

# *8.882 LHC Physics*

*Experimental Methods and Measurements*

*Search Strategies and Observations*

*[Lecture 14, March 30, 2009]*

# *Organizational Issues*

## Project 2

- due **April 9**

Project 3 is coming up quickly afterwards

- due **May 2**

Guest lecture planned on:

- Bayesian versus frequentistic approach to statistics
- Michael Betancourt

# Lecture Outline

## Search Strategies and Observations

- introduction
- general methods
- analysis biases and how to avoid them
- two examples
  - resonance searches:  $D^0 \rightarrow \mu^+ \mu^-$
  - search for an oscillation frequency:  $B_s$  oscillations

# *Introduction*

## Different types of analysis in particle physics

- measurement
  - yields a number with an uncertainty for a given observable
  - that number should say something about the Standard Model (SM)
  - at best: could be inconsistent with the SM: **something is wrong**
  - at least: **knowledge of SM improved**, better test: *bread and butter*
- **search**
  - assume some physics model (usually new physics or even SM)
  - determine some characteristic observable to verify the model
  - analyze data concerning observable: is model supported by data?
    - yes: a signal was found, turn the search into a measurement
    - no: search yields a limit with a confidence level
  - at best: **find physics beyond the Standard Model**
  - at least: set a new limit for a given physics model

# *Typical Design of a Search*

## Physics assumption

- search for SM Higgs at mass 115 GeV
- search for  $Z'$  (mass larger than  $Z$  mass)
- new physics (NP) appears in tail of high transverse momenta

## Choice of observable

- SM Higgs: best channel  $H \rightarrow \gamma\gamma$ , observable  $m_{\gamma\gamma}$
- $Z'$ : prime decay channel  $Z' \rightarrow \mu^+\mu^-$ , observable  $m_{\mu\mu}$
- NP: all final states with identifiable objects ( $j$  – jet)
  - $\mu\mu, \mu\mu\mu, \mu e, \mu\mu e, ee, eee, ee\mu, jj, jjj, \dots$
  - use events falling into the high  $p_T$  portion

# *Typical Design of a Search*

## Optimization with Monte Carlo / data

- find quantity  $Q_s$  which defines sensitivity of analysis
  - classical optimization quantities:  $S/\sqrt{B}$  or  $S/\sqrt{S+B}$
  - $S/\sqrt{B}$  used for search without knowledge of cross section
    - careful, unnatural behavior at low  $S$
  - $S/\sqrt{S+B}$  used when signal cross section is known
    - careful, it really optimizes the measurement if search successful
  - improved quantity:  $S/(1.5 + \sqrt{B})$
  - see [www.cmsaf.mit.edu/twiki/bin/view/Class8882/WebHome](http://www.cmsaf.mit.edu/twiki/bin/view/Class8882/WebHome)
- find optimal set of cuts maximizing  $Q_s$

## Include systematic uncertainties

- in many cases the effect is small
- for large uncertainties statistical methods complicated: see [www.cmsaf.mit.edu/twiki/bin/view/Class8882/WebHome](http://www.cmsaf.mit.edu/twiki/bin/view/Class8882/WebHome)

# *Typical Design of a Search*

Look at fully optimized data analysis

- find a signal: determine significance
  - method reasonably straight forward
  - see examples in what follows
- conventions
  - observation or discovery at **5 standard deviations**
  - evidence is to mark the turf, at about **3 standard deviations**
  - evidence became more popular since experiments take so long to accumulate statistics to have an observation
- set a limit at a given confidence level (usually 95%)
  - once sensitivity is known confidence level straight forward
  - usually implies Poisson statistics: few events are found
  - see examples in what follows

# Clever Hans Effect

## Astonishing horse (real story)

- Hans von Osten in early 1900 could do math!
- Hans added up numbers, result communicated in number of pawns on the ground
- trainer cues? no, trainer was send outside, hmmm?
- 1907 Oskar Pfungst proposed: *result should be unknown to people in the room*: Hans lost all ability to add numbers
- horse sensed cues from people who knew the answer
- really a clever horse
- imagine how 'clever' we could be ....





# *The Hawthorne Experiment\**

Study of illumination\*\* (optimal value for productivity)

- Study 1a: In the first experiment, there was no control group. The researchers experimented on three different departments; all showed an increase of productivity, whether illumination increased or decreased.
- Study 1b: A control group had no change in lighting, while the experimental group got a sequence of increasing light levels. Both groups substantially increased production, and there was no difference between the groups. This naturally piqued the researchers' curiosity.

\* Hawthorne experiments have been intensely discussed. Read up on the web if you like.

\*\* source is wikipedia

# *The Hawthorne Experiment*

## Study of illumination (optimal value for productivity)

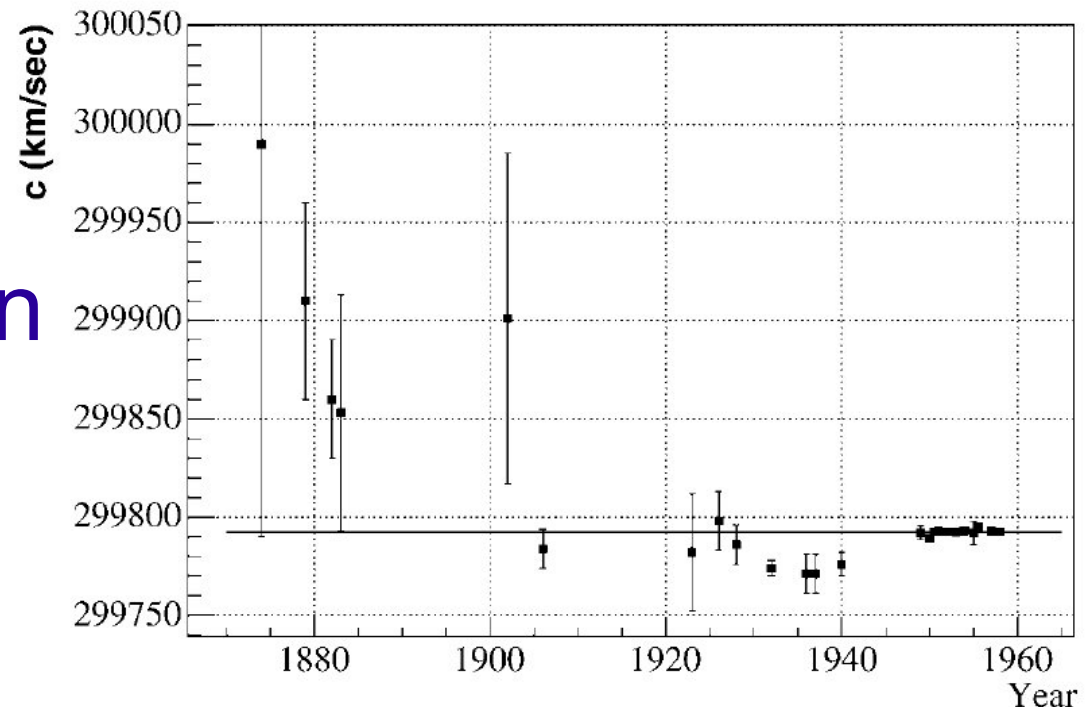
- Study 1c: The researchers decided to see what would happen if they decreased lighting. The control group got stable illumination; the other got a sequence of decreasing levels. **Surprisingly, both groups steadily increased production** until finally the light in experimental group got so low that they protested and production fell off.
- Study 1d: This was conducted on two women only. Their production stayed constant under widely varying light levels. It was found that if the experimenter said bright was good, they said **they preferred the light; the brighter they believed it to be, the more they liked it.** The same was true when he said dimmer was good. **If they were deceived about a change, they said they preferred it.** Researchers concluded that their preference on lighting level was completely subjective - if they were told it was good, they believed it was good and preferred it, and vice versa.

# Effects in Physics

## Measurements of the speed of light\*

- measurements around 1930-40 are clearly off
- investigating these measurements it was concluded:
  - the investigator searches for the source or sources of errors, and continues to search until he gets a result close to the accepted value. *Then he stops!*

Particle physics, even recently has many more examples



\* from <http://arjournals.annualreviews.org/doi/pdf/10.1146/annurev.nucl.55.090704.151521>

## *Feynman's Short Version*

“The first principle is that you must not fool yourself – and you are the easiest person to fool.”

So, what can we do?

# *Blind Analysis ....*

## Dangerous traps in searches

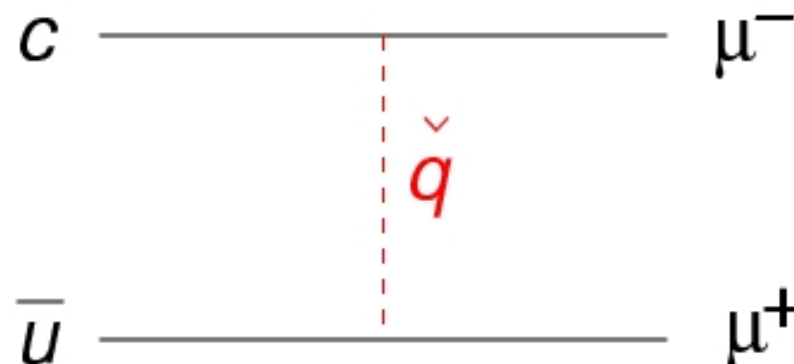
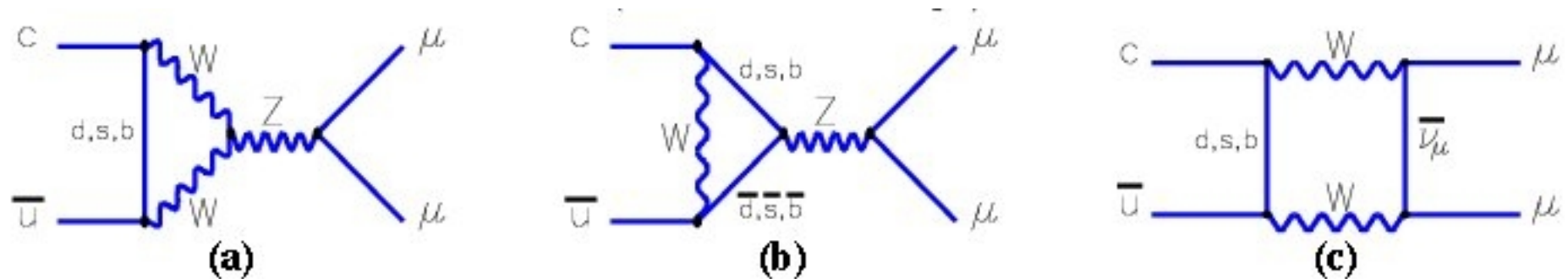
- optimizing the analysis looking at data in signal region
  - small number of events, makes one vulnerable to statistical fluctuations: many examples in history
  - **bias** introduced through optimization procedure and human being
  - simple way out, do not even look at data: jargon **blind analysis**
  - device strategy where data in signal area are not used
  - usually use: data for background area and MC for signal area



# Search for Rare Decay: $D^0 \rightarrow \mu^+ \mu^-$

## Physics interest\* (CDF analysis)

- reaction is a flavor changing neutral current reaction
- highly suppressed in Standard Model: rate exp.  $10^{-13}$
- SU(per)SY(mmetric) models can accommodate large branching ratio

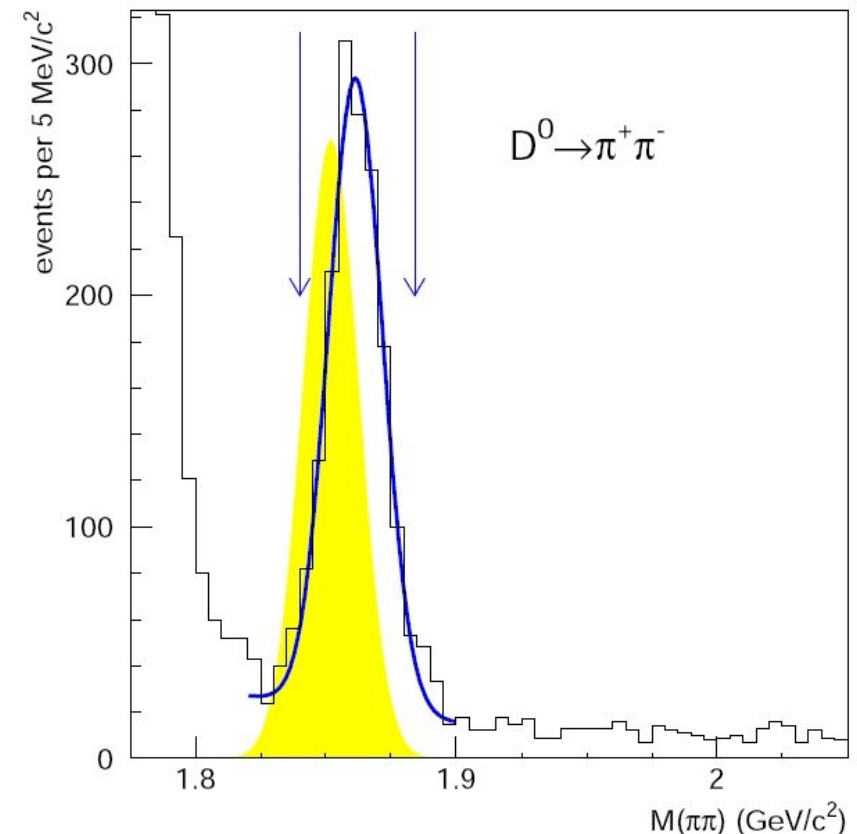
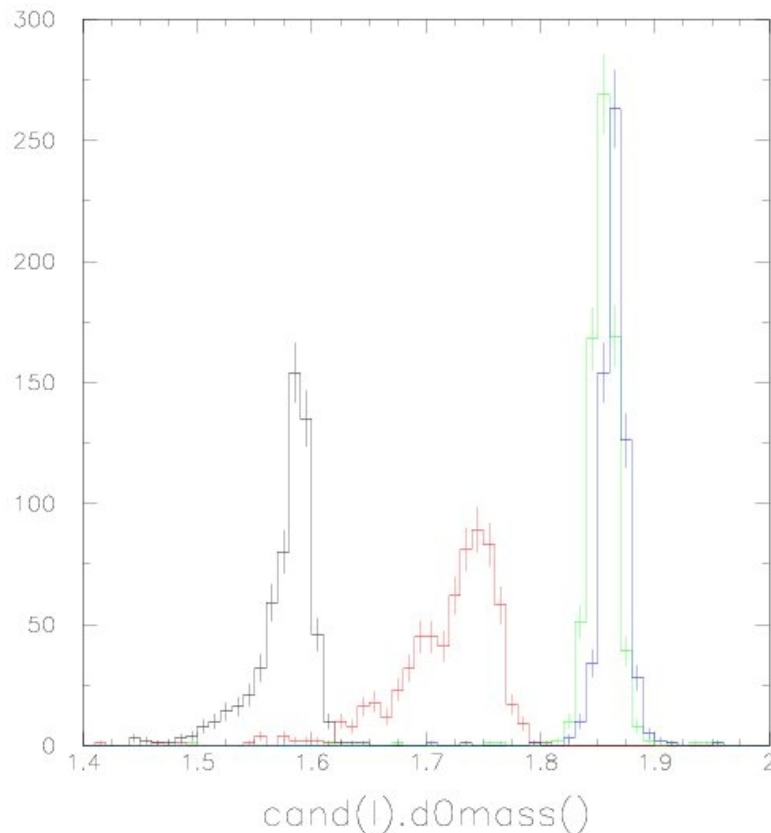


\* from CDF note 6273

# Search for Rare Decay: $D^0 \rightarrow \mu^+ \mu^-$

## Analysis outline

- use copiously produced:  $D^0 \rightarrow K^- \pi^+$  and  $D^0 \rightarrow \pi^- \pi^+$
- determine: acceptance, background, signal normalization
- signal looks like  $D^0 \rightarrow \pi^- \pi^+$  with two muon stubs attached



# Search for Rare Decay: $D^0 \rightarrow \mu^+ \mu^-$

## Analysis outline, continued

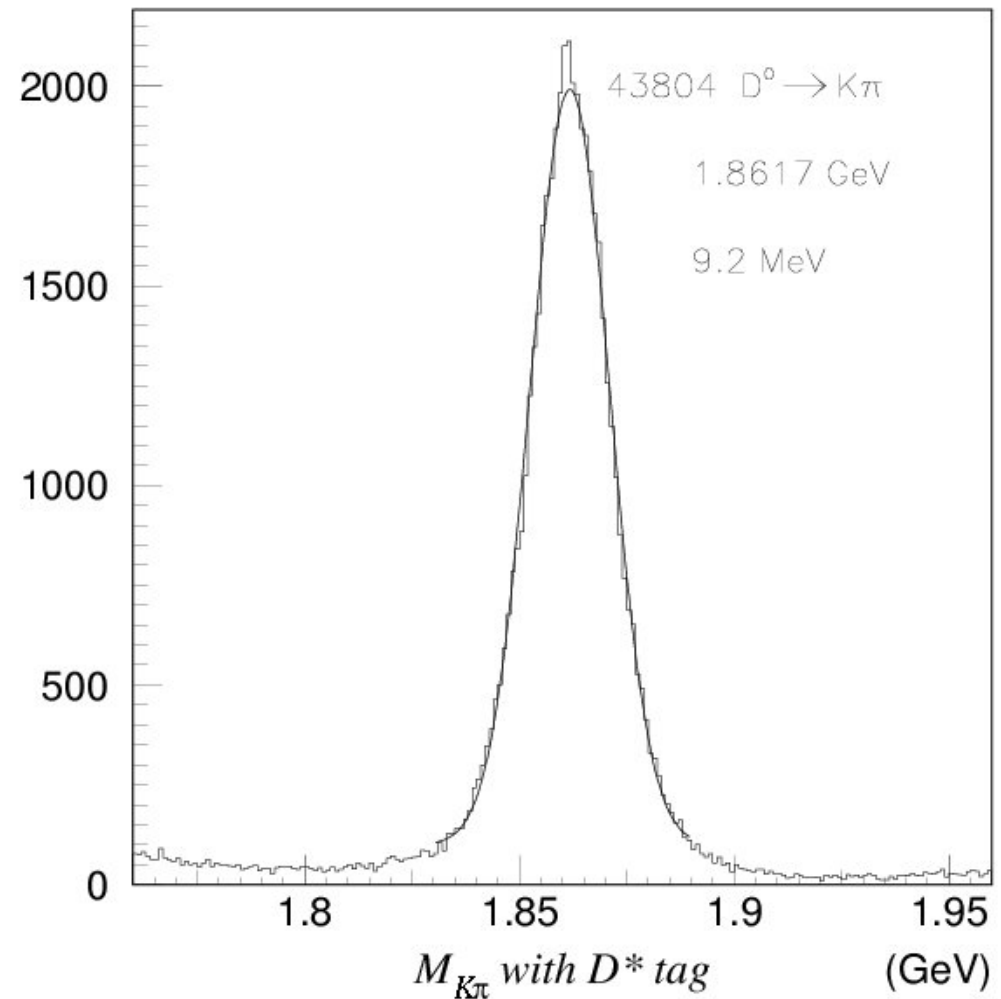
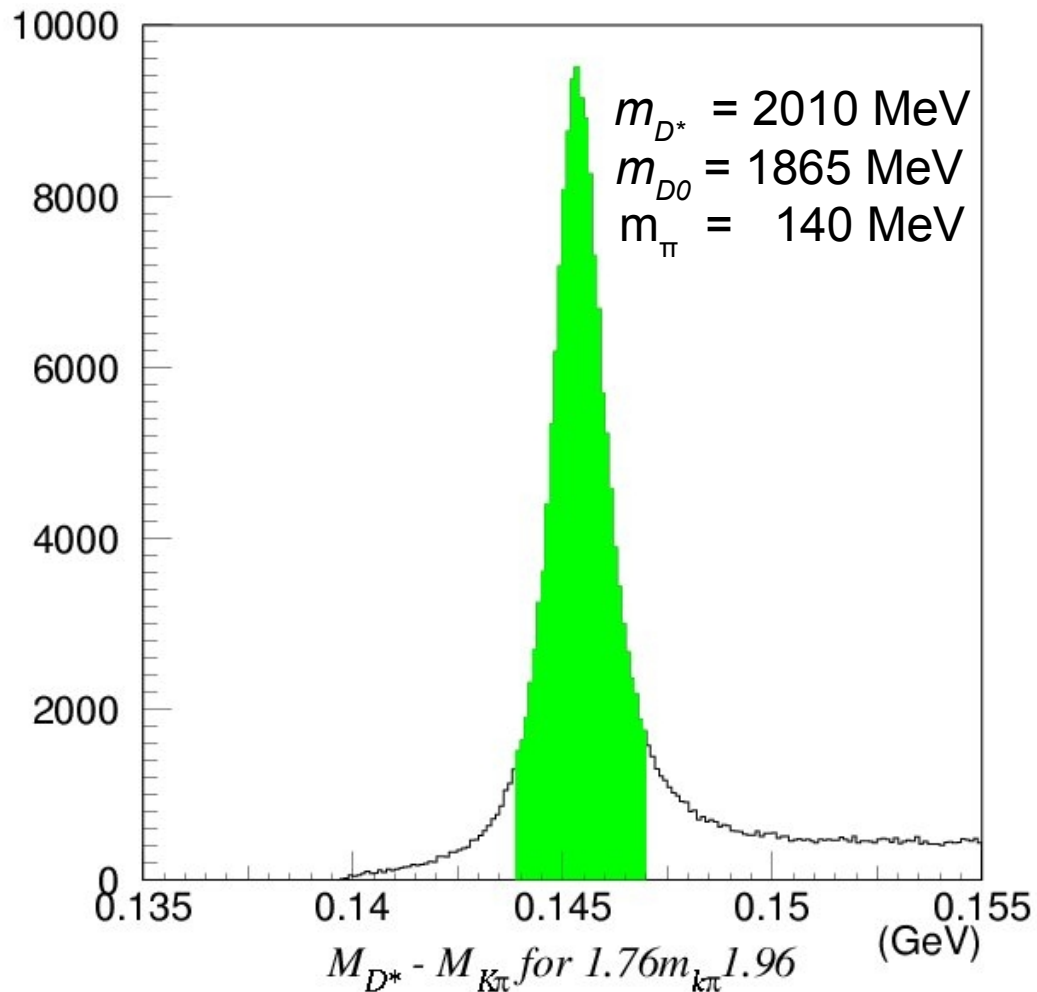
- branching ratio:  $Br(D^0 \rightarrow \mu^+ \mu^-) = Br(D^0 \rightarrow \pi^+ \pi^-) \frac{N(\mu\mu) \varepsilon(\pi\pi)}{N(\pi\pi) \varepsilon(\mu\mu)}$
- limit:  $Br_{95\%CL}(D^0 \rightarrow \mu^+ \mu^-) = Br(D^0 \rightarrow \pi^+ \pi^-) \frac{N(\mu\mu)_{95\%CL} \varepsilon(\pi\pi)}{N(\pi\pi) \varepsilon(\mu\mu)}$
- $N(\mu\mu)_{95\%CL}$  is upper limit of how many events could be there according to accumulated statistics we have and given number of events observed (covers 95% of cases)
  - often selection completely erases candidates, and mostly there are very few events
  - non-Gaussian statistics has to be applied, *Poisson* statistics:  $\text{TMath::Poisson}(n_{\text{exp}}, n_{\text{obs}})$  gives the probability to observe  $n_{\text{obs}}$  events when on average  $n_{\text{exp}}$  are expected
  - here  $N(\mu\mu)_{95\%CL}$  is 3 events when zero events are observed (from  $\text{TMath::Poisson}(0, 3) = 5\%$ )



# Search for Rare Decay: $D^0 \rightarrow \mu^+ \mu^-$

## Analysis outline, *continued*

- the  $D^*$  trick:  $D^{*+} \rightarrow D^0 \pi^+$  with very soft  $\pi^+$
- very clean  $D^0$  sample for studies (removes auto reflection)



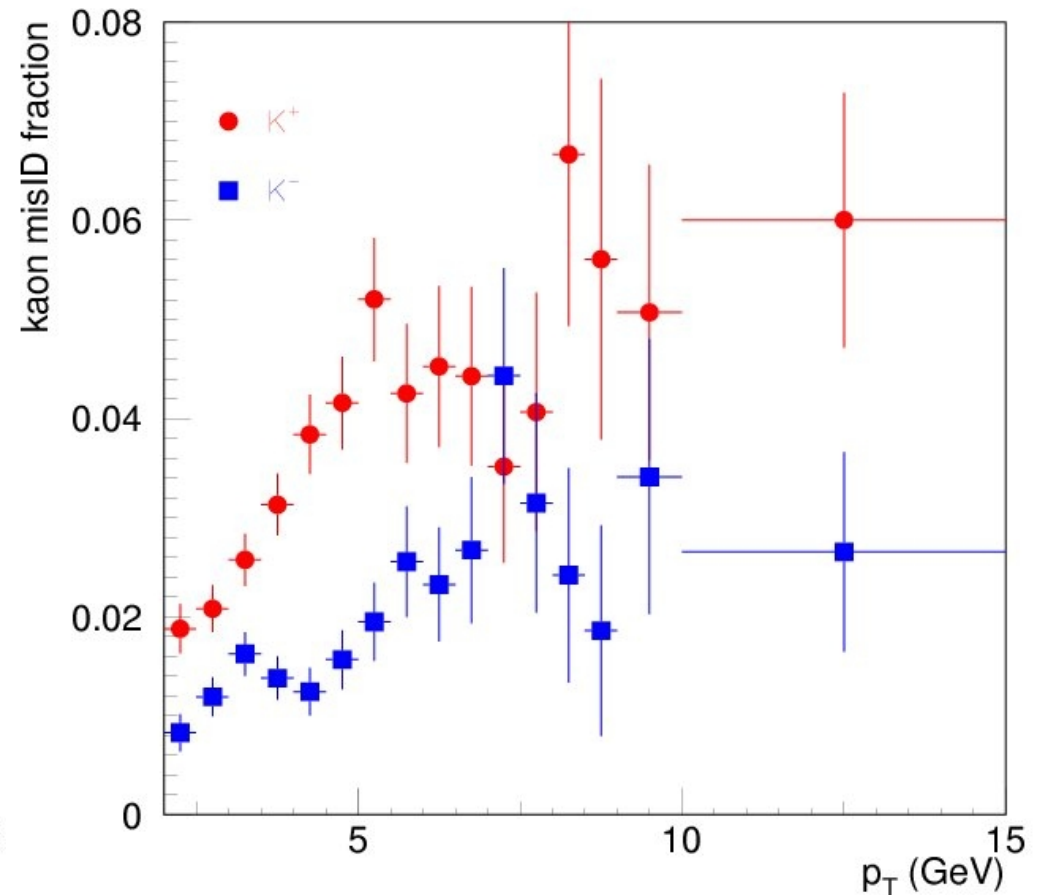
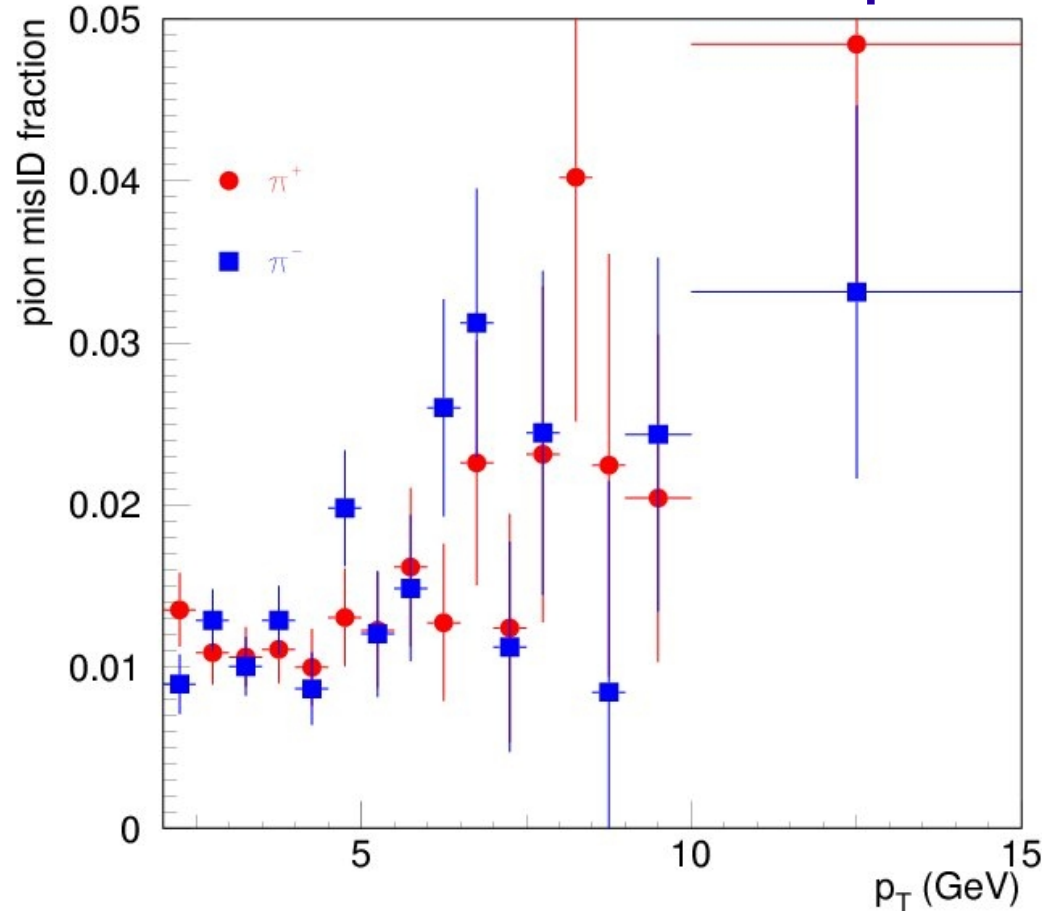
# Search for Rare Decay: $D^0 \rightarrow \mu^+ \mu^-$

## Selection requirement optimization

- use selection quality:  $Q_S = S/(1.5 + \sqrt{B})$
- blinding is implemented by not using the signal but a fake signal sample:
  - $S$  is obtained from the  $D^0 \rightarrow \pi^- \pi^+$  sample
  - $B$  is derived from the  $D^0 \rightarrow K^- \pi^+$  sample with both  $K$  and  $\pi$  have muons attached to them (no overlap with  $D^0 \rightarrow \mu^+ \mu^-$ )
  - cuts are varied to maximize  $Q_S$
- signal sample with  $D^0 \rightarrow \mu^+ \mu^-$  is never looked at

# Search for Rare Decay: $D^0 \rightarrow \mu^+ \mu^-$

Use our clean and copious  $D^0 \rightarrow K^- \pi^+$  sample



Probabilities to attach muons to pions (left) or kaons (right) [decay in flight or punch through]

# Search for Rare Decay: $D^0 \rightarrow \mu^+ \mu^-$

## background contributions

- **combinatorial background**: no real physics signal but simply random coincidence between muon stub and a track
- **physics background**: both pions decay to real muon or punch through the calorimeter and have a real muon signal (relevant only  $D^0 \rightarrow \pi \pi^+$ )

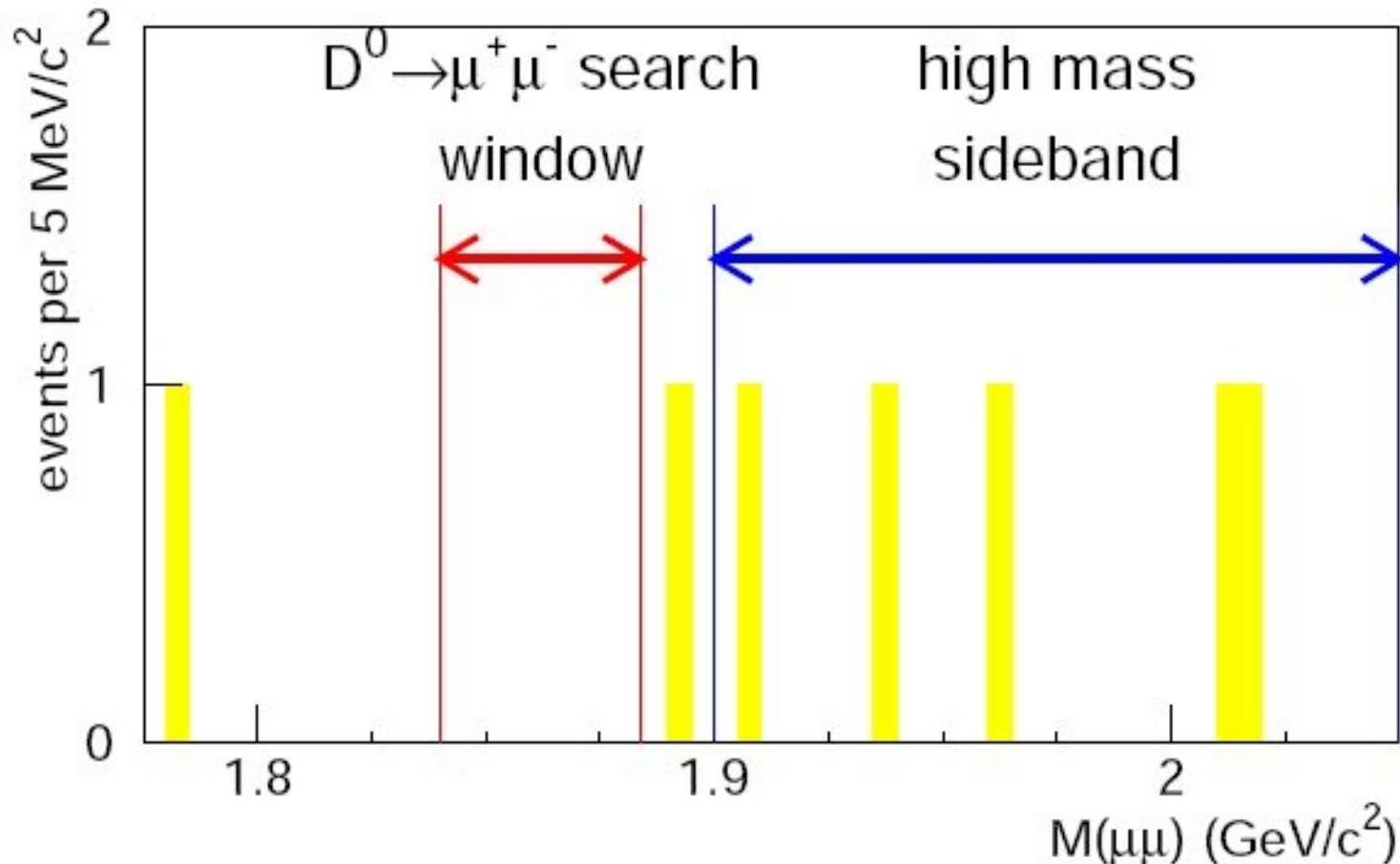
## determine the background

- combinatorial background: use high mass side band of  $D^0$  with both pions having muon stub attached (no physics contribution):  $1.6 \pm 0.7$  events
- physics background: use  $D^0 \rightarrow \pi^- \pi^+$  signal and multiply with muon attach probabilities for pions according to values measured in  $D^0 \rightarrow K^- \pi^+$  sample:  $0.22 \pm 0.02$  events
- total background:  $1.8 \pm 0.7$  events

# Search for Rare Decay: $D^0 \rightarrow \mu^+ \mu^-$

## Opening the box

- analysis with all systematic uncertainties is **completed**
- predicted number of events from background 1.8
- no events inside signal area



# Search for Rare Decay: $D^0 \rightarrow \mu^+ \mu^-$

## Deriving the limit

- no signal is found
- derive upper limit on the branching fraction for  $D^0 \rightarrow \mu^+ \mu^-$ 
  - assume that there is no background (conservative)
  - for 95% CL we can exclude a maximum of 3 events

$$Br(D^0 \rightarrow \mu^+ \mu^-) < 1.43 \cdot 10^{-3} \cdot \frac{3}{1412} \cdot 1.08 = 3.3 \times 10^{-6}$$

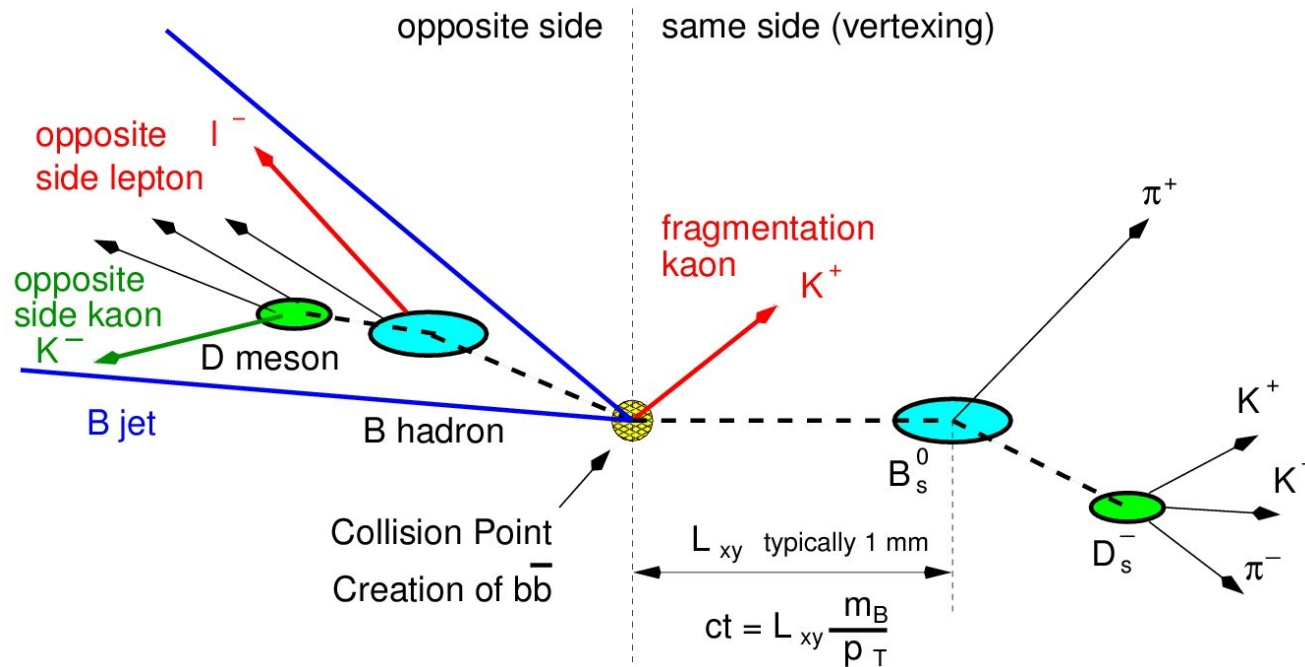
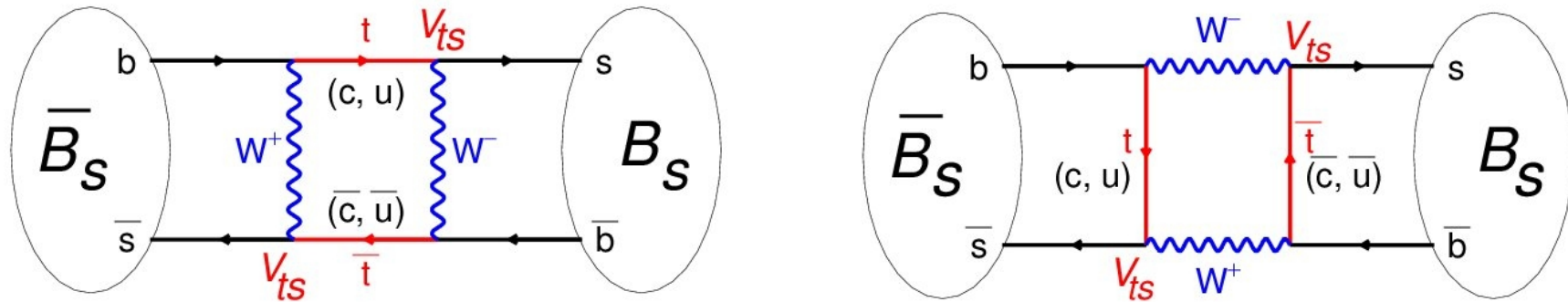
## Also included are systematic uncertainties

- all turned out to be negligible ( $\delta S < 5\%$ )

- formula for inclusion is:  $\Delta N_{95\%CL} = 0.5 \left( N_{95\%CL} \frac{\delta S}{S} \right)^2$

# $B_s$ Oscillations

Neutral Mesons like  $K^0$ ,  $B^0$ ,  $B_s$  can instantaneously switch into their anti particles (higher order)

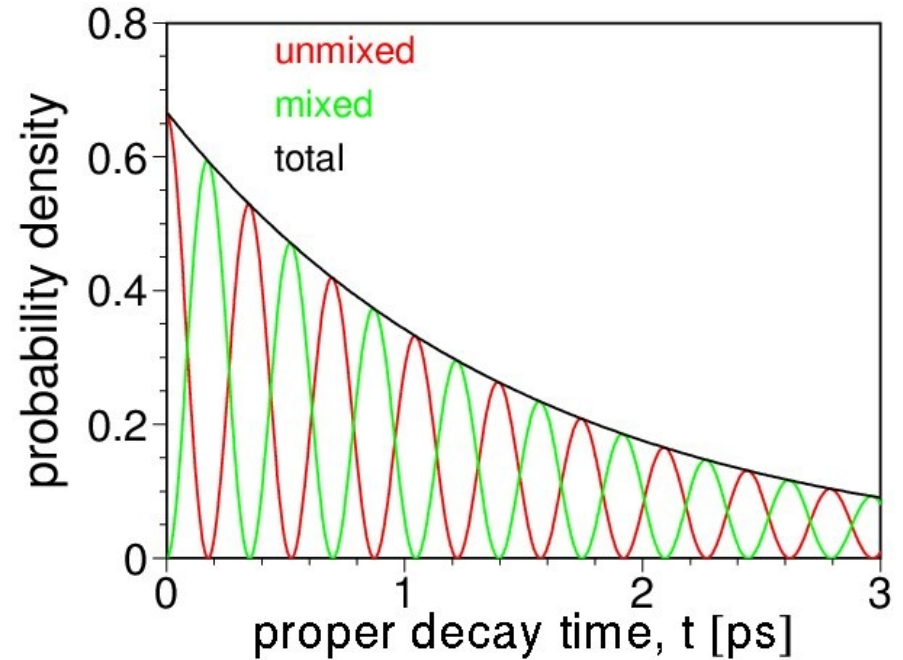


# $B_s$ Mixing Phenomenology

Behavior in proper time

$$P(t)_{B^0 \rightarrow B^0} = \frac{1}{2\tau} e^{-t/\tau} (1 + \cos \Delta m t)$$

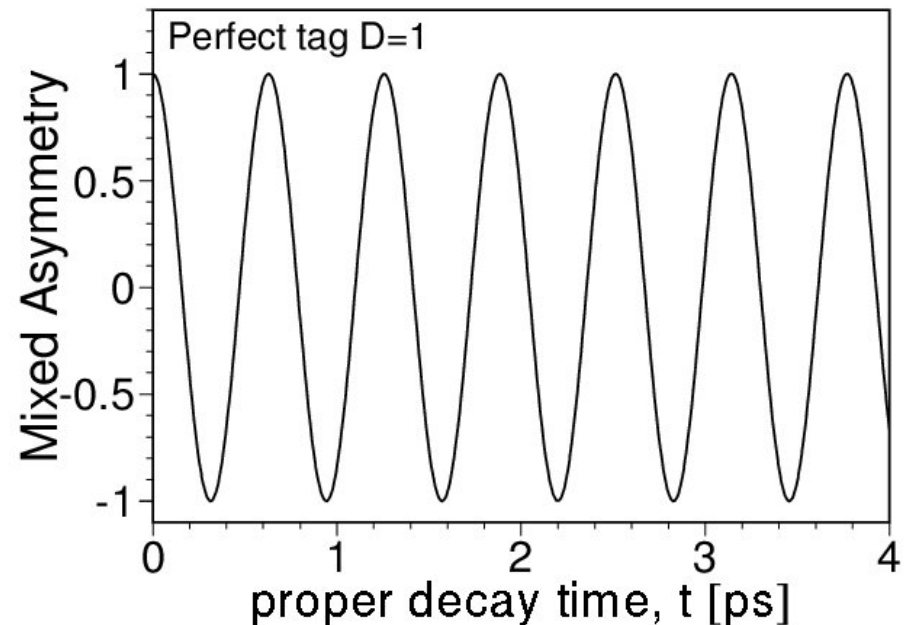
$$P(t)_{B^0 \rightarrow \bar{B}^0} = \frac{1}{2\tau} e^{-t/\tau} (1 - \cos \Delta m t)$$



Determine asymmetry

$$A_0(t) = \frac{N(t)_{unmixed} - N(t)_{mixed}}{N(t)_{unmixed} + N(t)_{mixed}} = \cos \Delta m t$$

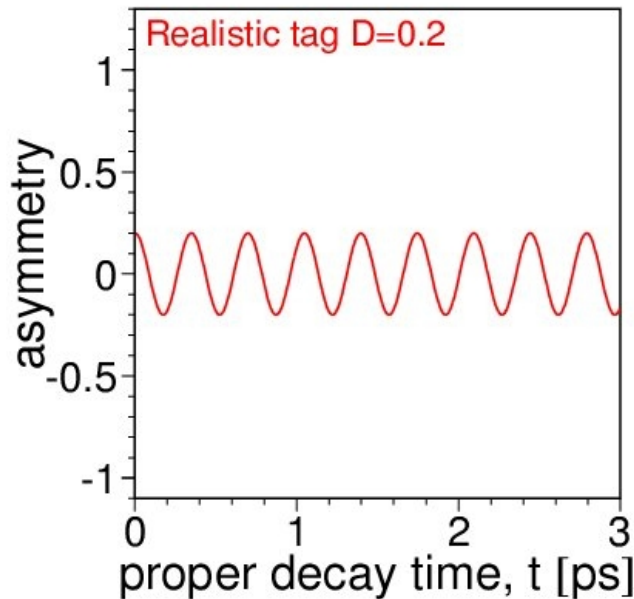
In a perfect world



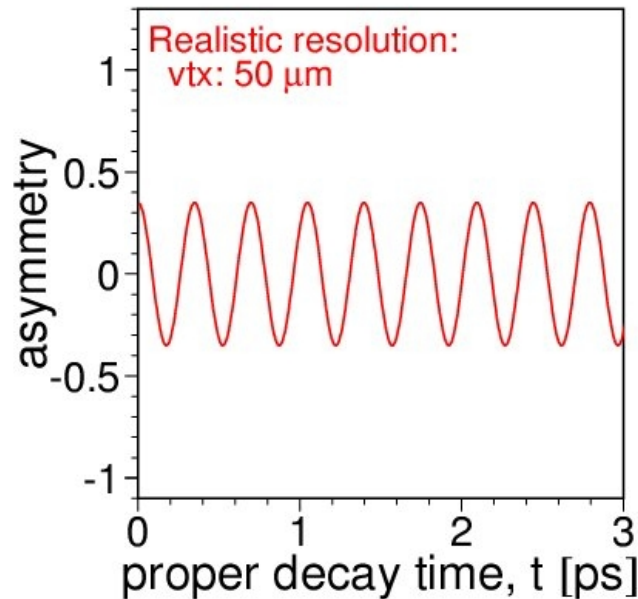


# What Do We See in the End?

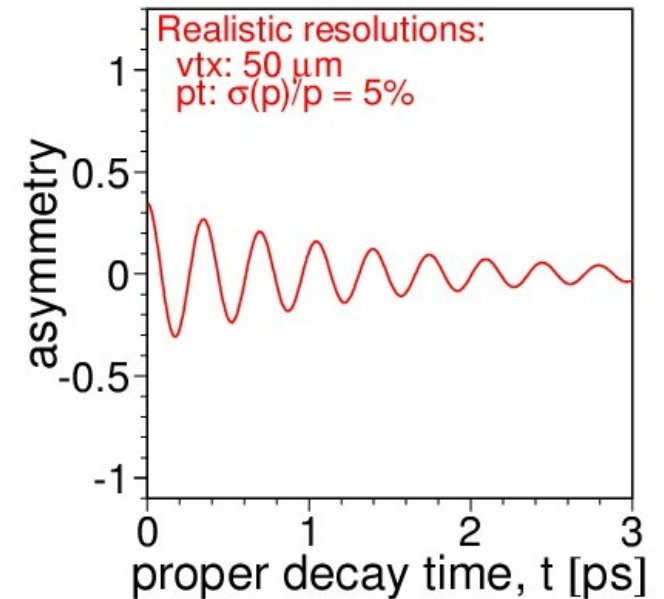
Flavor tagging



Vertex



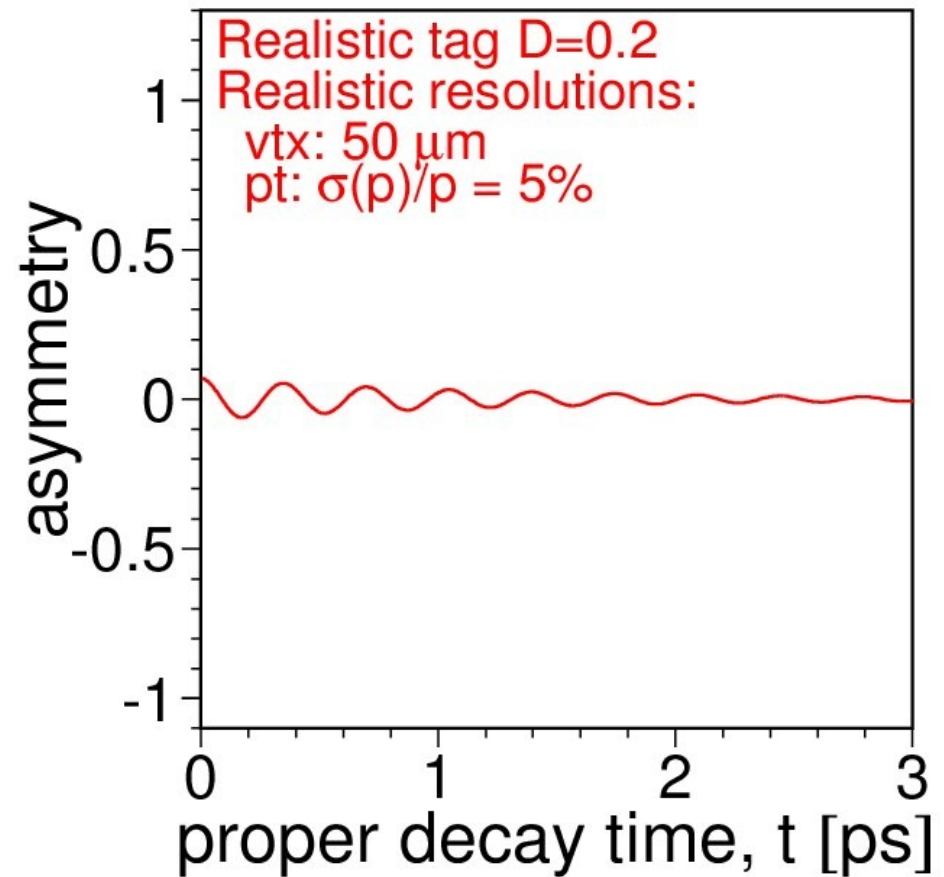
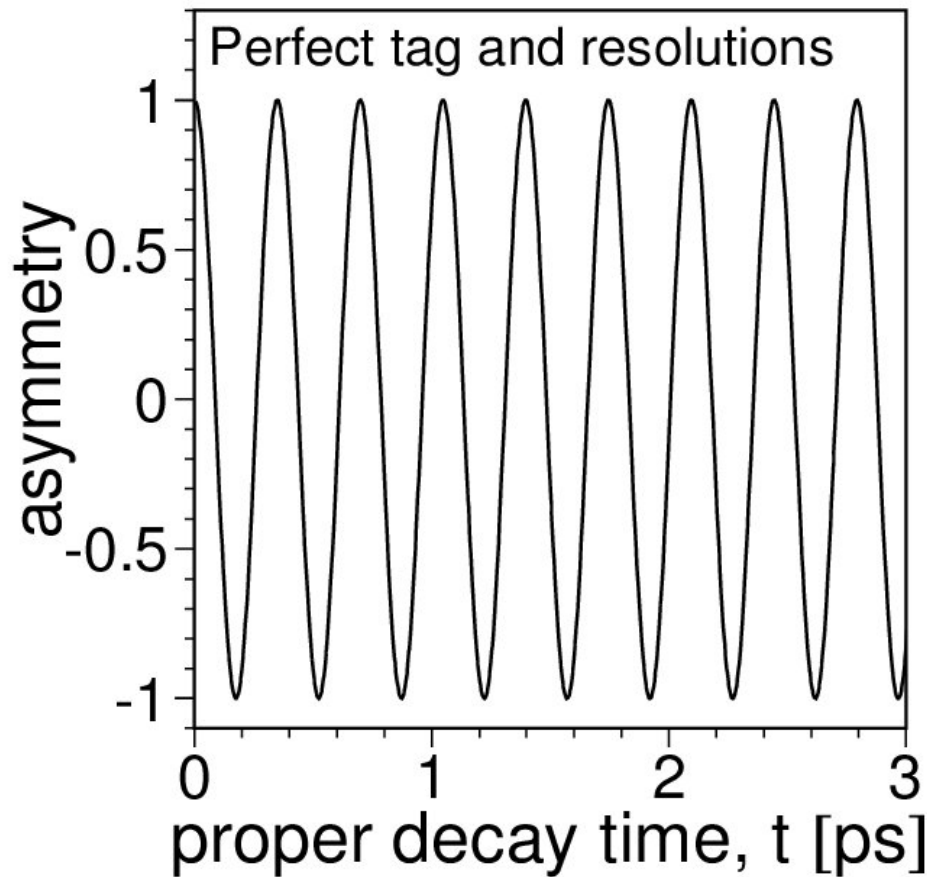
Vertex and Momentum



$$1/\sigma_A = \sqrt{\frac{n_S \epsilon D^2}{2}} \sqrt{\frac{n_S}{n_S + n_B}} \exp\left(-\frac{(\Delta m_S \sigma_{ct})^2}{2}\right)$$

$$\sigma_{ct} = \sqrt{(\sigma_{ct}^0)^2 + \left(ct \frac{\sigma_p}{p}\right)^2}$$

# Perfect World to the Real World

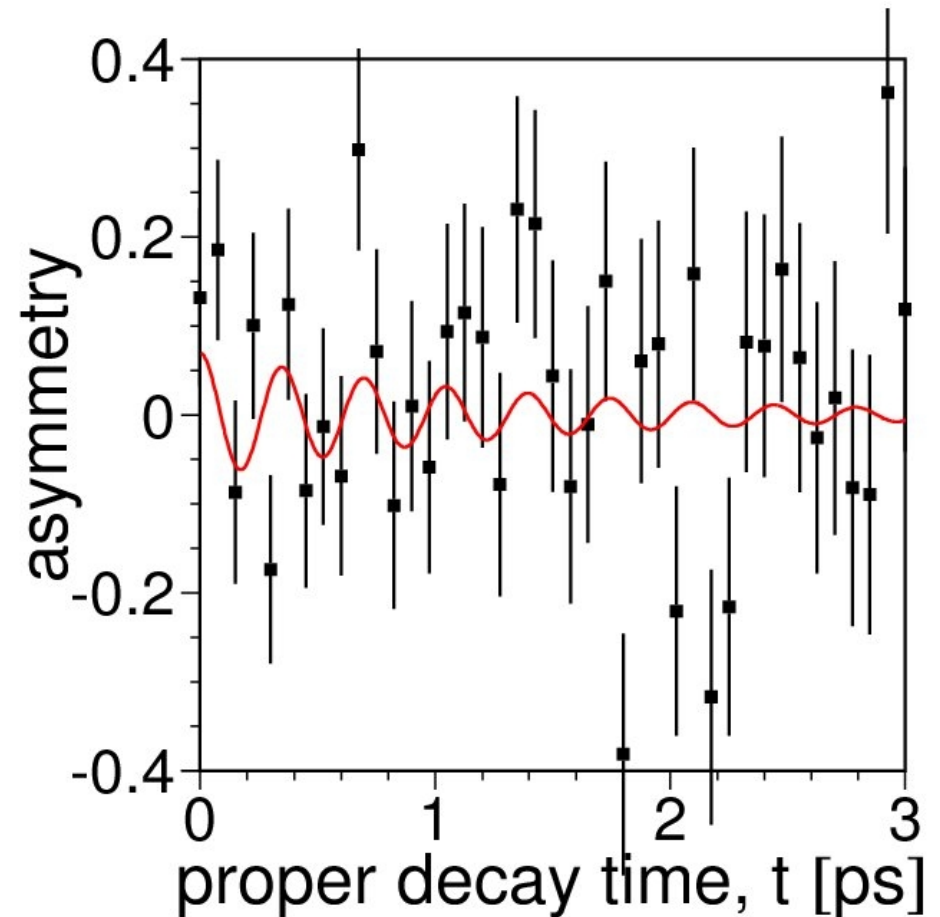
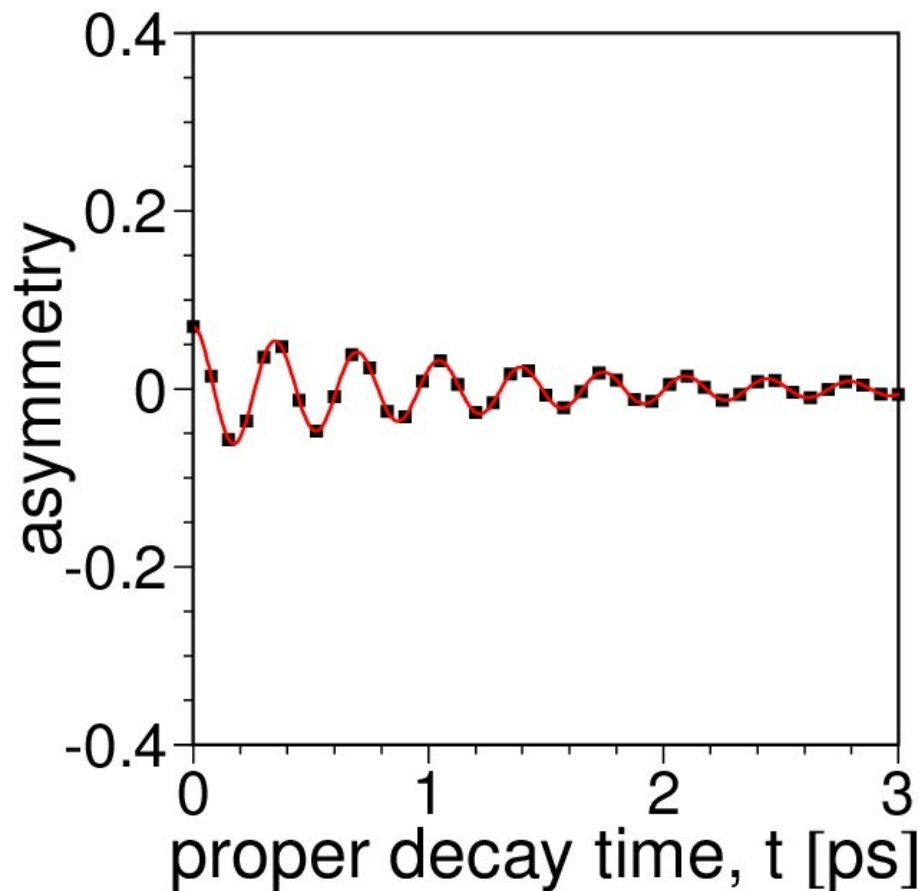


Unbinned likelihood fit:  $p \sim \exp(-t/\tau)(1 \pm AD \cos \Delta mt)$

- ▶ scan  $\Delta m_s$  for signal: determine amplitude,  $A$
- ▶ measure  $\Delta m_s$  with  $A = 1$

apply Fourier analysis

# How Would a Signal Look Like?

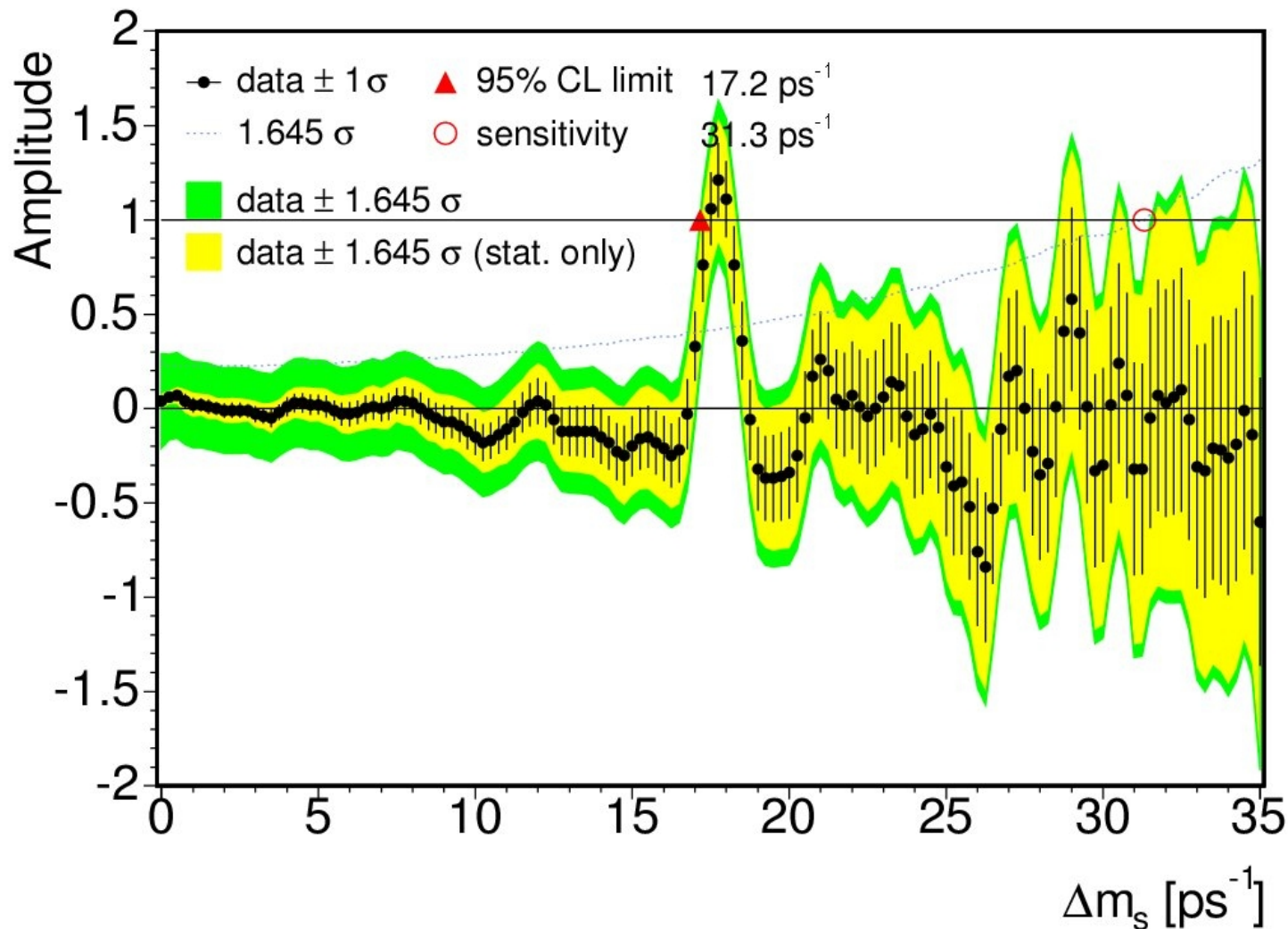


Unbinned likelihood fit:  $p \sim \exp(-t/\tau)(1 \pm AD \cos \Delta mt)$

▶ scan  $\Delta m$  for signal: determine amplitude,  $A$

▶ measure  $\Delta m_s$  with  $A = 1$

# *Bs* Mixing Signal



$A = 1.21 \pm 0.20(\text{stat})$  compatible with 1 for  $\Delta m_s \sim 17.75 \text{ ps}^{-1}$

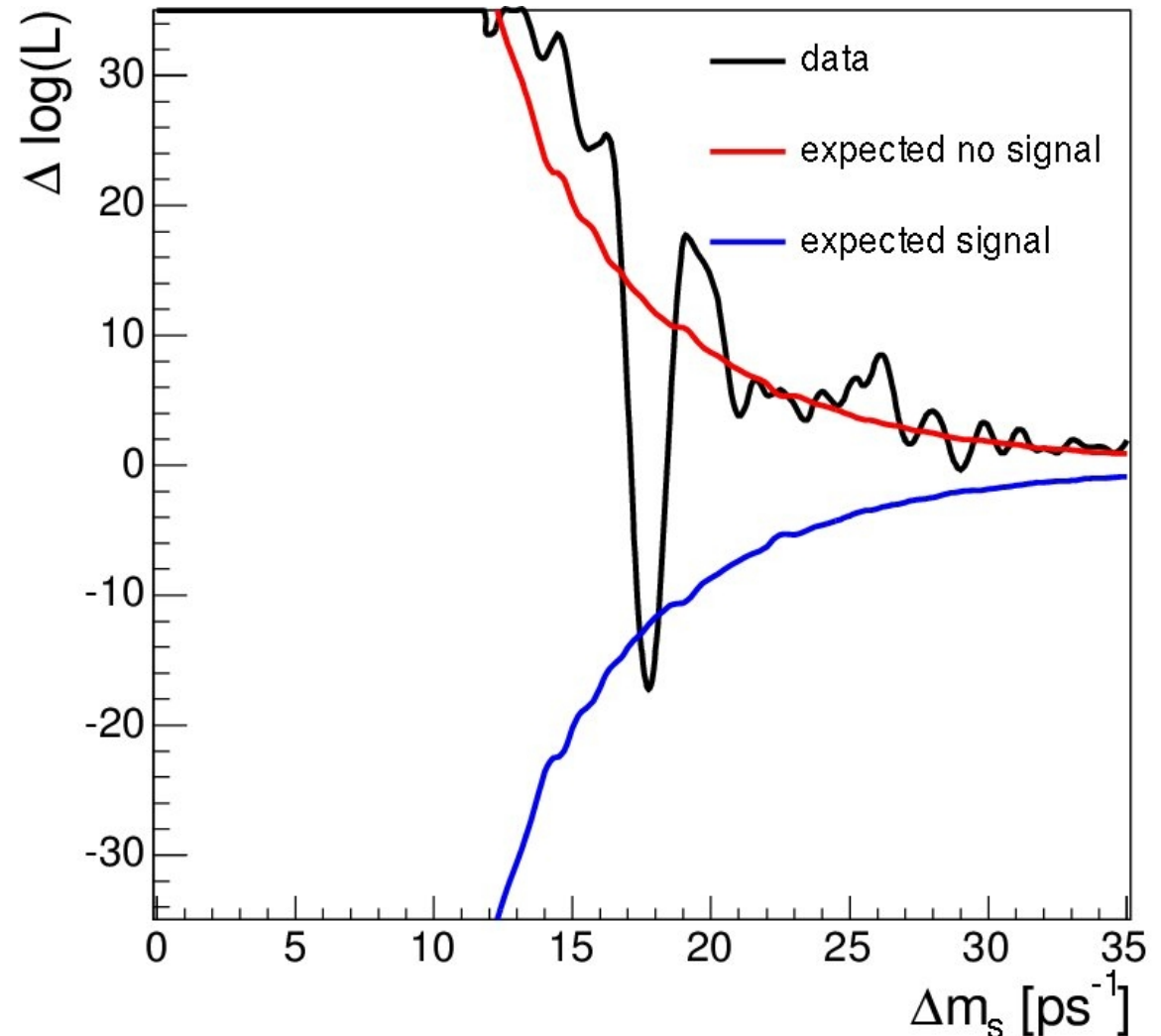
$A/\sigma_A(\Delta m_s = 17.75 \text{ ps}^{-1}) = 6.05$ , but **what is the  $p$ -value?**

# Likelihood Profile

Difference,  $-\Delta \log(L)$

$$\log(\mathcal{L}(A = 1)) - \log(\mathcal{L}(A = 0))$$

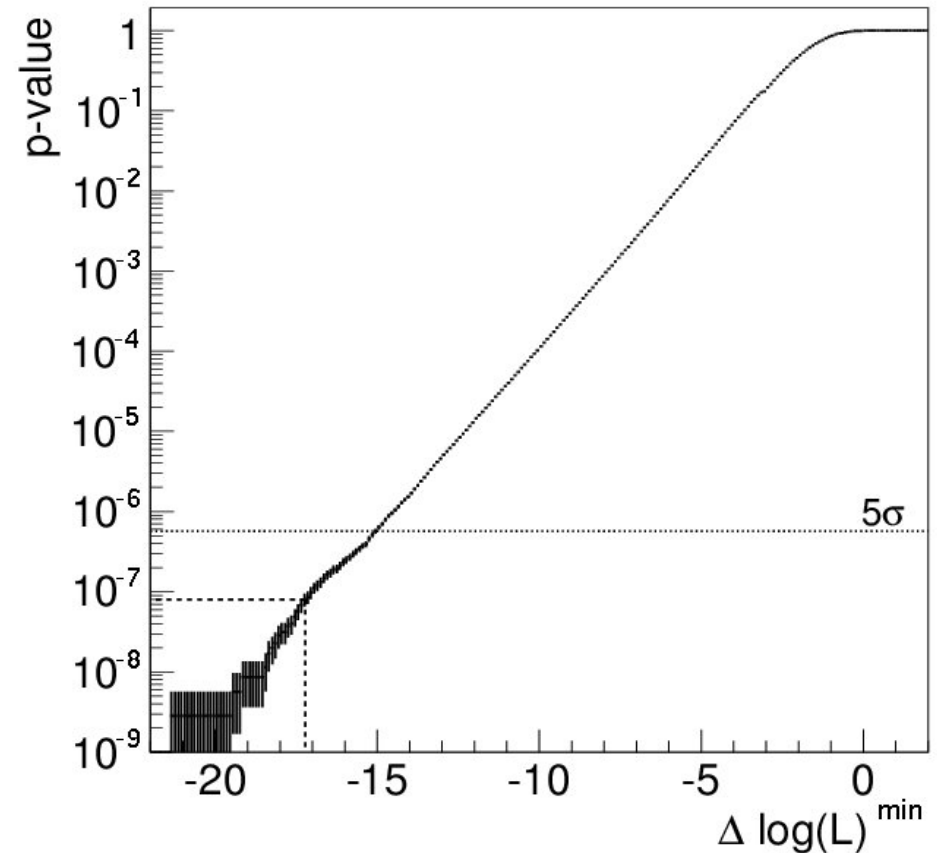
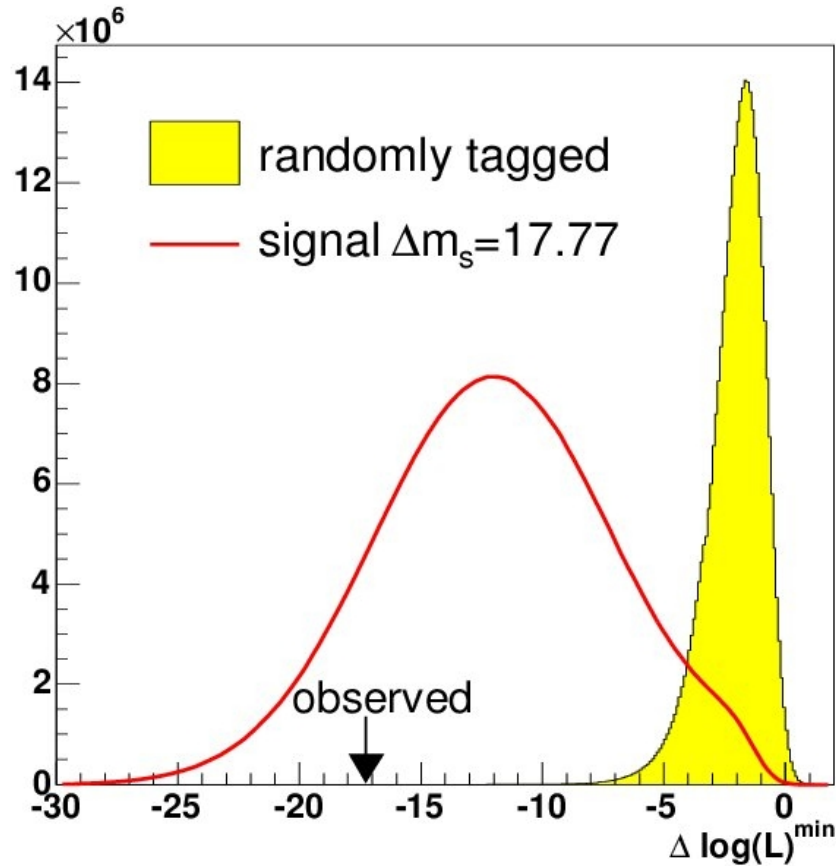
Minimum:  $-17.26$



Key question:

How often can random tags produce a minimum at least as deep?

# Likelihood Significance



28 trials out of 350 million

$p$ -value  $\approx 8 \times 10^{-8}$  corresponding to  $5.4\sigma$

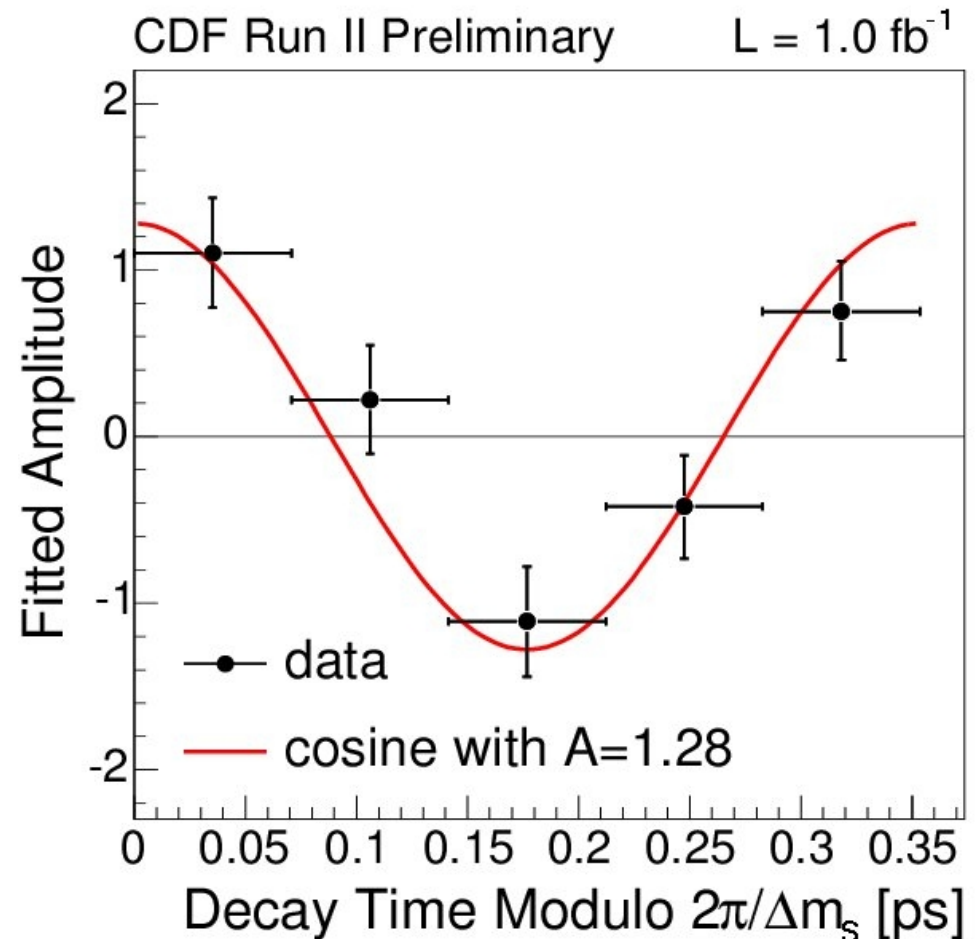
(5 standard deviations is  $= 5.7 \times 10^{-7}$ )

→ passed observation criterion

# *Folding All Oscillations on Top*

The eye is not good at Fourier transforms on the fly

- take events and fold them on top of each other with the measured frequency



# *Conclusion*

## Searches versus measurements

- search looks for a so far unobserved signal (most likely in a mass distribution but other distributions possible)
- searches start with some assumptions, they could be very general (more difficult) or very specific (gives more handles but is less general)
- signal is accepted as observed with five standard deviations, evidence around 3 standard deviations
- non-observation needs more work to evaluate carefully analysis sensitivity to be able to exclude a signal with some confidence level
- measurements analyze a known process and measure a quantity which is either a SM parameter or can be predicted in the Standard Model (test SM hypothesis)
- measurement has potential to find SM disagreement



# *Next Lecture*

## Efficiency and Acceptance