

# $F_+(D \rightarrow 4\pi)$ from $K_S\pi\pi$ vs $4\pi$ and $K_L\pi\pi$ vs $4\pi$

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**Claire Prouve**, Chris Thomas, Guy Wilkinson, Sneha Malde

# What is $F_+(D \rightarrow 4\pi)$ ?

$F_+$  is the CP-even fraction of the  $D \rightarrow 4\pi$  decay:

$$F_+(D \rightarrow K^+ K^-) = 1 \quad \text{CP even}$$

$$F_+(D \rightarrow K_S^0 \pi^0) = 0 \quad \text{CP odd}$$

$$F_+(D \rightarrow 2\pi^+ 2\pi^-) = [0, 1] \quad \text{CP mixture}$$

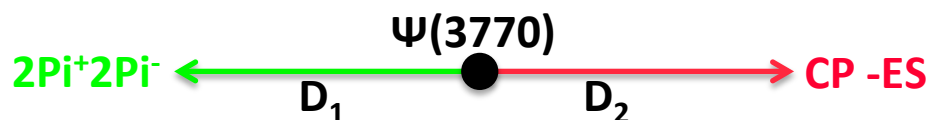
Measure  $F_+$  using the quantum correlated  $D^0/D^0$  decays at CLEO:

CLEO-c reminder:  $ee \rightarrow \Psi(3770) \rightarrow D^0 D^0 / D^- D^+$

- Straight forward measurement of  $F_+$ :

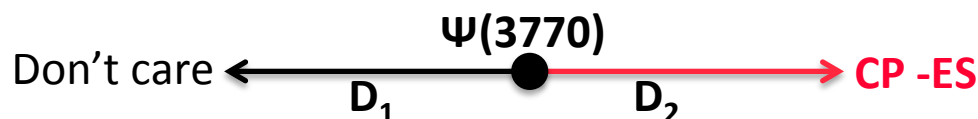
1. count  $4\pi$  vs. CP even/ CP odd eigenstates

$\Rightarrow M^+ / M^-$



2. count the total number of CP even/ CP odd decays in sample

$\Rightarrow S^+ / S^-$



3.

$$N^+ = M^+ / S^+ \quad \Rightarrow \quad F_+ \equiv \frac{N^+}{N^+ + N^-}$$

# Measuring $F_+$ with $K^0\text{PiPi}$

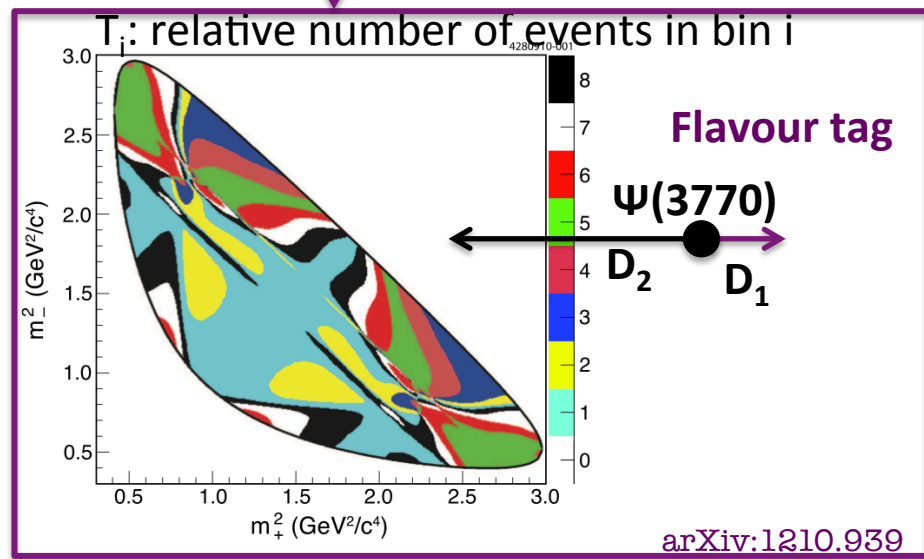
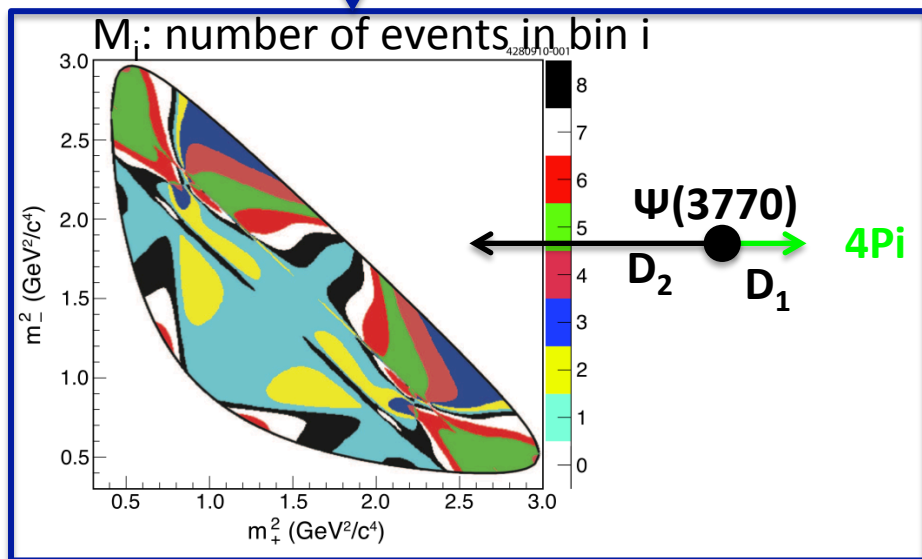
$F_+$  is related to the strong phase difference in the D-decay:

We know the **binned** strong phase difference of  $D \rightarrow K_S^0\text{PiPi}$  and  $D \rightarrow K_L^0\text{PiPi}$ !  
(arXiv:1010.2817)

Amplitude weighted cos of  
strong phase difference

Here is how to get  $F_+(4\text{Pi})$ :

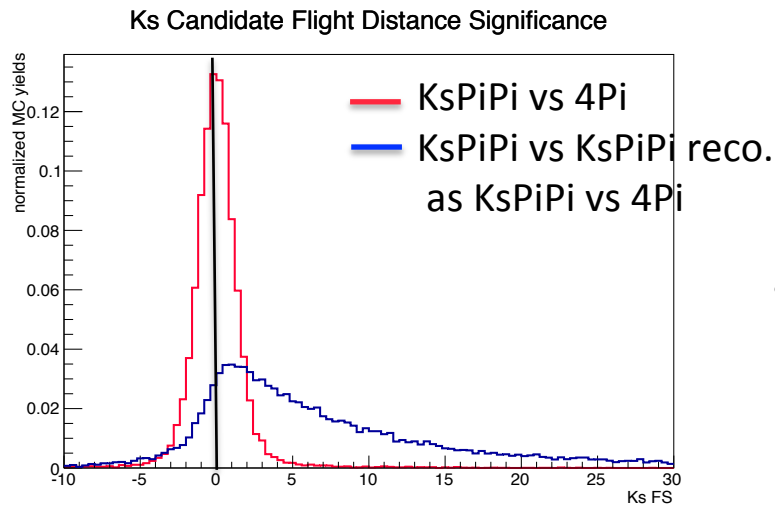
$$M_i^{K_S^0\pi\pi} = h_{K_S^0\pi\pi} (T_i^{K_S^0\pi\pi} + T_{-i}^{K_S^0\pi\pi} - 2c_i \sqrt{T_i^{K_S^0\pi\pi} T_{-i}^{K_S^0\pi\pi}} (2F_+^{4\pi} - 1))$$



➔ Select  $K^0\text{PiPi}$  vs  $4\text{Pi}$  events and count how many signal events there are in each bin!  
(and then fit the data with the above formula for  $F_+$ )

# $K_s\pi\pi$ vs $4\pi$

- Fairly boring selection (CLEO is just so clean)
- Fitting the final state particles to the D Mass
- Rejecting  $K_s\pi\pi$  in the  $4\pi$  side using the flight distance significance of a *possible*  $K_s$  formed of two opposite sign pions



cut of  $KS\_FS < 0$ :

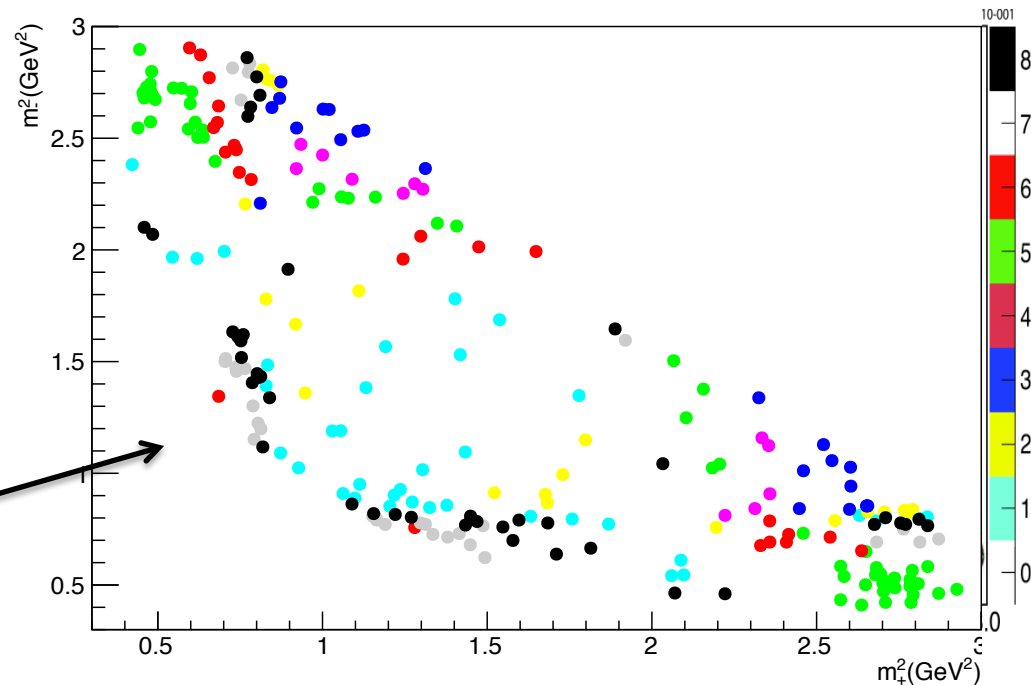
Signal efficiency: 85.1%

bkg efficiency: 9.6%

Getting the signal yield:

Signal yield = raw event yield

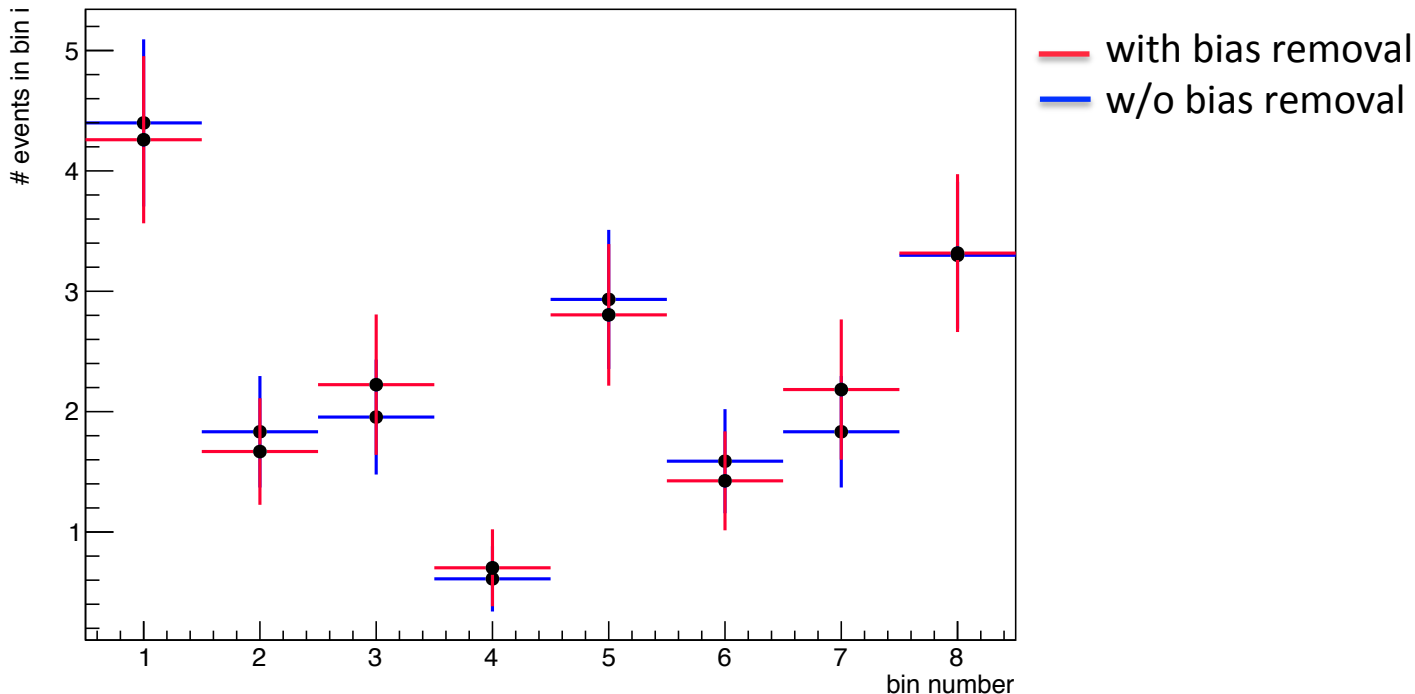
- peaking bkg yield
- flat bkg yield



# $K_S\text{PiPi}$ vs $4\text{Pi}$ – peaking bkg

Peaking bkg:  $K_S\text{PiPi}$  vs  $K_S\text{PiPi}$   $B_i^{\text{peak}} = B_{\text{tot}}^{\text{peak}} \cdot a_i^{\text{peak}}$

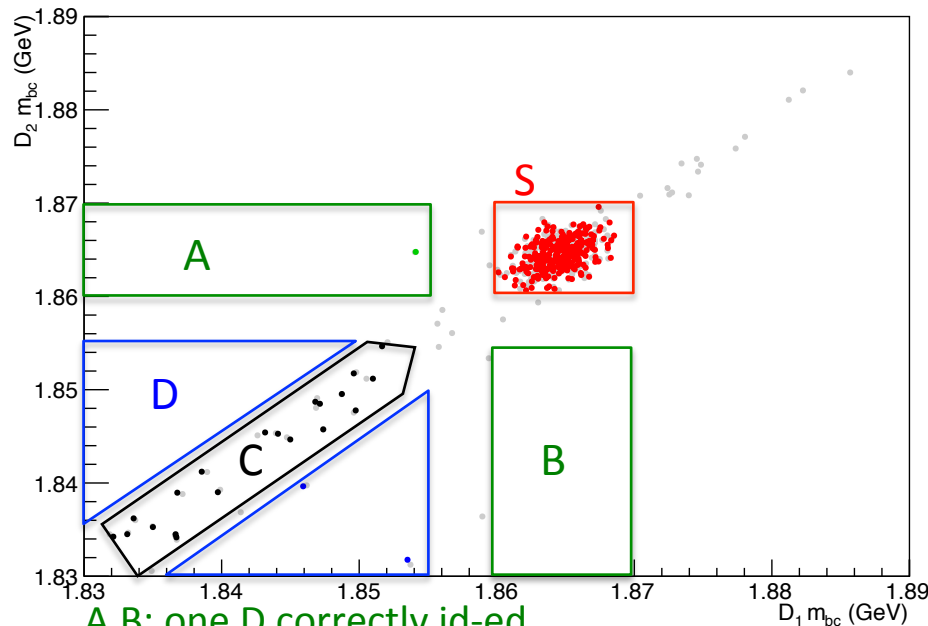
- total number of peaking bkg events from *generic MC*:  $18.45 \pm 1.13$
- distribution of peaking bkg events over bins from data:  
apply  $K_S\text{PiPi-Selection}$  cut on  $4\text{Pi}$  side to get a  $K_S\text{PiPi}$  vs  $K_S\text{PiPi}$  data sample
- Remove bias due to  $K_S$ -Selection instead of  $K_S$ -Veto with  $K_S\text{PiPi}$  vs  $K_S\text{PiPi}$  Signal MC



# $K_S\text{PiPi}$ vs $4\text{Pi}$ – flat bkg

Total number of bkg events determined extrapolating from the sidebands in data:

$$B_{tot}^{flat} = \frac{a_S}{a_D} N_D + \sum_{j=A,B,C} \frac{a_S}{a_j} \left( N_j - \frac{a_j}{a_D} N_D \right)$$



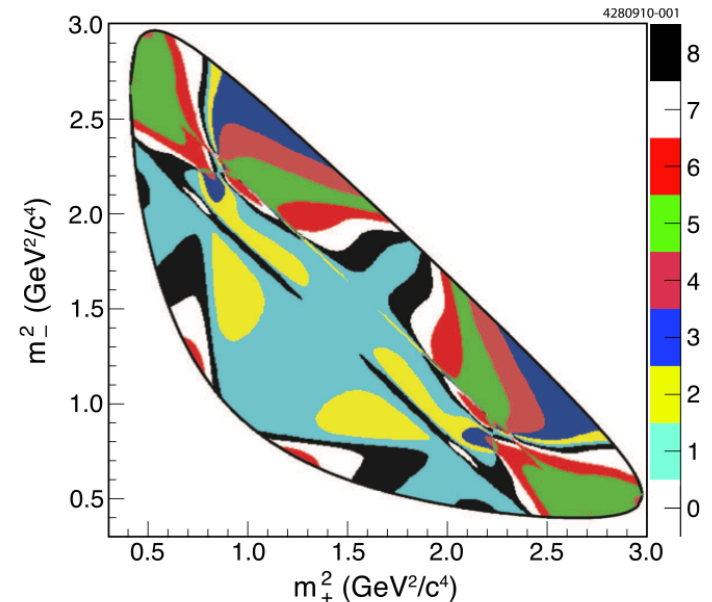
A,B: one D correctly id-ed

C: continuum bkg

D: pure combinatoric

$\Rightarrow 11.64 \pm 2.90$  flat bkg events in the signal region

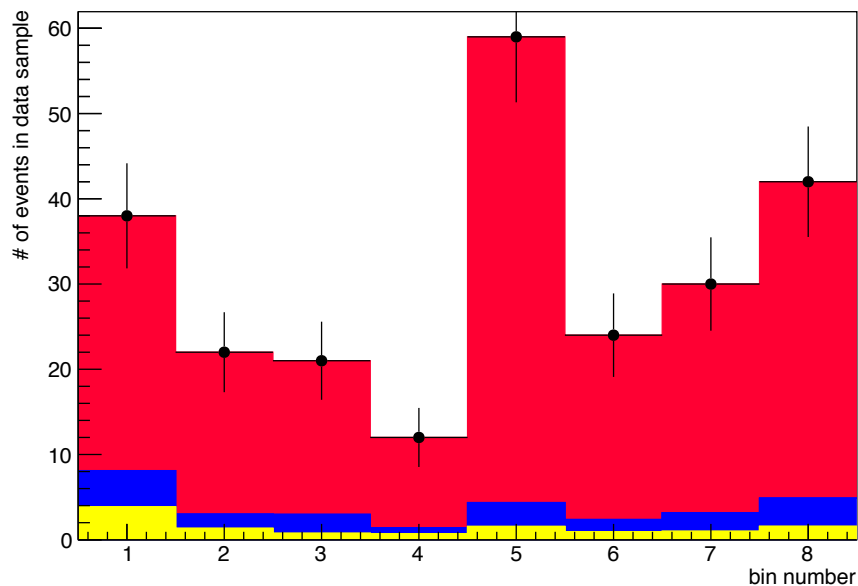
Distribute them 'flatly' across the bins  
= proportional to the area of the bin



# $K_S\text{PiPi}$ vs $4\text{Pi}$ – signal yields

$$M_i = (N_i^{\text{meas}} - B_i^{\text{peak}} - B_i^{\text{flat}})$$

$4\pi$  vs  $K_S\pi\pi$



## Efficiency corrected signal yields:

stat errors only, (first MC, second data)

bin	$MM_i$
1	$185.4 \pm 2.4058 \pm 39.054$
2	$119.24 \pm 1.8904 \pm 29.642$
3	$98.618 \pm 2.1205 \pm 25.254$
4	$61.551 \pm 1.1733 \pm 20.21$
5	$333.14 \pm 4.3184 \pm 47.018$
6	$127.08 \pm 2.1409 \pm 28.908$
7	$165.69 \pm 2.895 \pm 34.025$
8	$222.47 \pm 3.2254 \pm 39.089$

# $K_L\text{PiPi}$ vs $4\text{Pi}$ – peaking bkg

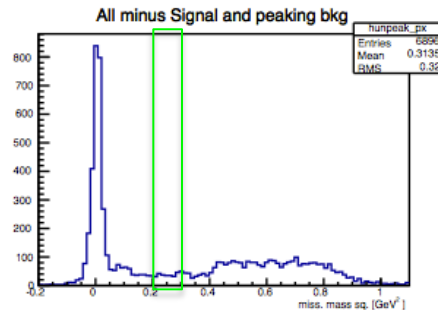
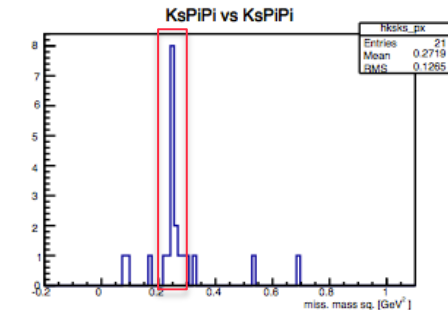
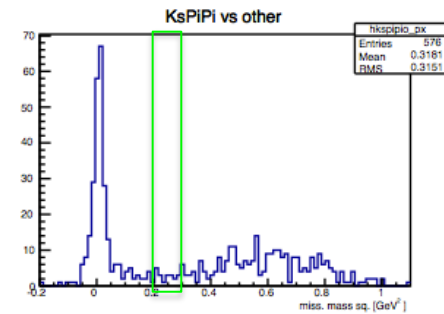
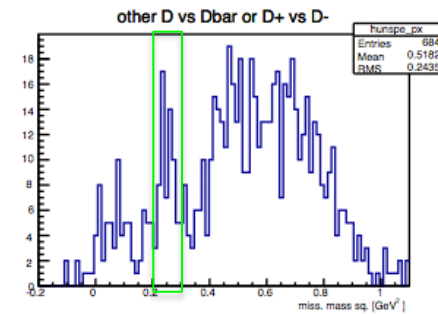
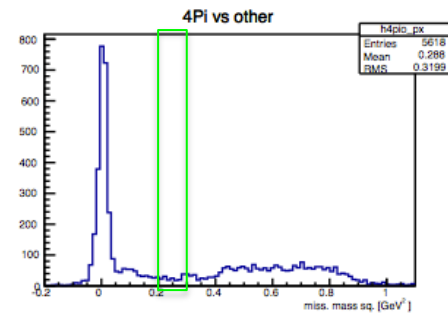
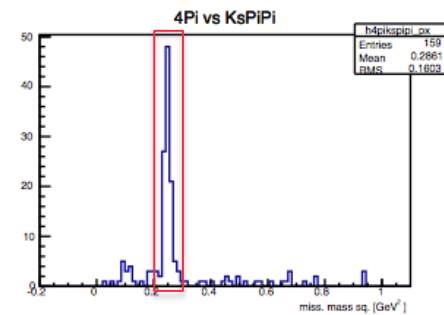
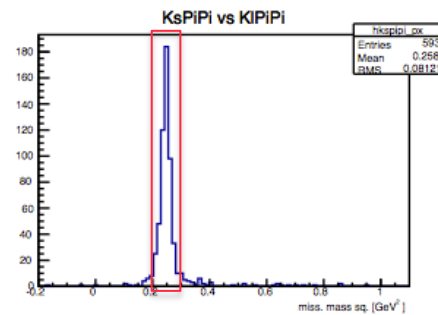
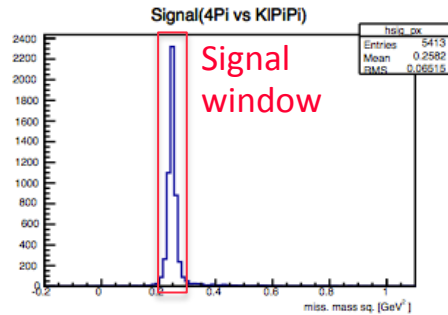
Same strategy as for  $K_L\text{PiPi}$  vs  $4\text{Pi}$ , but more tricky due to more and different bkg.

## Peaking Bkg:

Generic MC:  
Reconstruct  
 $K_L\text{PiPi}$  vs  $4\text{Pi}$  and see  
what we get →

## Peaking in SW:

- $K_L^0\pi\pi$  vs  $K_S^0\pi\pi$ : 8% of
- $K_S^0\pi\pi$  vs  $4\pi$ : 1.9% of
- ~~$K_S^0\pi\pi$  vs  $K_L^0\pi\pi$ : 0.3%~~





# $K_L \text{PiPi}$ vs $4\text{Pi}$ – peaking bkg

## 1. Scale the total number of peaking bkg events from generic MC:

$$B_{tot}^{peak}(K_L^0 \pi\pi \text{ vs } K_S^0 \pi\pi) = 31.24 \pm 1.46$$

$$B_{tot}^{peak}(K_S^0 \pi\pi \text{ vs } 4\pi) = 18.45 \pm 0.62$$

## 2. Get $K_L \text{PiPi}$ vs $K_S \text{PiPi}$ distribution over bin using data with $K_S$ -Selection on the $4\text{Pi}$ side:

generic MC tells us that the data sample with  $K_S$ -Selection contains:

- ➔ Subtract the flat distribution from the data to get the peaking bkg
- ➔ Remove the bias introduced by the  $K_S$ -Selection

- $K_L^0 \pi\pi$  vs  $K_S^0 \pi\pi$ : 92%

- flat events: 5%

- ~~$K_L^0 \pi\pi$  vs  $4\pi$ : 1.5%~~

- ~~$K_S^0 \pi\pi$  vs  $K_S^0 \pi\pi$ : 1.5%~~  
Put a systematic on it later

## 3. Get $K_S \text{PiPi}$ vs $4\text{Pi}$ distribution over bin using data from $K_S \text{PiPi}$ vs $4\text{Pi}$ analysis :

- ➔ Take results from  $K_S \text{PiPi}$  vs  $4\text{Pi}$  and reweight according to the bin-dependent efficiency in  $K_L \text{PiPi}$  vs  $4\text{Pi}$

The efficiency for reconstructing and selecting  $K_S \text{PiPi}$  vs  $4\text{Pi}$  events as  $K_L \text{PiPi}$  vs  $4\text{Pi}$  events is taken to be the same as that for genuine  $K_L \text{PiPi}$  vs  $4\text{Pi}$  events - under the assumption the  $K_S$  is 'lost' which should result in the same kinematics of the two decays.

# $K_L\text{PiPi}$ vs 4Pi – flat+cont bkg

Since for  $K_L\text{PiPi}$  vs 4Pi only 4 pions are reconstructed, the amount of continuum bkg is considerable.

Use the *Powell method* to determine the total number of flat and cont. bkg event

- Divide the MissMassSq region into lower sideband, signal region, upper sideband
- Assume that the ratio of events between these regions is correctly modeled by MC:

$$\text{for example } N_{\text{sig}}^{K_L\text{PiPivs4Pi}} = MC_{\text{sig}}^{K_L\text{PiPivs4Pi}} / MC_{\text{LU}}^{K_L\text{PiPivs4Pi}} N_{\text{LU}}^{K_L\text{PiPivs4Pi}}$$

- 3 equations with 3 unknowns:  $D_i = T_{Si} + T_{Pi} + T_{Ci} + T_{Fi}$

Raw data yields

Signal yields

Peaking bkg

Continuum bkg

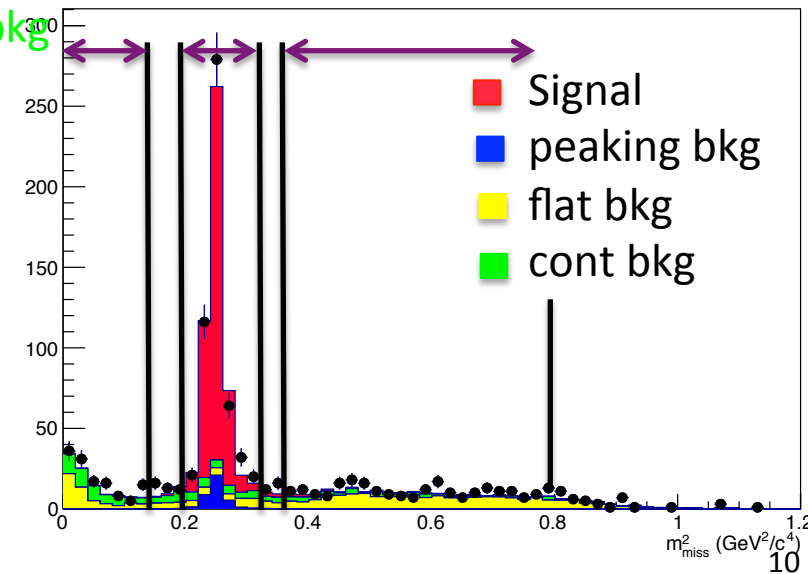
Flat bkg

- Solve and get:

$$B^{\text{flat}} = 28.42 \pm 10.80 \pm 6.80$$

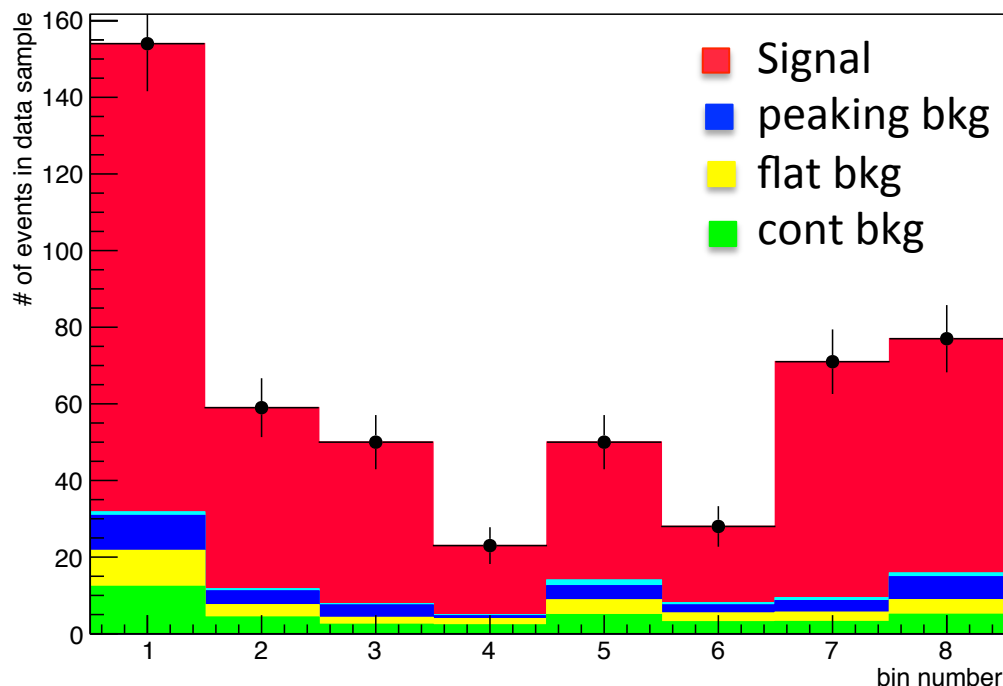
$$B^{\text{cont}} = 37.15 \pm 10.85 \pm 9.45$$

- Distribute 'flatly' over bins, i.e. according to bin area.



# $K_L\text{PiPi}$ vs $4\text{Pi}$ – signal yield

$4\pi$  vs  $K_L\pi\pi$



Preliminary: we are still fiddling around with some yields here...

## Efficiency corrected signal yields:

stat errors only, (first MC, second data)

bin	$MM_i$		
1	512.33	$\pm 24.16$	$\pm 54.52$
2	204.17	$\pm 10.19$	$\pm 33.69$
3	202.33	$\pm 10.02$	$\pm 34.23$
4	78.20	$\pm 4.89$	$\pm 20.92$
5	164.73	$\pm 11.46$	$\pm 33.18$
6	82.38	$\pm 6.30$	$\pm 22.26$
7	268.10	$\pm 10.49$	$\pm 36.93$
8	277.09	$\pm 13.55$	$\pm 40.47$

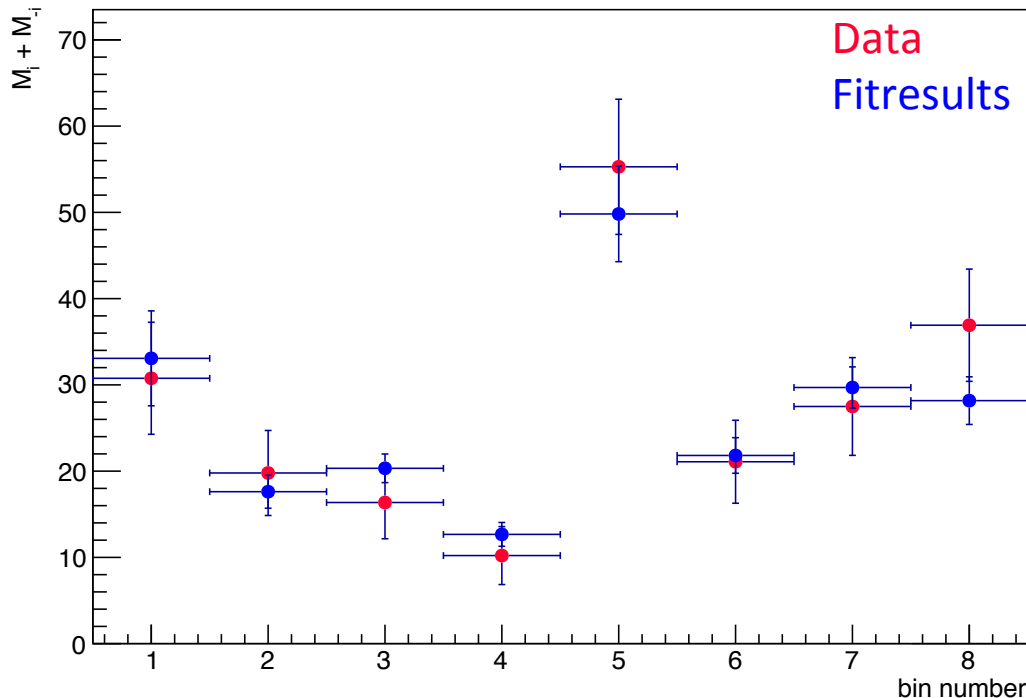
# $F_+$ - from $K_S\pi\pi$ vs $4\pi$

Reminder:  $M_i^{K_S^0\pi\pi} = h_{K_S^0\pi\pi} (T_i^{K_S^0\pi\pi} + T_{-i}^{K_S^0\pi\pi} - 2c_i \sqrt{T_i^{K_S^0\pi\pi} T_{-i}^{K_S^0\pi\pi}} (2F_+^{4\pi} - 1))$

Just measured  $\rightarrow$  External input  $\rightarrow$  Want this

In the Fit: Gaussian constrain all input parameters, take into account correlation between  $c_i$ , fit for  $F_+$  and  $h$

$K_S\pi\pi$  vs  $4\pi$



Thanks to  
Jeremy's  
tutorial

$F_+ = 0.830 \pm 0.071$   
Chi2/ndof = 0.7

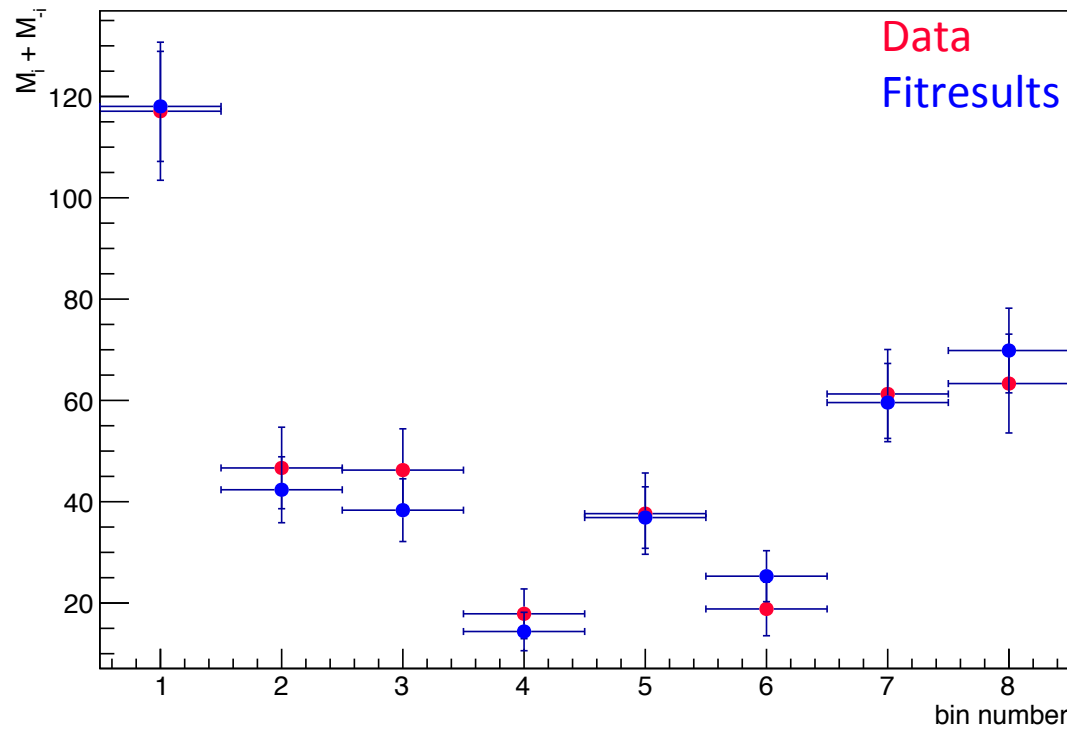
# $F_+$ - from $K_L \text{PiPi}$ vs $4\text{Pi}$

Reminder:  $M_i^{K_L^0 \pi \pi} = h_{K_L^0 \pi \pi} (T_i^{K_L^0 \pi \pi} + T_{-i}^{K_L^0 \pi \pi} + 2c'_i \sqrt{T_i^{K_L^0 \pi \pi} T_{-i}^{K_L^0 \pi \pi}} (2F_+^{4\pi} - 1))$

Just measured  $\rightarrow$  External input  $\rightarrow$  Want this

In the Fit: Gaussian constrain all input parameters, take into account correlation between  $c_i$ , fit for  $F_+$  and  $h$

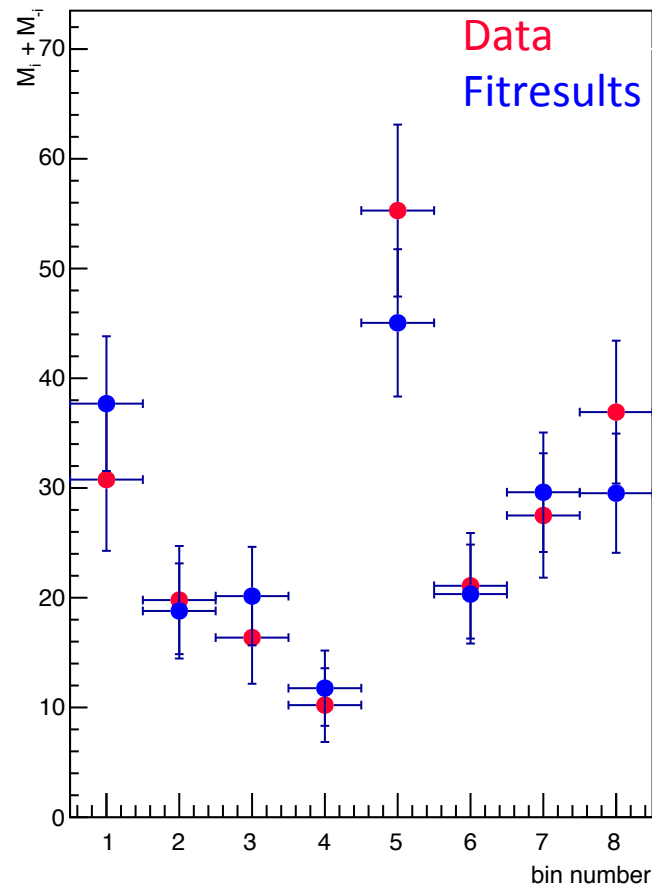
$K_L \pi \pi$  vs  $4\pi$



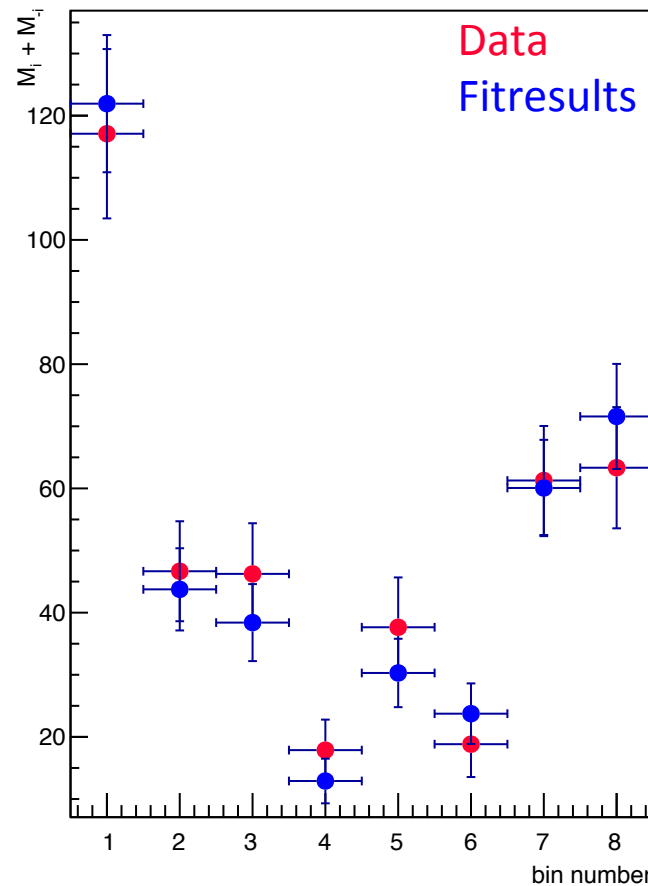
$F_+ = 0.678 \pm 0.069$   
 $\text{Chi2/ndof} = 0.62$

# $F_+$ from simultaneous fit

$K_S\pi\pi$  vs  $4\pi$



$K_L\pi\pi$  vs  $4\pi$



$$F_+ = 0.753 \pm 0.052$$
$$\text{Chi2/ndof} = 0.77$$

$K_S\text{PiPi}$  vs  $4\text{Pi}$ :

$$F_+ = 0.830 \pm 0.071$$
$$\text{Chi2/ndof} = 0.7$$

$K_L\text{PiPi}$  vs  $4\text{Pi}$ :

$$F_+ = 0.678 \pm 0.069$$
$$\text{Chi2/ndof} = 0.62$$

As I said: more fiddling has to be done on the  $K_L\text{PiPi}$  part...

# Question 1

When distributing the total number of flat bkg events over the bins  
no error on the area of the bin is assumed.

→ The error on the bkg yield per bin is smaller than  $\sqrt{N_i}$   
Is that a problem? How to fix it?

For the distribution of peaking bkg I only take the error from

distribution in data (multinomial error)

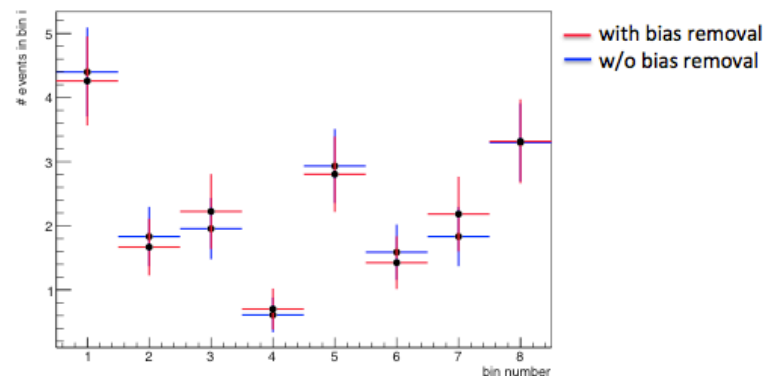
rescaling from MC (ratio of two efficiencies)

→ Does that not introduce the same problem as above?

## $K_S \text{PiPi}$ vs $4\text{Pi}$ – peaking bkg

Peaking bkg:  $K_S \text{PiPi}$  vs  $K_S \text{PiPi}$   $B_i^{\text{peak}} = B_{\text{tot}}^{\text{peak}} \cdot a_i^{\text{peak}}$

- total number of peaking bkg events from *generic MC*:  $18.45 \pm 1.13$
- distribution of peaking bkg events over bins from data:  
apply  $K_S \text{PiPi}$ -Selection cut on  $4\text{Pi}$  side to get a  $K_S \text{PiPi}$  vs  $K_S \text{PiPi}$  data sample
- Remove bias due to  $K_S$ -Selection instead of  $K_S$ -Veto with  $K_S \text{PiPi}$  vs  $K_S \text{PiPi}$  Signal MC



# Question 2

## $K_L\pi^+\pi^-$ vs $4\pi$ – error on signal yields

Done using a toy study:

1. in the Powell method Poisson/Gauss vary all inputs (MC and data independently)

$$D_1 - T_{P1} = \frac{m_{S1}}{m_{S2}} T_{S2} + T_{C1} + \frac{m_{F1}}{m_{F3}} T_{F3}$$

$$D_2 - T_{P2} = T_{S2} + \frac{m_{C2}}{m_{C1}} T_{C1} + \frac{m_{F2}}{m_{F3}} T_{F3}$$

$$D_3 - T_{P3} = \frac{m_{S1}}{m_{S3}} T_{S2} + \frac{m_{C3}}{m_{C1}} T_{C1} + T_{F3}$$

2. Still from within the toy study calculate the flat bkg, cont bkg yields per bin

3. Subtract from data yield per bin to get signal yields per bin

The data yield is needed twice: in step 1 and in step 2:

I generate the data yields for each bin (which I am going to use in step 2), and use the sum as input for step 1. ➔ I am using the same data yield in the entire toy. Is that correct?



# Summary

- Calculated the  $K_S\text{PiPi}$  vs  $4\text{Pi}$  signal yield per bin
- Almost finished calculating the  $K_L\text{PiPi}$  vs  $4\text{Pi}$  signal yield per bin
- Fitter is fully setup and functional

## Still to do:

- Some kinks to iron out in the  $K_L\text{PiPi}$  vs  $4\text{Pi}$  part
- Generate more  $K_L\text{PiPi}$  vs  $4\text{Pi}$  signal MC for the efficiency correction
- Calculate ALL the systematics

# Backup

Binned strong phase difference

$$c_i \equiv \frac{1}{\sqrt{T_i T_{-i}}} \int_{\mathcal{D}_i} a_{xy} a_{yx} \cos(\delta_{xy} - \delta_{yx}) dx dy,$$

Amplitude of D going to point x,y in the Dalitz Plot

# Backup

Powell method:

$$D_1 - T_{P1} = \frac{m_{S1}}{m_{S2}} T_{S2} + T_{C1} + \frac{m_{F1}}{m_{F3}} T_{F3}$$

$$D_2 - T_{P2} = T_{S2} + \frac{m_{C2}}{m_{C1}} T_{C1} + \frac{m_{F2}}{m_{F3}} T_{F3}$$

$$D_3 - T_{P3} = \frac{m_{S1}}{m_{S3}} T_{S2} + \frac{m_{C3}}{m_{C1}} T_{C1} + T_{F3}$$