

# $\Lambda_b^0 \rightarrow \Lambda_c^+ \pi^-$ Lifetime Measurement

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## Talk Outline

- Lifetime fit in slices of  $\sigma_{ct}$
- Lifetime fit in trigger slices
- Signal Monte Carlo review
- Baseline version of the lifetime fit
- Conclusion and Outlook

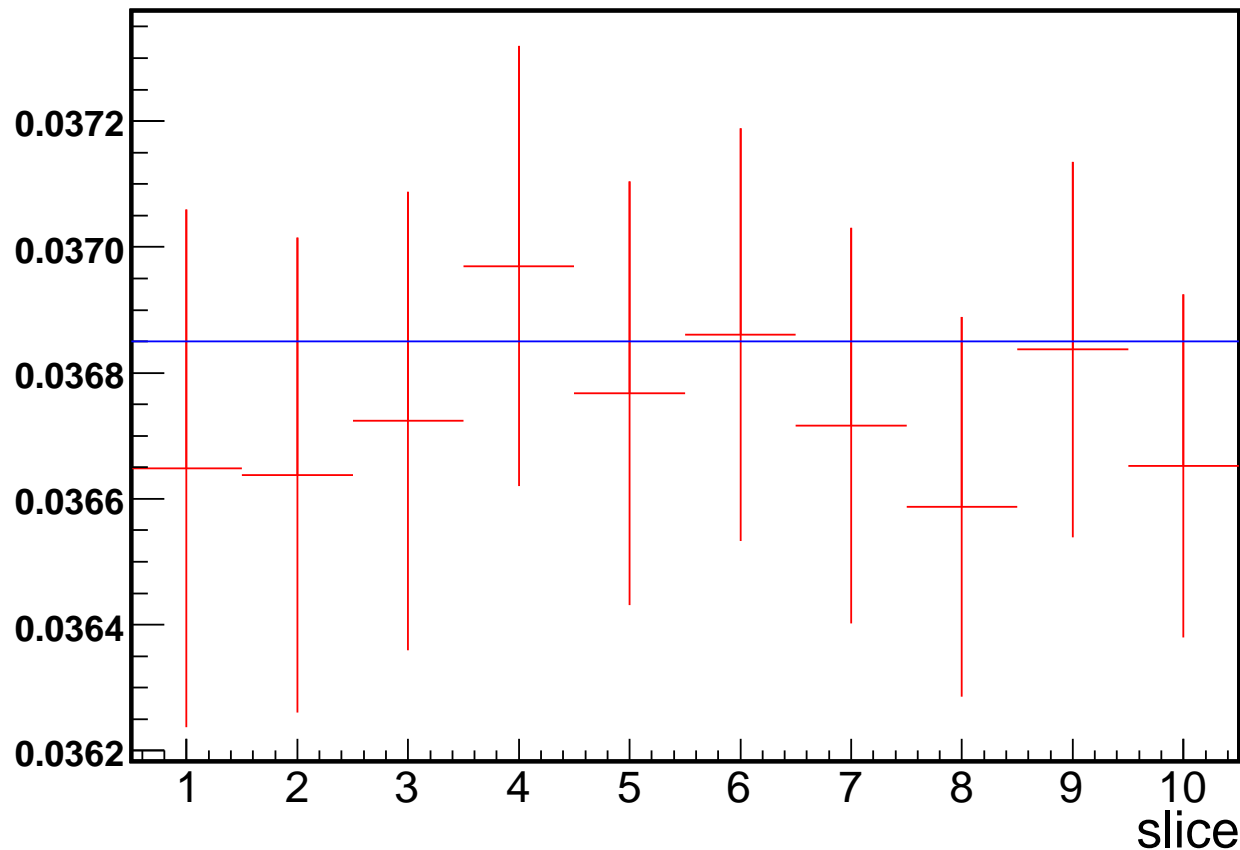
# Lifetime Fit in $\sigma_{ct}$ Slices

In August, we were cutting our  $\sim 330k$  event Signal MC into 10  $\sigma_{ct}$  slices.

A separate SVT efficiency was calculated for each slice.

MC fits in each  $\sigma_{ct}$  slice agreed w/ the generated  $\Lambda_b^0$  lifetime ( $368.5\mu m$ ).

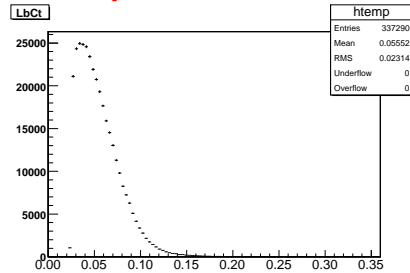
**ct( $\Lambda_b$ ) Signal MC fits in slices of  $\sigma_{ct}$**



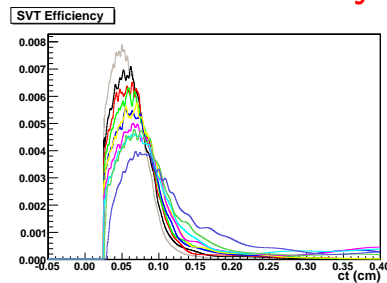
# Lifetime Fit in $\sigma_{ct}$ Slices II

When fitting the entire MC sample, the efficiency slice was chosen dynamically based on  $\sigma_{ct}$  of the event.

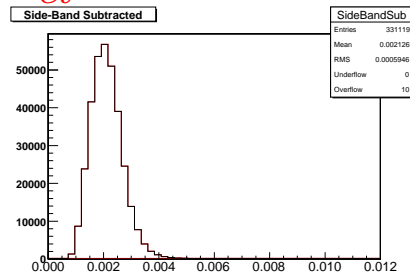
Sample



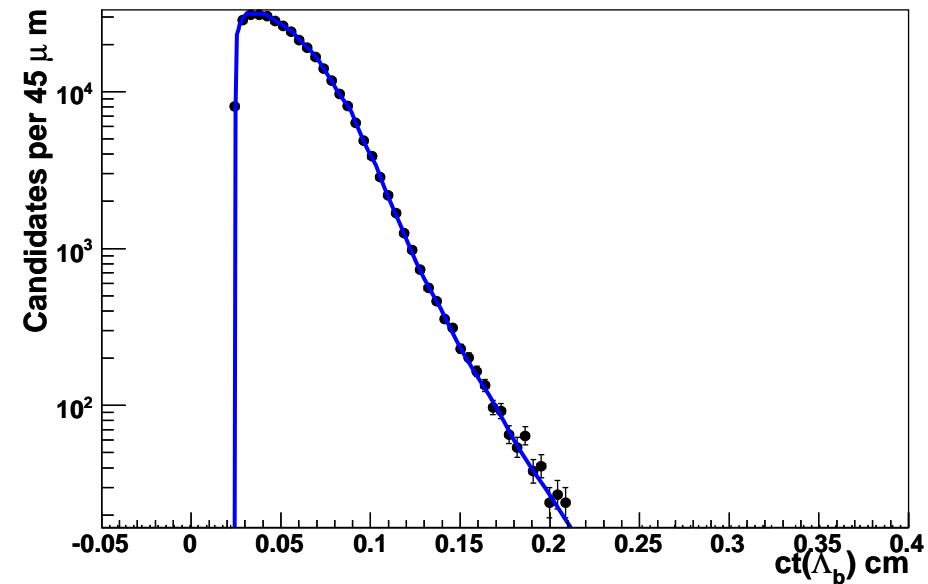
Sliced Efficiency



$\sigma_{ct}$  Distribution



CDF II Preliminary, L=1.1 fb<sup>-1</sup>



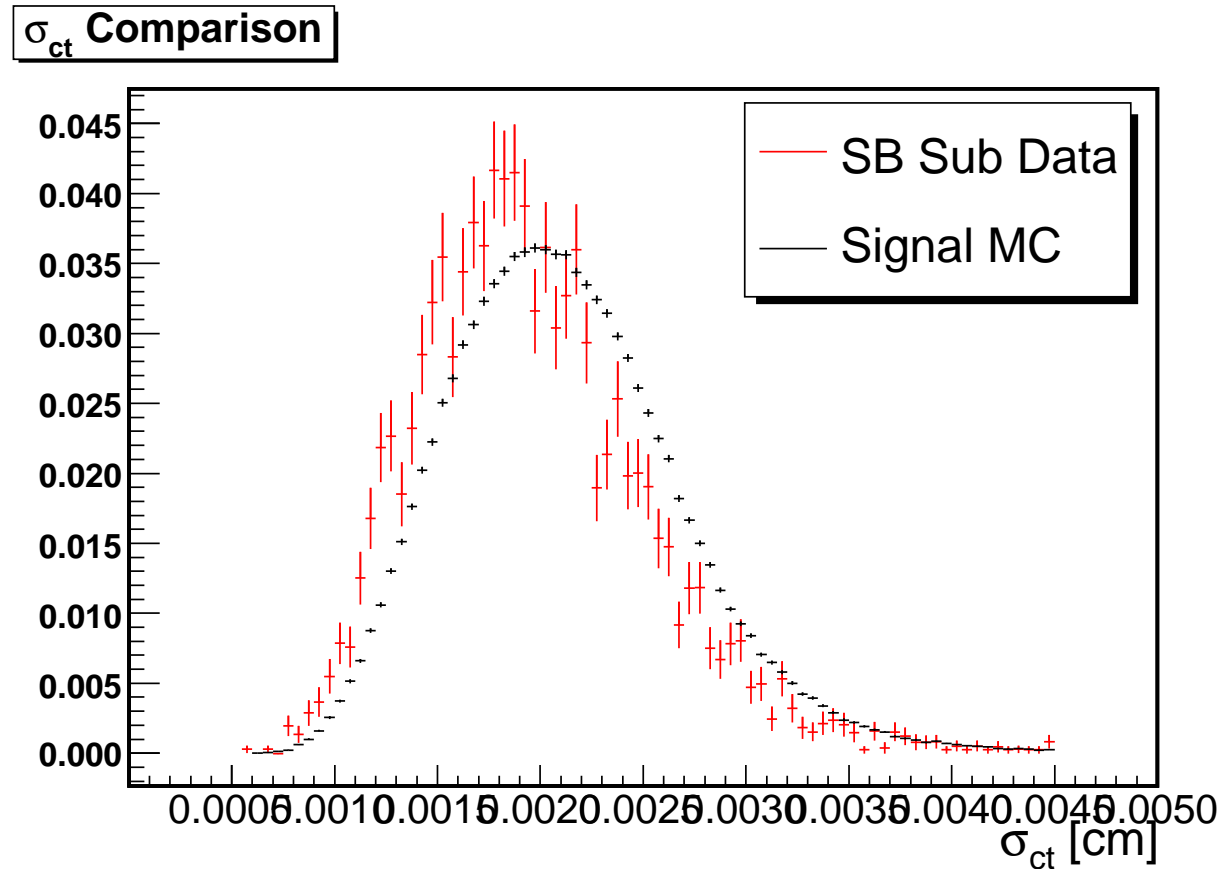
$$ct(\Lambda_b^0) = 367.2 \pm 1.0 \mu\text{m}$$

Consistent w/ Generated and Stand-Alone lifetime.

Method works great in MC.

# Complications with $\sigma_{ct}$ Slice Fits

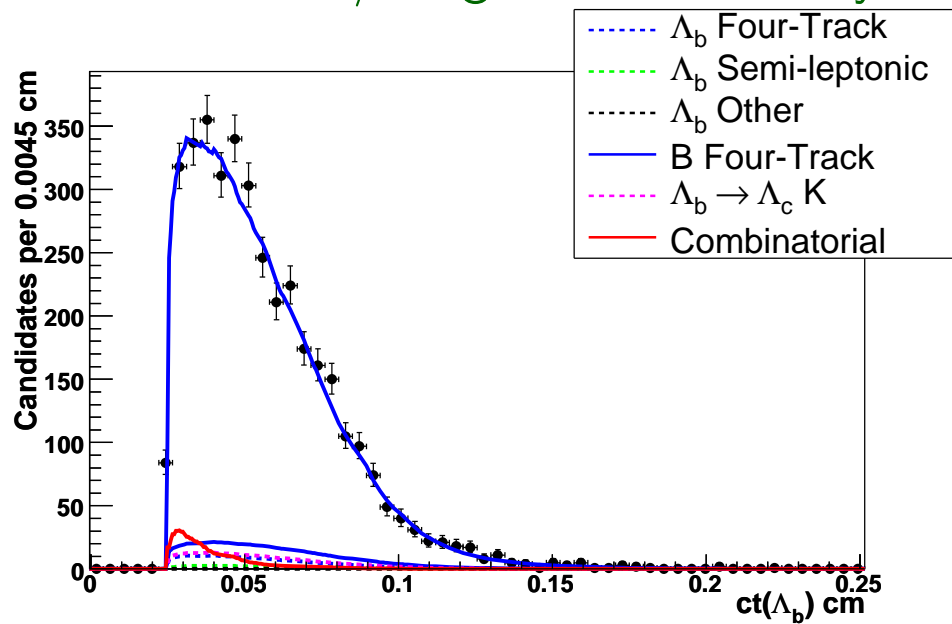
Problems arise when moving from realistic MC to fitting the blinded data.  
How to map the  $\sigma_{ct}$ ? What value do we use for  $\sigma_{ct}$  scale factor?



A comparison of the  $\sigma_{ct}$  distributions from data and MC.

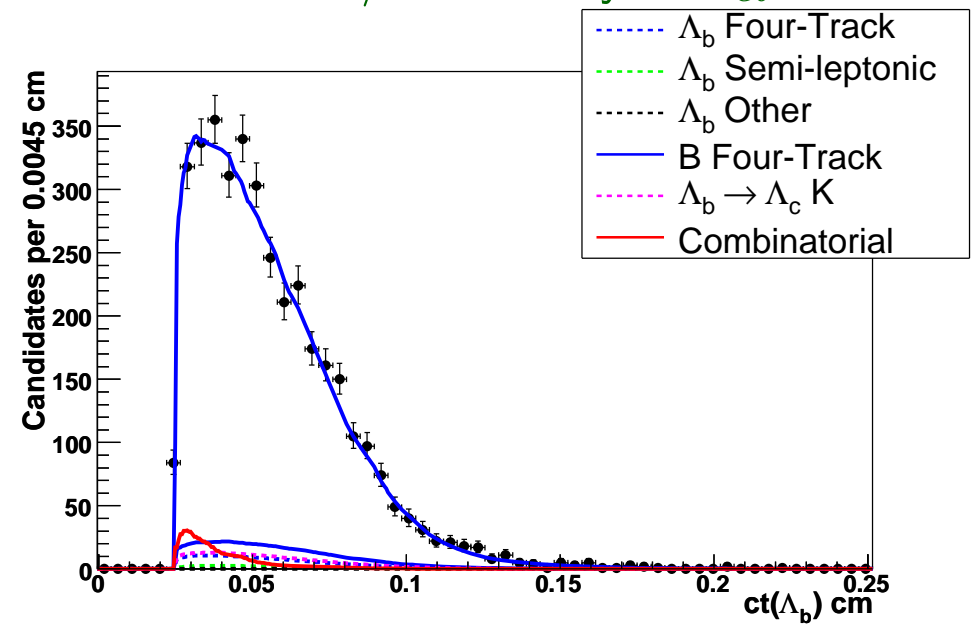
# Abandoning $\sigma_{ct}$ Slice Fits

Blinded Data w/ single SVT efficiency



$$c\tau(\Lambda_b^0) = 390.1 \pm 10.5 \mu\text{m}$$

Blinded Data w/ Efficiency in  $\sigma_{ct}$  Slices



$$c\tau(\Lambda_b^0) = 409.4 \pm 11.9 \mu\text{m}$$

Small changes in  $\sigma_{ct}$  scale factor result in large lifetime shifts.

Because there is no way for us to measure the  $\sigma_{ct}$  scale factor, mapping the slices in  $\sigma_{ct}$  between data and MC artificially increases our sensitivity to getting the scale factor correct.

Because of this, we abandoned the idea of fitting in  $\sigma_{ct}$  slices.

# A New Approach

Instead of  $\sigma_{ct}$  slice fits, we tried another method.

- Match MC kinematics to data: (*i.e.* fit  $\Lambda_c$  Dalitz structure,  $\Lambda_b^0$  polarization, etc.)
- Fit  $\Lambda_b^0$  lifetime in slices of Trigger Type.

Trigger Type flags which combination of tracks fires the TTT and is critically depending on the kinematics of the event.

# Matching MC to Data: $\Lambda_c$ Dalitz

Realized that our old signal MC included only  $\Lambda_c \rightarrow pK^*$  and  $\Lambda_c \rightarrow \Delta^{++}K$  Dalitz modes.

2 missing Dalitz modes were generated and added to our existing MC sample.

- $\Lambda_c \rightarrow pK^*$  ( $\sim 22.6\%$ )
- $\Lambda_c \rightarrow \Delta^{++}K$  ( $\sim 12.2\%$ )
- $\Lambda_c \rightarrow \Lambda(1520)\pi$  ( $\sim 25.5\%$ )
- $\Lambda_c \rightarrow pK\pi$  ( $\sim 39.7\%$ )

(Dalitz modes are weighted to match PDG values)

# Matching MC to Data: $\Lambda_b^0$ Polarization

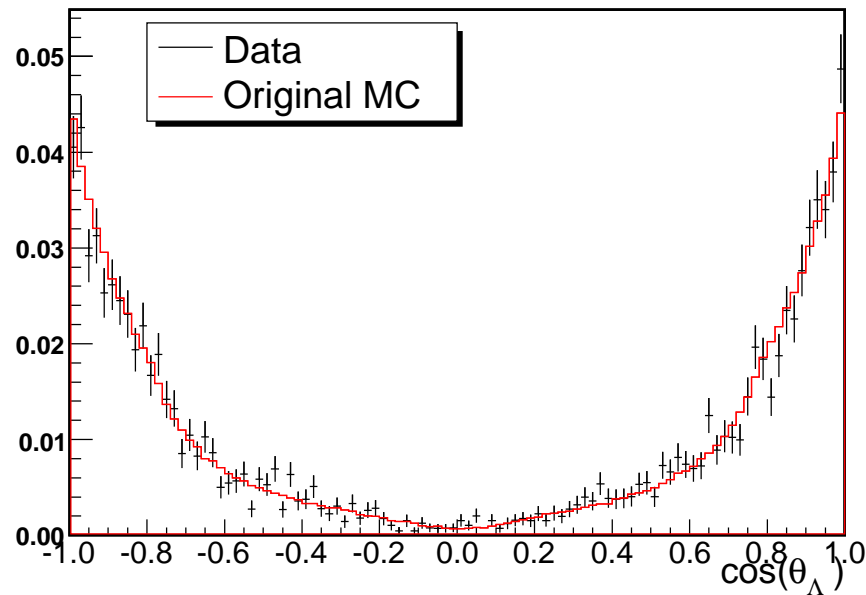
The basic variables for our polarization parameterization are two angles;

$\cos(\theta_{\Lambda_c})$ : Angle between  $\Lambda_b^0$  production (beam direction) and the  $\Lambda_c$  momentum in  $\Lambda_b^0$  rest frame.

$\cos(\theta_p)$ : Angle between  $\Lambda_c$  momentum in  $\Lambda_b^0$  rest frame and the  $p$  momentum in the  $\Lambda_c$  rest frame.

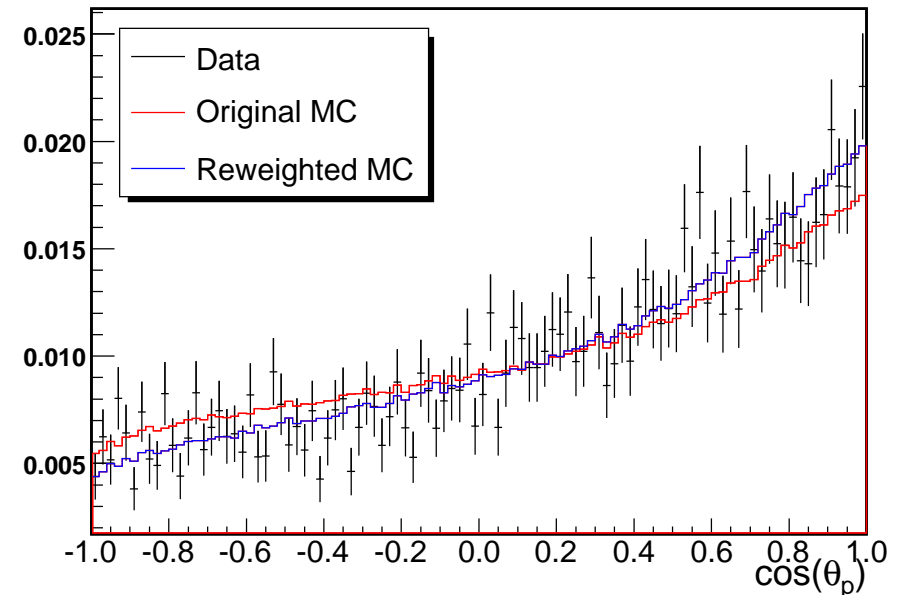
$\cos(\theta_{\Lambda_c})$  distribution matches well.  $\cos(\theta_p)$  is re-weighted to match data.

$\Lambda_c$  production angle



$\cos(\theta_{\Lambda_c})$

proton production angle



$\cos(\theta_p)$

# Matching MC to Data: Trigger Code

Pairs of stable tracks are tested for Two Track Trigger satisfaction.

**Labeling Convention:**  $\Lambda_b^0 \rightarrow \Lambda_c \pi_2; \Lambda_c \rightarrow pK \pi_1$

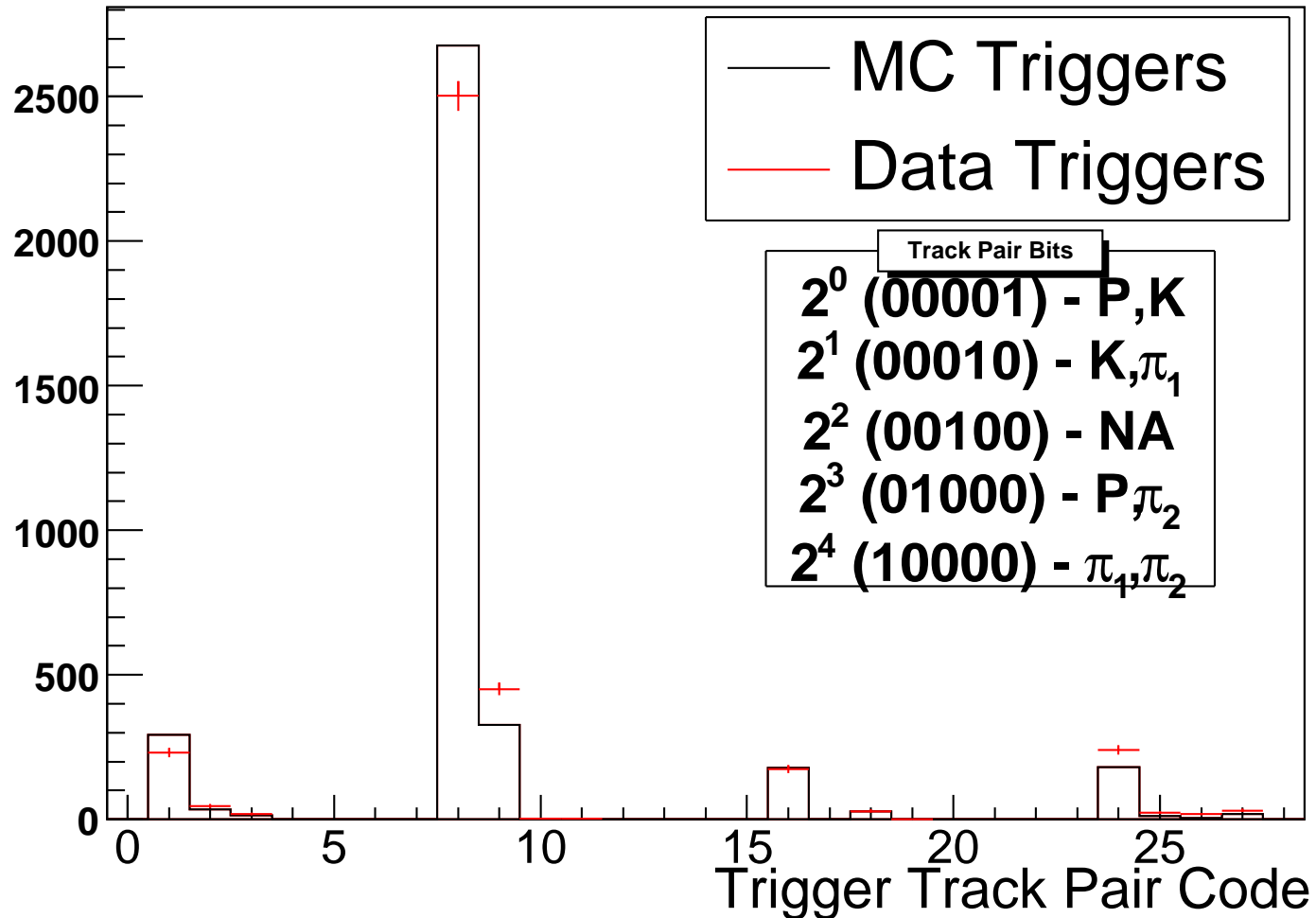
Code	Bits	Description	Nevts
1	00001	$pK$	26,071
2	00010	$K\pi_1$	3,216
3	00011	$pK + K\pi_1$	1,428
8	01000	$p\pi_1$	237,704
9	01001	$p\pi_1 + pK$	29,206
10	01010	$p\pi_1 + K\pi_1$	60
11	01011	$p\pi_1 + pK + K\pi_1$	164
16	10000	$\pi_1\pi_2$	17,711
17	10001	$\pi_1\pi_2 + pK$	17
18	10010	$\pi_1\pi_2 + K\pi_1$	2454
19	10011	$\pi_1\pi_2 + pK + K\pi_1$	16
24	11000	$p\pi_1 + \pi_1\pi_2$	17,398
25	11001	$p\pi_1 + \pi_1\pi_2 + pK$	1,212
26	11010	$p\pi_1 + \pi_1\pi_2 + K\pi_1$	533
27	11011	$p\pi_1 + \pi_1\pi_2 + pK + K\pi_1$	1,664

For the lifetime fit,  
Decide to split sample into 6 categories:

Code	Description	Nevts
1	$pK$	26,071
8	$p\pi_1$	237,704
9	$p\pi_1 + pK$	29,206
16	$\pi_1\pi_2$	17,711
24	$p\pi_1 + \pi_1\pi_2$	17,398
99	All other combinations	10,402

# Monte Carlo vs. Data Trigger Codes

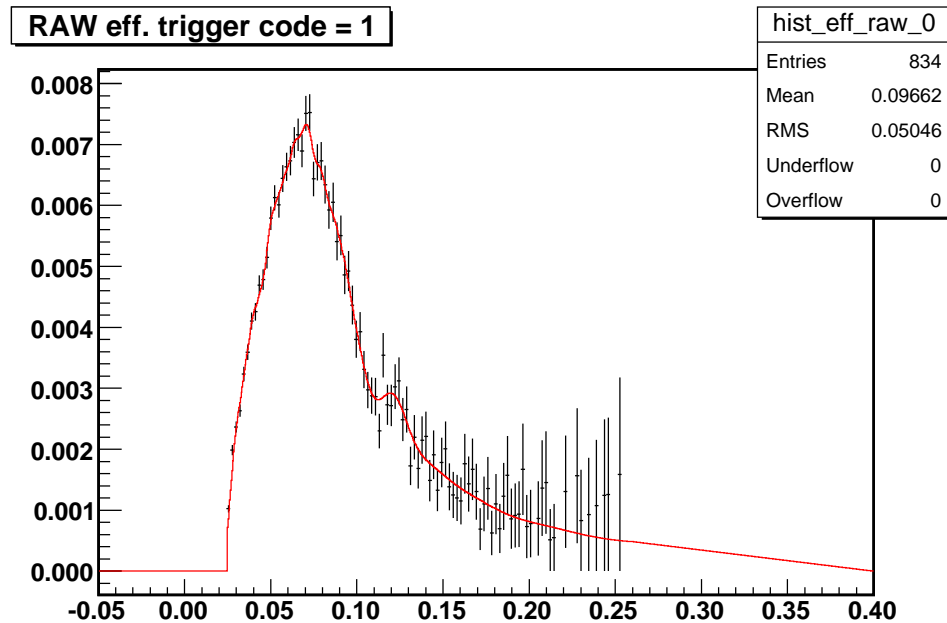
Trigger Composition



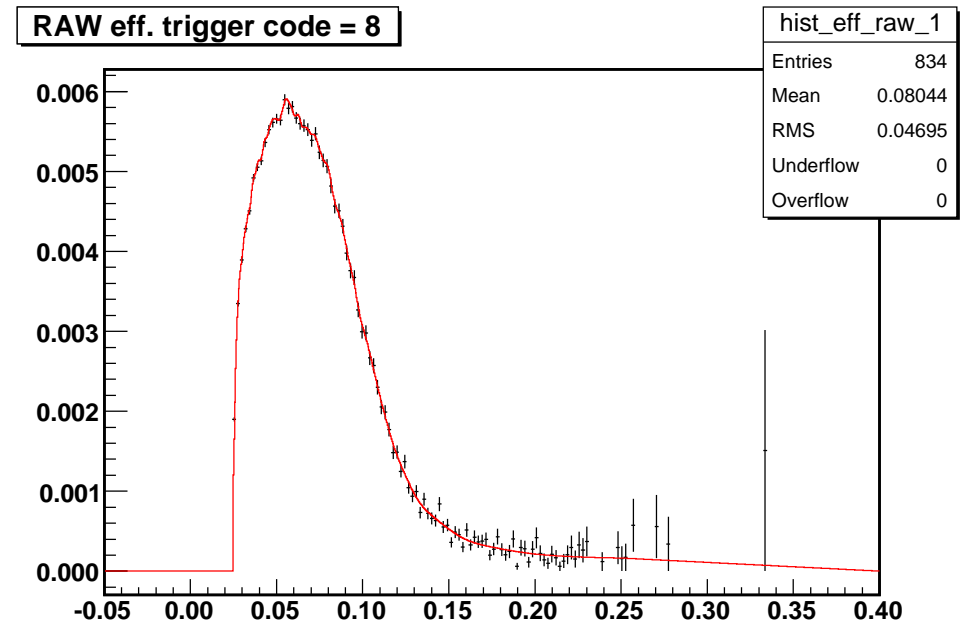
# Efficiencies by TrigCode I

For each of the 6 Trigger Codes, efficiencies are calculated.

We use the same, RooKeysPdf histogram smoothing method that we've been using all along to generate the efficiencies.

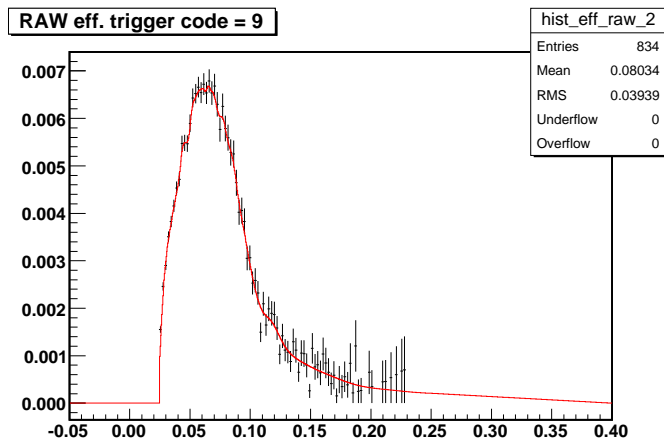


TrigCode=1 ( $pK$ )

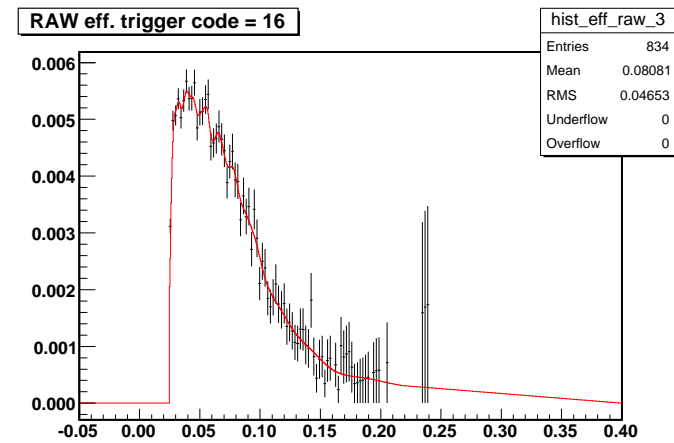


TrigCode=8 ( $p\pi_1$ )

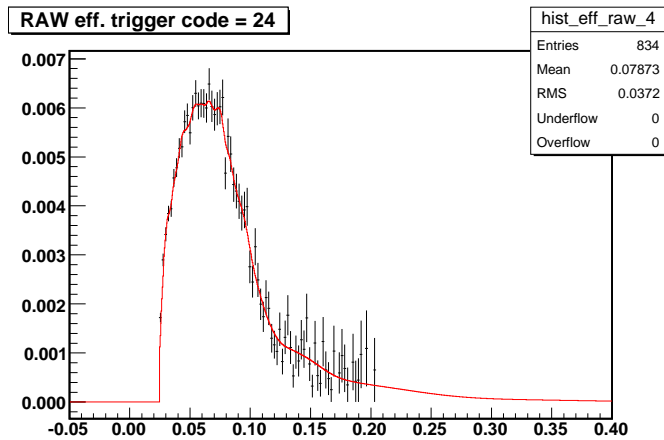
# Efficiencies by TrigCode II



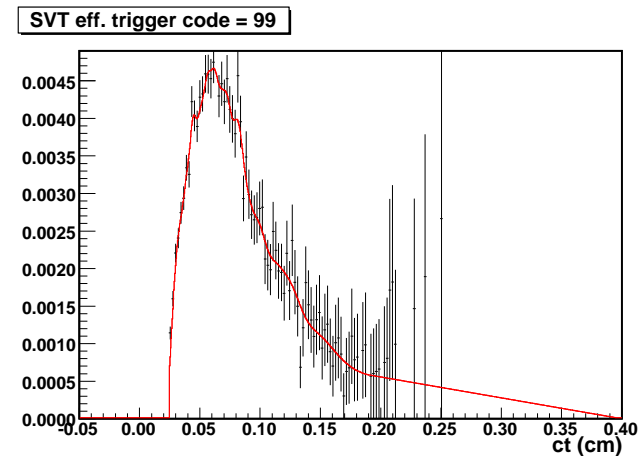
TrigCode=9 ( $p\pi_1 + pK$ )



TrigCode=16 ( $\pi_1\pi_2$ )



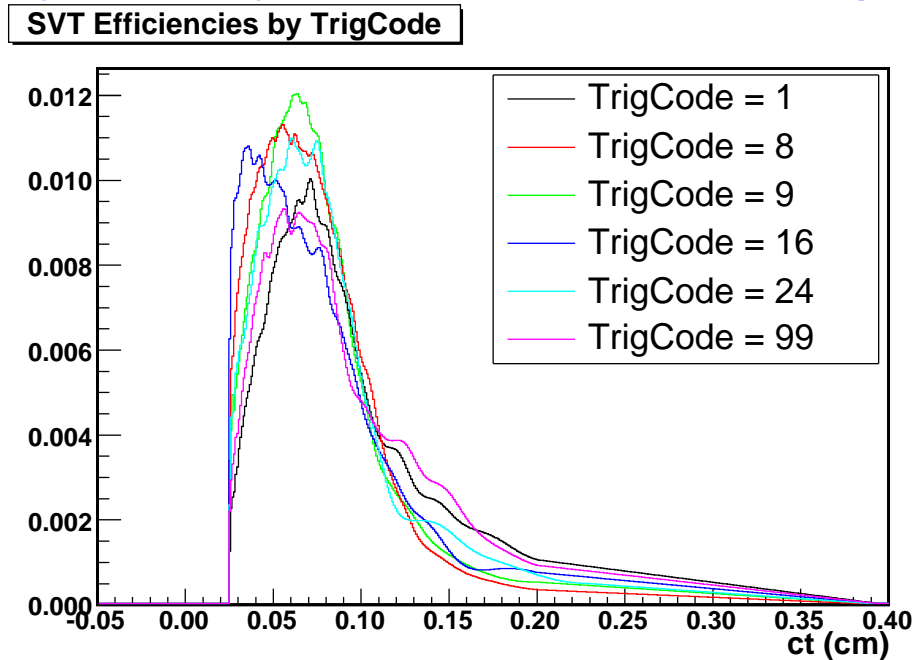
TrigCode=24 ( $p\pi_1 + \pi_1\pi_2$ )



TrigCode=99 (All Other)

# Fitting $\Lambda_b^0$ Lifetime with Trigger Codes

Efficiencies are dynamically chosen based on the TrigCode of each event.



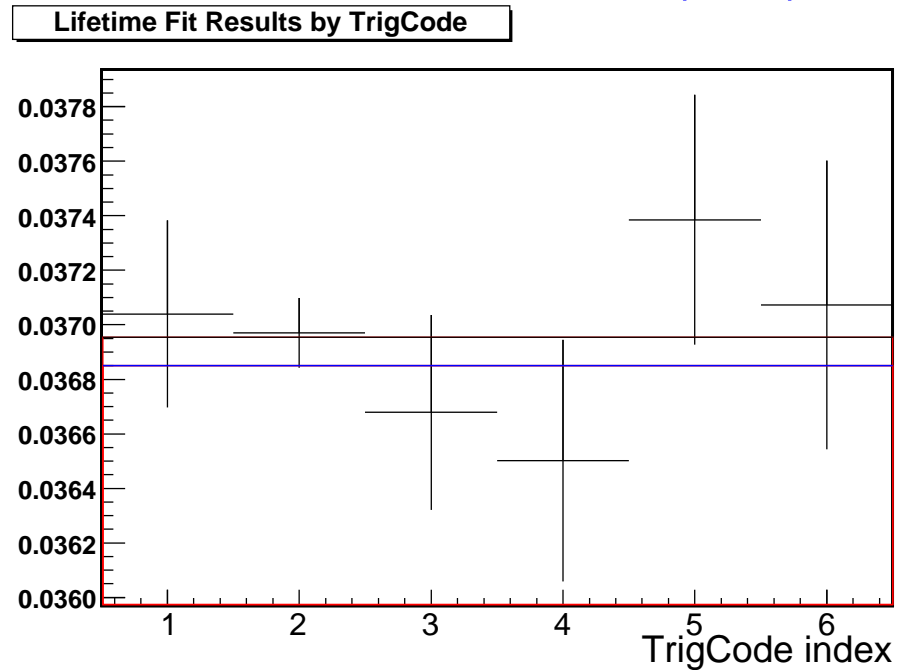
Punzi term is now a 2-D histogram in terms of TrigCode and  $\sigma_{ct}$ . The likelihood looks like this;

$$\mathcal{L}(ct, \sigma_{ct}) = P(ct|\sigma_{ct}, T) \cdot P(\sigma_{ct}, T)$$

The corresponding  $\sigma_{ct}$  distribution is also dynamically chosen.

# TrigCode Slice $\Lambda_b^0$ Lifetime Results

Simultaneous (black) and per-slice(points) TrigCode,  $\Lambda_b^0$  lifetime fit results all agree with generated value (blue).



**However, we have decided to abandon the fit in TrigCode slices because;**

- The statistics in our data sample ( $\sim 3,000$  events) doesn't support a fit sliced in terms of TrigCode (sample is  $\sim 80\%$  TrigCode=8)
- The overall fit improvement does not justify the increased fit complexity.

# *New* Baseline $\Lambda_b^0$ Lifetime Fit

For our “new” baseline  $\Lambda_b^0$  lifetime fit  
We return to fit technology that we were using several months ago.

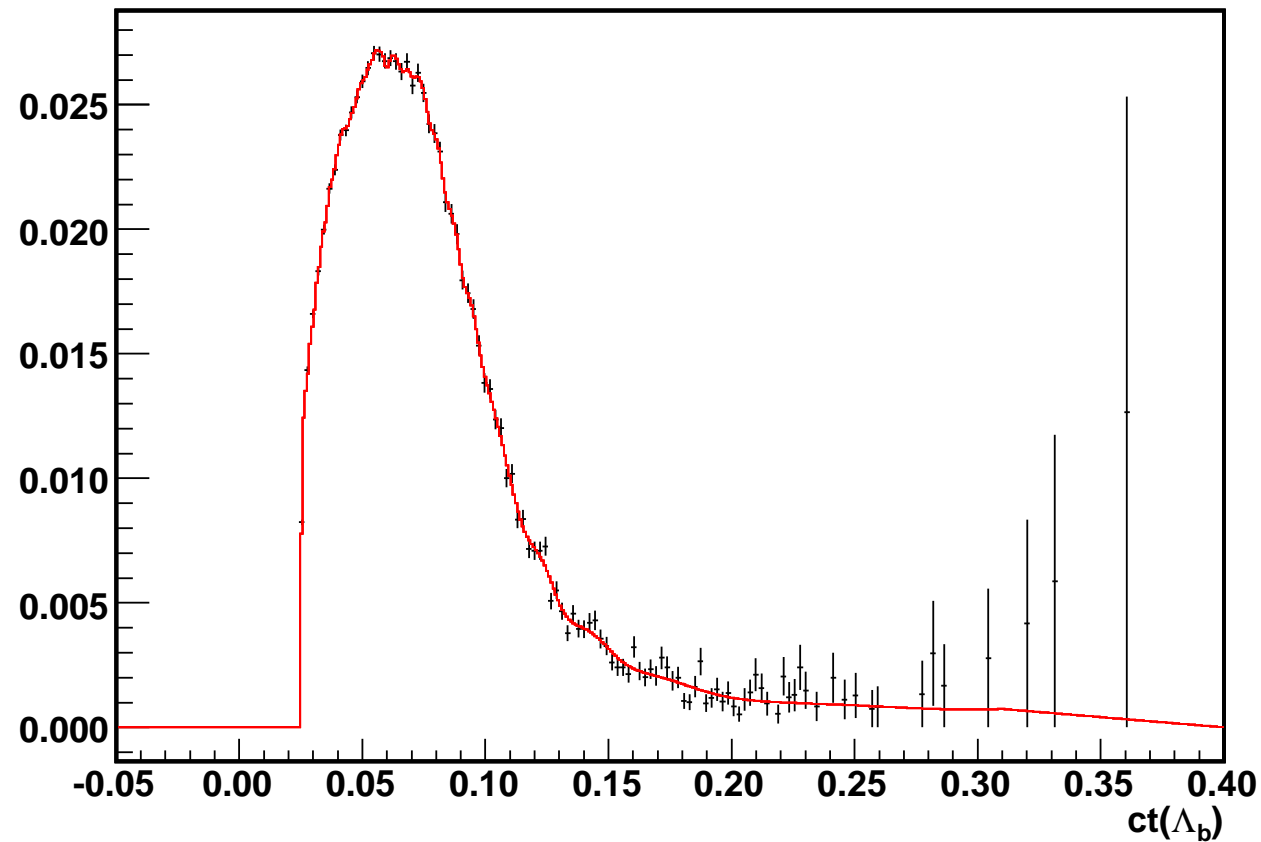
- MC re-weighted in Dalitz, Polarization, and TrigCode.
- monolithic SVT efficiency based on the re-weighted MC.
- 1-D, side-band subtracted,  $\sigma_{ct}$  Punzi term.

# Baseline SVT Efficiency

Generated from signal MC sample re-weighted in  $\Lambda_c$  Dalitz (to PDG fractions), Polarization (from Data), and TrigCode (from Data).

Signal MC sample of  $\sim 234,000$  events used to compute efficiency.

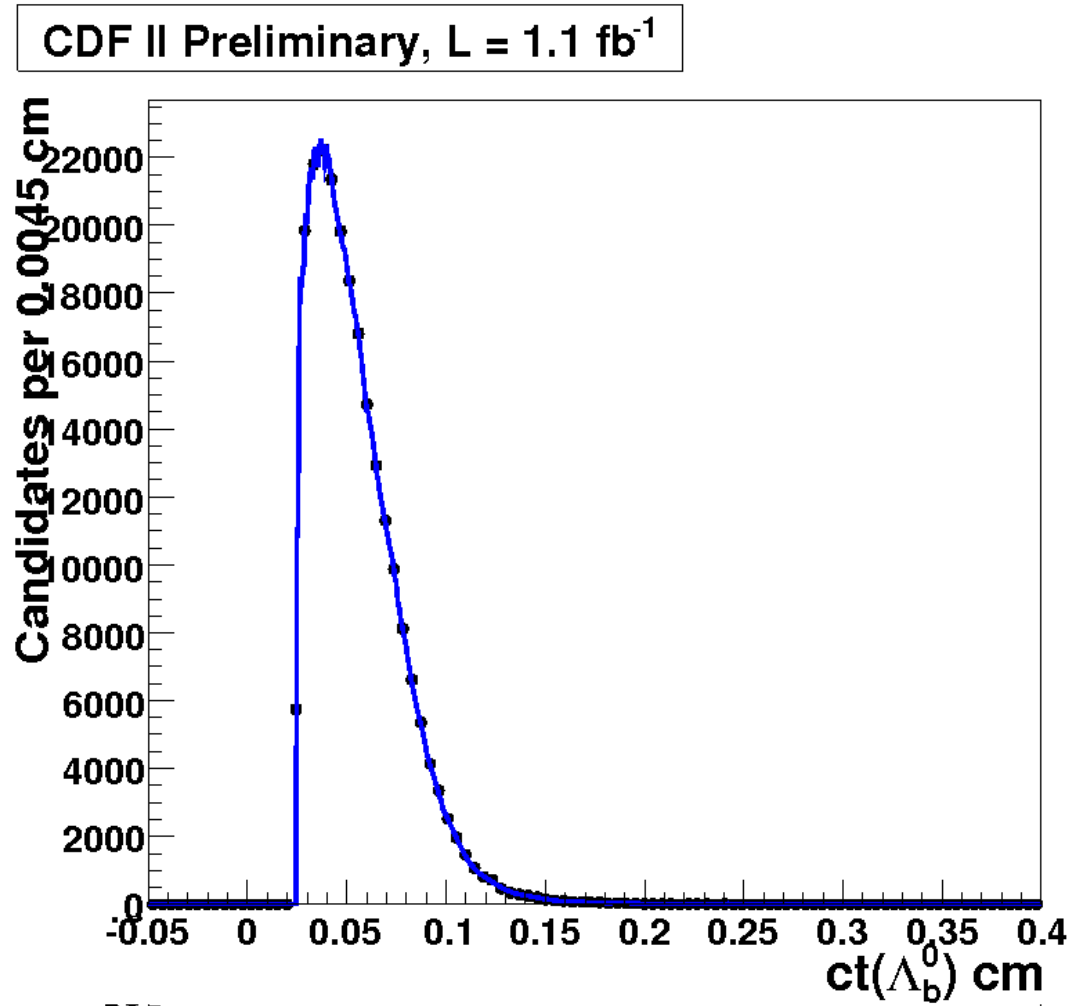
SVT Efficiency



# $\Lambda_b^0$ Lifetime Result: Signal MC Fit

Fitting the same realistic MC sample used to generate the efficiency.

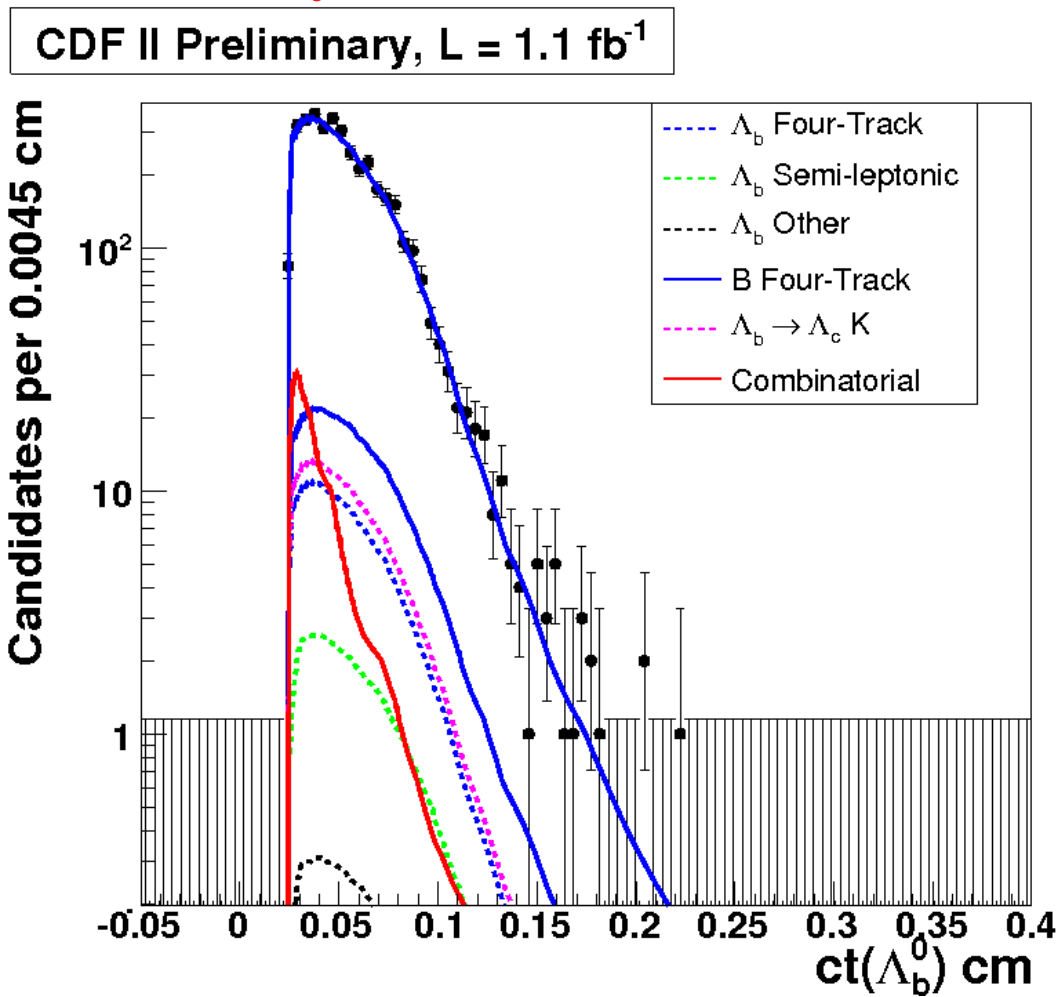
$$c\tau(\Lambda_b^0) = 369.8 \pm 1.2 \mu m$$



# $\Lambda_b^0$ Lifetime Result: Blinded Data Fit

Fit results on blinded data.

$$c\tau(\Lambda_b^0) = XXX.X \pm 11.4\mu\text{m}$$



# Systematics

We have converged on our baseline fit.

Most of the systematics have been setup before, but will need to be re-evaluated with the new baseline fit.

Systematic	Status
Alignment	$2\mu m$ (cdfnote 8524)
SVT-SVX d0 correlation	$1\mu m$ (cdfnote 7386)
Fitter bias	Setup (negligible)
Sample Composition	Setup ( $4\mu m$ )
$\Lambda_c$ Dalitz structure	Setup ( $4\mu m$ )
$ct$ background ( $B$ $\tau$ -parameter)	Setup ( $< 1\mu m$ )
Global scale factor	Setup (negligible)
$ct$ parameterization	Setup (negligible)
Combinatorial $ct$ distribution	Setup (negligible)
$\sigma_{ct}$ binning	Setup (negligible)
$\Lambda_b^0$ polarization	Setup ( $1\mu m$ )
Data-MC comparison: $pt(\Lambda_b^0)$ spectrum	To Do
Data-MC comparison:TrigCode re-weighting	To Do
Primary Vertices	To Do

# Conclusion and Outlook

- We have abandoned lifetime fit in  $\sigma_{ct}$  slices.
- We have developed (and abandoned) a lifetime fit in trigger code slices.
- We have settled on our Baseline fit.
- We expect  $11.4\mu m$  statistical error on our result.
- We are evaluating the systematic errors.
- We plan to pre-bless ASAP.