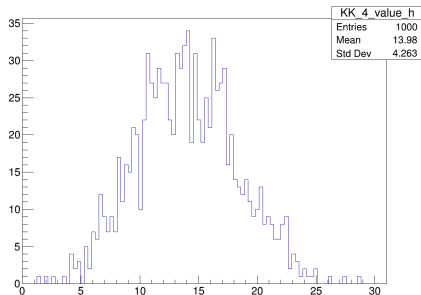
- 1 Check fit convergence
- 2 Check error coverage
- 3 Correct any biases in fitted parameters

Unfortunately, in this analysis we cannot simply generate Poisson-distributed DT yields:

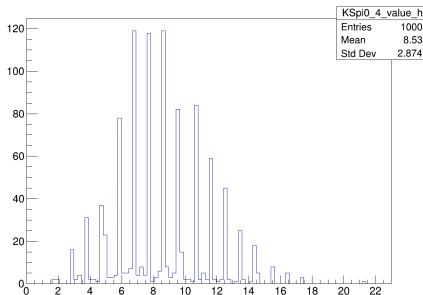
- 1 Asymmetric uncertainties
- 2 Large background-to-signal
- 3 Multi-body tags with low yields require a full unbinned likelihood

Solution: Generate toy datasets for each double tag fit

Input-output checks



(a) Input yield: 13.9

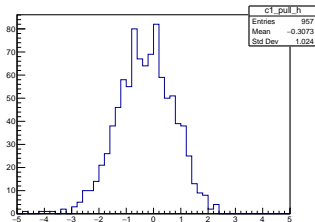


(b) Input yield: 8.6

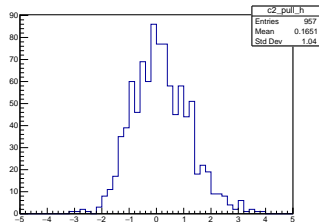
Figure 12: Fitted yields in toy datasets for the (left) KK and (right) $K_S\pi^0$ tags

- 1 No biases are observed
- 2 Distributions are asymmetric \implies uncertainties are asymmetric
- 3 KK uncertainty is non-Poisson because of large backgrounds from $q\bar{q}$
- 4 $K_S\pi^0$ has small backgrounds, so observed yields Poisson distributed

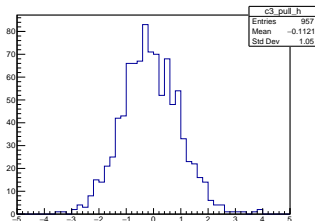
Input-output checks



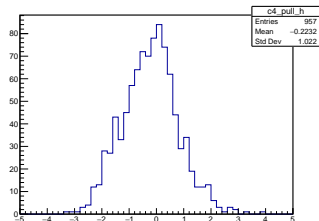
(a) c_1 pulls



(b) c_2 pulls

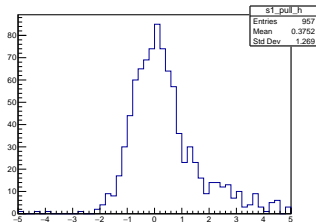


(c) c_3 pulls

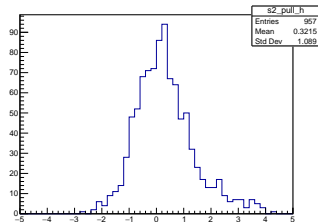


(d) c_4 pulls

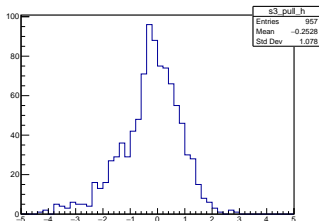
Input-output checks



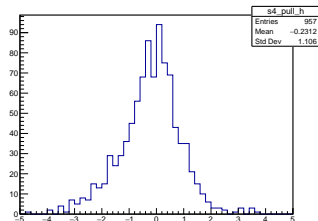
(a) s_1 pulls



(b) s_2 pulls



(c) s_3 pulls



(d) s_4 pulls

What do the toy fits tell us?

- 1 Both c_i and s_i pulls has asymmetric tails
- 2 Effect is small in c_i and a small bias correction will be sufficient
- 3 s_i has a small overcoverage, but a large tail, so special care is required

Fit results

$$c_1 = -0.40 \pm 0.13$$

$$s_1 = -0.18 \pm 0.32$$

$$c_2 = 0.79 \pm 0.06$$

$$s_2 = -0.21 \pm 0.24$$

$$c_3 = 0.85 \pm 0.06$$

$$s_3 = 0.53 \pm 0.24$$

$$c_4 = -0.36 \pm 0.13$$

$$s_4 = -0.38 \pm 0.31$$

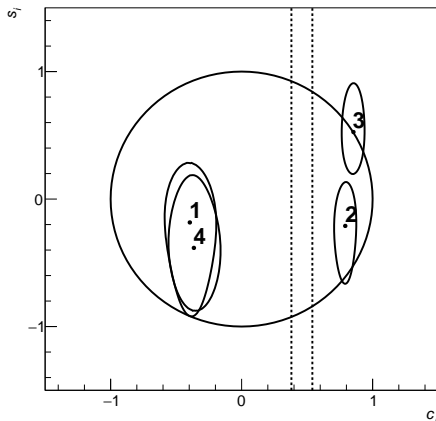


Figure 15: Contours of c_i vs s_i , corresponding to $\Delta \log(\mathcal{L}) = 2.30$, or 68% confidence level. Dashed lines show F_+ measurement.

Fit results

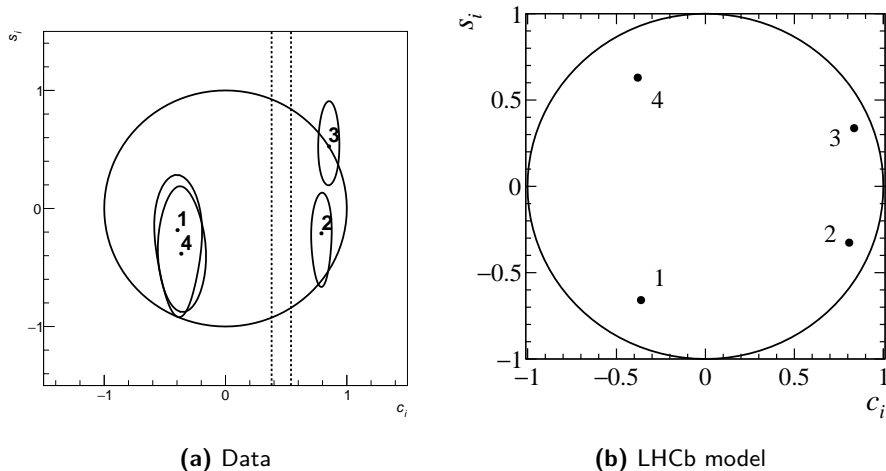


Figure 16: Comparison between model and model-independent measurements

- K_i , which are constrained by flavour tags, are free parameters
- Corrections for DCS decays, which depend on the strong phases δ_D , are part of the fit
- We could instead treat δ_D as a free parameter, and make a simultaneous measurement
- In this analysis we can measure $\delta_{K\pi}$ with negligible reduction in sensitivity to c_i and s_i

$$\hat{N}_i^{\text{DT}} = N^{\text{ST}} \mathcal{B} \epsilon_{ij} \left(K_{-j} + r_D^2 K_j - 2r_D \sqrt{K_j K_{-j}} (c_j \cos(\delta_D) + s_j \sin(\delta_D)) \right)$$

- Free parameters: $r_D \cos(\delta_{K\pi})$ and $r_D \sin(\delta_{K\pi})$

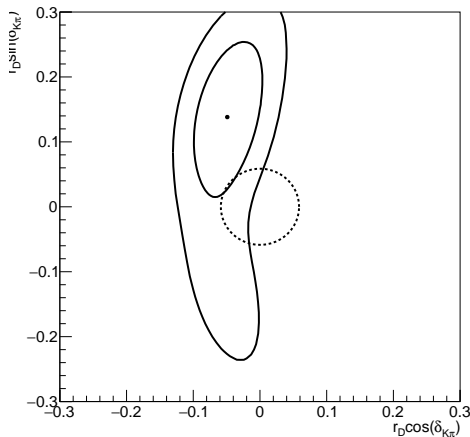


Figure 17: Contours of $r_D \cos(\delta_{K\pi})$ vs $r_D \sin(\delta_{K\pi})$, corresponding to $\Delta \log(\mathcal{L}) = 2.30$ and 6.18 , or 68% and 95% confidence level. Dashed line indicates measured value of r_D .

Branching fraction measurement

In addition, we have another nuisance parameter: The $D^0 \rightarrow K^+ K^- \pi^+ \pi^-$ branching fraction

Current PDG value:

$$\mathcal{B} = (2.47 \pm 0.11) \times 10^{-3}$$

Fit result:

$$\mathcal{B} = (2.765 \pm 0.046) \times 10^{-3}$$

Much better precision than the current world average!

Do we have the necessary estimates of tracking and PID systematics to also publish this measurement?

Summary

- A phase-space binned measurement of the $D^0 \rightarrow K^+ K^- \pi^+ \pi^-$ strong-phases has been performed
 - Analysis uses a 2×4 binning scheme
 - Measurement will improve significantly with full 20 fb^{-1} dataset
 - New strategy: Treat c_i , s_i and K_i on an equal footing
 - Additionally, $\delta_D^{K\pi}$ can also be determined
 - The $D^0 \rightarrow K^+ K^- \pi^+ \pi^-$ branching fraction more than 2 times more precise than the current PDG value
- Next steps:
 - 1 Perform systematics studies
 - 2 Write up MEMO

Thank you!