

# BESIII Oxford Group Meeting

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# Introduction

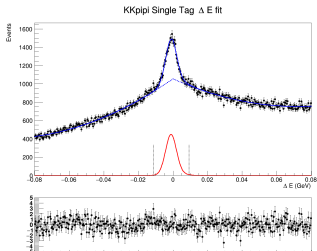
- Strong-phase analysis of  $D^0 \rightarrow K^+ K^- \pi^+ \pi^-$
- Measure  $c_i$ ,  $s_i$  (and  $K_i$ ) using double-tags
- Strategy:
  - 1 Determine single tag yields for normalization
  - 2 Determine  $KK\pi\pi$  vs flavour tag yields to obtain  $K_i$
  - 3 Determine  $KK\pi\pi$  vs CP and self-conjugate tag yields
  - 4 Maximum likelihood fit to obtain  $c_i$  and  $s_i$
- Focused a lot on covering *all* tag modes previously...
- Current plan is to first obtain  $K_i$  and make sure everything works
- Then run toys with yields extrapolated to  $20 \text{ fb}^{-1}$  in fit to  $c_i$  and  $s_i$

## Previous work (before May 2021)

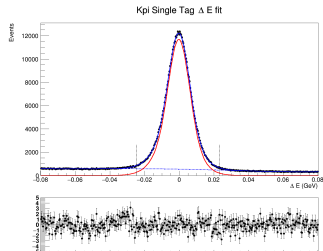
- Fitted single tag  $\Delta E$  distributions
  - Double Gaussian + 2nd order polynomial
  - Apply cut at  $\pm 3\sigma$
- Fitted single tag  $m_{BC}$  distributions
  - Signal shape from MC, Argus for combinatorial background, Gaussian for peaking backgrounds
  - Fit quality very poor in flavour tag modes because of higher statistics
- Double tag yield using sideband subtraction (same as  $K_S KK$  strong-phase analysis)

- Signal: Double Gaussian
- Background: Chebychev polynomials of arbitrary order for two- and three-body modes
- For  $KK\pi\pi$  and  $K\pi\pi\pi$  use two independent polynomials on each side of  $\Delta E = 0$

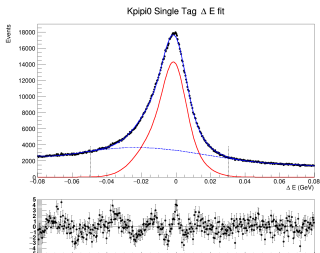
# New $\Delta E$ fits



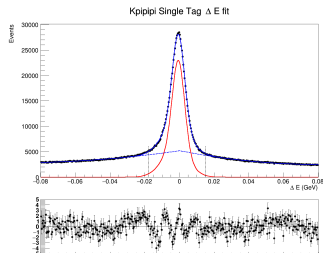
(a)  $KK\pi\pi$



(b)  $K\pi$



(c)  $K\pi\pi^0$



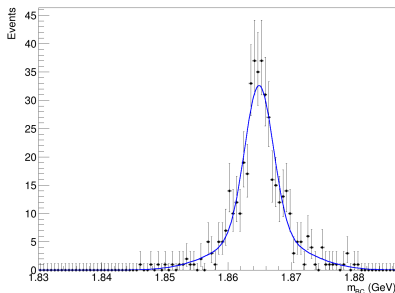
(d)  $K\pi\pi\pi$

# New single tag $m_{BC}$ fits

- Signal: Signal MC shape (after truth matching)
- Resolution: Convolve with double Gaussian
- Combinatorial background: Argus shape
- Can add arbitrary shapes for peaking backgrounds
- New strategy for peaking backgrounds:
  - 1 Identify backgrounds with inclusive MC
  - 2 Generate signal MC and fit shape to this sample
  - 3 Calculate yield using relative efficiencies and branching fractions

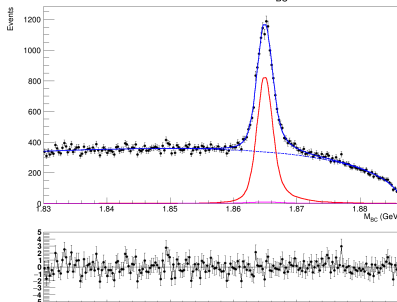
# New single tag $m_{BC}$ fits

KSKK peaking background in KKpipi single tag



(a)  $K_S KK \rightarrow KK\pi\pi$

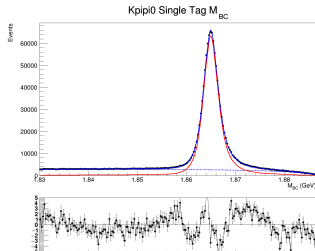
KKpipi Single Tag  $M_{BC}$



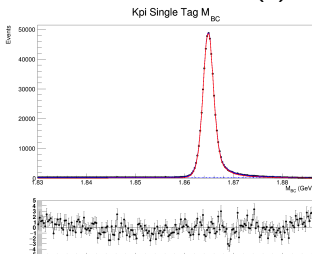
(b)  $KK\pi\pi$   $m_{BC}$  fit

Yield:  $10\,821 \pm 198$

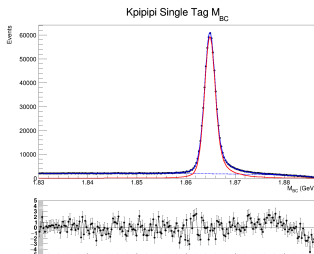
# New single tag $m_{BC}$ fits



(a)  $K\pi\pi^0$   $m_{BC}$  fit



(b)  $K\pi$   $m_{BC}$  fit



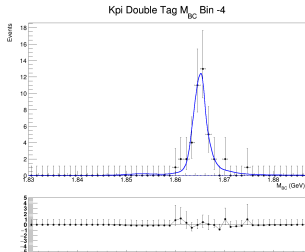
(c)  $K\pi\pi\pi$   $m_{BC}$  fit



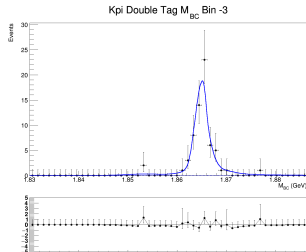
# Double tag yield simultaneous fit ( $KK\pi\pi$ vs $K\pi$ )

- Should have enough events to perform fit of  $KK\pi\pi$   $m_{BC}$ 
  - Same strategy as  $\pi\pi\pi\pi$  strong-phase analysis
  - Use  $2 \times 4$  bins for now
  - Scrap sideband subtraction
- Fit shapes:
  - Signal shape: Signal MC shape (after truth matching)
  - Resolution: Single Gaussian
  - Combinatorial background: Argus shape
- Fit strategy:
  - 1 For each double tag mode, perform simultaneous fit of all bins
  - 2 Float signal and combinatorial yield in each bin
  - 3 Gaussian shape and Argus slope are floated
  - 4 Then fix all shapes and any combinatorial background less than 0.5
  - 5 Fit a second time to obtain accurate signal yields in each bin

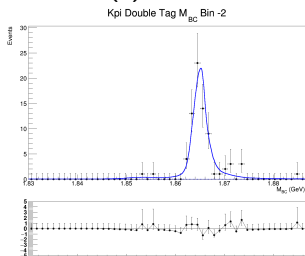
# Double tag yield simultaneous fit ( $KK\pi\pi$ vs $K\pi$ )



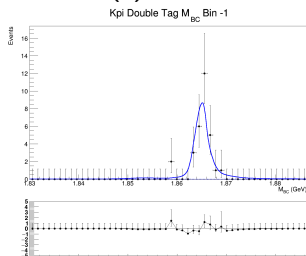
(a) Bin -4



(b) Bin -3

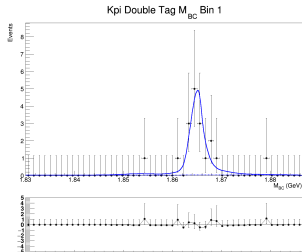


(c) Bin -2

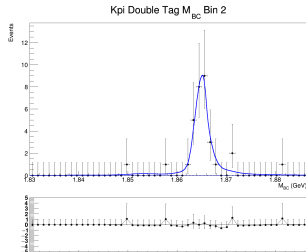


(d) Bin -1

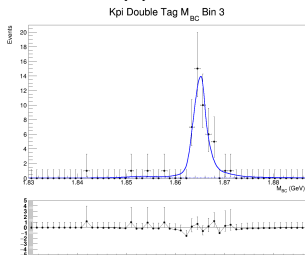
# Double tag yield simultaneous fit ( $KK\pi\pi$ vs $K\pi$ )



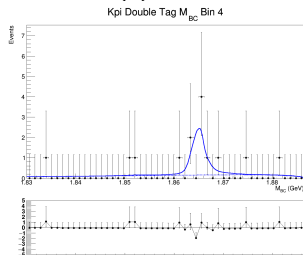
(a) Bin 1



(b) Bin 2



(c) Bin 3

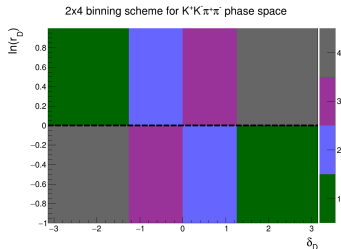


(d) Bin 4

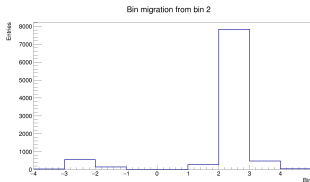
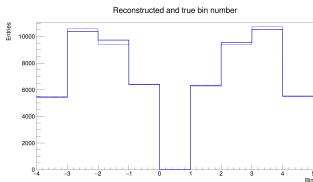
- Single tag efficiencies trivially corrected by dividing the yields
- For double tags, construct an efficiency matrix to account for both efficiency and bin migration
- For events reconstructed in bin  $i$  and generated in bin  $j$ :

$$\epsilon_{ij} = \frac{N_{ij}^{\text{reconstructed}}}{N_j^{\text{generated}}}$$

# Efficiency corrections



(a) Binning scheme



(b) Reconstructed bin vs true bin (c) Bin migration from bin 2

Bin migration much higher than expected (10%-15%)

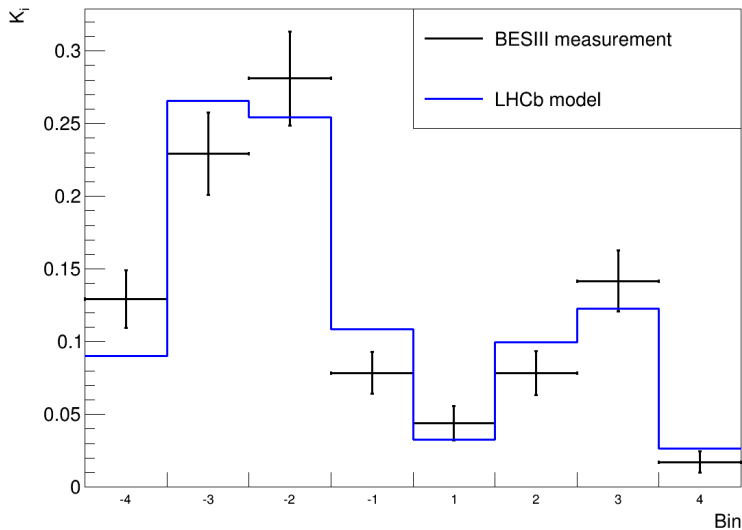
But net migration is relatively small

Calculate using  $c_i$ ,  $s_i$ ,  $K_i$  from LHCb model

Bin	Correction
-4	$1.0091 \pm 0.0017$
-3	$0.9538 \pm 0.0007$
-2	$0.9669 \pm 0.0014$
-1	$1.0166 \pm 0.0016$
+1	$1.0599 \pm 0.0228$
+2	$0.7833 \pm 0.0027$
+3	$0.8263 \pm 0.0060$
+4	$1.1850 \pm 0.0229$

# $K_i$ results for $KK\pi\pi$ vs $K\pi$

$K_i$  from  $KK\pi\pi$  vs  $K\pi$  tags



# Conclusion and next steps

- Single tag  $m_{BC}$  show acceptable fit quality now
- Double tag simultaneous fit works great!
- $K_i$  from  $KK\pi\pi$  vs  $K\pi$  tag looks encouraging, but not perfect
- Next steps:
  - Include  $K\pi\pi^0$ ,  $K\pi\pi\pi$  and  $Ke\nu$  tags in  $K_i$  determination
  - Calculate peaking backgrounds in each bin with LHCb model
  - Develop likelihood fitter for  $c_i/s_i$  and run some toys
  - Start looking at CP tag fits to data (but yields will be very low!)
  - Remove  $K_S\phi$ , include  $K_S KK$  and  $K_L KK$  as self-conjugate tags