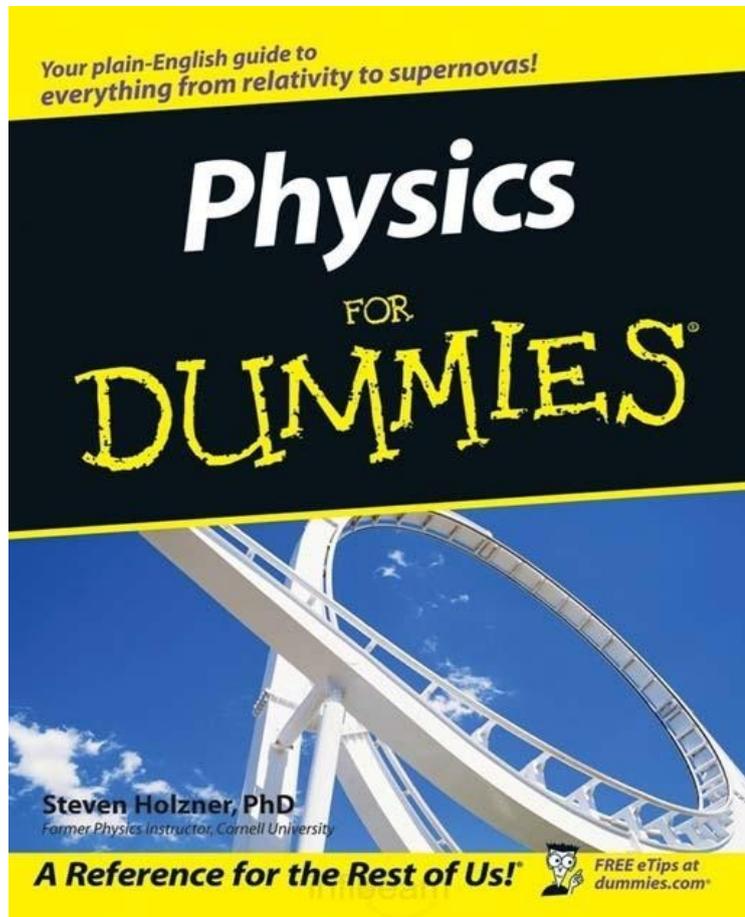


# Branching Fraction Measurements for Dummies

$$(B^0 \rightarrow \pi^+ \pi^-)$$

Kolja Prothmann



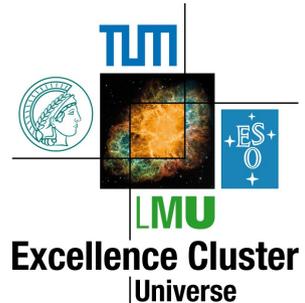
- Motivate basic procedure
- Reconstruct a  $B \rightarrow \pi\pi$  event
- Extract the variables



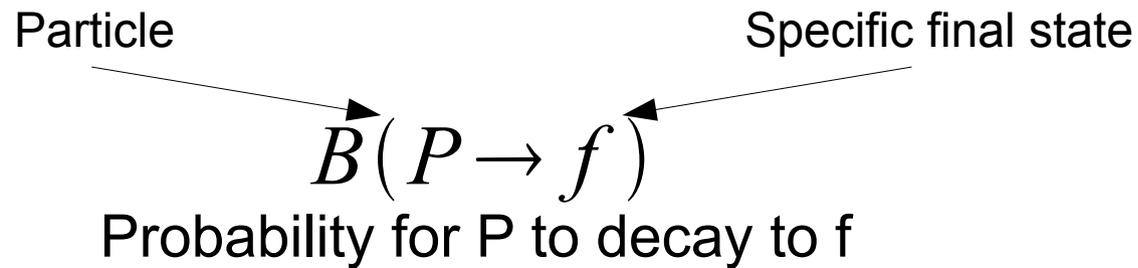
Max-Planck-Institut für Physik  
(Werner-Heisenberg-Institut)



MAX-PLANCK-GESELLSCHAFT

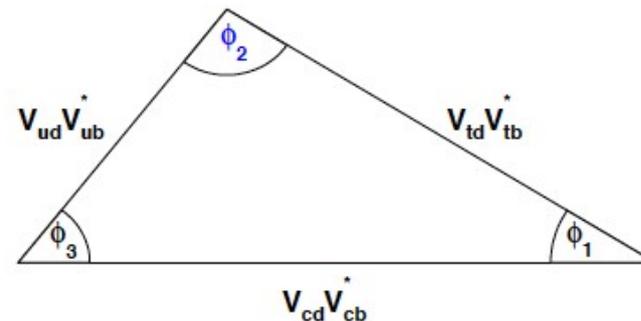


- What is the Branching fraction?

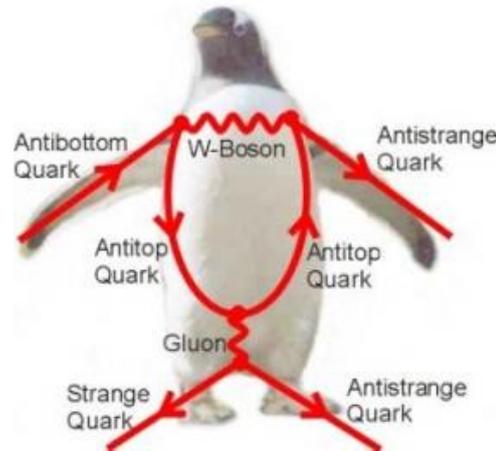


- Why is the branching fraction interesting?
  - Probability itself interesting
  - Can measure CKM  $|V_{ij}|$
  - Needed on my way to measure CP violation
  - Needed for isospin analysis to remove penguin contributions

$$V_{\text{CKM}} \equiv \begin{pmatrix} V_{ud} & V_{us} & V_{ub} \\ V_{cd} & V_{cs} & V_{cb} \\ V_{td} & V_{ts} & V_{tb} \end{pmatrix}$$



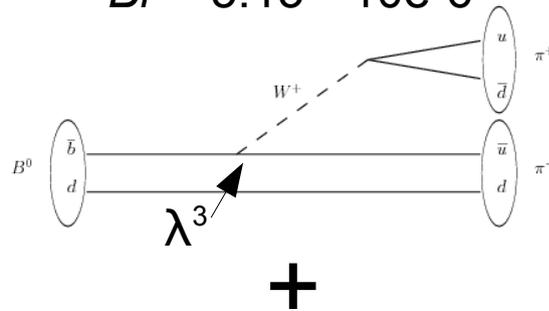
Feynman diagram:  
Shape of a penguin



Tree diagram

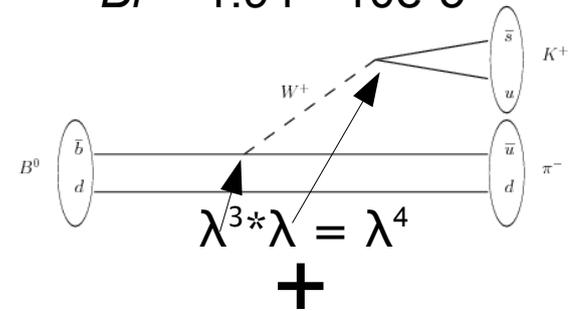
$$B^0 \rightarrow \pi^+ \pi^-$$

$$Br = 5.13 * 10e-6$$

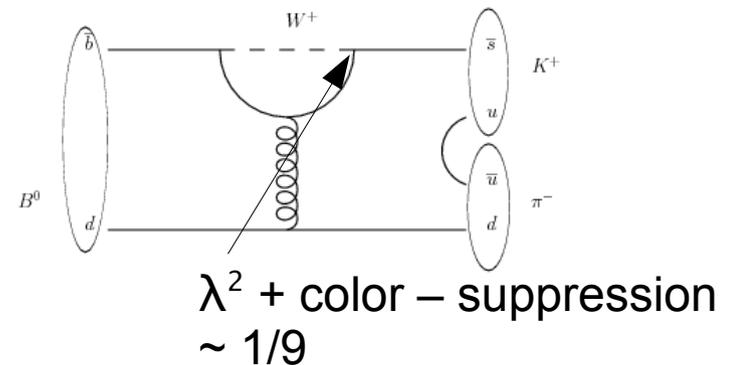
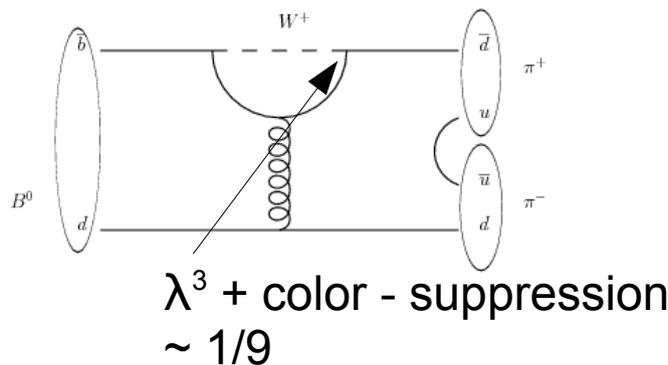


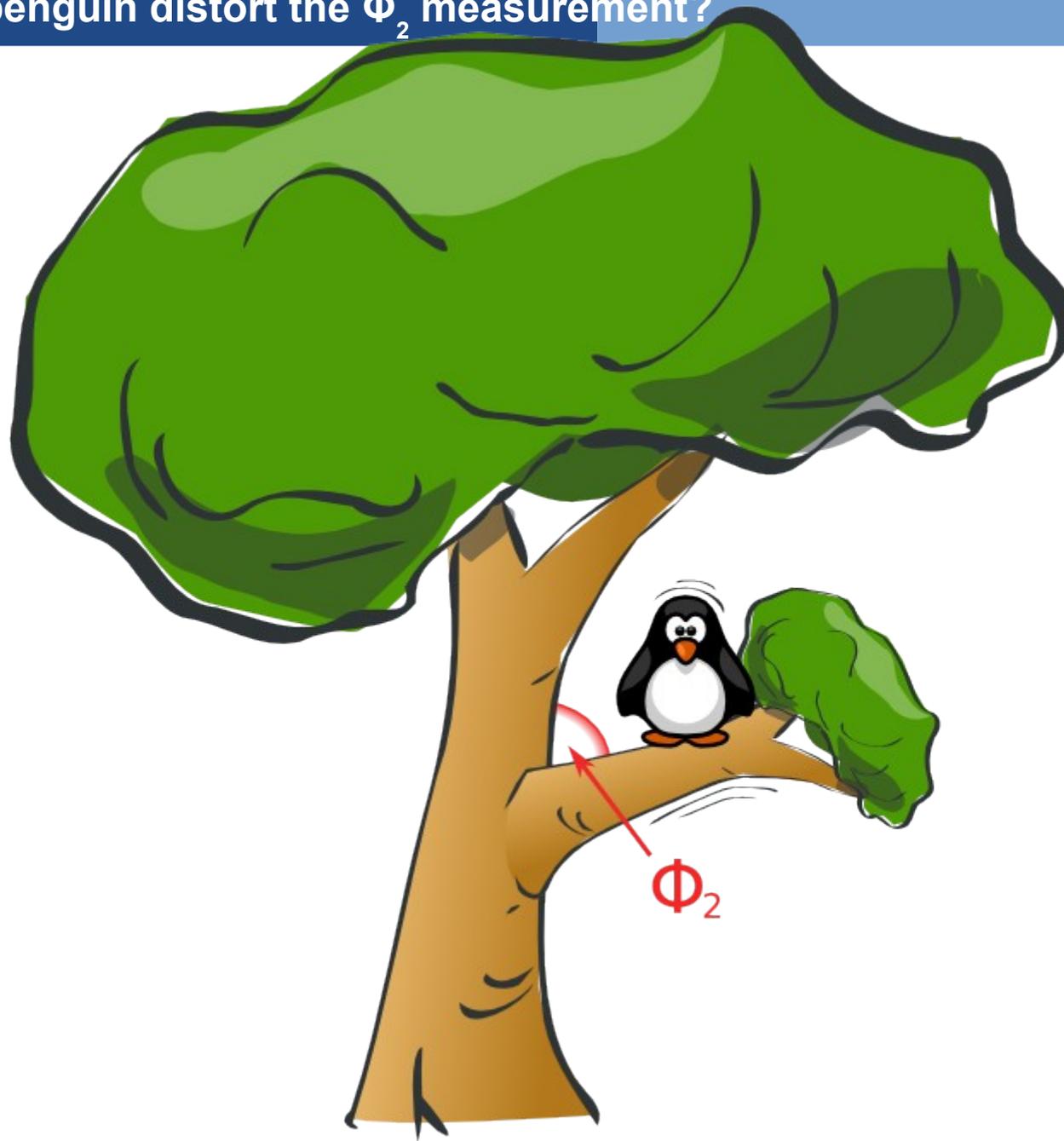
$$B^0 \rightarrow K^+ \pi^-$$

$$Br = 1.94 * 10e-5$$



Penguin diagram



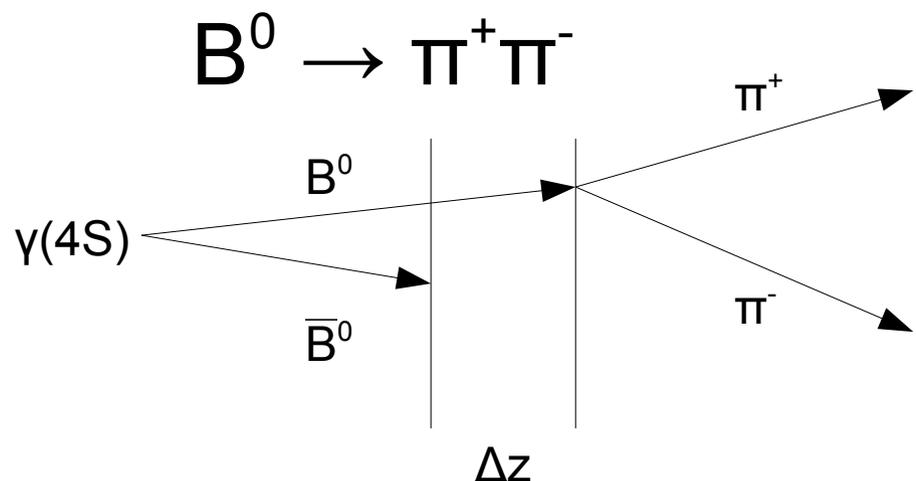


How the penguin distorts the tree level measurement

How to measure branching fractions for  $\Upsilon(4S) \rightarrow B^0 \bar{B}^0$  ?

$$B(B^0 \rightarrow f) = \frac{Y(B^0 \rightarrow f)}{N(BB) \epsilon(B^0 \rightarrow f)}$$

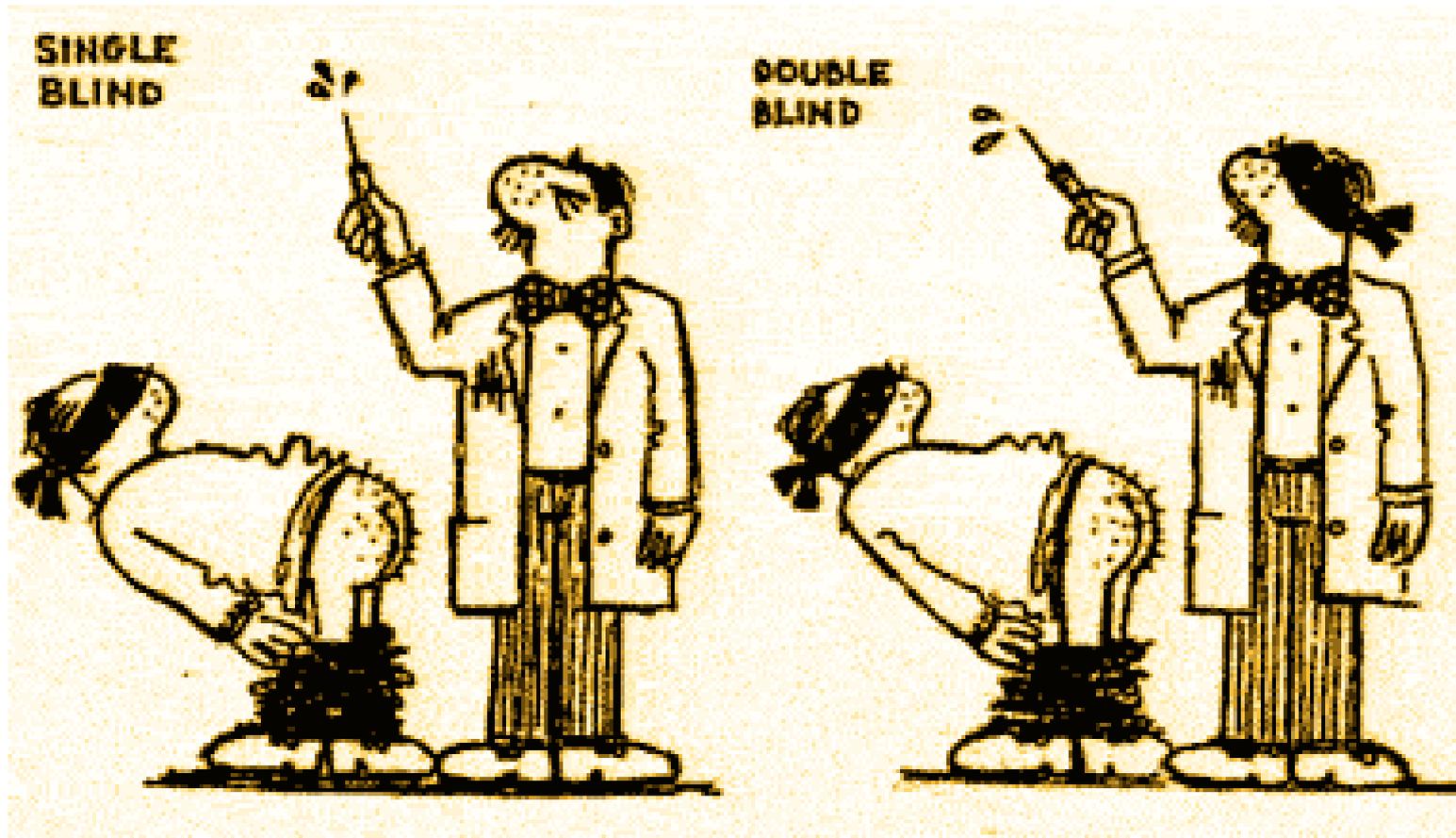
- Yield  $\longleftarrow$  Result of my analysis
- $N(BB)$   $\longleftarrow$  Proportional to Luminosity delivered from KEKB accelerator (removed QED, Bhabha, ... )
- Efficiency  $\longleftarrow$  Result of my analysis convoluted with detector effects



Topology similar to  $B^0 \rightarrow K^+ \pi^-$   
- it is a strong background  
- therefore we want to include it in the fit

- Belle recorded  $1 \text{ ab}^{-1}$  (nearly two times the BaBar dataset)
- About 770 mio.  $B\bar{B}$  pairs
- Final goal: measure CP violation and branching fractions in one fit

**Task: Find the  $\sim 2000$  events where  $B^0 \rightarrow \pi^+ \pi^-$   
out of 770 mio. other events**



## 1. “Blind”: **do not look at the data!**

Everything is done on Monte Carlo simulation before

This way you cannot fall in the trap of “producing” a signal

## 2. “Double”

- Cross check with other analysis
  - Independent reconstruction by Jeremy Dalseno
  - Independent reconstruction by me (Kolja)

→ Agreement within floating point accuracy!

- Reconstruct event
  - Find 4 vectors of the decay products
  - Combine 4 vectors to obtain mother particles

- Variables for event discrimination:

–  $\Delta E$  ~ reconstructed energy

–  $M_{bc}$  ~ reconstructed mass

$$\Delta E = E_B^{CMS} - E_{beam}^{CMS}$$

$$M_{bc} = \sqrt{(E_{beam}^{CMS})^2 - (p_B^{CMS})^2}$$

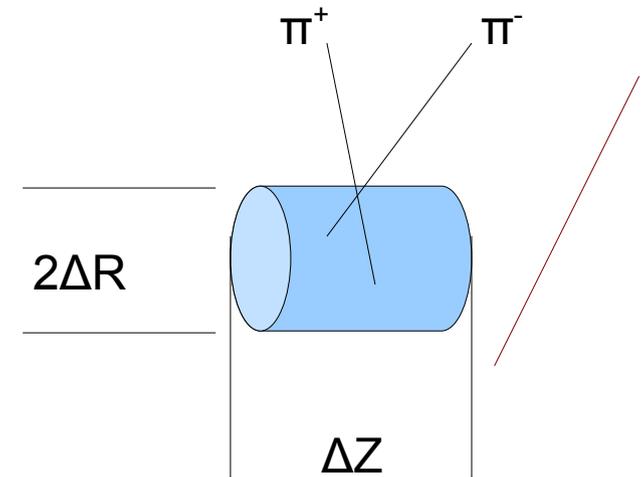
–  $L_{K/\pi}(\pi^+)$

–  $L_{K/\pi}(\pi^-)$

} Likelihood from the particle identification system (PID)

- tracks:

- veto electrons
  - veto protons
- } Cuts on PID Likelihood
- impact parameter cut (loose)
    - $\text{abs}(\Delta R) < 4 \text{ cm}$
    - $\text{abs}(\Delta Z) < 6 \text{ cm}$

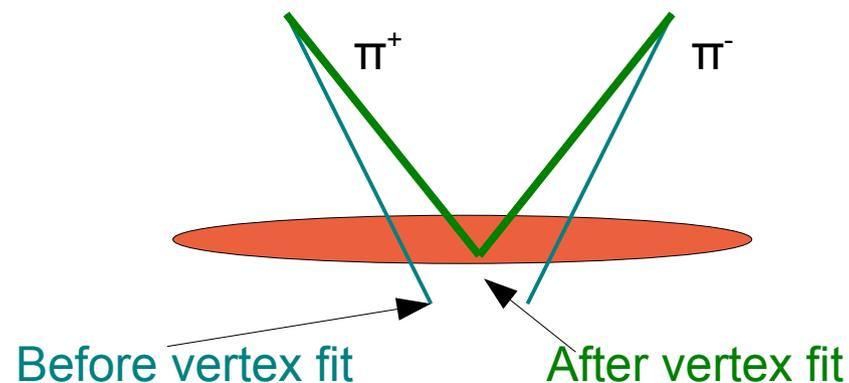


- vertexing:

- use only good tracks (needed to use resolution function)
  - $n_{\text{rphi\_hits}} \geq 1 \ \&\& \ n_{\text{Z\_hits}} \geq 2$
- beam tube constraint

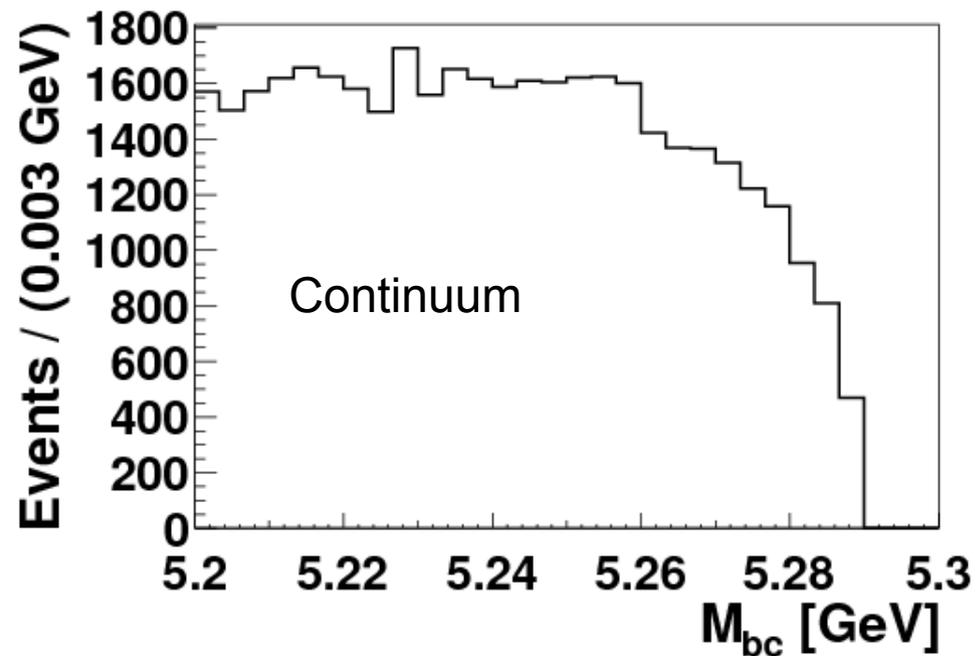
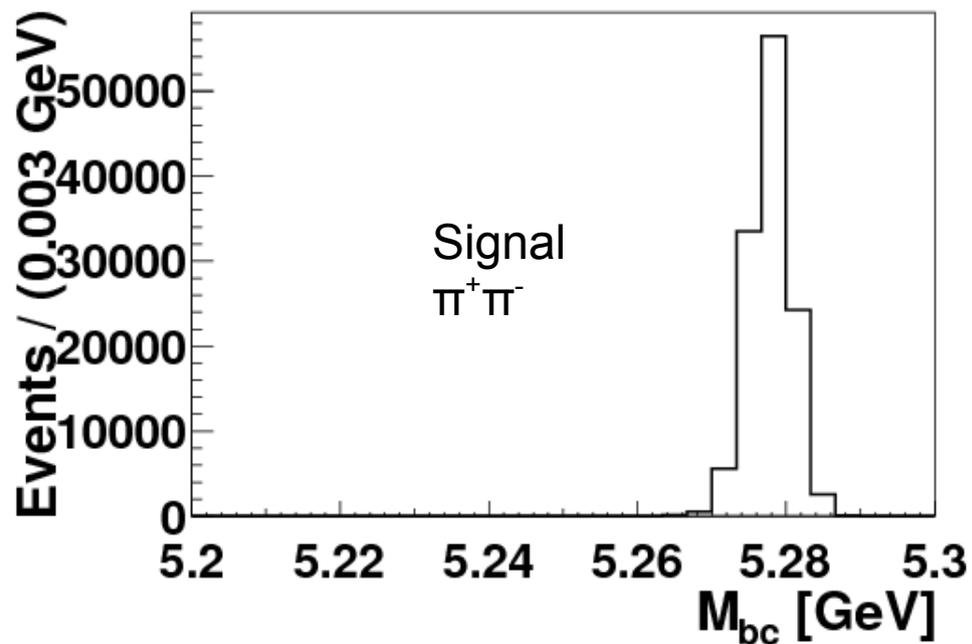
- define analysis window:

- $M_{bc} > 5.2$
- $-0.15 < \Delta E < 0.15$



Mbc = mass beam constraint

$$M_{bc} = \sqrt{(E_{beam}^{CMS})^2 - (p_B^{CMS})^2}$$

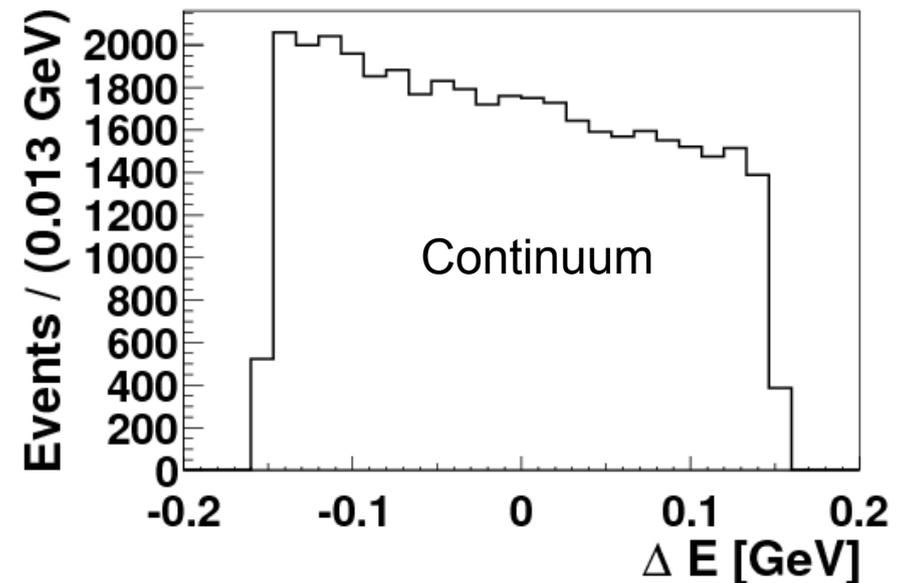
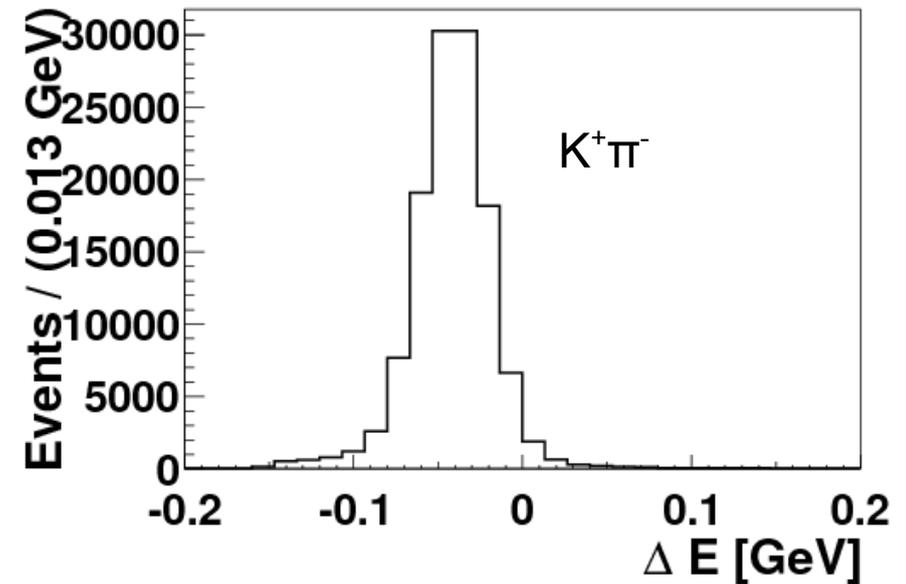
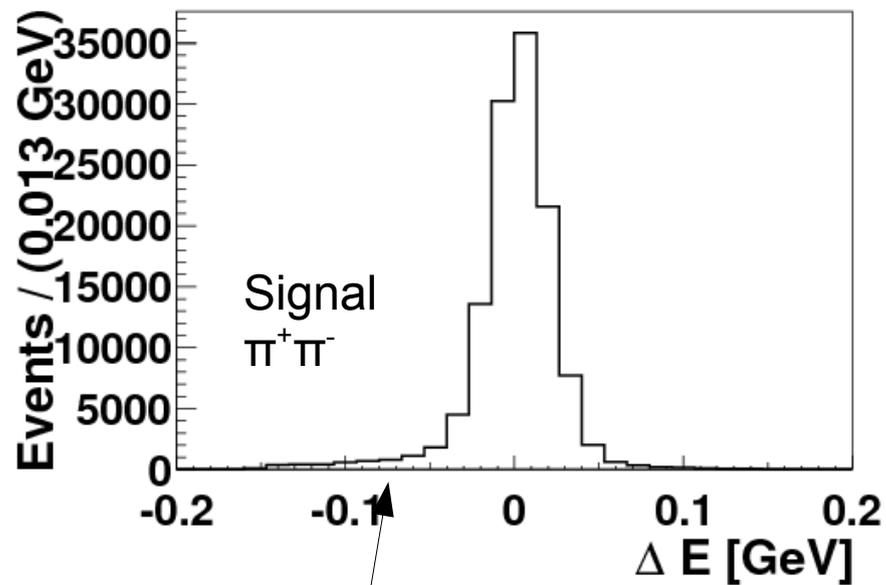


define analysis window:  $M_{bc} > 5.2$

the upper bound is given by the beam energy (event by event)

$$\Delta E = E_B^{CMS} - E_{beam}^{CMS}$$

$\Delta E =$  Energy difference between beam and reconstructed particles

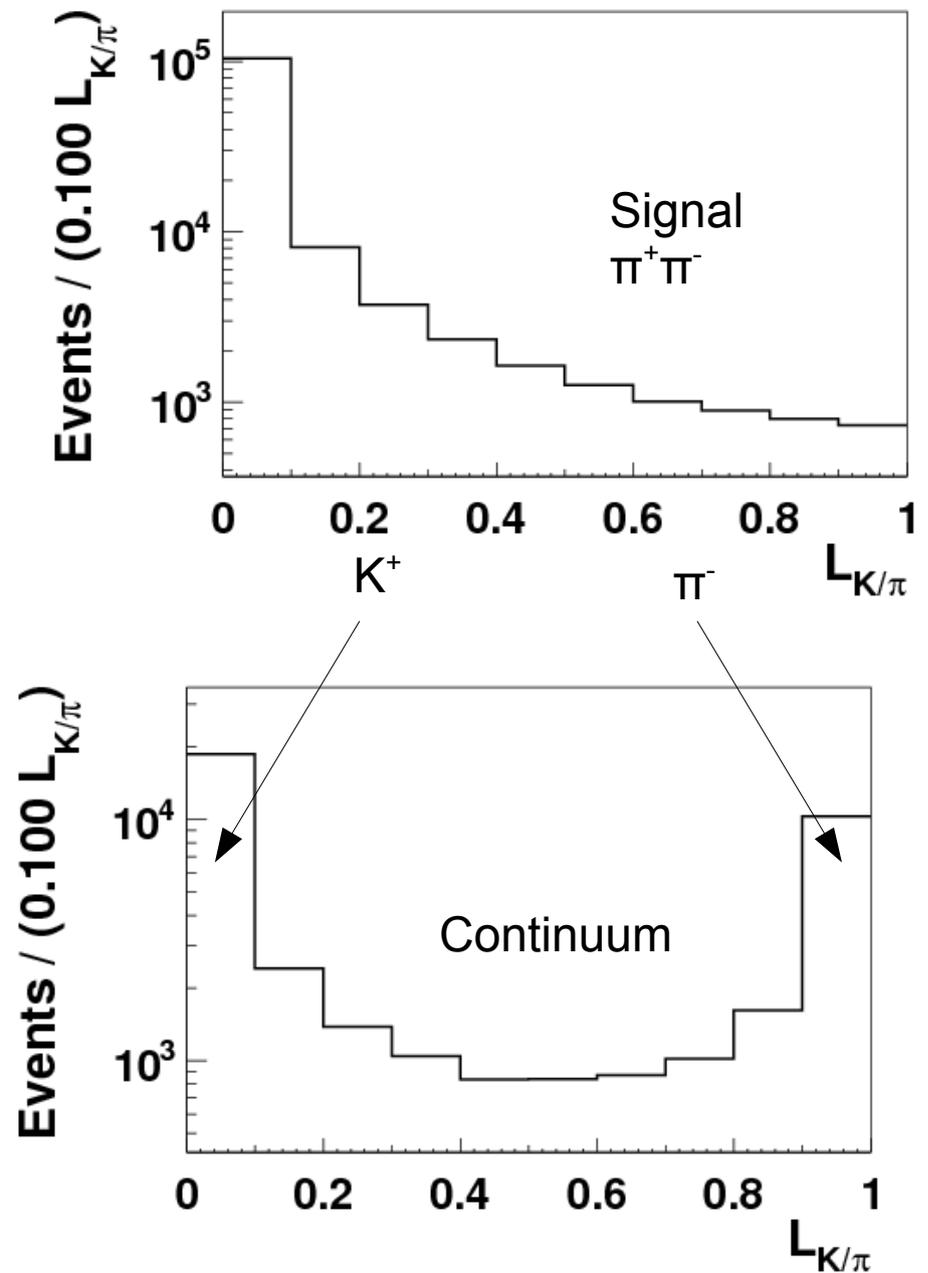
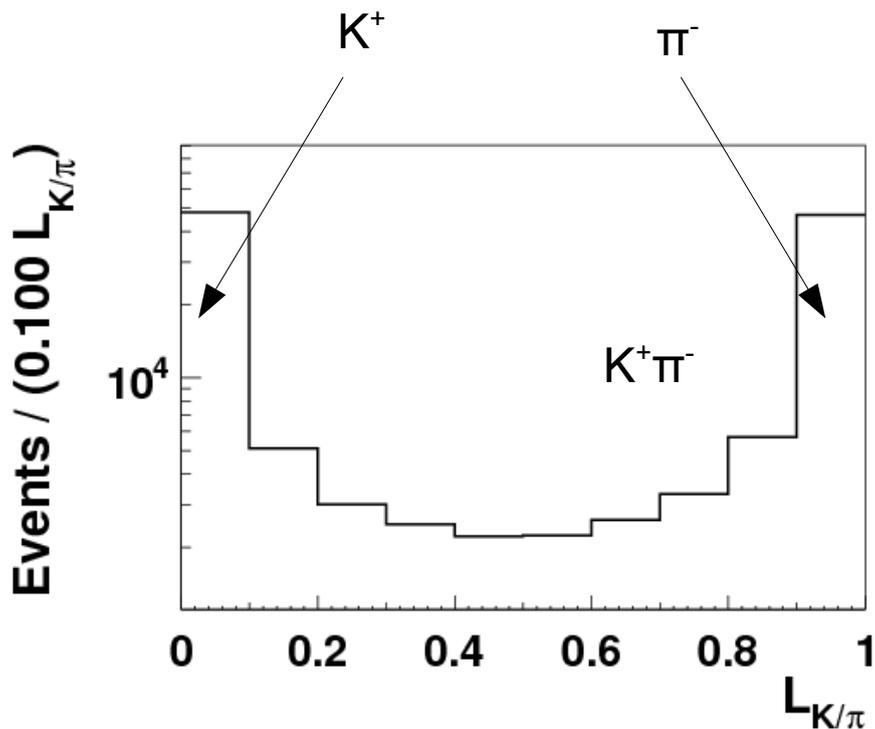


Effect of final state radiation: tails to lower energies

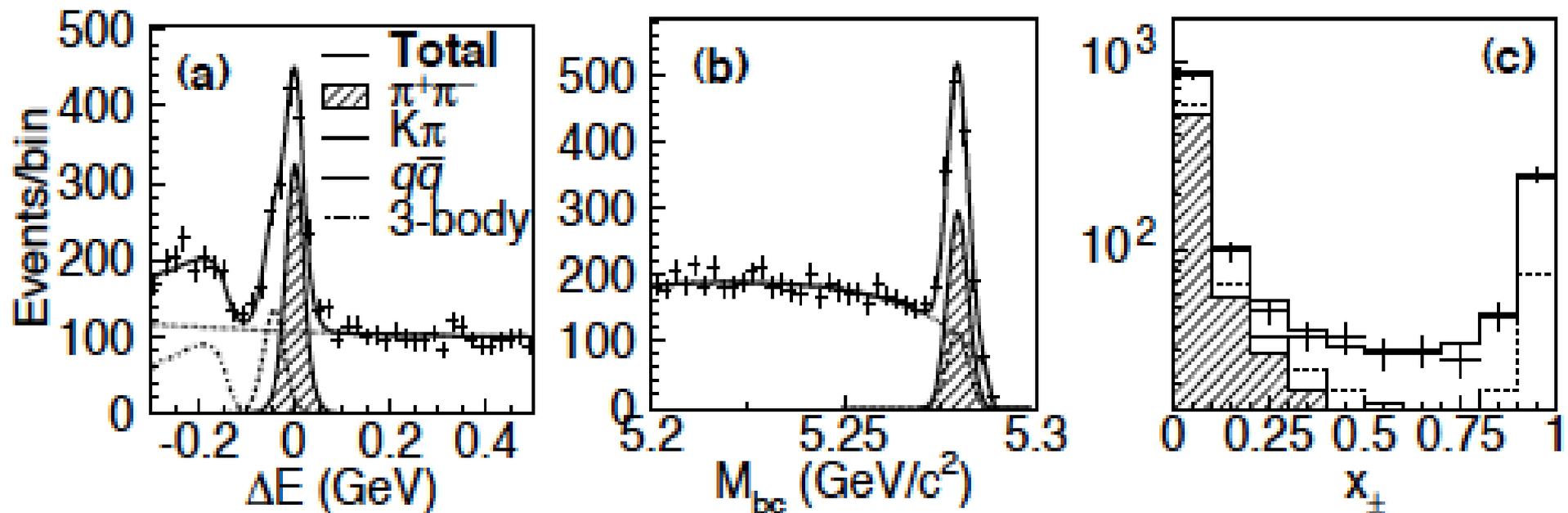
define analysis window:  $-0.15 < \Delta E < 0.15$

Likelihood to be a Kaon or Pion:

PID: Central drift chamber (CDC)  
Time of flight (TOF)  
Aerogel Cherenkov Counter (ACL)



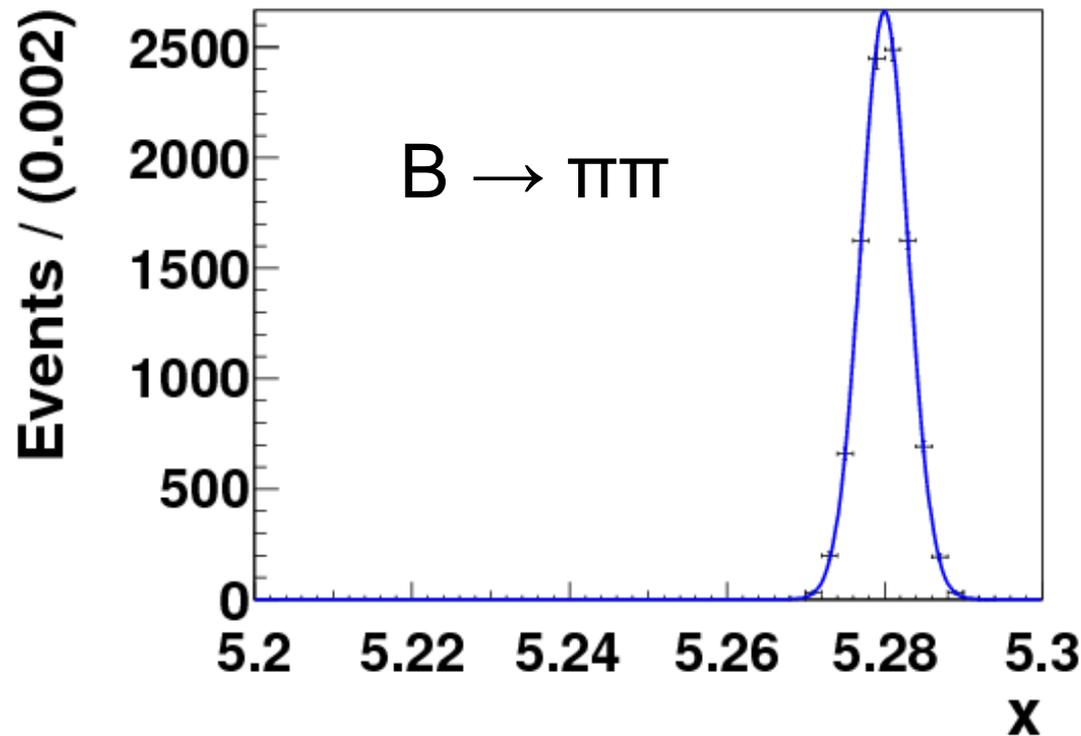
- Model Signal contributions for all possible sources
  - signal
  - generic decays ( $b \rightarrow c$ )
  - rare decays ( $b \rightarrow uds$ )
  - off-resonance (continuum  $q\bar{q}$ )
  - special contributions ( $K^+\pi^-$ )





## Fitting recipe

- Describe all variables by a function (or histogram)
  - PDF (Probability Density Function)
- Perform Unbinned Maximum Likelihood Fit
- Extracted the “Signal Shape”



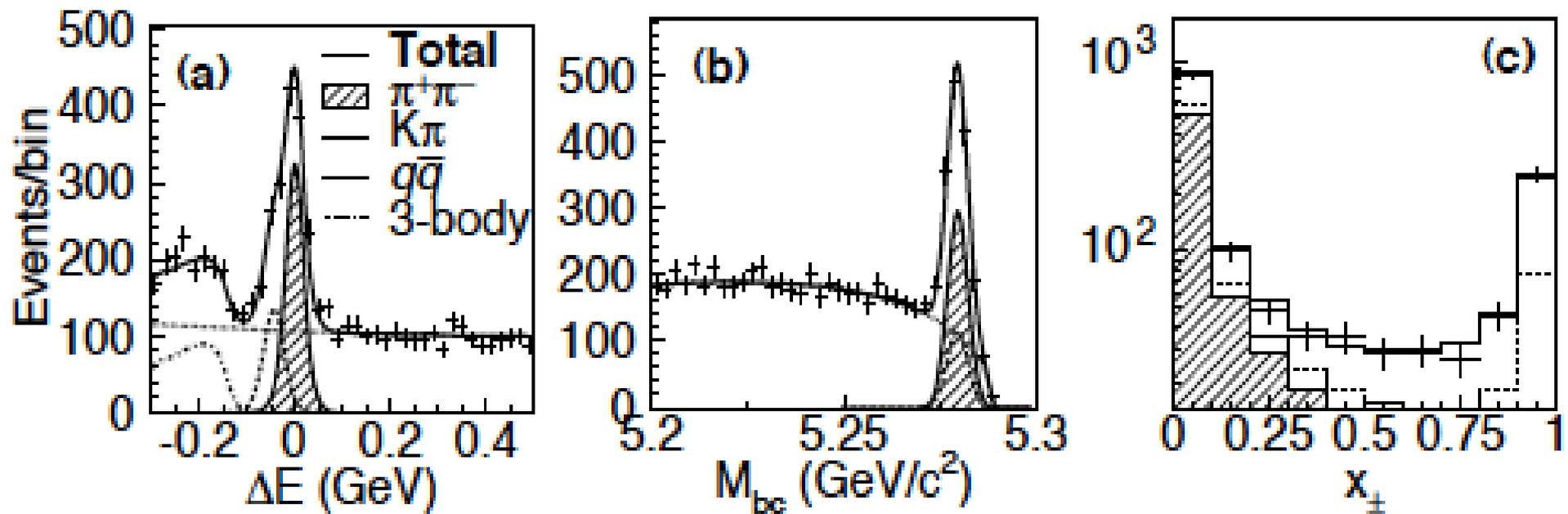
Variable:	Mbc
PDF:	Gaussian
Free parameters:	Mean Sigma

$$L = \prod_i P(M_{bc}^i; \vec{\Theta}) \quad \text{Maximize Likelihood for } \theta$$

$$-2 \log L = -2 \sum \log P(M_{bc}^i; \vec{\Theta}) \quad \text{Minimize } -2 \log L \text{ with software Minuit2}$$

- Fix the signal shapes
- let the ratios between the different components float in the fit

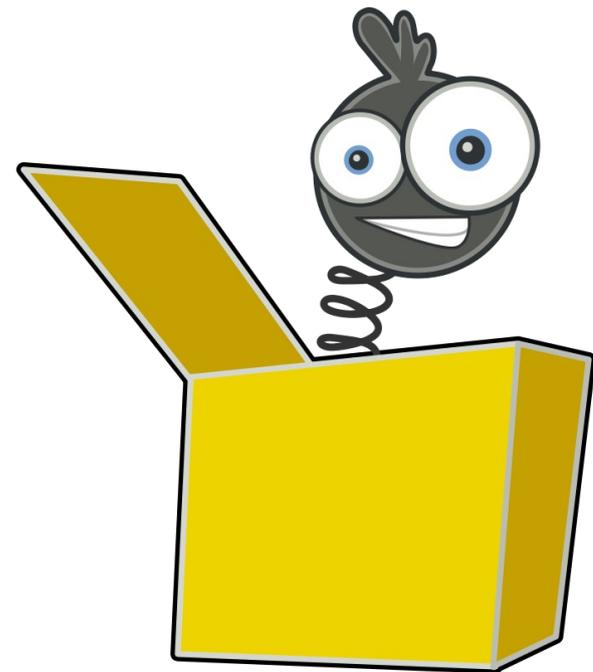
$$Yield = \frac{\text{signal events}}{\text{total number of events}}$$



- Toy experiments to find out correlations and biases
  - Vary the expected branching ratio in the simulation
    - Get systematic uncertainty
  - Check if you obtain the simulated branching ratio

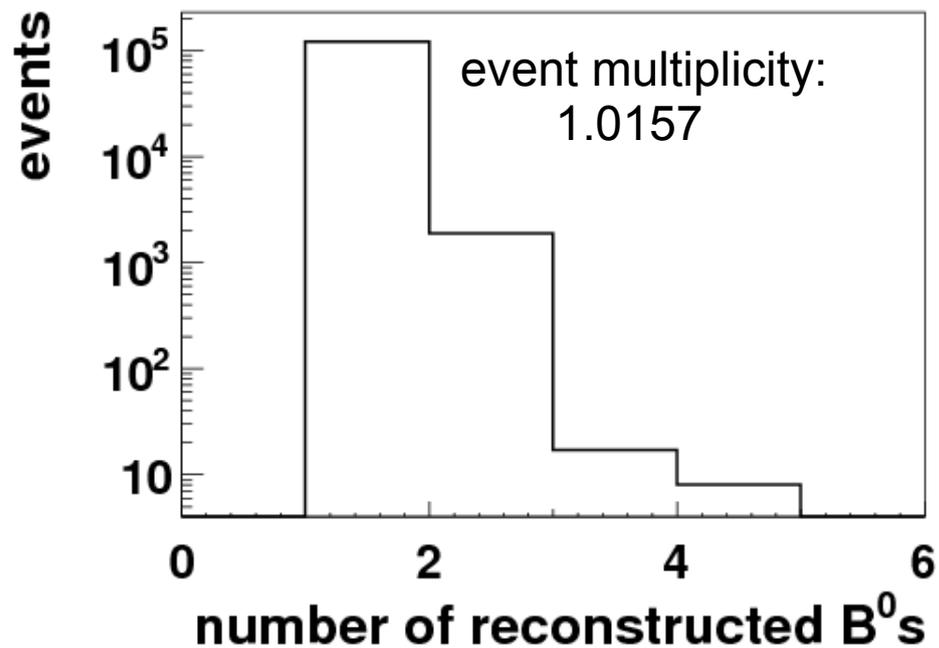
Open the box!

(Run your reconstruction and fitting on the real data)

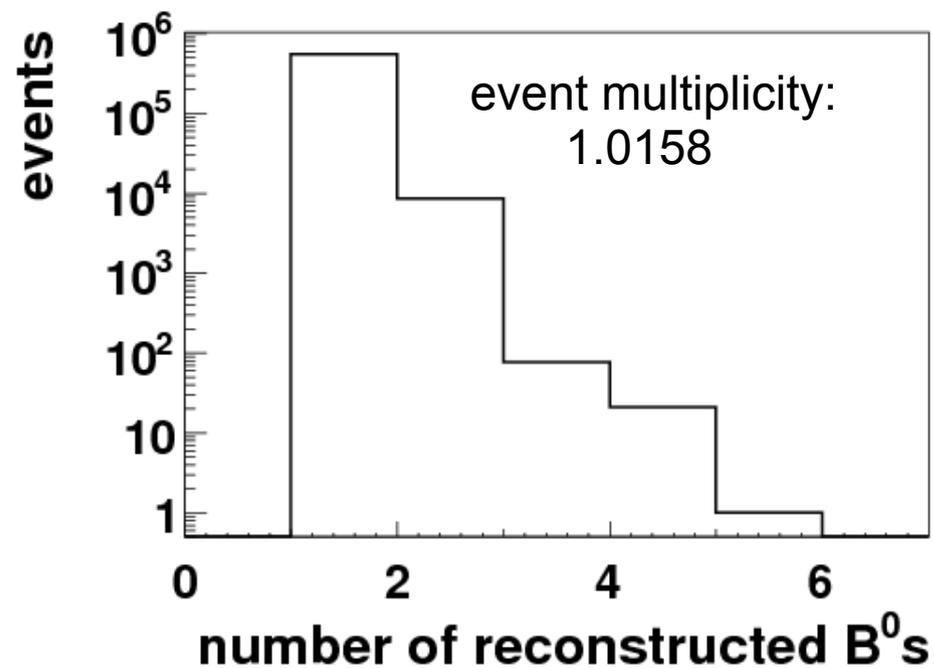


# Backup

SVD1



SVD2



Best B selection → take first from list

reconstruction efficiency:  
66.7 %

reconstruction efficiency:  
73.6 %

How to measure branching fractions for ?

Assuming no direct CP-violation:

$$B(Y \rightarrow f) = \frac{N(B^0 \rightarrow f) + N(\bar{B}^0 \rightarrow \bar{f})}{N(B^0) + N(\bar{B}^0)}$$

$B^0$  and  $\bar{B}^0$  are produced in pairs:

$$B(Y \rightarrow f) = \frac{N(B^0 \rightarrow f) + N(\bar{B}^0 \rightarrow \bar{f})}{2N(B^0 \bar{B}^0)}$$

Assuming equal production of  $B^0 \bar{B}^0$  and  $B^+ B^-$

$$B(Y \rightarrow f) = \frac{N(B^0 \rightarrow f) + N(\bar{B}^0 \rightarrow \bar{f})}{2 \cdot \frac{1}{2} \cdot N(B^0 \bar{B}^0)} = \frac{N(B^0 \rightarrow f) + N(\bar{B}^0 \rightarrow \bar{f})}{N(B \bar{B})}$$

Not all  $B^0 \rightarrow f$  can be observed  $\rightarrow$  introduce Signal Yield (Number of detected events)

$$Y(B^0 \rightarrow f) = \epsilon(B^0 \rightarrow f) (N(B^0 \rightarrow f) + N(\bar{B}^0 \rightarrow \bar{f}))$$

  
 Efficiency

$$H_l = \sum_{i,j} \frac{|p_i||p_j|}{E_{vis}^2} P_l(\cos(\theta_{ij}))$$

$\theta_{ij}$ : angle between hadron i and j

$E_{vis}$ : total visible energy

Normalization:

$$H_{0l} = H_l / H_0$$

→ balanced Momentum:

$$H_1 = 0$$

→ 2 jet events:

$$H_l = 1 \quad \text{for } l \text{ is even}$$

Legendre polynomials:

$$P_0(x) = 1,$$

$$P_1(x) = x,$$

$$P_2(x) = \frac{1}{2}(3x^2 - 1),$$

$$P_3(x) = \frac{1}{2}(5x^3 - 3x),$$

$$P_4(x) = \frac{1}{8}(35x^4 - 30x^2 + 3),$$

- Other idea: thrust and sphericity (minimization procedure)  
80s (PEP/PETRA)
- Belle`s invention: **Super-Fox-Wolfram-Moments** (R. Enomoto)
- Combine merits of FW moments and thrust axis
  - separate tracks of Signal B and other B

$$R_i^{s0} = \sum_{j,k} |p_j||p_k| P_i(\cos \theta_{jk}) / \sum_{j,k} |p_j||p_k|, \left[ \begin{array}{l} j \text{ over signal B tracks} \\ k \text{ over other B tracks} \end{array} \right]$$

$$R_i^{00} = \sum_{j,k} |p_j||p_k| P_i(\cos \theta_{jk}) / \sum_{j,k} |p_j||p_k|, (j, k \text{ over other B tracks})$$

- H. Kakuno: **Kakuno-Super-Fox-Wolfram** Moments (KSFW)

→ include charge

$$H_i = \sum_{j,k} Q_j Q_k |p_k| P_i(\cos(\theta_{ij}))$$

→ separate treatment of charged and neutral tracks

- g=0: only charged tracks of other B
- g=1: only photons of other B
- g=2: only missing momentum of other B

→ include missing momentum vector

- $E_T$ : scalar sum of transverse energy
- $MM^2$ : missing momentum squared

- 17 variables

$E_T$  — scalar sum of transverse energy of all particles

$H_i^{SO,c}$  (for  $i=0$  to 4), using only charged tracks of other B

$H_i^{SO,n}$  (for  $i=0,2,4$ ), using only photons of other B

$H_i^{SO,\nu}$  (for  $i=0,2,4$ ), using only missing momentum of other B

$H_i^{OO}$  (for  $i=0$  to 4)

- Strong dependence on  $MM^2$ , separate Fisher in 7  $MM^2$  bins

$$F = \sum_{i=1}^N \alpha_i x_i$$

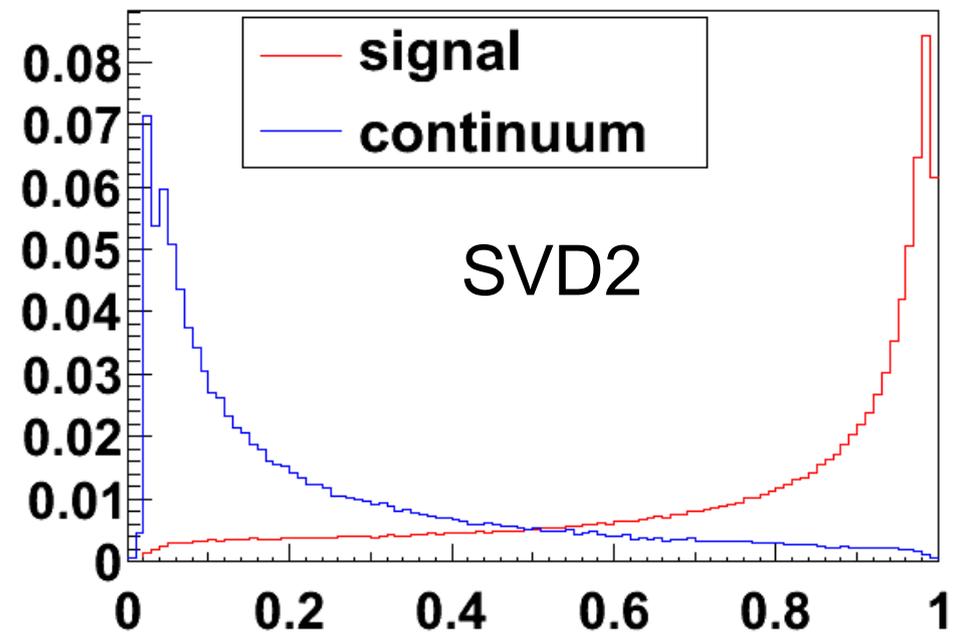
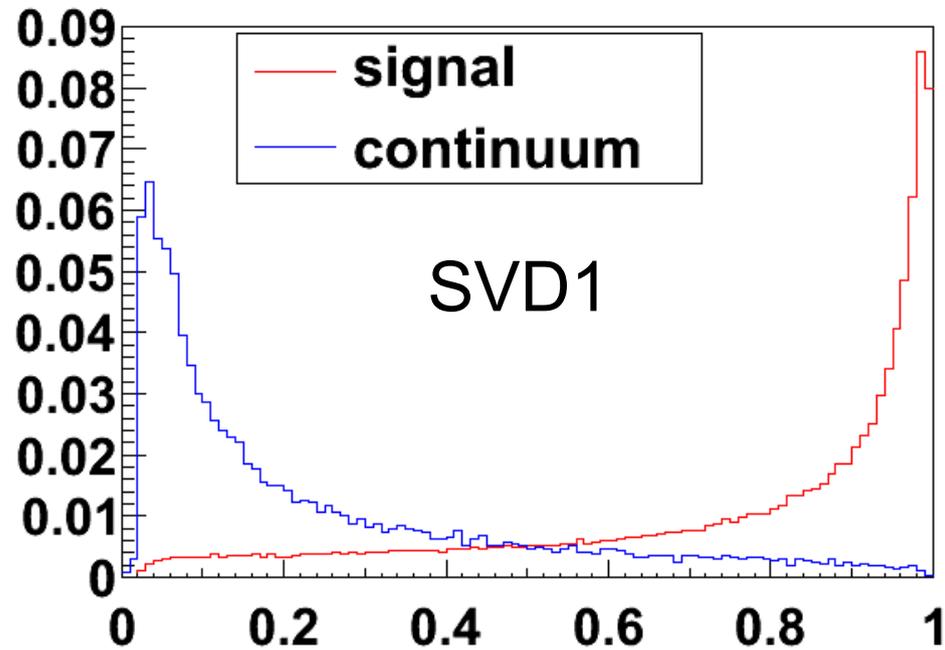
- Find the set of  $\alpha_i$  where the distance between the signal and the background events is maximal
- Procedure (including matrix inversion):

$$\alpha_i = \sum_{j=1}^N (U_{ij}^b + U_{ij}^s)^{-1} (\mu_j^b - \mu_j^s)$$

$\mu_{ij}^{b,s}$  : mean for background, signal

$U_{ij}^{b,s}$  : covariant matrix background, signal

introduced by CLEO (PRD53, 1039(1996)) implemented in brutus\_f at Belle



- Signal and Continuum are normalized
- Full off-resonance dataset used
- very smooth curves

SVD1

SVD2

