

A 3D Track Finder for the Belle II CDC L1 Trigger

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Outline

Introduction

Background

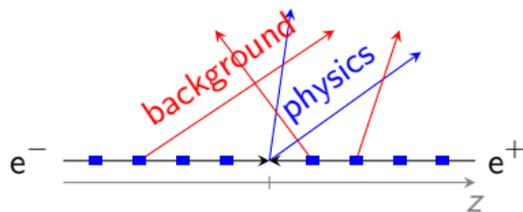
3D Hough Finder

Algorithm

Accuracy



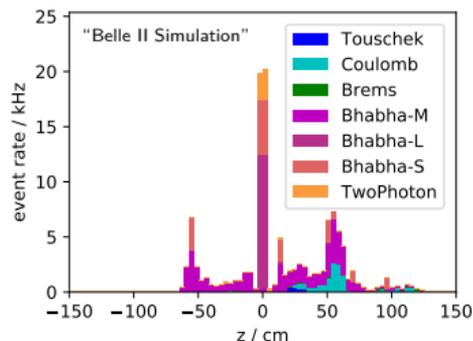
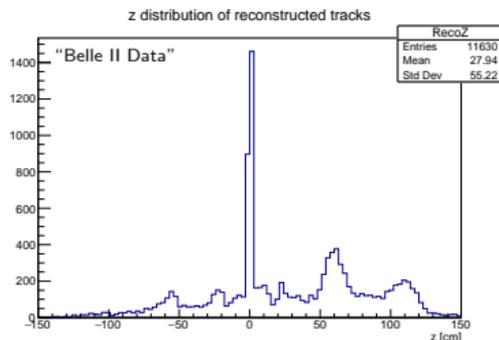
Beam Background Tracks



- ▶ tracks generated at the beam-line & -wall with vertices $z \neq 0$ cm
- ▶ increase with luminosity
- ▶ main processes:
 - Touschek effect
 - radiative Bhabha back scatters
 - beam gas

NeuroTrigger Goals

- ▶ reject tracks from $z \neq 0$ cm
- ▶ single track z-vertex resolution < 2 cm
- ▶ latency $< 1 \mu\text{s}$

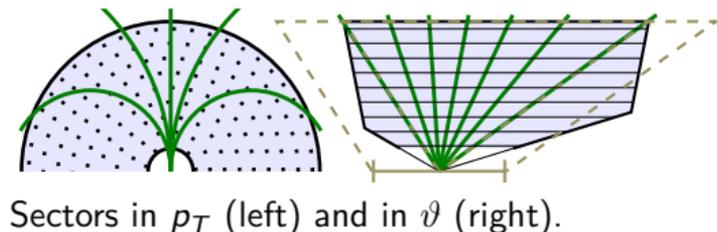


⇒ need z vertex reconstruction at 1st trigger level

3D Hough Finder (p_T, φ, ϑ)

Motivation

- ▶ include CDC stereo hits
- ▶ improve track finding efficiency
- ▶ get NN hit selection in one step (axial & stereo)
- ▶ estimate ϑ (allow NN sectorization)



Track Finder Concept

Bayes'ian estimation

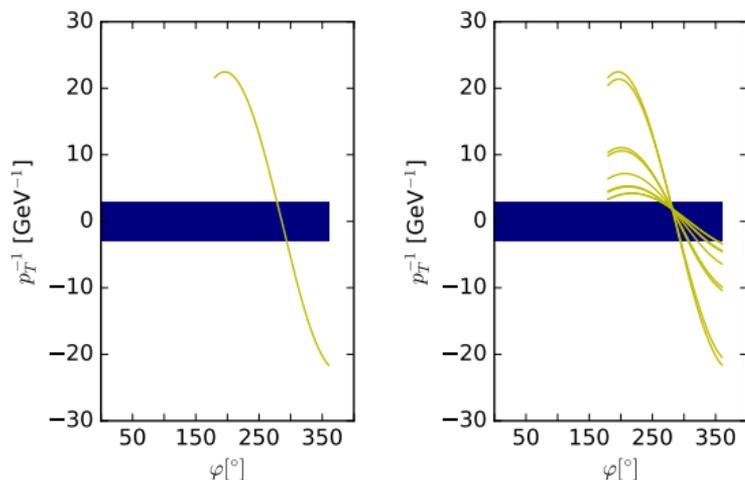
$$P(\text{tracks}|\text{hits}) = \frac{P(\text{hits}|\text{tracks}) \cdot P(\text{tracks})}{P(\text{hits})}$$

with a set *tracks* and a set *hits*.

- ▶ general approach
- ▶ allows easy change of the track and hit parametrization
- ▶ results equivalent to a Hough transformation

Hits in Parameter Space

1. conformal mapping: $x' = \frac{2x}{x^2 + y^2}; y' = \frac{2y}{x^2 + y^2}$
2. Hough transform: $p_T^{-1}(\varphi) = C \cdot (x' \cos(\varphi) + y' \sin(\varphi))$



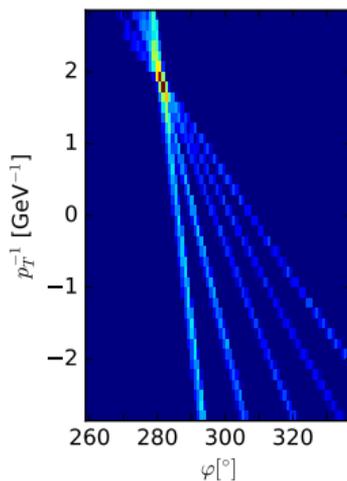
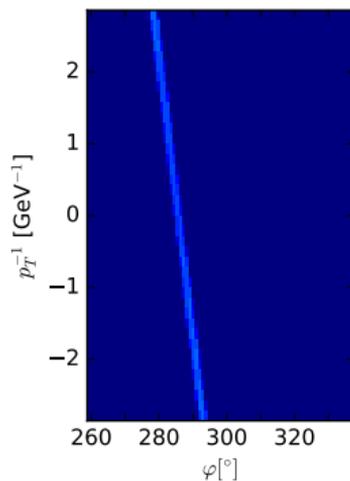
- ▶ hits are sine curves in the p_T^{-1}, φ space
- ▶ tracks are intersections
- ▶ blue region: $p_T > 350$ MeV

binning of track parameters (φ, p_T)

Construct Houghplane

$$H(t|hits) = \sum_{h \in hits} P(t|h)$$

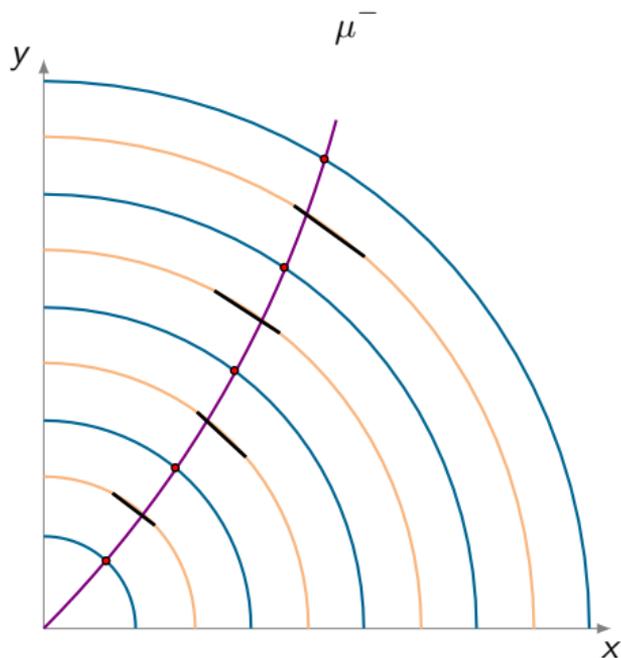
$P(t|h)$ single hit contributions. $H(t|hits)$: Houghplane for all hits.



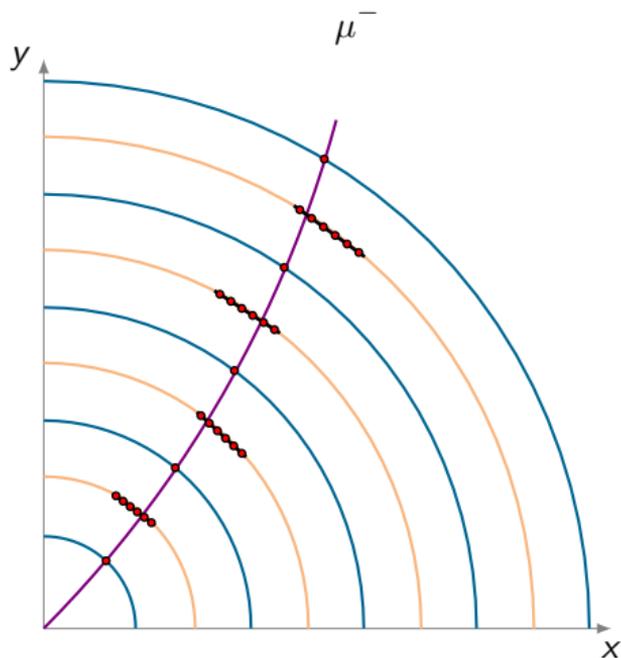
Cluster Peaks

- ▶ identify tracks
- ▶ are local maxima
- ▶ have a minimum weight

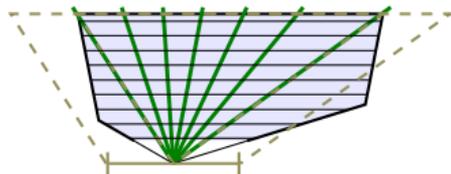
Transverse Hit Positions



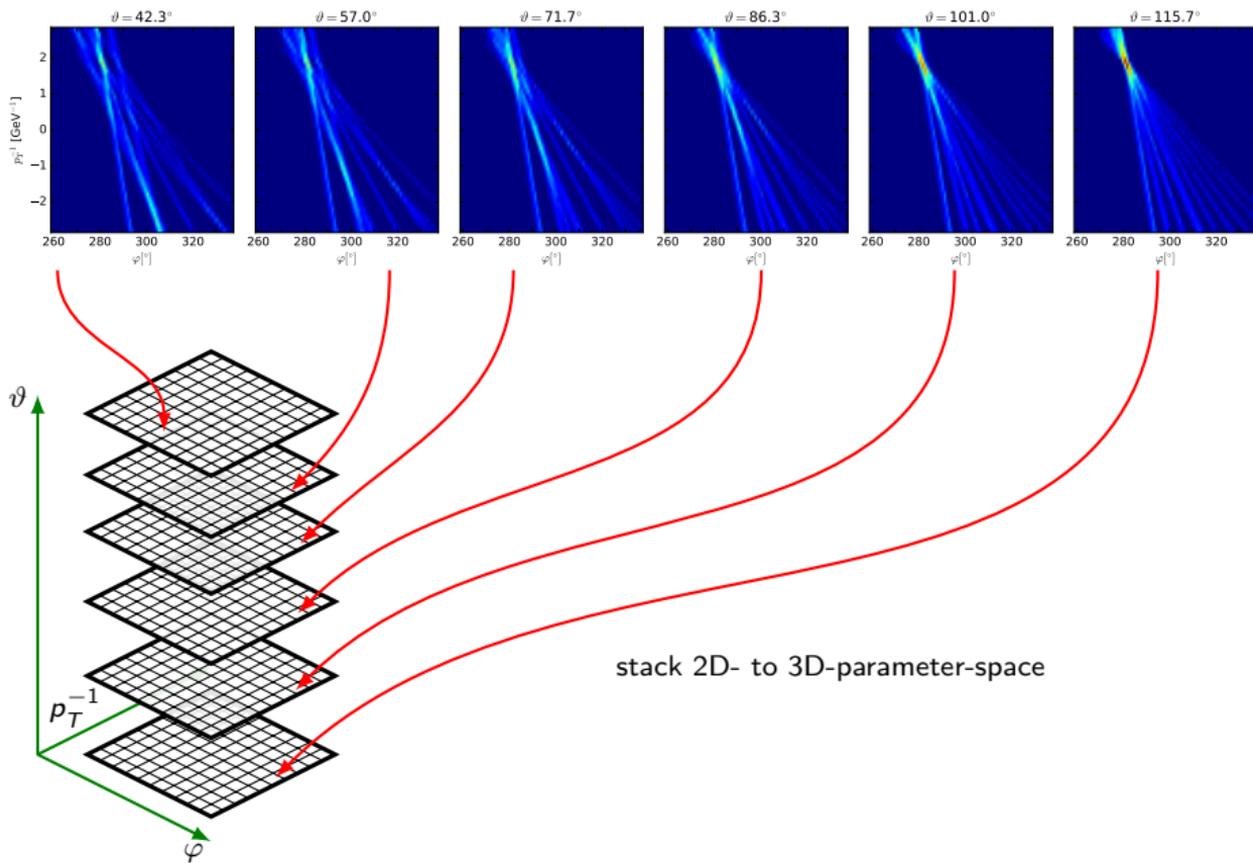
- ▶ axial hits appear as points
- ▶ stereo hits as line segments
- ▶ ϑ binning allows to represent stereo hits as points



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3D Hough Finding

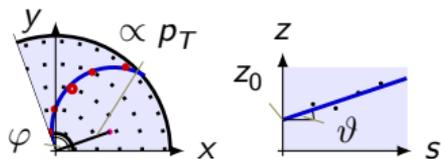


$$H(t|hits) = \sum_{h \in hits} P(t|h)$$

weights for all possible tracks t given a set $hits$.

Track Phase Space

- ▶ $p_T^{-1}, \varphi, \vartheta$

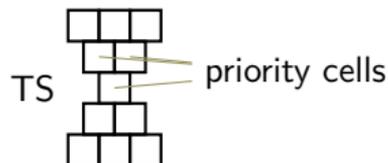


$P(t|h)$

- ▶ approximated by a 5D array A (stored as lookup table)
- ▶ A can be trained using Monte Carlo

Hit Phase Space

- ▶ TS-id, priority



	p_T	φ	ϑ	id	prio
bins	40	384	6	2336	3

Table: size of the array A

3D Finder Training



Filling

for each track

1. find related hits: h
2. bin track parameters: t
3. increment $A[t, h]$ for all pairs $[t, h]$

Normalization

normalize A for all tracks t (\equiv all tracks are equally probable)

$$A[t, h] = \frac{A[t, h]}{\sum_{\text{all } h} A[t, h]}$$

Set Bit Width

- ▶ adjust maximum bit width of each cell in A
- ▶ currently 3 bits are used



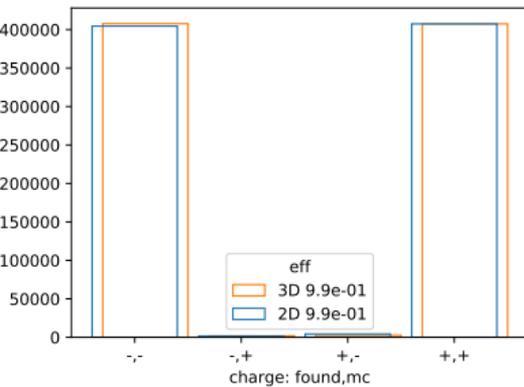
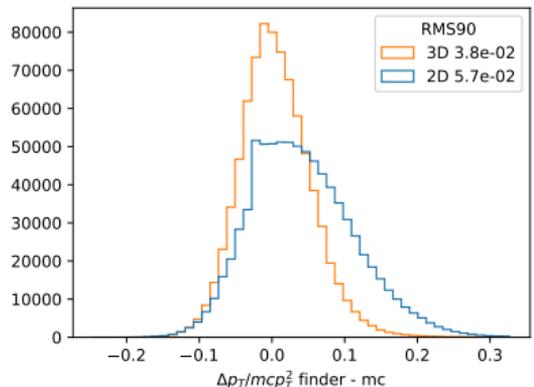
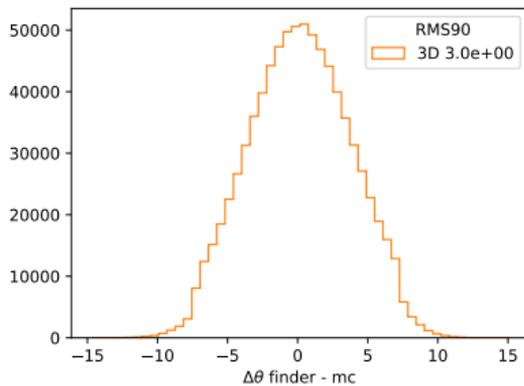
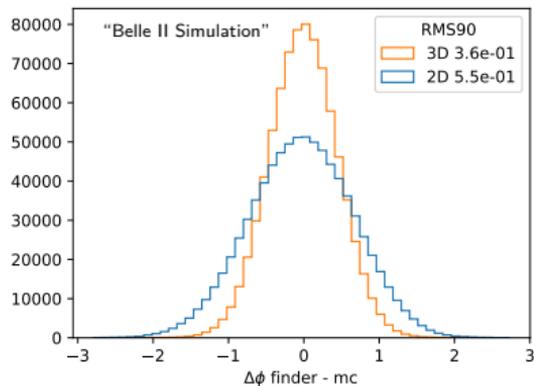
Construct “Houghplane”

$$H[\text{tracks}] = \sum_{h \in \text{hits}} A[\text{tracks}, h]$$

for an event with a set *hits*, *tracks* are peaks in *H*

Clustering and Track Estimation

1. find clusters
density based clustering algorithm (DBSCAN)
require cluster cells to have a minimum weight: minweight = 24
2. select contributing hits
hits with high weight contribution to the cluster
require a minimum number of hits related to a cluster: minhits = 4
3. calculate track parameters
weighted mean of cluster cells close to the peak
require cells to have a minimum weight >thresh × peakweight: thresh = 0.85

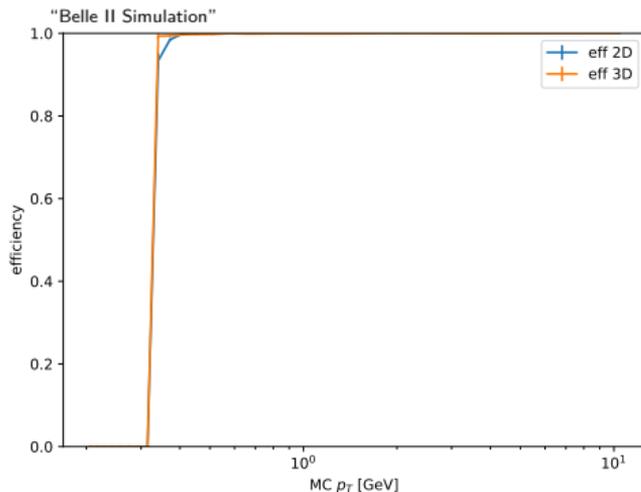


- ▶ single μ^\pm IP tracks

var	range
p_T	[0.35, 7] GeV
ϕ	[0, 360]°
z	[-1, 1] cm
ϑ	[35, 123]°

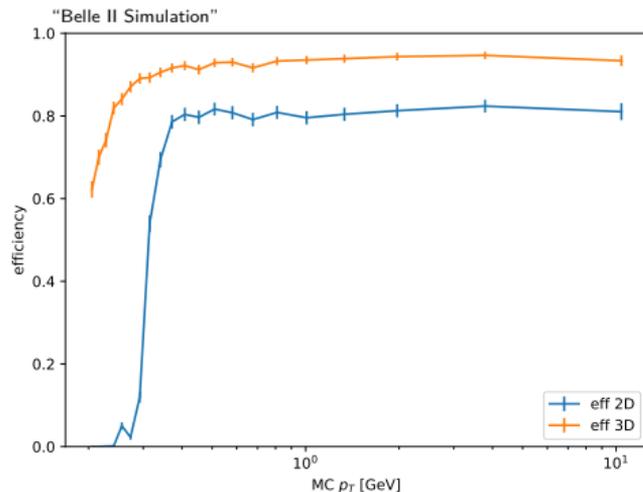
- ▶ 3D resolution $\Delta\vartheta \approx 3^\circ$

Track Finding Efficiency



- ▶ tracks in the acceptance region
- ▶ μ^\pm track parameters

p_T [GeV]	ϕ [°]	z [cm]	ϑ [°]
[0.35, 7]	360	[-1, 1]	[35, 123]



- ▶ tracks with low p_T and ϑ
- ▶ μ^\pm track parameters

p_T [GeV]	ϕ [°]	z [cm]	ϑ [°]
[0.2, 7]	360	[-1, 1]	[19, 140]

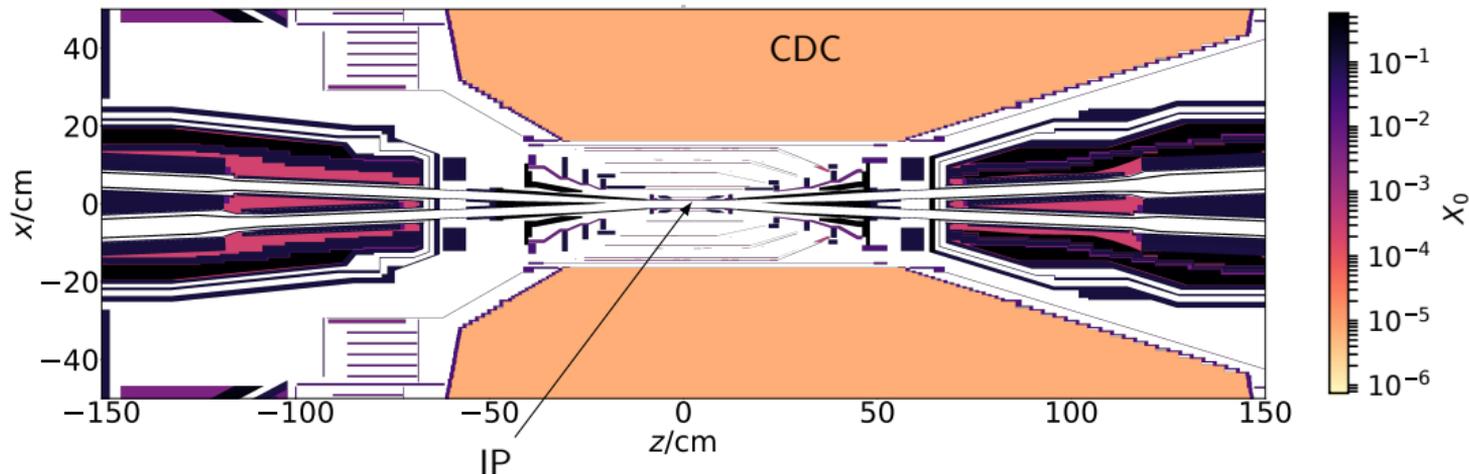


3D Track Finder

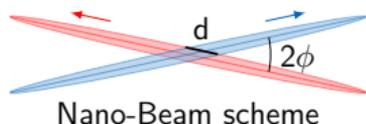
- ▶ high track finding efficiency
- ▶ improves efficiency for tracks with low p_T and flat θ angles
- ▶ improves 2D track parameter accuracy
- ▶ provides ϑ estimate
- ▶ directly relates stereo hits to tracks
- ▶ Hough map construction implemented in HW
- ▶ HW clustering under investigation

Backup

- ▶ scattering at material \rightarrow background tracks
- ▶ two separate rings with different energies

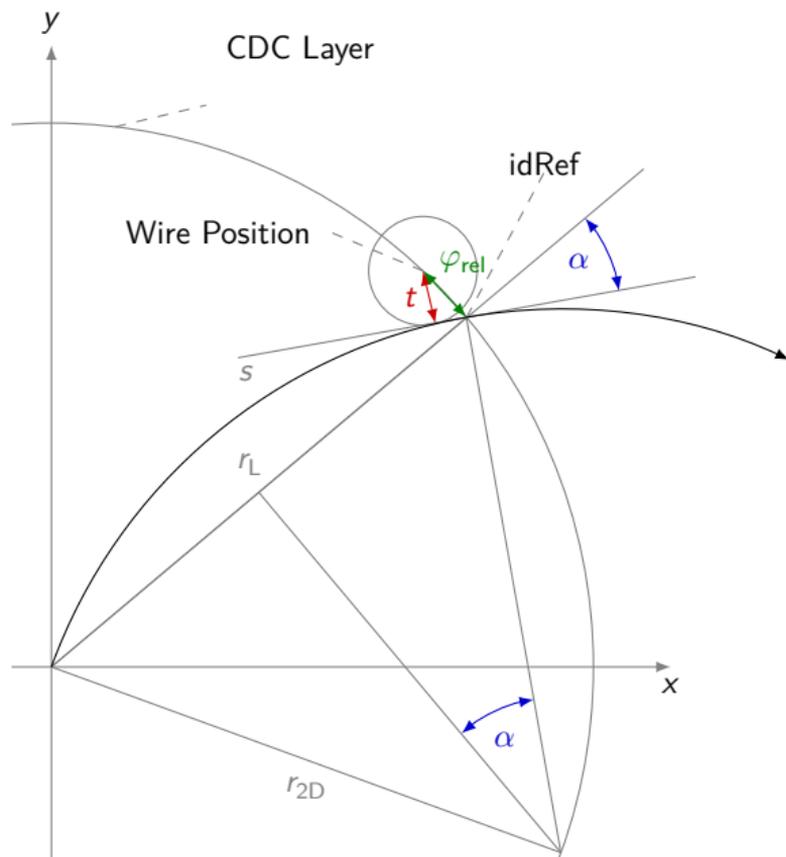


Bunch crossing:



$\sigma_{x,y,z}$	beam size	$\sigma_y \approx 60 \text{ nm}$
ϕ	crossing angle	$\phi \approx 40 \text{ mrad}$
$d = \frac{\sigma_x}{\phi}$	eff. bunch length	$d \approx 0.2 \text{ mm}$

NeuroTrigger - Input Representation

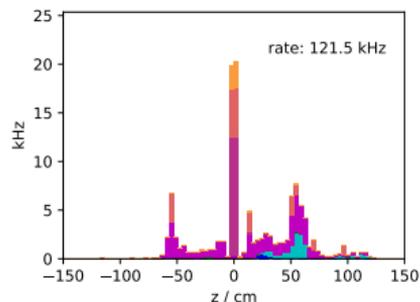
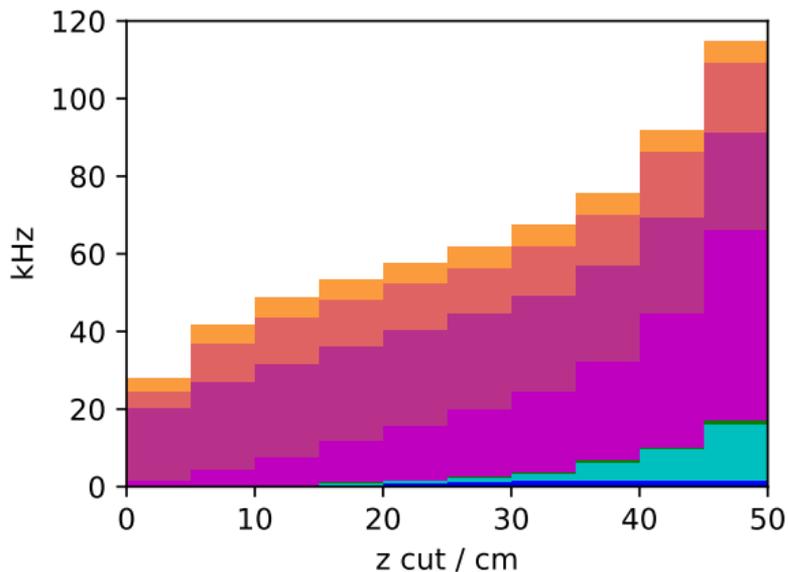


- ▶ $idRef$: crossing point of the track with the layer
- ▶ α : crossing angle of the track with the layer
- ▶ φ_{rel} : distance of the wire position to $idRef$
- ▶ t : drift time

Background - Suppression

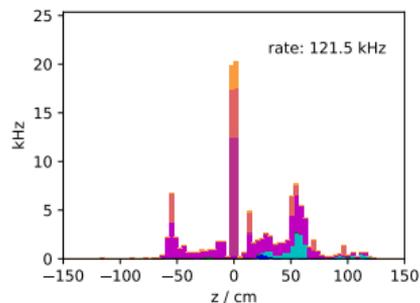
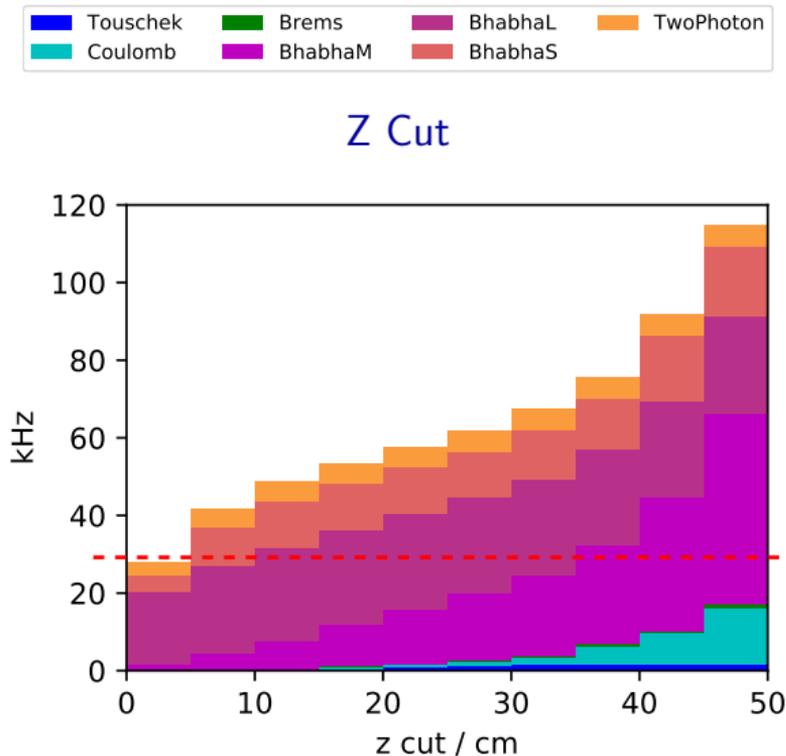


Z Cut



- ▶ cumulative bkg rate after a cut on the neural network z
- ▶ z_{cut} is varied in 5 cm steps

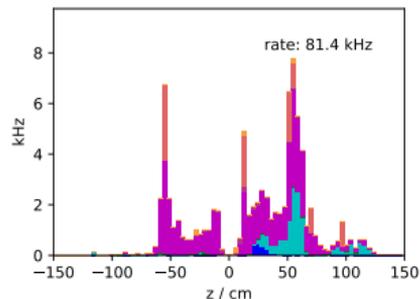
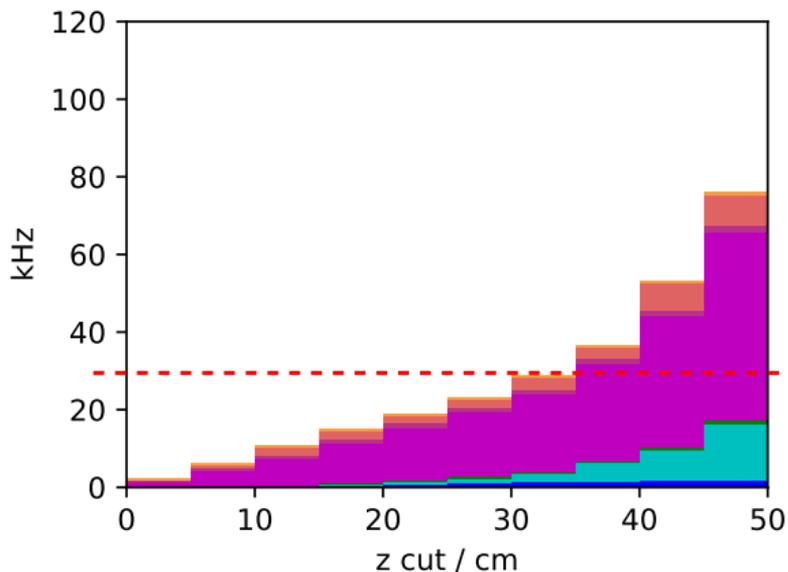
Background - Suppression



- ▶ cumulative bkg rate after a cut on the neural network z
- ▶ z_{cut} is varied in 5 cm steps



Z Cut (Tracks not from IP)



- ▶ only tracks with $|z_{MC}| \geq 1$ cm
- ▶ cumulative bkg rate after a cut on the neural network z
- ▶ z_{cut} is varied in 5 cm steps