

# A z-Vertex Trigger for Belle II

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# Outline

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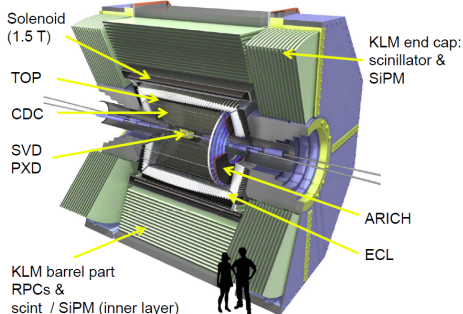
Efficiency Analysis

## Conclusion

## Neuro Team

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## Belle II Detector



# Introduction

## Goals

- ▶ build a z-vertex track trigger for the Belle II experiment
- ▶ achieve high precision (spatial resolution  $\approx 1$  cm)
- ▶ get a fast decision ( $< 1 \mu\text{s}$ )

## Methods

- ▶ CDC Track Segment data as input [IDs & clock cycle (2 ns timing)]
- ▶ current approaches:
  - ▶ MLP - Multi Layer Perceptron
  - ▶ Cascade prediction

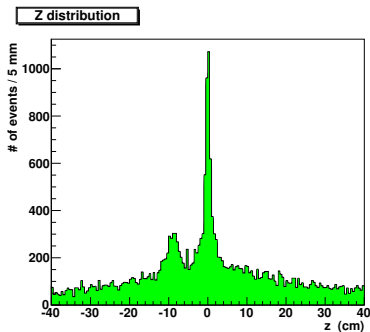


Figure: Offline z distribution in the Belle Experiment<sup>a</sup>.

a) T. Abe et al., *Belle II Technical Design Report*, KEK-REPORT-2010-1, arXiv:1011.0352v1 [physics.ins-det] (2010).

## Main approach

- ▶ sectorize input in  $(p_T, \phi, \theta)$ -Sectors
- ▶ many MLPs (one for each Sector)
- ▶ Hits are TS-IDs & drift times (2 ns resolution from TSF clock)

### Look Up Table - Bayes

$$P(\text{Sector}|\text{Hits}) = P(\text{Hits}|\text{Sector}) \cdot \frac{P(\text{Sector})}{P(\text{Hits})} \quad (1)$$

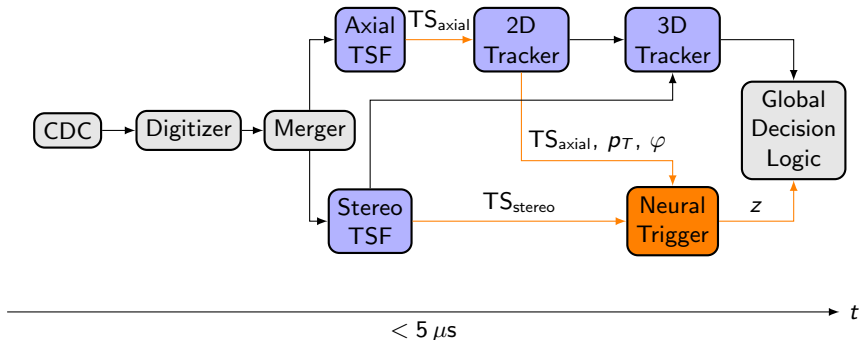
- ▶ modeled as nD array
- ▶ used to predict Sector & used to generate NN input

### MLP

$$z(\text{Hits}, p_T, \phi, \theta) = NN_{p_T, \phi, \theta}(\text{Hits}) \quad (2)$$

- ▶ output float value interpreted as scaled z-position

# Signal flow in the CDC Trigger



→ The neural network trigger will be implemented on a Virtex 7 FPGA

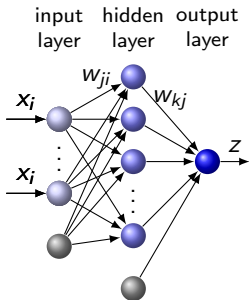
# MLP - Multi Layer Perceptron

## Motivation

- ▶ supervised machine learning
- ▶ universal real valued function approximation<sup>b</sup>
- ▶ short deterministic runtime

## Basic Setup

- ▶ one neuron:  $y = \tanh\left(\sum_{i=1} w_i \cdot x_i + w_0\right)$
- ▶ 3 layers, fully forward connected, hidden layer:  $N_{hidden} = 3 \cdot N_{input}$
- ▶ 1 output node, output value interpreted as scaled z-vertex position
- ▶ train size  $\approx N_{dof} \cdot 10$
- ▶ training with rprop algorithm (backpropagation)

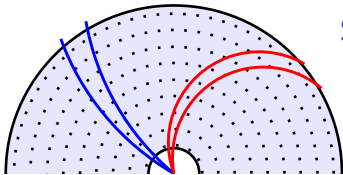


# Sectorization & TS selection

Sectorize input in  $p_T$ ,  $\phi$ ,  $\theta$

all 2336 Track Segments (TS) as input is too much for one MLP

- ▶ constraint in  $(\phi, \theta)$  to shoot always in the same direction
  - ▶ constraint in  $p_T$  to have the same curvature for the tracks
- subset of relevant TS for each  $(\phi, \theta, p_T)$  region



## Selection criteria

- ▶ require  $>p\%$  of the events to be covered by TS selection (chosen on a per superlayer basis)
- per sector  $\approx 20$  TS are used as input

Figure: Two different  $p_T - \phi$  sectors.

# “Relevant” Track Segments

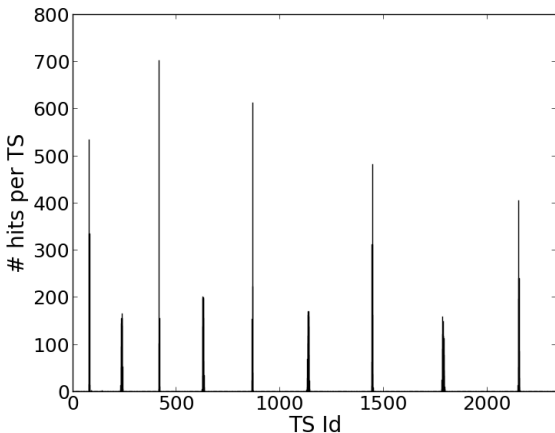


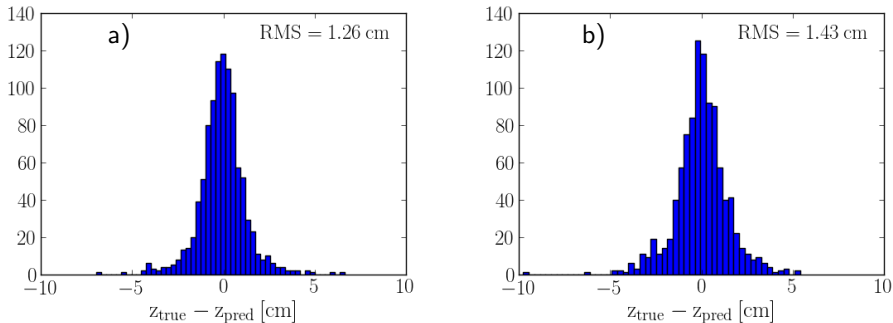
Figure: Hit distribution for events within a  $p_T - \phi$  sector.

## Track Parameters

$p_T \in [1.43, 1.67] \text{ GeV}$	$\phi \in [180, 181]^\circ$	$\theta \in [35, 123]^\circ$	$z \in [-50, 50] \text{ cm}$
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# “Expert” MLP - Capabilities

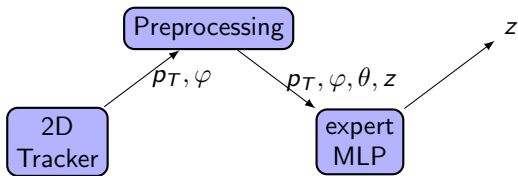


**Figure:** z-vertex prediction with an “expert” MLP in two  $p_T$  regions with  $\phi \in [180, 181]^\circ$ ,  $\theta \in [56, 62]^\circ$  and  $z \in [-10, 10]$  cm.

a)  $p_T \in [1.6, 1.67]$  GeV. b)  $p_T \in [4.2, 4.8]$  GeV.

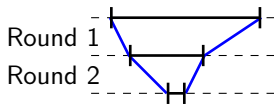
- ! high accuracy on the z-vertex within a small sector
- construct one MLP for each sector (“expert”)
- find optimal solution to determine the sectors (preprocessing)

# Preprocessing



**Figure:** The  $p_T - \phi$  from the 2D Tracker is enriched with  $\theta - z$  information by the preprocessing to allow the selection of an “expert” MLP.

- ▶  $\mathcal{O}(10^6)$  sectors in total
- ▶ preprocessing determines correct sector
- ▶ “expert” MLP provides  $z$ -vertex value
- ▶ MLP is main candidate for preprocessing
- ▶ several rounds in the preprocessing might be required (ideal: 1 round)



**Figure:** Incrementally shrink the sector size in the preprocessing.

# $\theta$ Sector Finding with a MLP

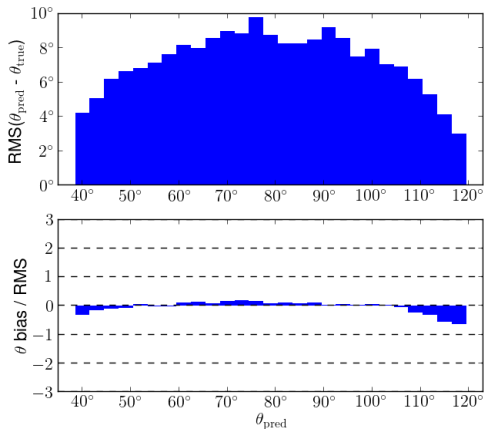


Figure: RMS of  $\theta$  prediction in  $\theta$ -bins.

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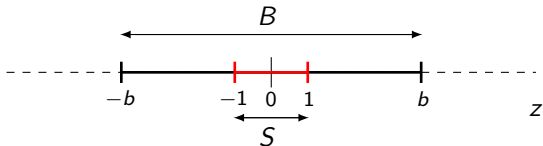
- ▶ bins are assigned using the prediction  $\theta_{\text{pred}}$
- RMS is  $\theta$  dependent:  $\text{RMS}(\theta)$
- MLP suitable for  $\theta$  prediction

## z Sector Finding with a MLP - Efficiency Analysis

a high efficiency for Interaction Region (IR) events is required  
 → for the experiment, define two z-bins around IR ( $z = 0$  cm):

small  $S = [-1, 1]$  cm

big  $B = [-b, b]$  cm



$b$  is varied in the experiment

### Contingency Table

	$z_{\text{true}} \in S$	$z_{\text{true}} \notin S$
$z_{\text{pred}} \in B$	TP	FP
$z_{\text{pred}} \notin B$	FN	TN

Table: The prediction  $z_{\text{pred}}$  and true value  $z_{\text{true}}$  are evaluated in two different bins.

### Efficiency

► efficiency

$$EFF = \frac{TP}{TP + FN} \quad (3)$$

► false positive rate

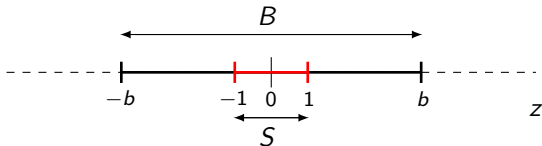
$$FPR = \frac{FP}{FP + TN} \quad (4)$$

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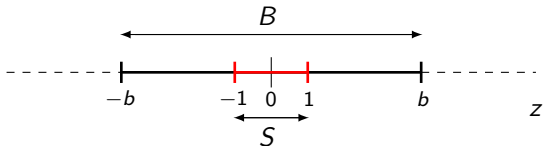
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## 3 Round $z$ Prediction Chain Experiment

In each round  $b$  is set such that  $EFF > 99\%$

→ use more specialized predictor in next round

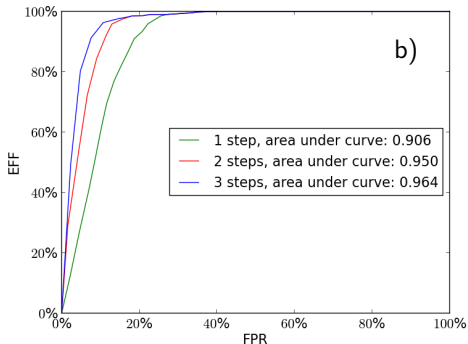
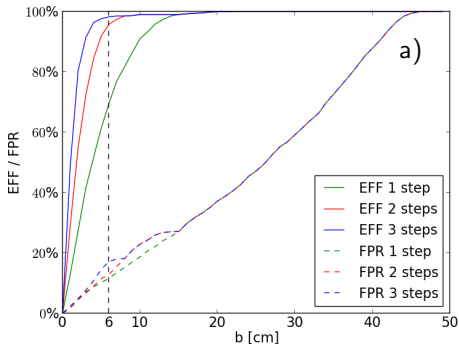
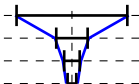


Figure: a)  $EFF$  and  $FPR$  of all 3 rounds vs. the half bin size  $b$ .

b)  $EFF$  vs.  $FPR$  as overall measure of the classification.

### Track Parameters

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# Conclusion

## “expert” MLP

- ▶ “experts” are specialized to phase space sectors
- ▶ high  $z$ -resolution with  $RMS < 2$  cm
- ▶ sectorized solution requires  $\mathcal{O}(10^6)$  “expert” MLP

## Preprocessing

- ▶ preprocessing necessary for the sector finding
- ▶ MLP is good candidate for preprocessing  
(can predict  $\theta$  and  $z$  in given  $p_T - \phi$  sectors)

## Outlook

- ▶ need generalization experiment to full detector acceptance region
- ▶ for hyper parameters and efficiency in the preprocessing further optimization is required
- ▶ Virtex 7 FPGA implementation anticipated



Thank you for your attention!