

# Search for $Z_c(3900)$ via Quantum Computing Machine Learning

Pan Huang

August 26, 2024

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# Introduction

- $Z_c(3900)$  decay chain  
 $e^+e^- \rightarrow Z_c(3900)^\pm \pi^\mp$   
 $Z_c(3900)^\pm \rightarrow J/\psi \pi^\pm$   
 $J/\psi \rightarrow e^+e^-(\mu^+\mu^-)$

- Signal MC sample (BOSS 7.0.3)

decay tree	decay model
$e^+e^- \rightarrow Z_c(3900)^\pm \pi^\mp$	PHSP
$Z_c(3900)^\pm \rightarrow J/\psi \pi^\pm$	PHSP
$J/\psi \rightarrow e^+e^-(\mu^+\mu^-)$	VLL

- Data Sample (BOSS 7.0.3)

## 4.260 GeV Data at BESIII of 2013

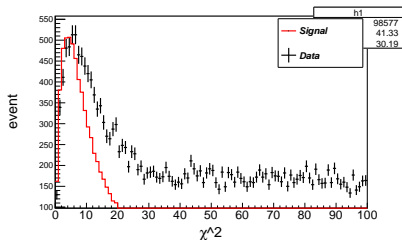
sample	luminance	center-mass energy	Run number
4260	$828.4 \pm 0.1 \pm 5.5$	$4257.97 \pm 0.04 \pm 0.66$	29677-30367 31561-31981

# Classical Data Processing Procedure

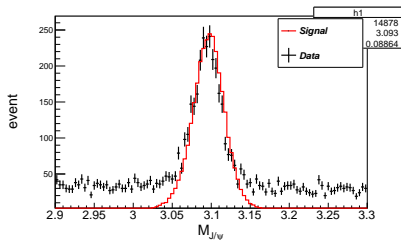
- Good charged track selection
  - $|\cos\theta| < 0.93, |V_z| < 10\text{cm}, |V_r| < 1\text{cm}$
  - Four good tracks and zero net charge
- particle identification
  - $p > 1\text{ GeV}/c$  identified as lepton
  - $p < 1\text{ GeV}/c$  identified as  $\pi$
  - The number of pions and that of leptons should be two in each event with zero net charge.
  - $E_{EMC} > 1.1\text{ GeV}$  identified as e
  - $E_{EMC} < 0.35\text{ GeV}$  identified as  $\mu$
- remove gamma-conversion background
  - $\cos(\pi^+\pi^-) < 0.98$
  - $\cos(\pi^\pm e^\mp) < 0.98$

# Classical Data Processing Procedure

- 4C kinematic Fit:  $\chi^2 < 60$
- $J/\psi$  mass window :  
 $3.08 < M(I^+I^-) < 3.12 \text{ GeV}/c^2$
- sideband :  
 $3.0 < M(I^+I^-) < 3.06$  and  $3.14 < M(I^+I^-) < 3.20 \text{ GeV}/c^2$



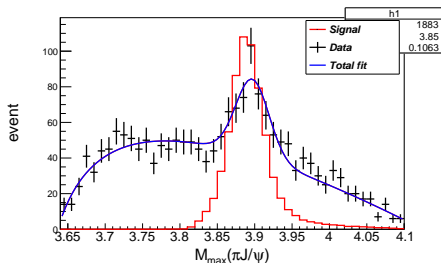
4C Kinematic fit  $\chi^2$  distribution



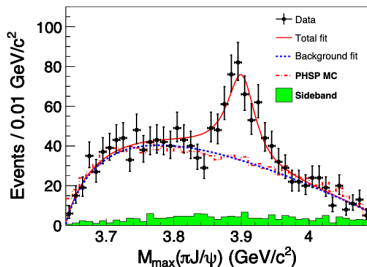
$M(I^+I^-)$  invariant mass distributions.

# Classical Data Processing Procedure

We use events from the  $J/\psi$  mass window and select the maximum of  $M(\pi^+ J/\psi)$  and  $M(\pi^- J/\psi)$  as variables, then fit this to obtain the signal



$M_{\max}(\pi J/\psi)$  invariant mass distributions.



$M_{\max}(\pi J/\psi)$  @BESIII

# Quantum Transform Introduction

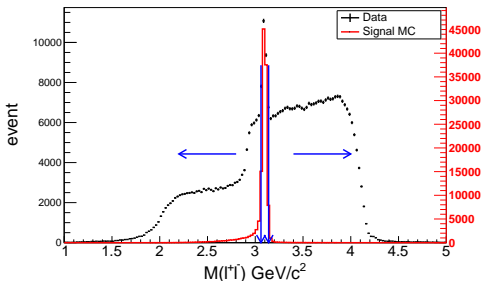


## Preliminary Selection

- Four good tracks, zero net charge
- $\pi$ :  $p < 1 \text{ GeV}/c$
- lepton :  $p > 1 \text{ GeV}/c$ 
  - $e$  :  $p > 1 \text{ GeV}/c$  and  $E_{EMC} > 0.8 \text{ GeV}$
  - $\mu$ :  $p > 1 \text{ GeV}/c$  and  $E_{EMC} < 0.6 \text{ GeV}$

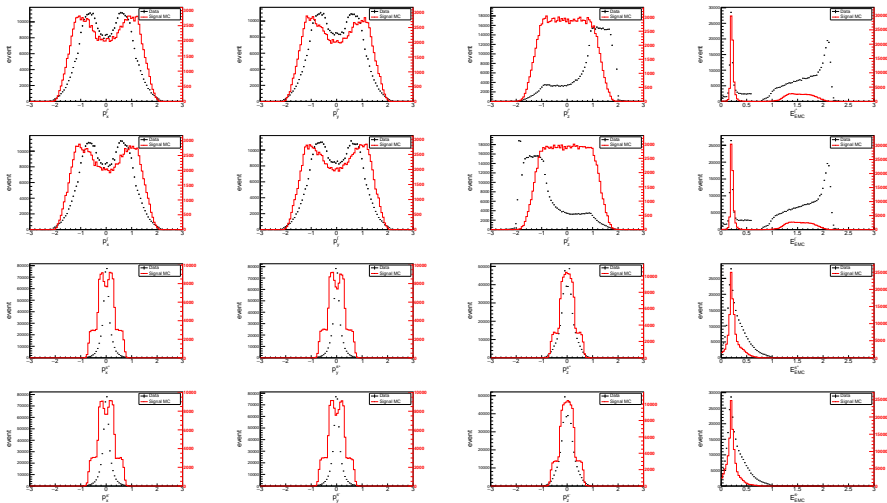
# Quantum Transform

- Model : Transform and Quantum Transform
- Parameter Set: 17 variables
  - 16 variables:  $p_x, p_y, p_z$  and  $E_{EMC}$  of 4 charge tracks
  - $\chi^2$  (4C kinematic fit)
- Signal : Signal MC
- Background is from data :  
 $M(I^+I^-) \in (0, 3.06) \cup (3.14, 5) \text{ GeV}/c^2$

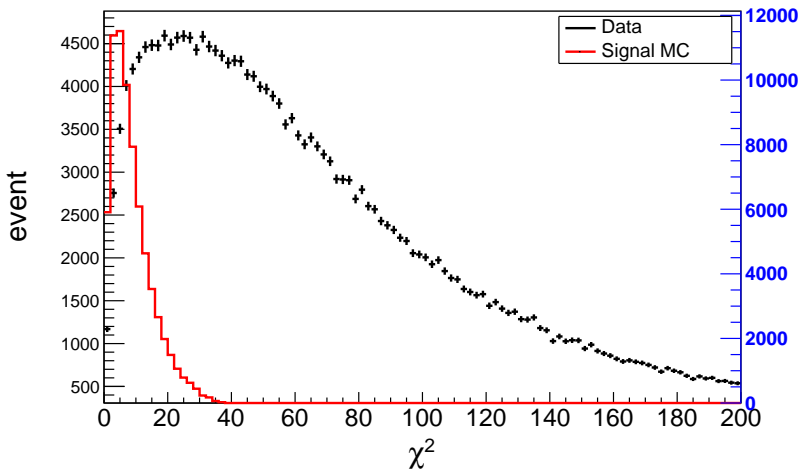


- Signal: background = 1:1
- Classical Transform :
  - 130k events (65k for train, 65k for validation)
  - 300 epochs
  - Learning rate : 0.0007 or 0.0009
- Quantum Transform :
  - 20k events (10k for train , 10k for validation)
  - 8 epochs
  - Learning rate : 0.0007 or 0.0009

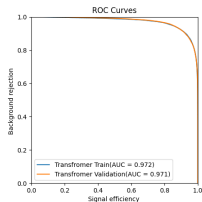
# Distribution of variables



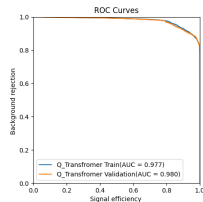
# Distribution of variables



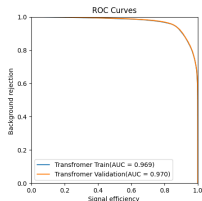
# ROC Curve



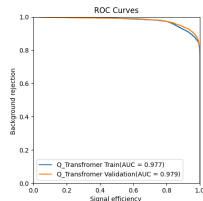
Classical, Lr: 0.0007



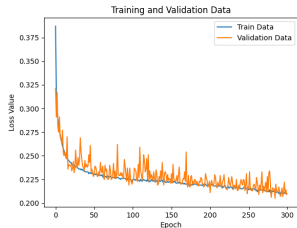
Quantum, Lr: 0.0007



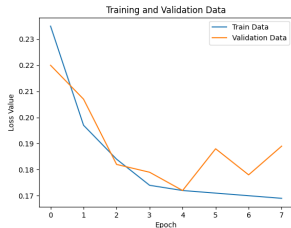
Classical, Lr: 0.0009



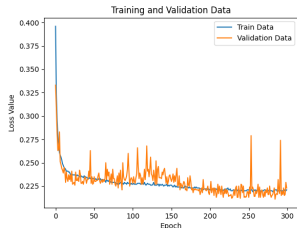
Quantum, Lr: 0.0009



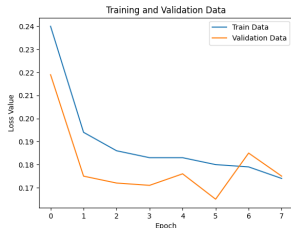
Classical, Lr: 0.0007



Quantum, Lr: 0.0007

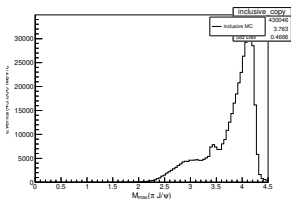


Classical, Lr: 0.0009

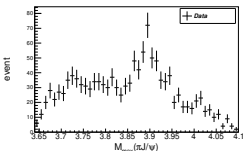


Quantum, Lr: 0.0009

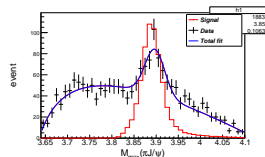
# Real data distribution



data with preliminary cut



data with all cut

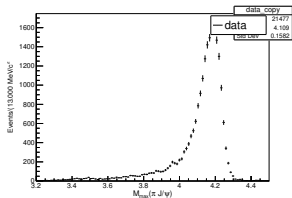


Signal with all cut

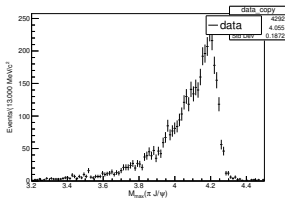


# Test model

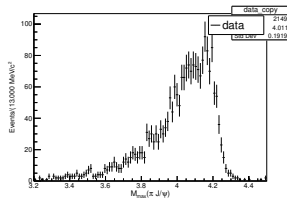
We apply the model to real data and set different background rejection rates.



bkg rejection rate :  
95%

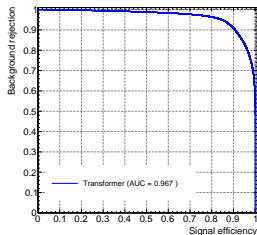


bkg rejection rate :  
99%

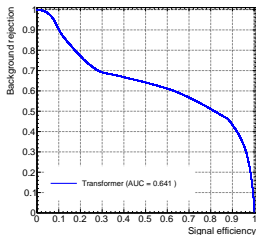


bkg rejection rate :  
99.5%

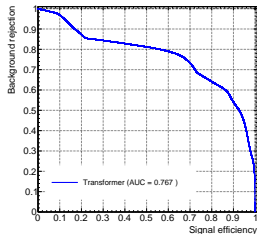
# ROC curve



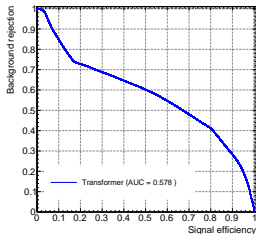
Sig:130k; data:130k



Sig:130k; data:250k



Sig:130k; data:200k



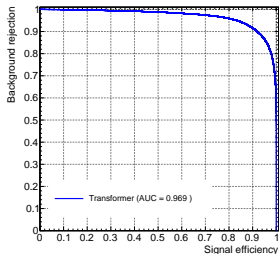
Sig:130k; data:430k

# ROC curve

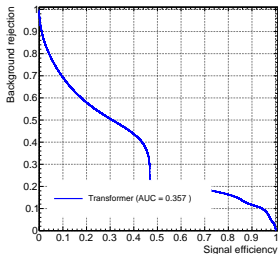
We divided the data into three parts and then applied the model to each part separately.

Signal : 130k

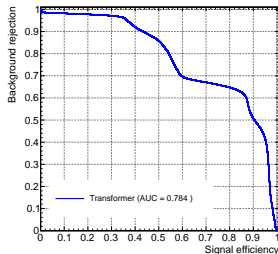
Data : 130k



Part 1 : 130k



Part 2 : 130k



Part 3 : 130k

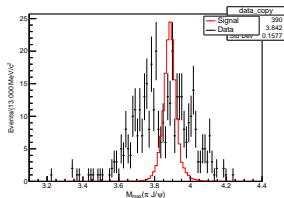
# Test model

We divided the data into three parts and then applied the model to each part separately.

Signal : 130k

Data : 130k

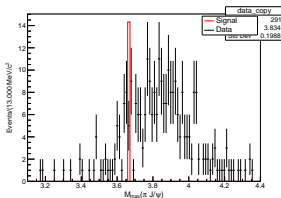
Background rejection rate :99.7%



Part 1

Data : 390

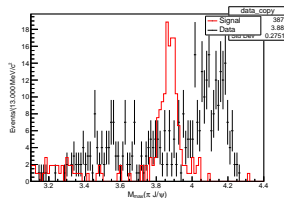
Signal :25173



Part 2

Data : 291

Signal : 1



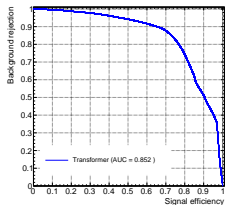
Part 3

Data : 387

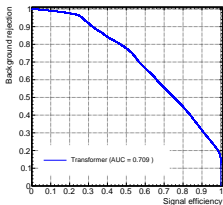
Signal : 210

# Test model

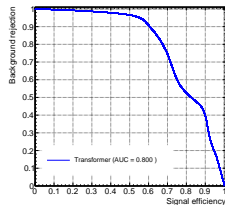
We reduce the number of epoch to train the model, and then apply the model.



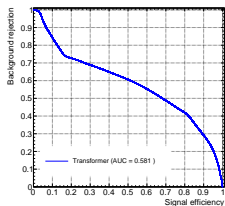
epoch :10



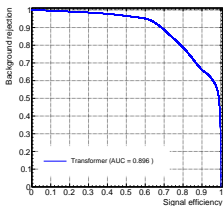
epoch :20



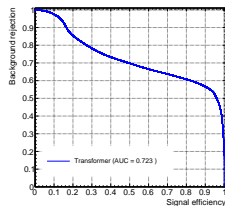
epoch :30



epoch :40



epoch :50



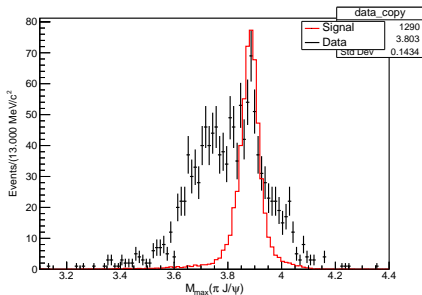
epoch :60

# Test Model

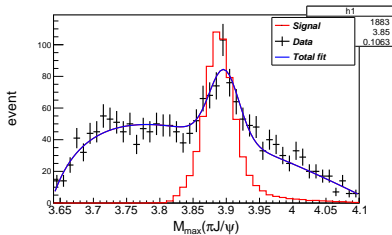
Signal : 130k

Data : 430k

Background rejection rate :99.7%



Epoch : 50  
Data : 1290  
Signal :20146



by Cut  
Data : 1883  
Signal : 55474

Transform 1 : use 130k sideband data to train.

Transform 2 : use all sideband data (400k) to train.

	<b>Transform 1</b>	<b>Transform 2</b>	<b>Cut</b>
signal	7442(5.7%)	20146(15.5%)	55474(42.6%)
Data	1290(99.7%)	1290(99.7%)	1883(99.5%)