

# The rediscovery of Zc(3900) with QSVM

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2026.04.16

# Outline

- Background Introduction
- QSVM/SVM Study
  - Dataset selection
  - Training and optimization
  - Results
- Summary and Outlook

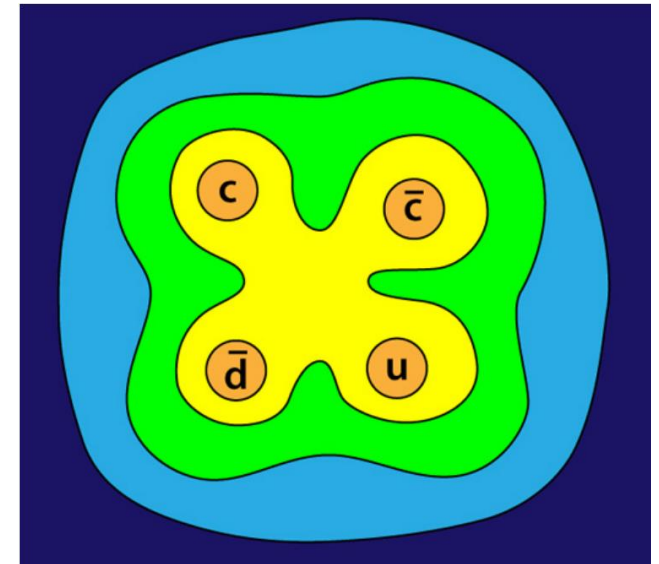
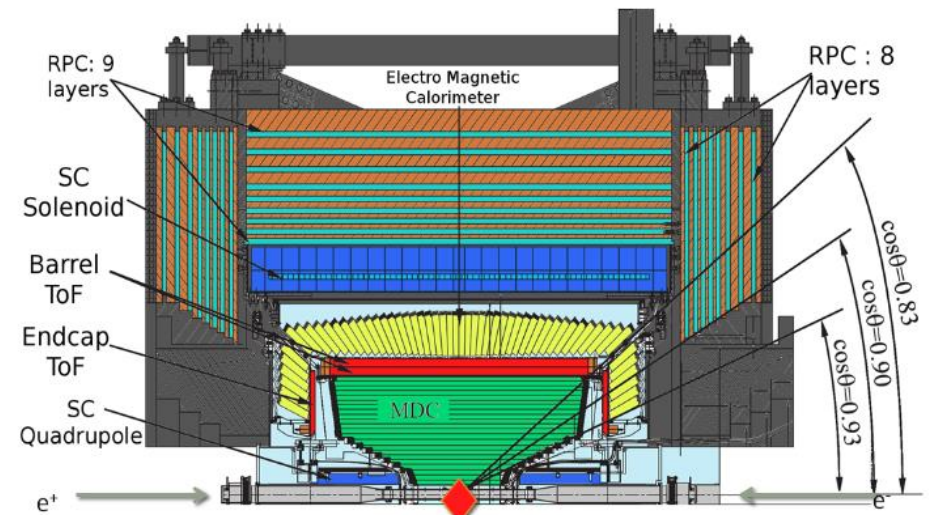
# Zc(3900) @ BESIII

- The BESIII experiment

- BEPCII running at a cms of 2.0-5.0GeV
- Sub-detectors: MDC, TOF, EMC, MUC
- $\tau$ -charm physics

- The discovery of Zc(3900)

- Initially discovered by BESIII at 2013
- Possibly the first tetra-quark state observed in experiment
- 美国物理学会《物理》2013年国际物理学领域十一项重要成果之首
- 2013年度“中国科学十大进展”



# The Traditional Physics Analysis Method

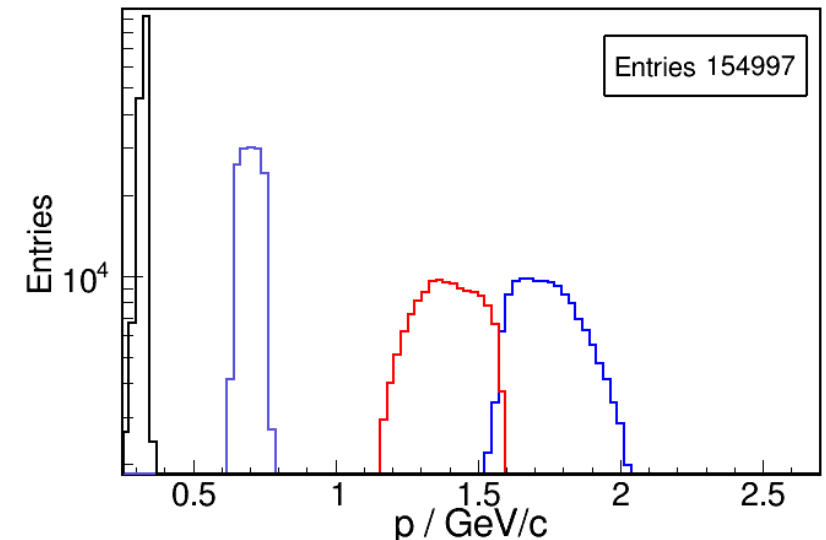
- Initially observed via the following decay mode with 4.26GeV data

$$e^+e^- \rightarrow Zc(3900)\pi^\pm$$

$$Zc(3900) \rightarrow J/\psi \pi^\mp$$

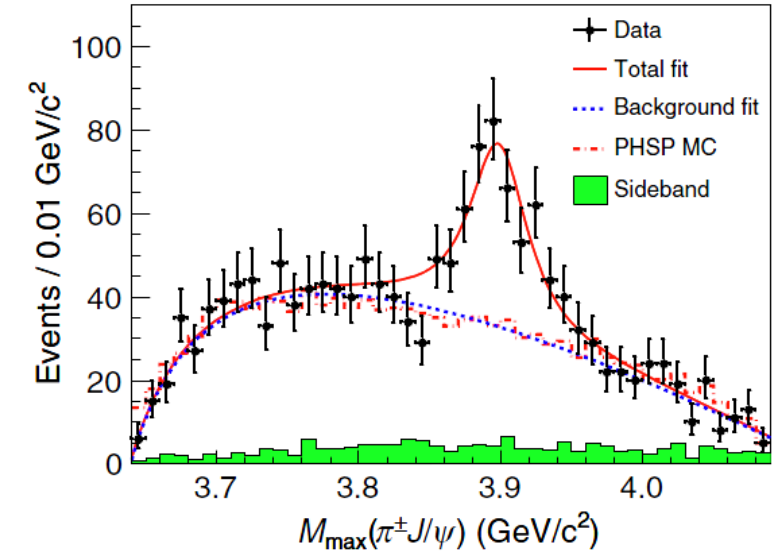
$$J/\psi \rightarrow e^+e^- \quad \text{or} \quad J/\psi \rightarrow \mu^+\mu^-$$

- The final state consists of 4 charged tracks
  - The leptons and pions are separated clearly by the momentum
  - As shown in the picture (MC data)



# Cut-based Event Selection

- 4 good charged tracks with netcharge = 0
  - Sorted in descending order by momentum
  - Track 1 and 2 are assumed to be the leptons
  - Momentums
    - The 2 leptons:  $p_1 > 1.0\text{GeV}$ ,  $p_2 > 1.0\text{GeV}$
    - The 2 pions:  $p_3 < 1.0\text{GeV}$ ,  $p_4 < 1.0\text{GeV}$
  - Energy deposition in EMC
    - ee mode:  $E_{1\text{emc}} > 1.1\text{GeV}$  &  $E_{2\text{emc}} > 0.8\text{GeV}$ , muon mode:  $E_{1\text{emc}} < 0.35\text{GeV}$  &  $E_{2\text{emc}} < 0.35\text{GeV}$
  - Angles between the tracks
    - For ee mode:  $\cos(\theta_{e^+\pi^-}) < 0.98$  &&  $\cos(\theta_{e^-\pi^+}) < 0.98$  &&  $\cos(\theta_{\pi^+\pi^-}) < 0.98$
    - For muon mode:  $\cos(\theta_{\pi^+\pi^-}) < 0.98$
- Kinematic fit  $\text{chisq} < 60$
- $3.077 < \text{InvariantMass}(1, 2) < 3.117$



# Support Vector Machine (SVM & QSVM)

- Supervised machine learning for binary classification
- Find the optimal hyperplane that separates data points in the feature space

$$\frac{1}{2} \|w\|^2 \quad \text{subject to } y_i(w \cdot x_i + b) \geq 1, \dots, m,$$

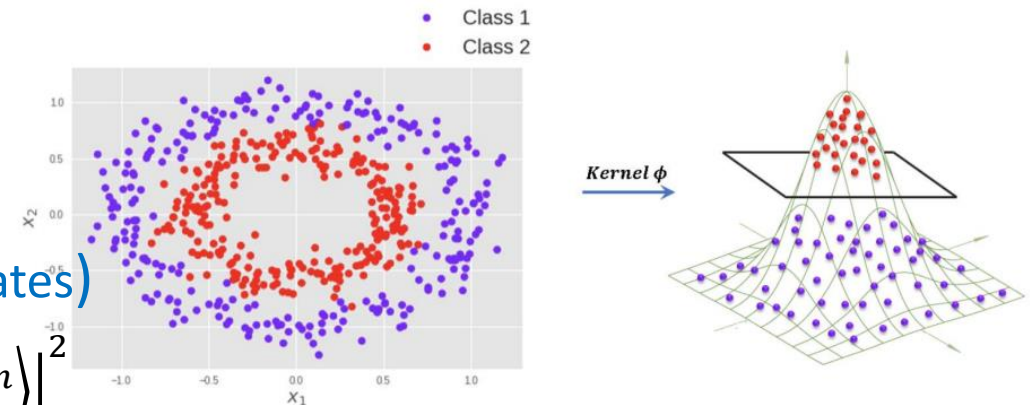
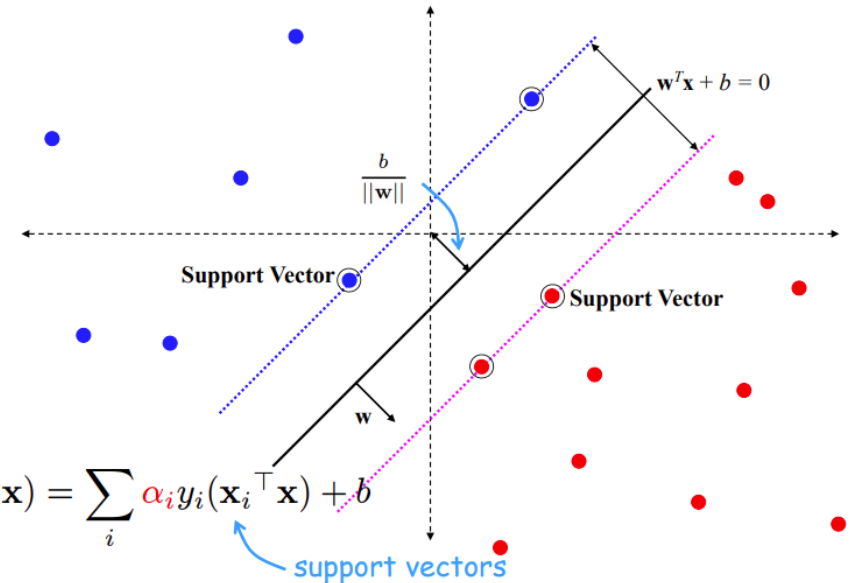
- Kernel method is used to map the original features to a higher dimensional feature space to separate non-linear data

- Traditional kernel

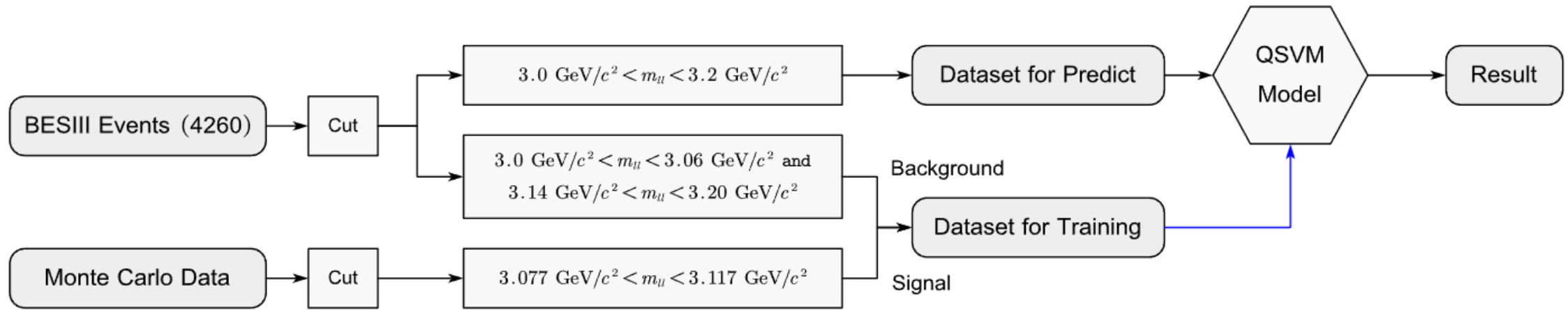
- Linear kernel, polynomial kernel, Gaussian kernel

- Quantum kernel (the inner product of the quantum states)

$$K(\vec{x}_i, \vec{x}_j) = |\langle \Phi(\vec{x}_i) | \Phi(\vec{x}_j) \rangle|^2 = \left| \langle 0^{\otimes n} | u_{\Phi(\vec{x}_j)}^\dagger u_{\Phi(\vec{x}_i)} | 0^{\otimes n} \rangle \right|^2$$



# Event Classification with QSVM (and SVM)



## 1. Event preselection

- 4 good charged tracks with netcharge = 0 (the same as the traditional cut-based method)
- Kinematic fit  $\text{chisq} < 200$
- $3.0 < \text{InvariantMass}(1, 2) < 3.2$

## 2. Training dataset selection

## 3. QSVM/SVM model training and optimization

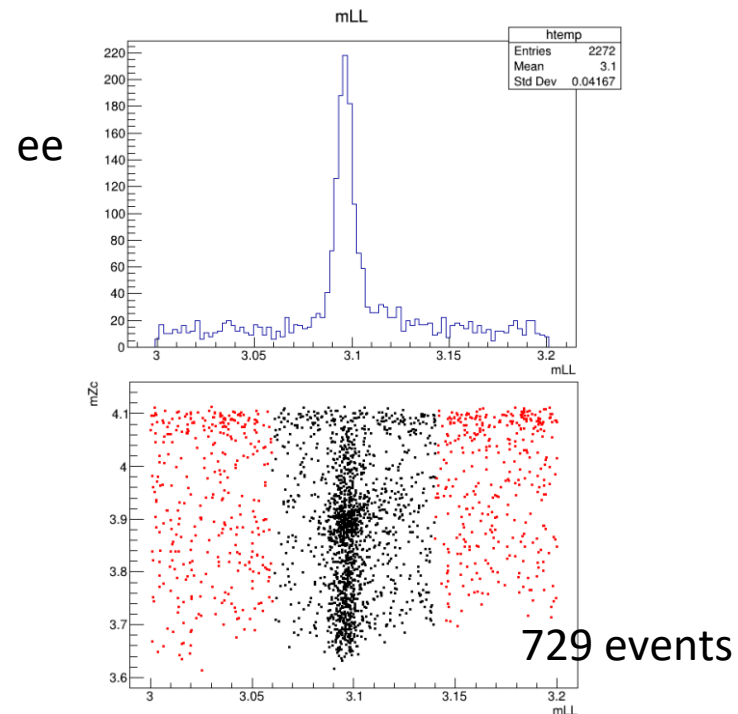
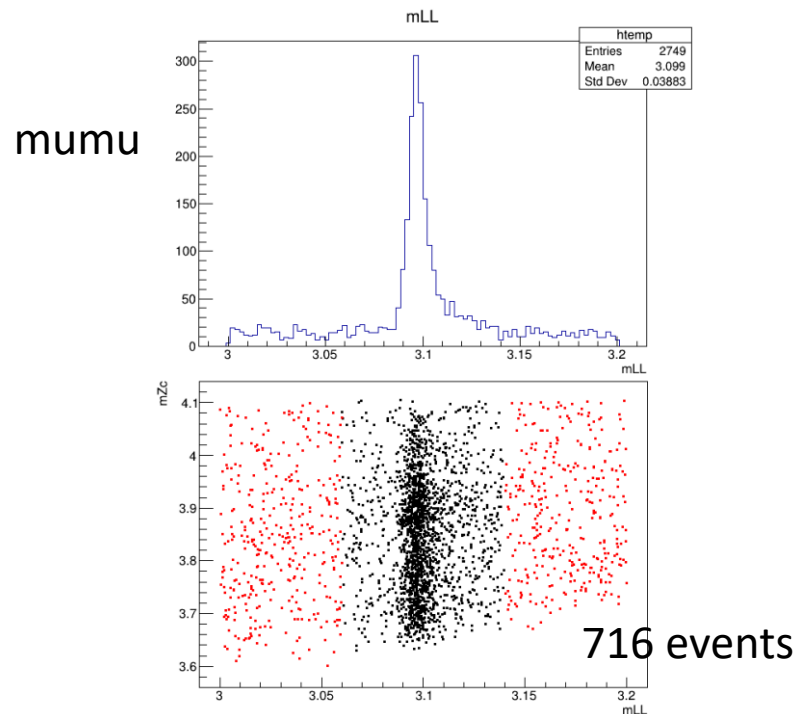
## 4. Apply the trained model to the real data

- The events after event preselection and  $m_{j/\psi}$  cut

Traditional Computing  
+  
Quantum Machine Learning

# Dataset Selection

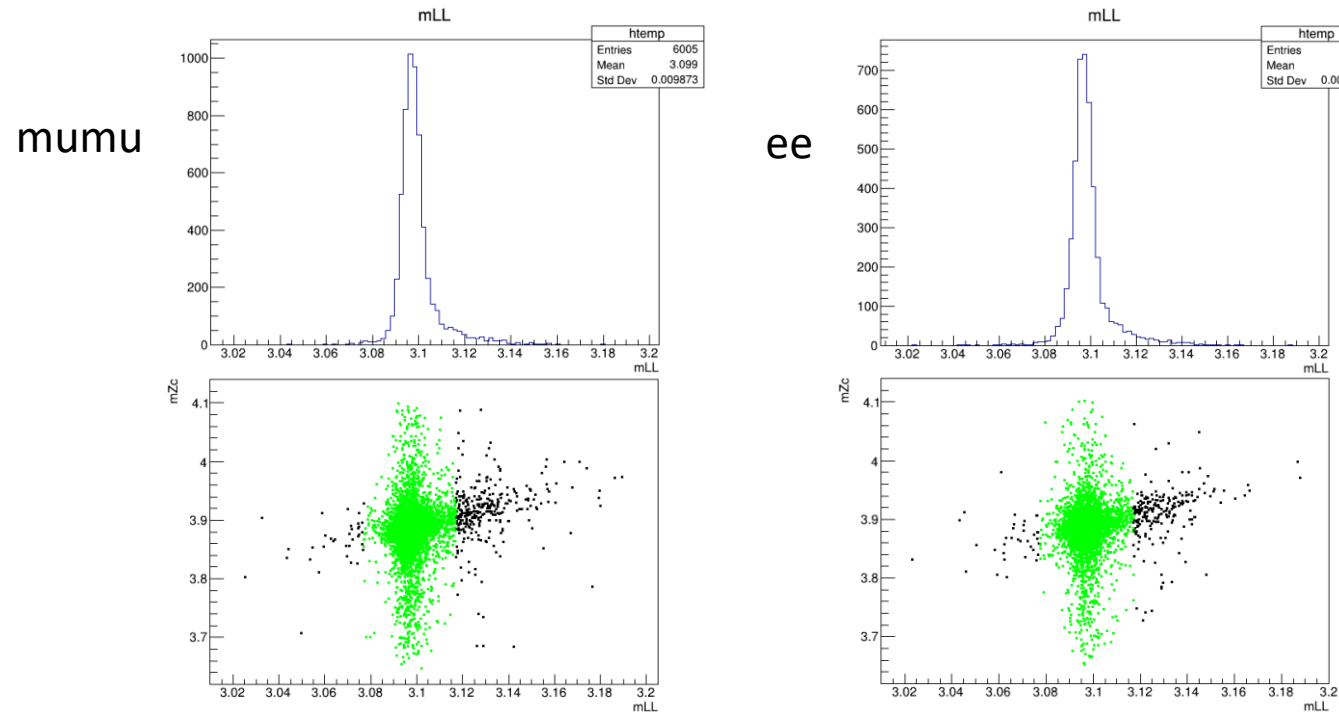
- Background dataset
  - The background can not be well described by the inclusive MC data
  - The background dataset is selected from the real data that after event preselection
  - Limit  $m_{J/\psi}$  in the range of (3.0, 3.06) and (3.14, 3.20) GeV



# Dataset Selection

- Signal dataset
  - Zc(3900) MC data after event preselection
  - And no Zc(3900) MC data (PHSP) after event preselection
  - Limit  $m_{J/\psi}$  in the range of (3.077, 3.117) GeV

$$e^+ e^- \rightarrow J/\psi \pi^+ \pi^-$$



Zc(3900) and no Zc(3900) events are mixed 1:1  
Signal and background events are mixed 1:1  
Randomly select 700 + 700 for each mode  
50% **training** dataset + 50% **testing** dataset

# Data Features and Encoding

- Totally 28 data features are considered to classify the signal and background events

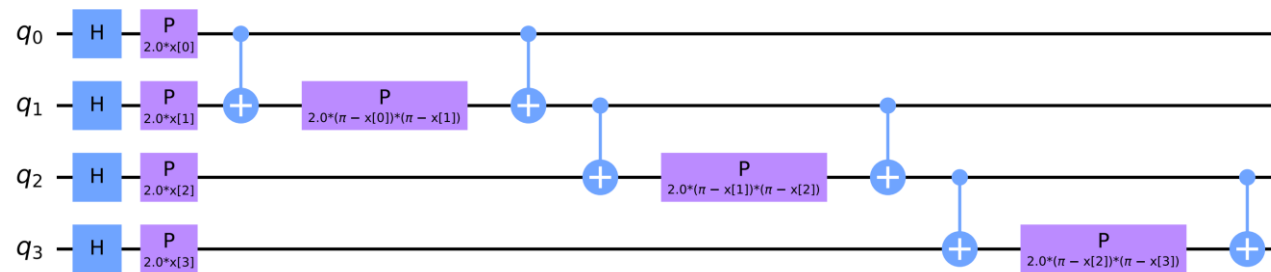
$P, P_x, P_y, P_z$  of the 4 final particles

$E_{emc}$  of the 4 final particles

$\theta_{12}, \theta_{13}, \theta_{14}, \theta_{23}, \theta_{24}, \theta_{34}$

$N_\gamma, E_\gamma$

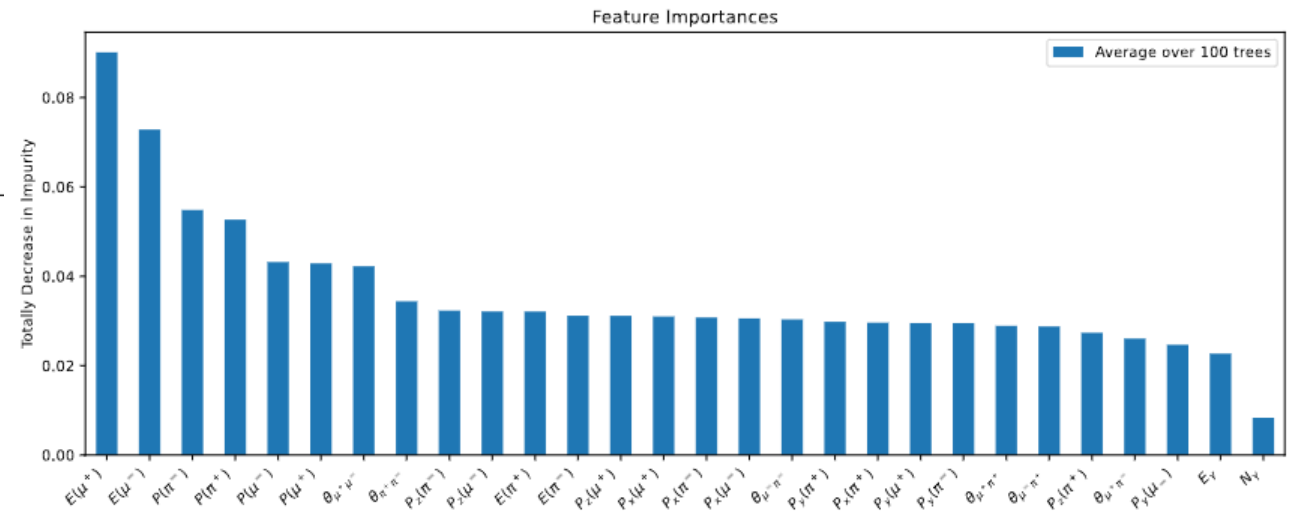
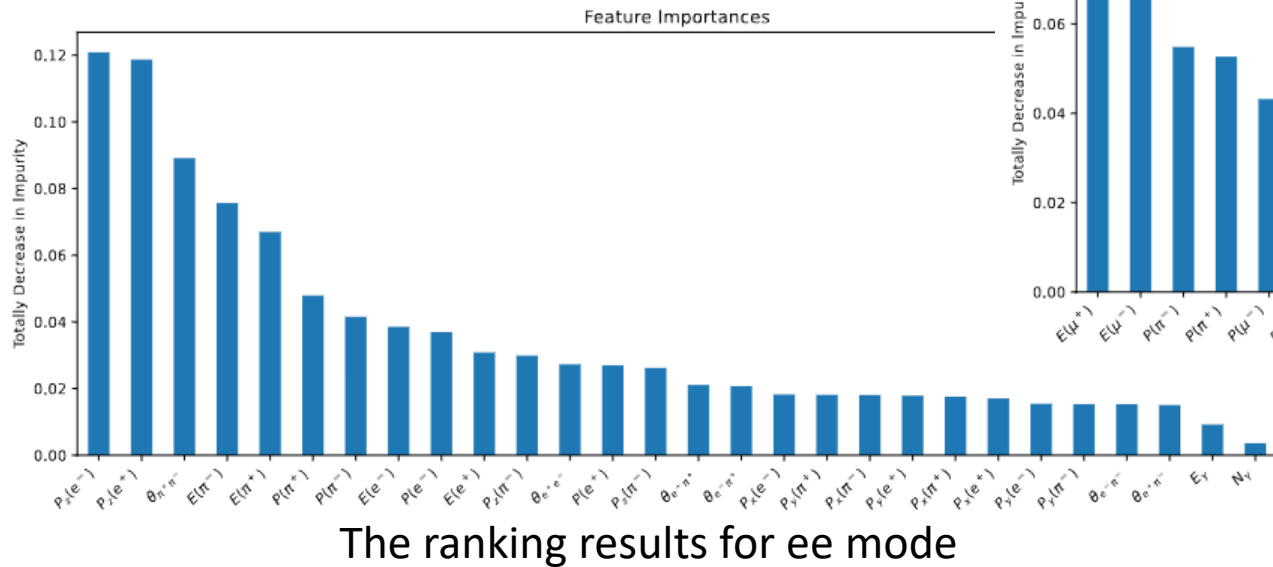
- The embedded ZZFeatureMap in **Qiskit** is used for data encoding



An example of a 4 qubits ZZFeatureMap quantum circuit

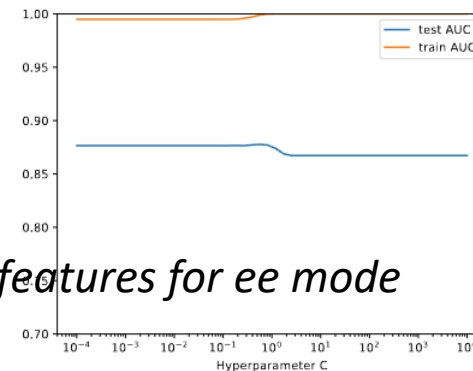
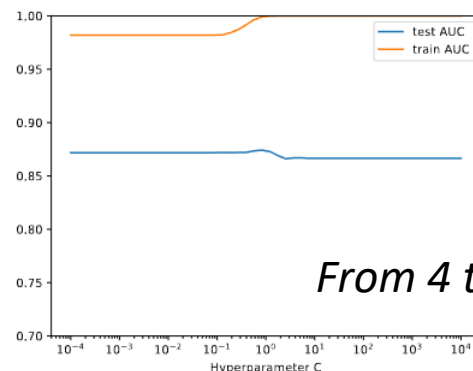
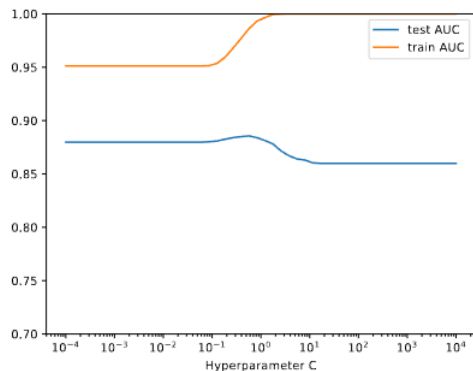
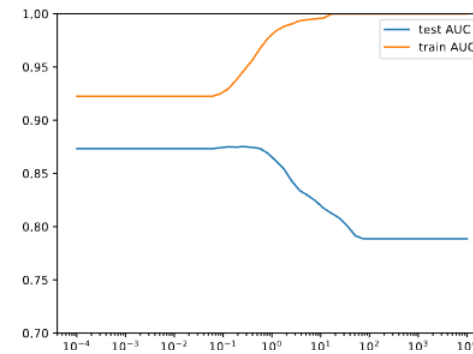
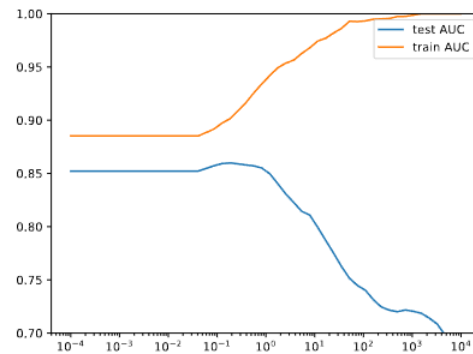
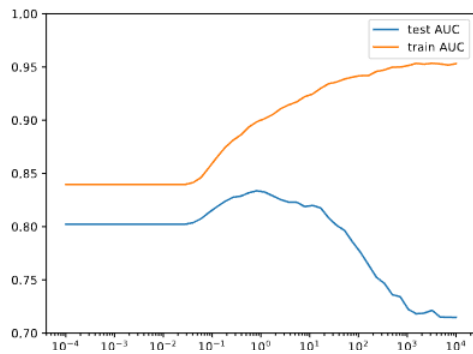
# Feature Ranking

- Optimize the feature set to get the best result
  - Start with the most discriminative features and gradually expand the feature set
- The ee and  $\mu\mu$  modes are processed separately



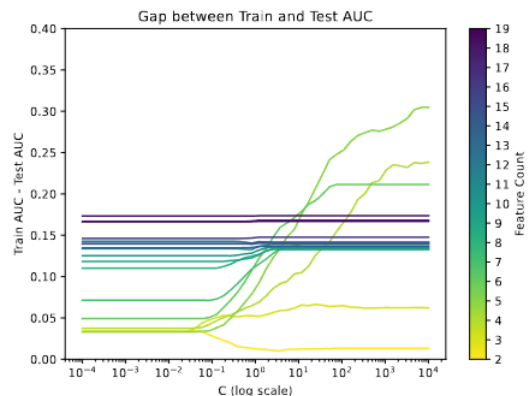
# QSVM Training Optimization

- Takes AUC as the main metric, with no obvious overfitting
  - Feature set selection (more features often lead to more severe overfitting)
  - Feature value normalization ( $0, \pi$ )
  - Tuning of the hyperparameter C

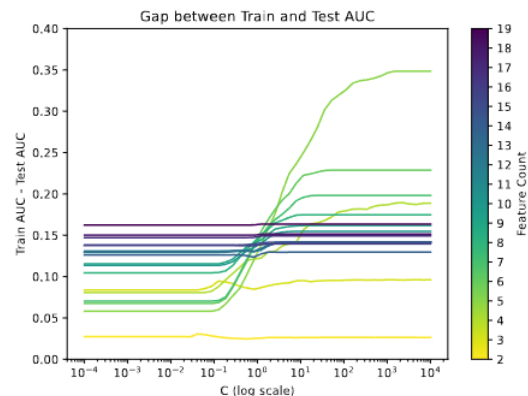


*From 4 to 9 data features for ee mode*

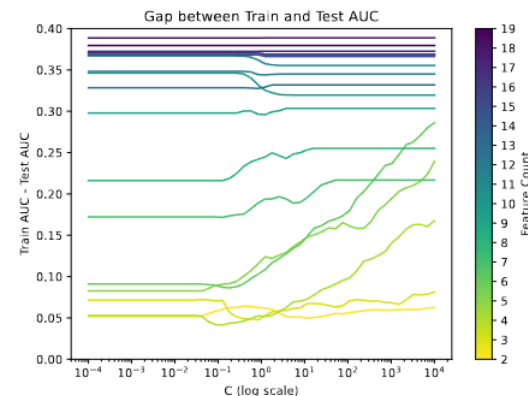
# QSVM Training Optimization



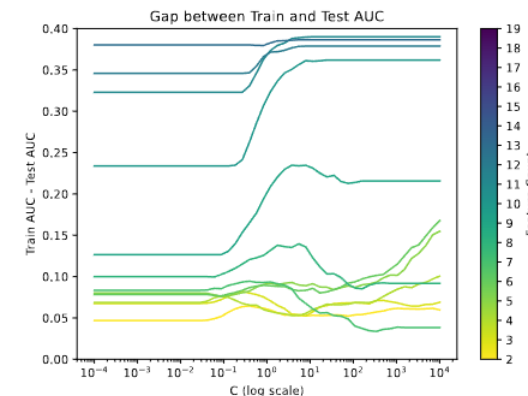
ee with normalizaiton



ee without normalizaiton



$\mu\mu$  with normalizaiton



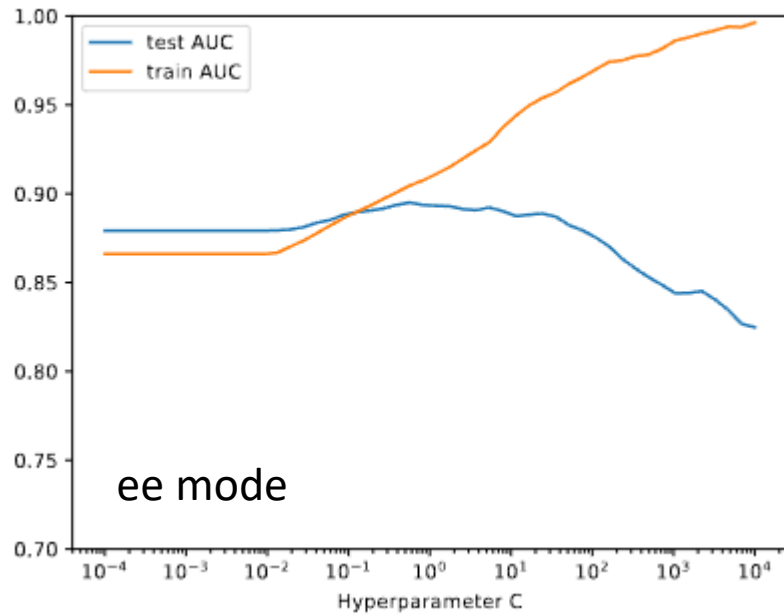
$\mu\mu$  without normalizaiton

The final criteria

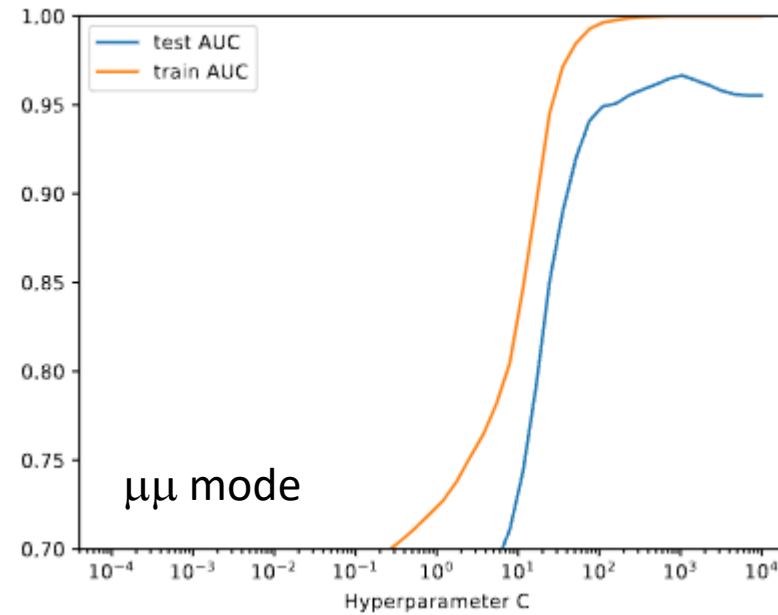
衰变道	选取特征	超参数 $C$
$ee$	$P_z(e^-), P_z(e^+), \theta_{\pi^+\pi^-}, E(\pi^-), E(\pi^+), P(\pi^+)$	$2.683 \times 10^{-1}$
$\mu\mu$	$E(\mu^+), E(\mu^-), P(\pi^-), P(\pi^+), P(\mu^-), P(\mu^+), \theta_{\mu^+\mu^-}$	$3.393 \times 10^2$

# Classical SVM Optimization

- Use the same features as QSVM
- Feature values are normalized to  $(0, \pi)$



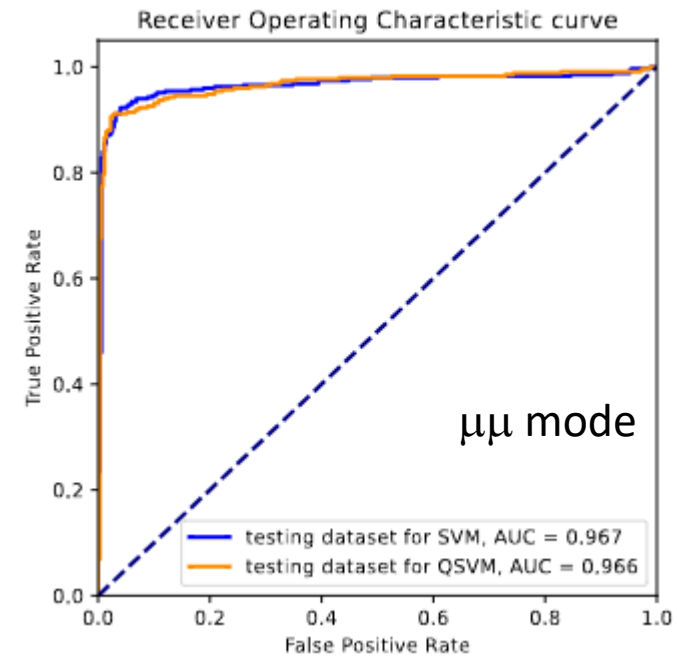
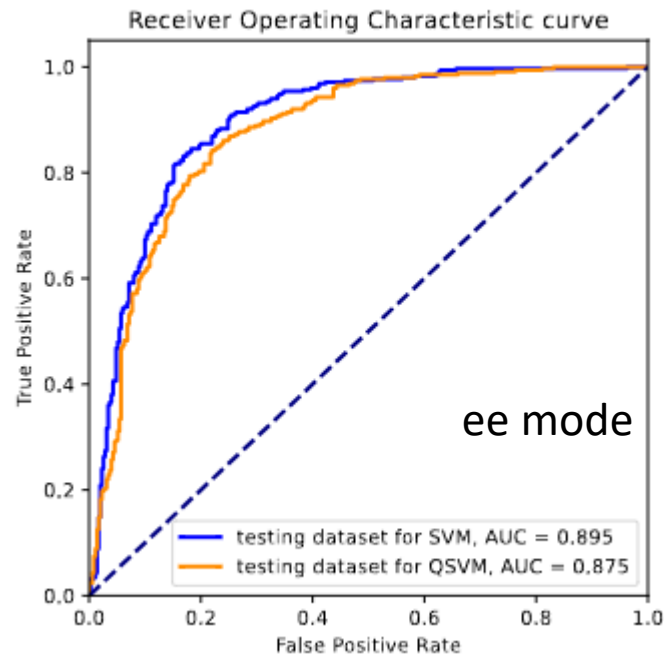
$$C = 5.690 \times 10^{-1}$$



$$C = 1.048 \times 10^3$$

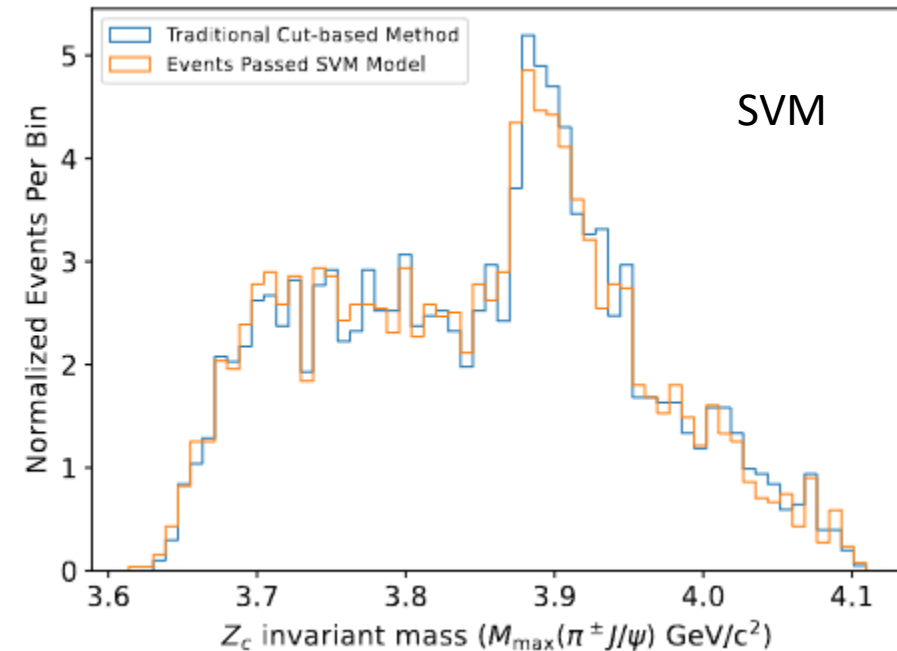
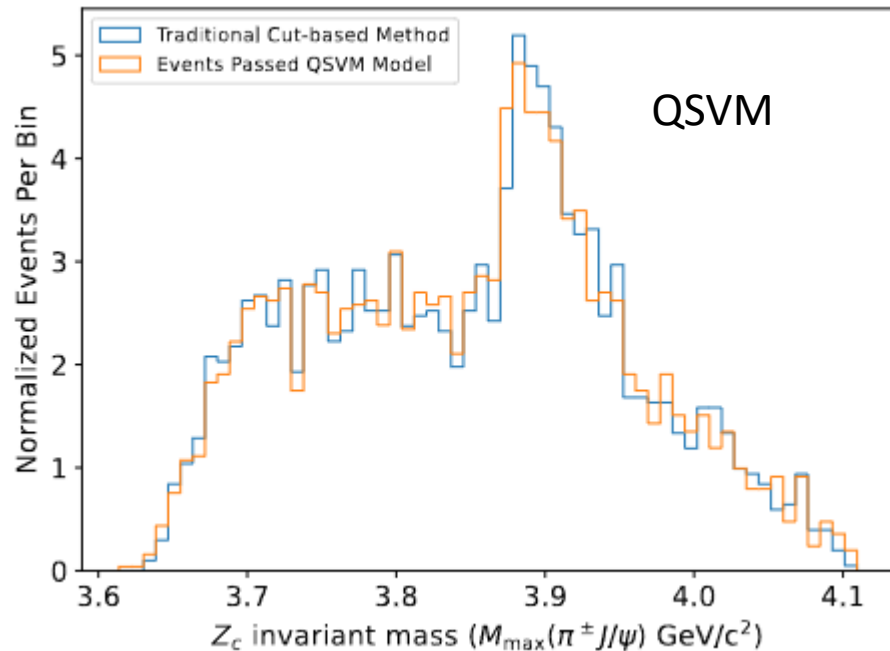
# The Performance of QSVM and SVM

- The ROC of the optimized QSVM and SVM models



# An Informal Result

- The  $M_{\max}(\pi^+J/\psi)$  distribution
  - Compare with the traditional result
  - The number of events is normalized to enable comparison of the differences in distributions
    - The QSVM/SVM method actually obtains more events since the dataset used is different to the traditional method



# Summary and Outlook

- The performance comparison
  - In this study, the performance of QSVM is very close to that of the classical SVM
  - The results of the QSVM/SVM method are also very close to those of the traditional method
- Try on the real quantum equipment
  - 国盾量子
  - 正在协调和研究解决的问题
    - Qiskit量子门的转换
    - 可用机时
    - 保真度对结果的影响

***Thanks!***