

Update on the dilepton analysis on RGA: TCS and photoproduction of J/ψ

Pierre Chatagnon, for the dilepton group
CLAS collaboration meeting – Jefferson Lab

8th of November 2023

Outline

I

Motivations, general considerations and planning

II

Spring 2019 Pass 2: Comparison with pass I and first look at MC/data comparison

III

Lepton PID using machine learning

IV

J/ψ event selection and resolution

V

Maximum likelihood fit for the extraction of the TCS parameters

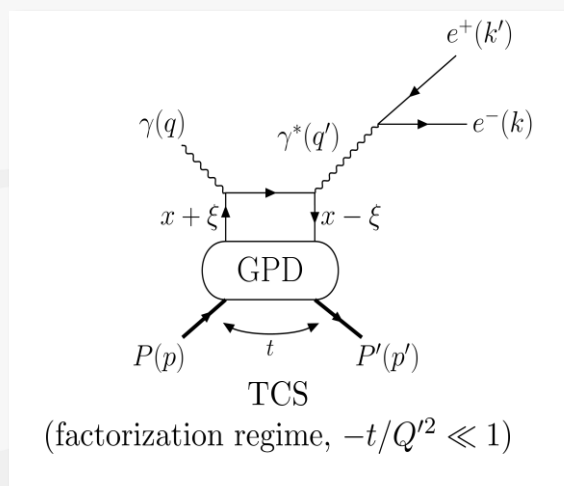
VI

Take-aways and timeline for future work

Motivations for dilepton final state measurement

Timelike Compton Scattering

$$\text{TCS: } \gamma p \rightarrow e^+ e^- p'$$

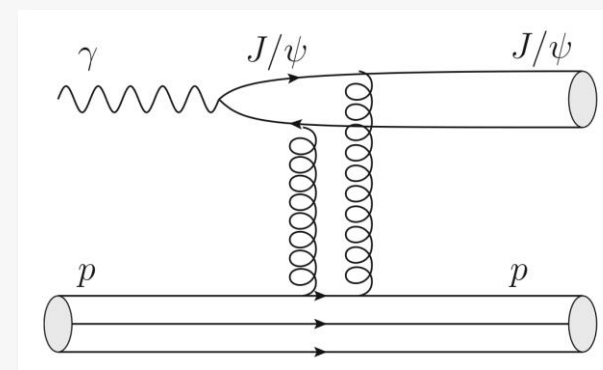


$$\frac{d^4 \sigma_{INT}}{dQ'^2 dt d\Omega} \propto \frac{L_0}{L} \left[\cos(\phi) \frac{1 + \cos^2(\theta)}{\sin(\theta)} \text{Re} \mathcal{H} + \dots \right]$$

$$\frac{d^4 \sigma_{INT}}{dQ'^2 dt d\Omega} = \frac{d^4 \sigma_{INT} |_{\text{unpol.}}}{dQ'^2 dt d\Omega} - \nu \cdot A \frac{L_0}{L} \left[\sin(\phi) \frac{1 + \cos^2(\theta)}{\sin(\theta)} \text{Im} \mathcal{H} + \dots \right]$$

J/ψ photoproduction at threshold

$$\gamma p \rightarrow J/\psi p \rightarrow e^+ e^- p'$$

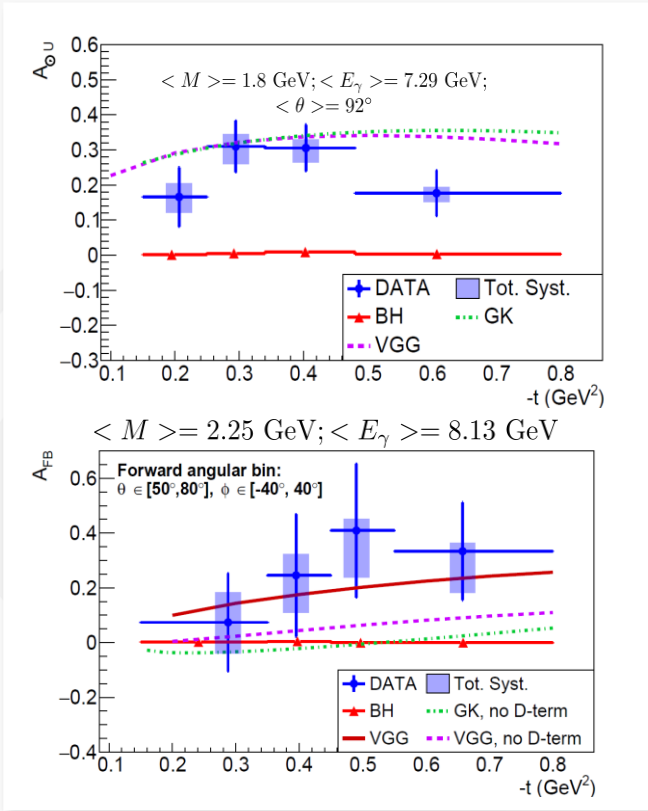


- The t -dependence of the cross-section allow to access gluon Gravitational Form Factors (GFFs), mass radius of the nucleon and gluon GPDs (under 2-gluon exchange assumption and no open-charm contributions)
- Model-dependent limit on the branching ration of the Pc pentaquark.

Publication status

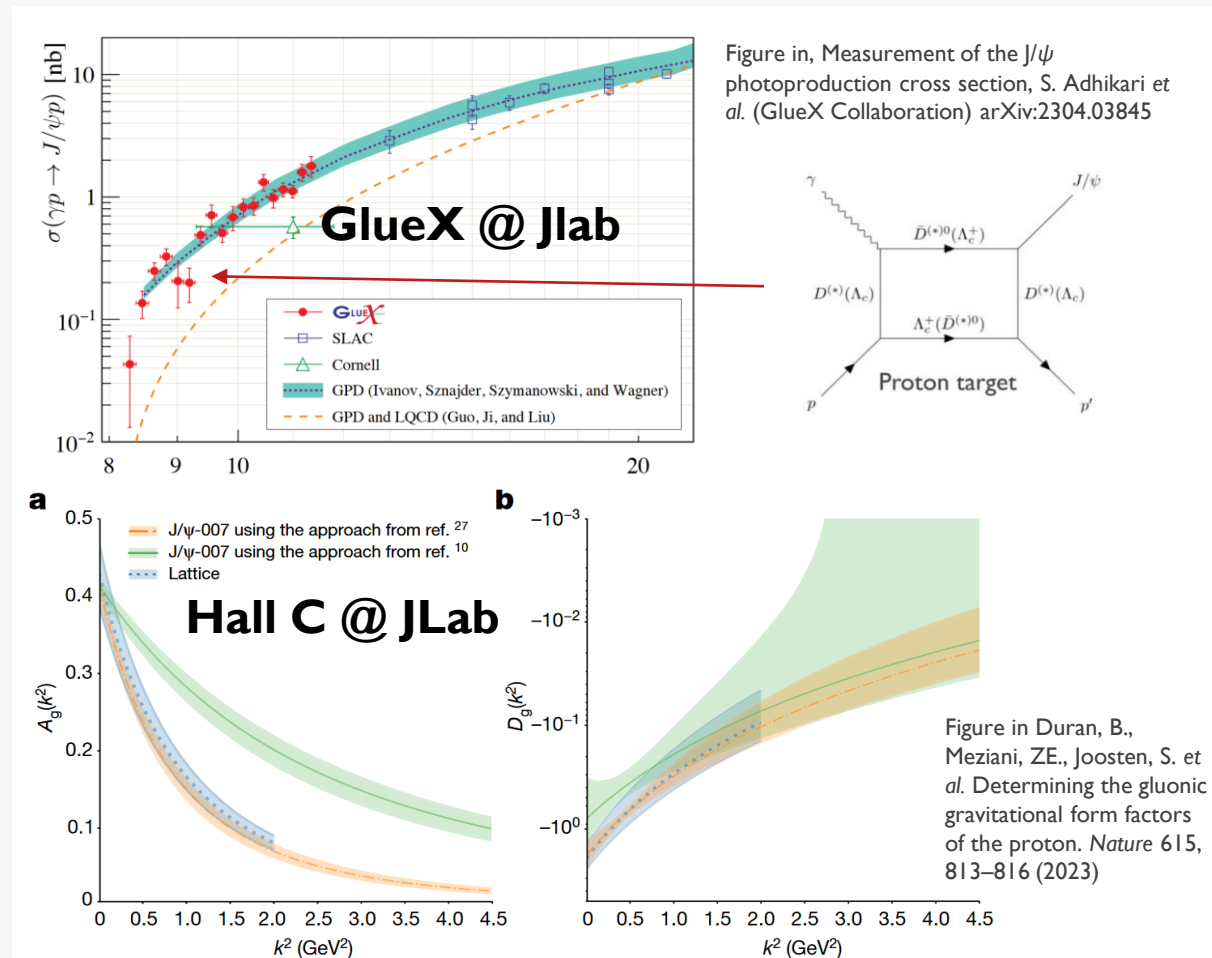
Timelike Compton Scattering

First Measurement of Timelike Compton Scattering, P. Chatagnon *et al.* (CLAS Collaboration), Phys. Rev. Lett. 127, 262501 (2021)



- Hint for the universality of GPDs
- Importance of the D-term in the GPD parametrization

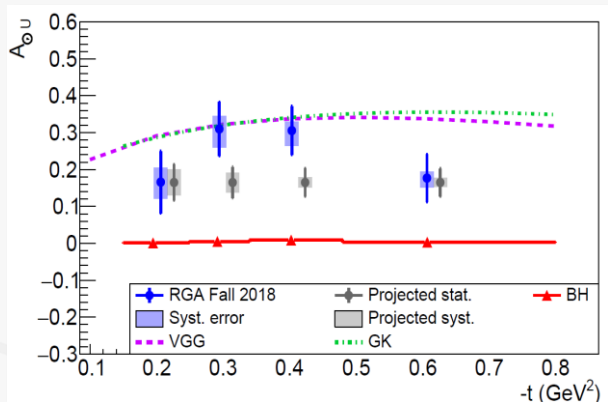
J/ ψ photoproduction at threshold



Goals and plans

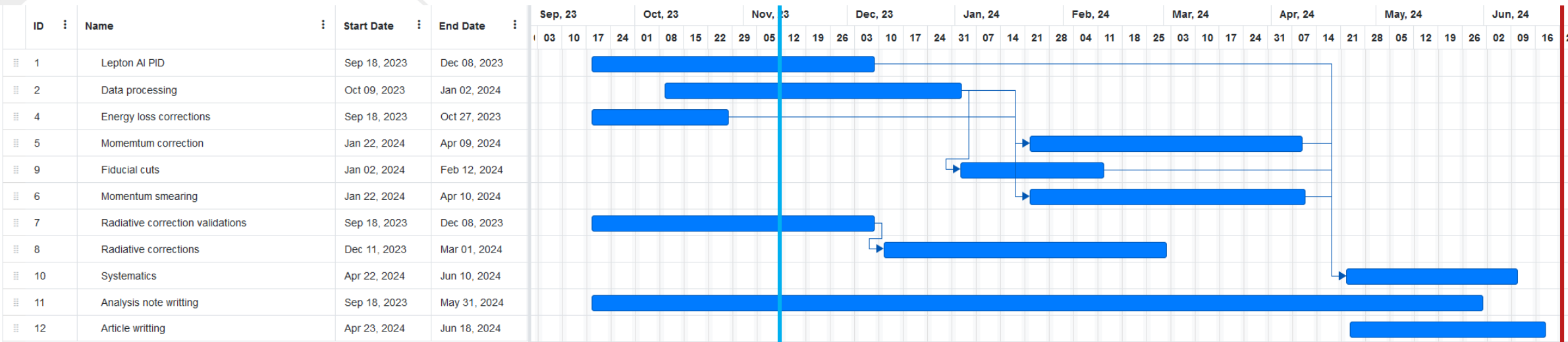
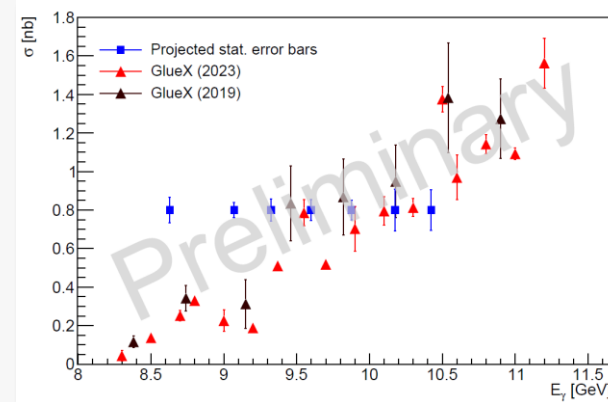
Timelike Compton Scattering

- Full statistics of RG-A will improve statistical accuracy by factor 2.
- Goal: use these results in GPD fits.



J/ψ photoproduction at threshold

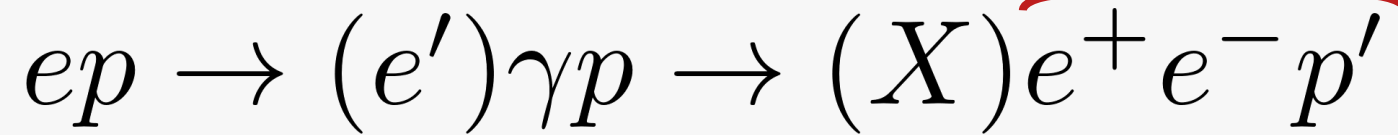
- Statistics comparable with GlueX 2019 analysis
- Independent cross-check of the ~9 GeV cusp
- Enough statistics to extract t-dependence and GFFs



Objective: summer 2024

General analysis strategy

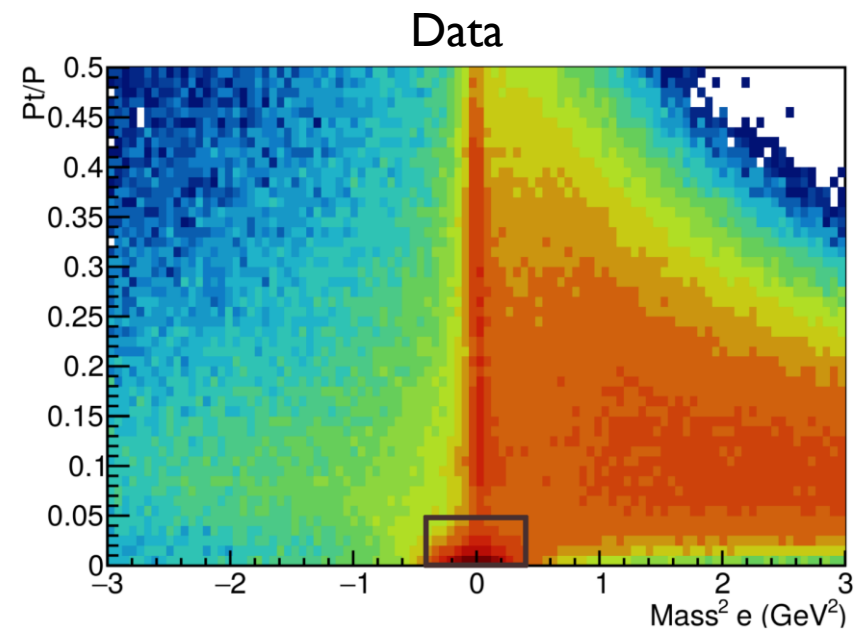
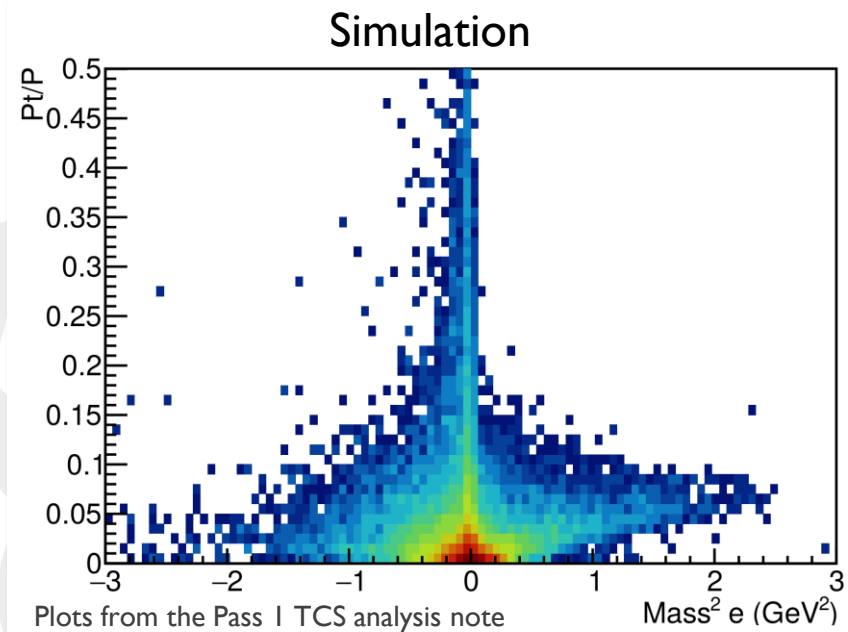
1) CLAS12 PID + Positron NN PID



$$p_X = p_{beam} + p_p - p_{e^+} - p_{e^-} - p_{p'}$$

2) $|M_X^2| < 0.4 \text{ GeV}^2$
Missing particle is an electron...

3) $|\frac{Pt_X}{P_X}| < 0.05$ or $Q^2 < 0.1 \text{ GeV}^2$
...scattered at small angle



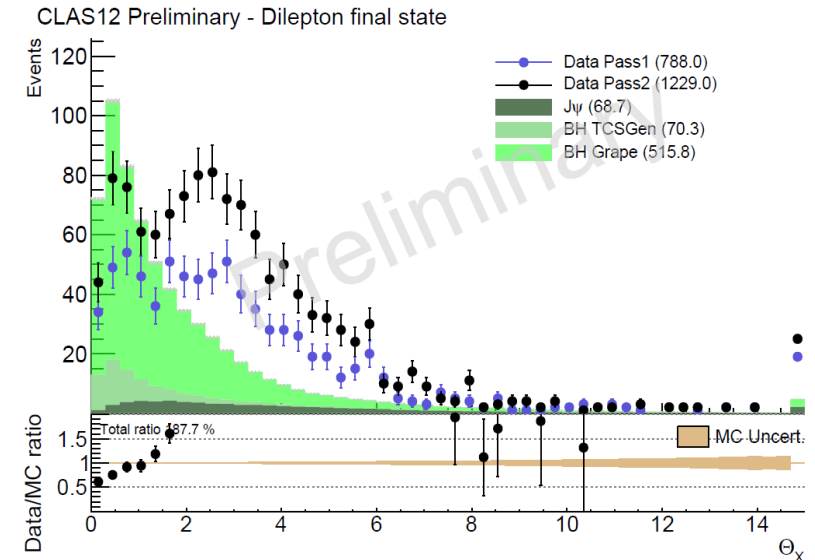
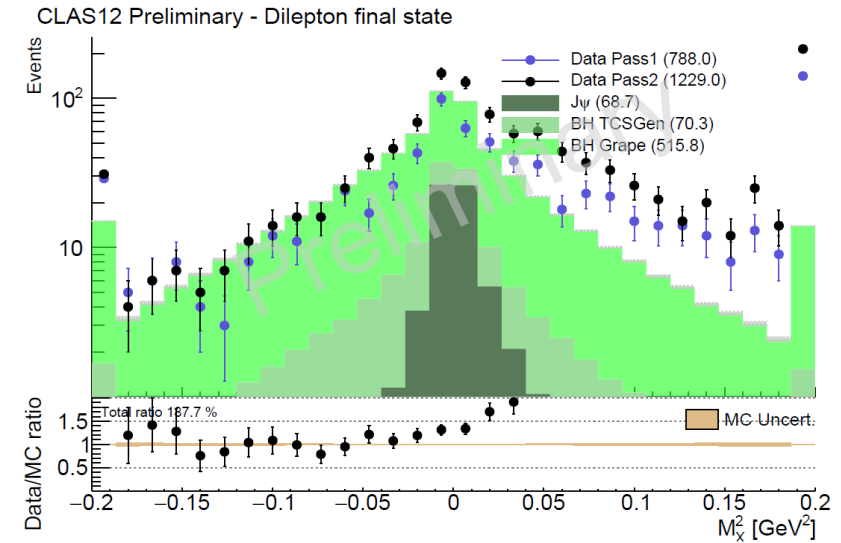
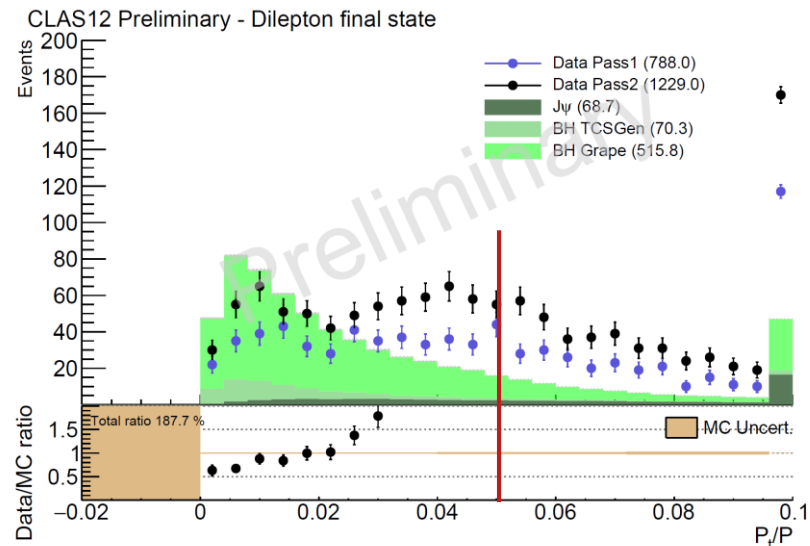
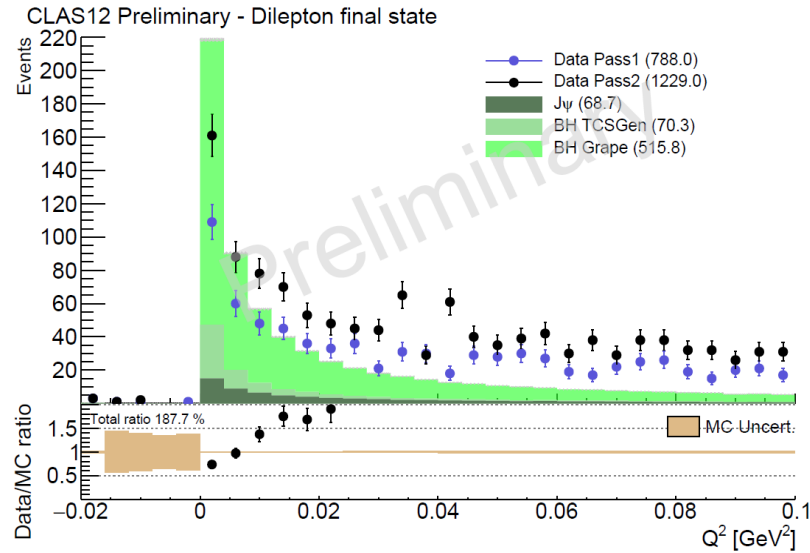
II - Pass 2 data: first look at Spring 2019



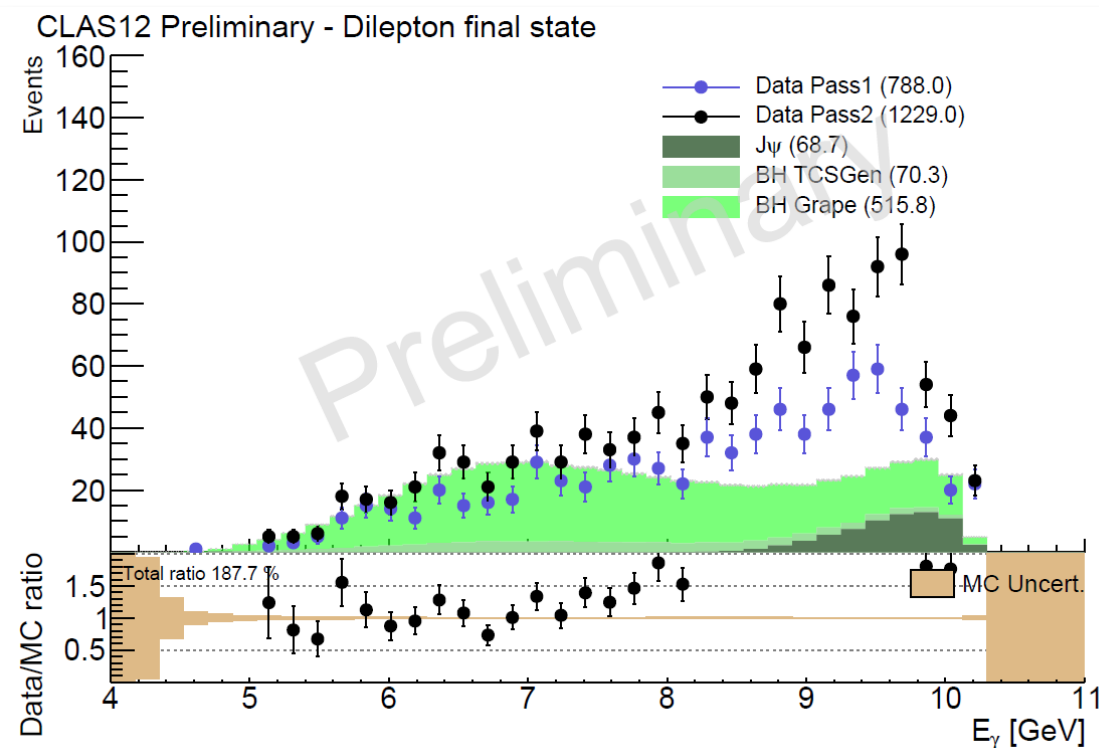
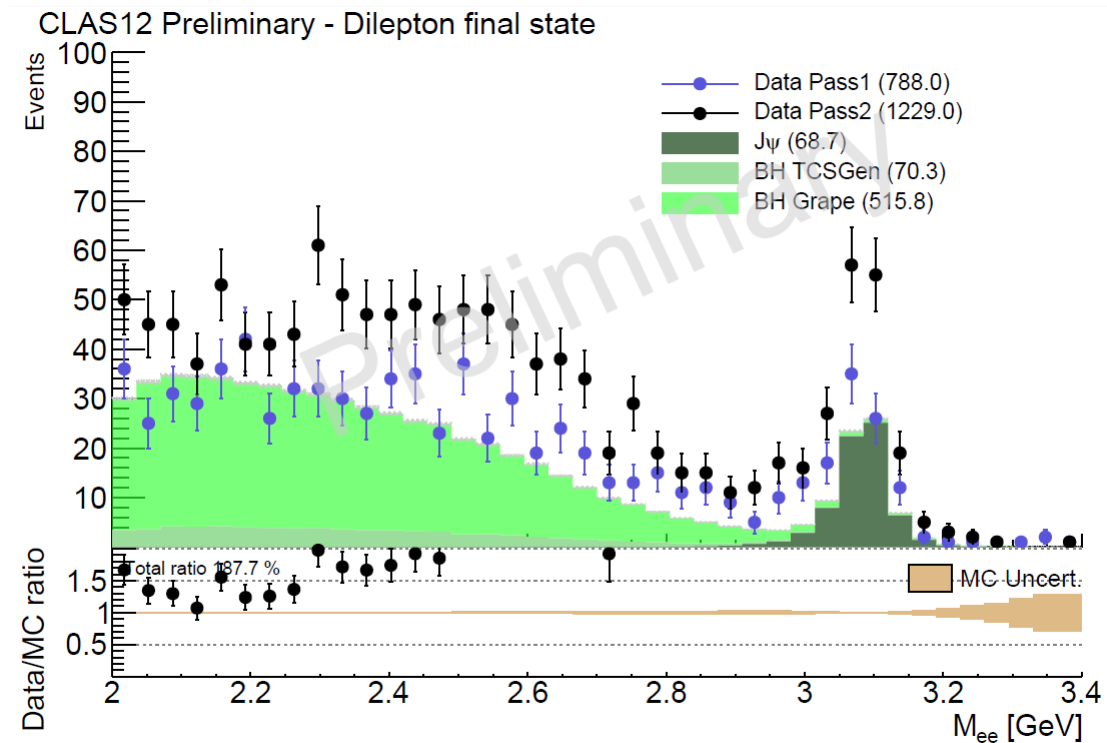
Comparison of data and MC (I)

Event selection

- Event topology:
 - exactly one electron in FD
 - exactly one positron in FD
 - exactly one proton
 - anything else
- Lepton momenta > 1.7 GeV
- Sampling Fraction > 0.15
- Lepton AI PID score > 0.05 (trained on pass I simulation)
- Exclusivity cuts:
 - $|MM^2| < 0.4$ GeV²
 - $|Q^2| < 0.1$ GeV²

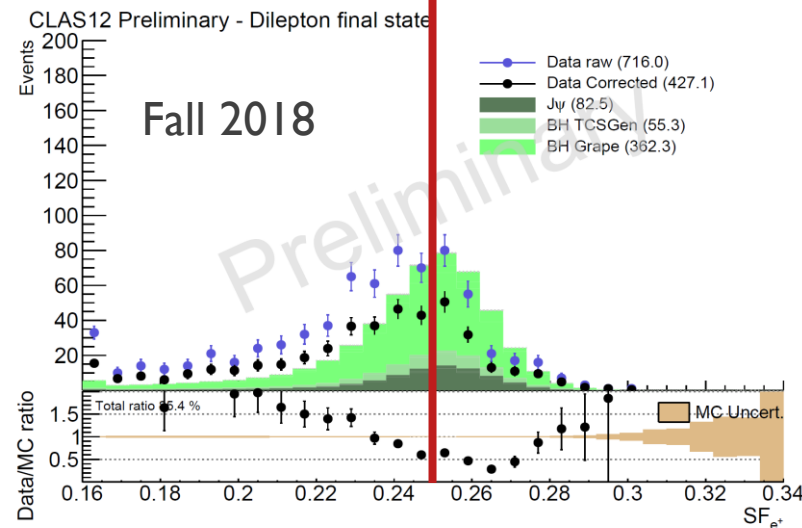
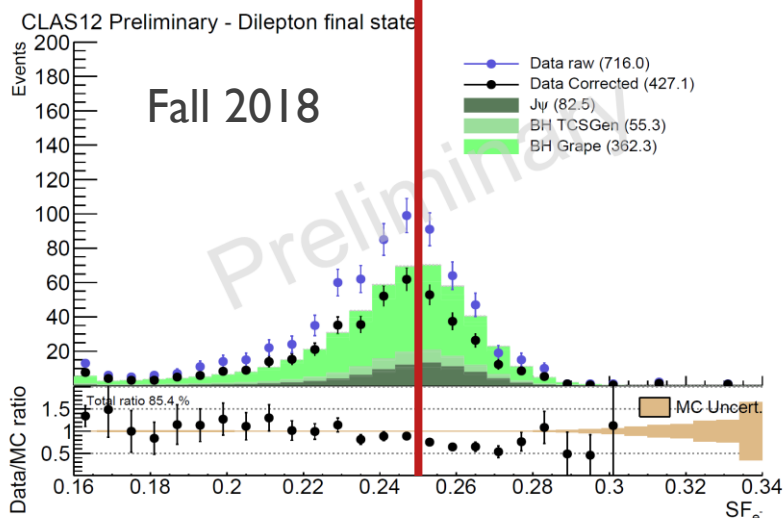
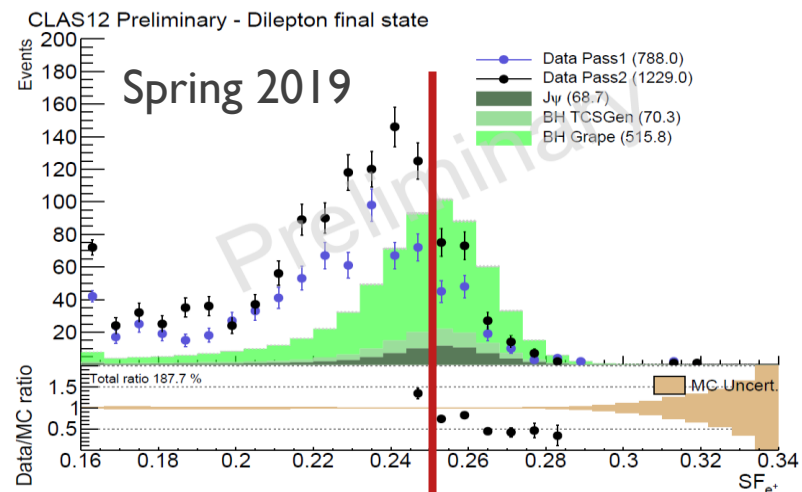
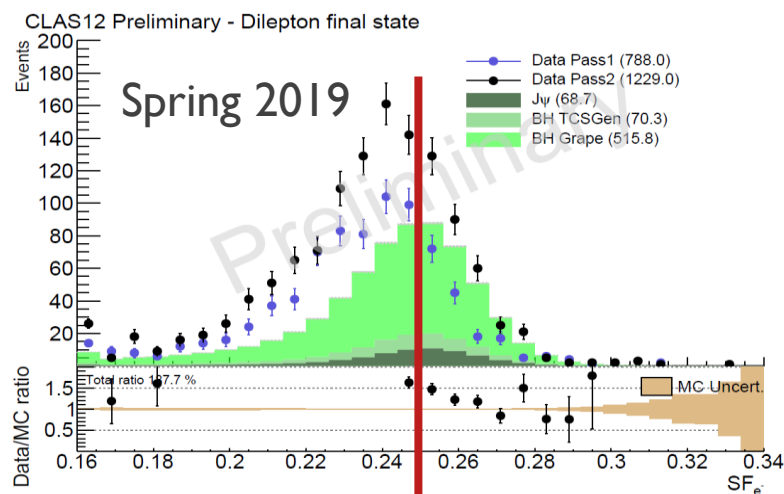


Comparison of data and MC (2)



- Same behavior is seen in Spring 19 and Fall 18 data: the large Q^2 background must be subtracted before calculating any cross-section
- We will use the same-charge lepton event method to do so
→ outbending dataset is essential

Sampling fraction MC/Data mismatch



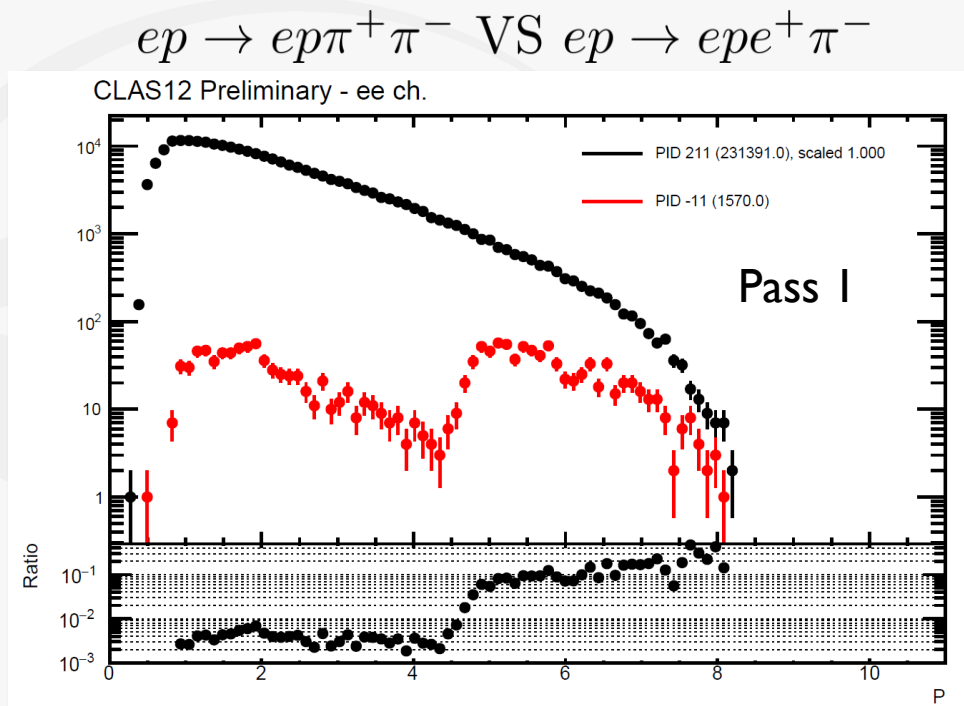
III - Lepton PID using machine learning



Motivations and previous work

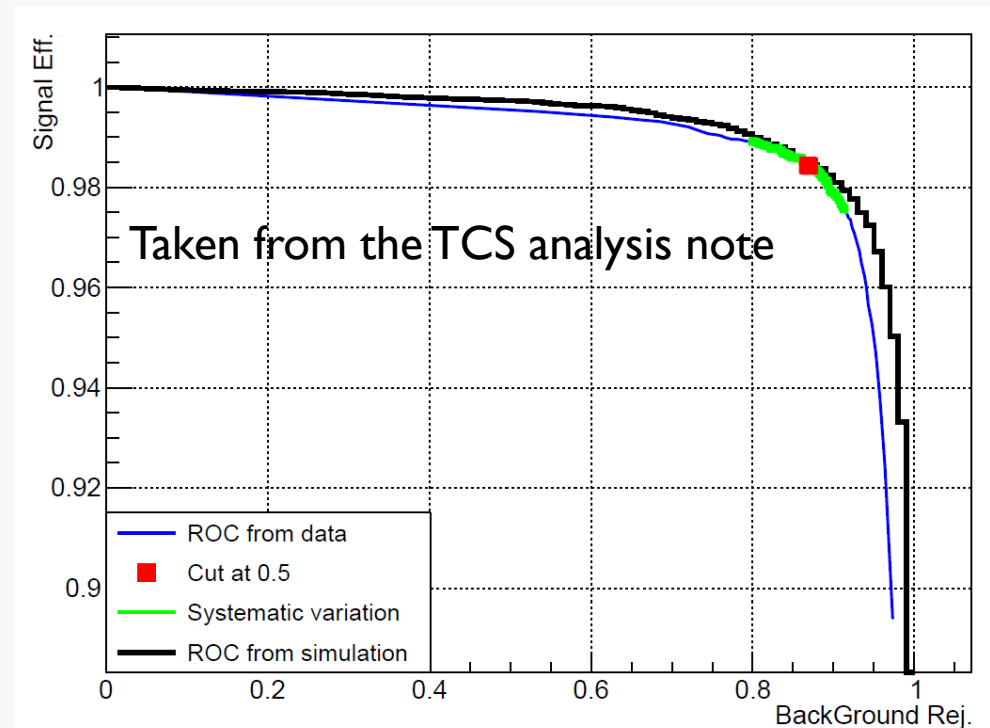
Motivations

- Above the HTCC, threshold both pions and leptons produce a HTCC signal. In the EB, only ECAL provide a separation between the two.
- $ep \rightarrow ep\pi^+\pi^-$ is a large background at large positron momenta



Previous work and motivations

- Long standing feature, already solved for the TCS publication
- Use the layer segmentation of the ECAL to provide separation
Variables used: SFs and m_2 of PCAL, ECIN, ECOUT
Method tested: **NN**, BDT



Current status

All material of this section provided by M.Tenorio Pita

Approach

- For both electrons and positrons, and for each RGA configuration:

$2 (e^+/e^-) \times 3 (\text{Spring19/Fall18 in/out}) = 6$ classifiers

- Use the layer segmentation of the ECAL to provide separation

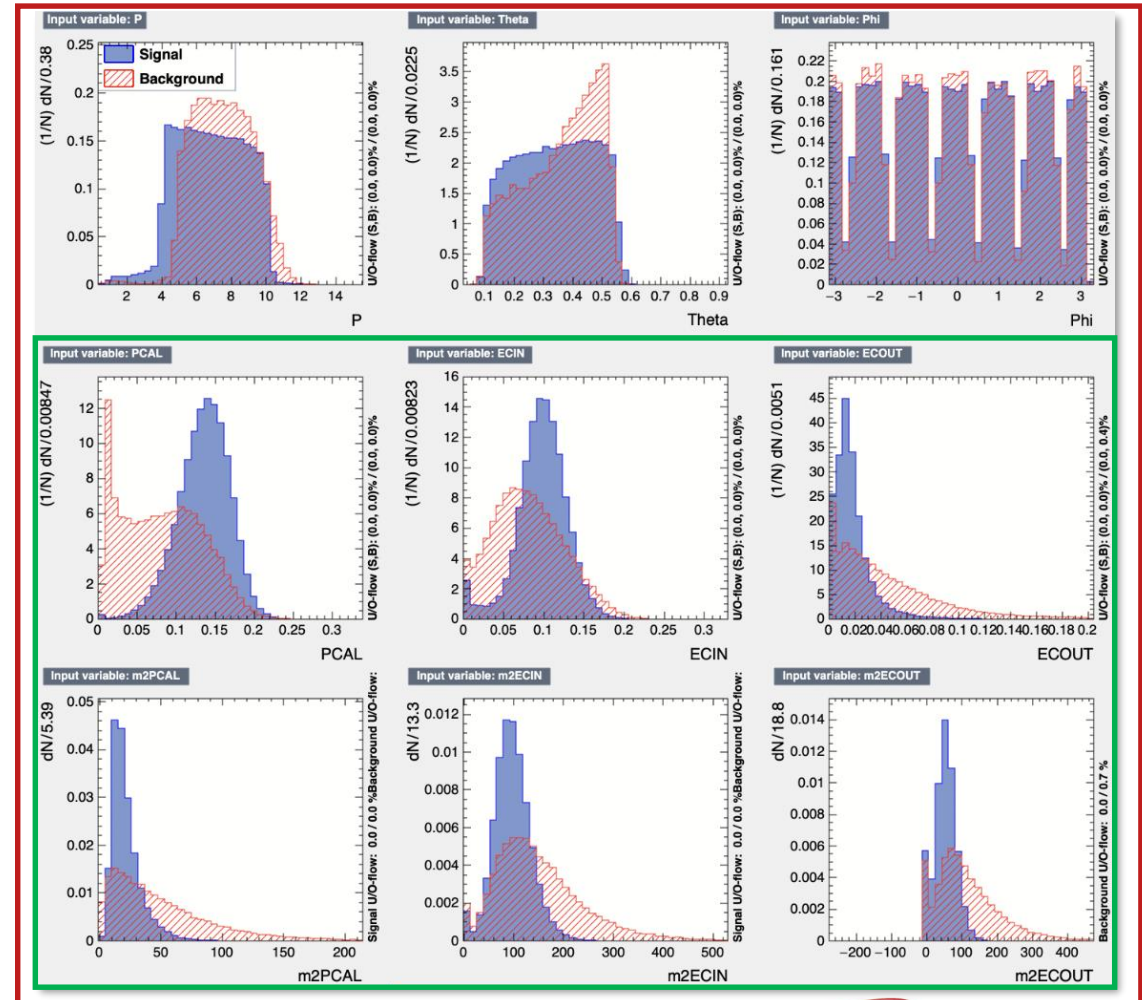
Variables used: P , θ , ϕ , SFs and m_2 of PCAL, ECIN, ECOUT

Method tested: NN, BDT

- Trained on simulation:
Signal: flat $e^{+/-}$ distribution, reconstructed as $e^{+/-}$
Background: flat $\pi^{+/-}$ distribution, reconstructed as $e^{+/-}$

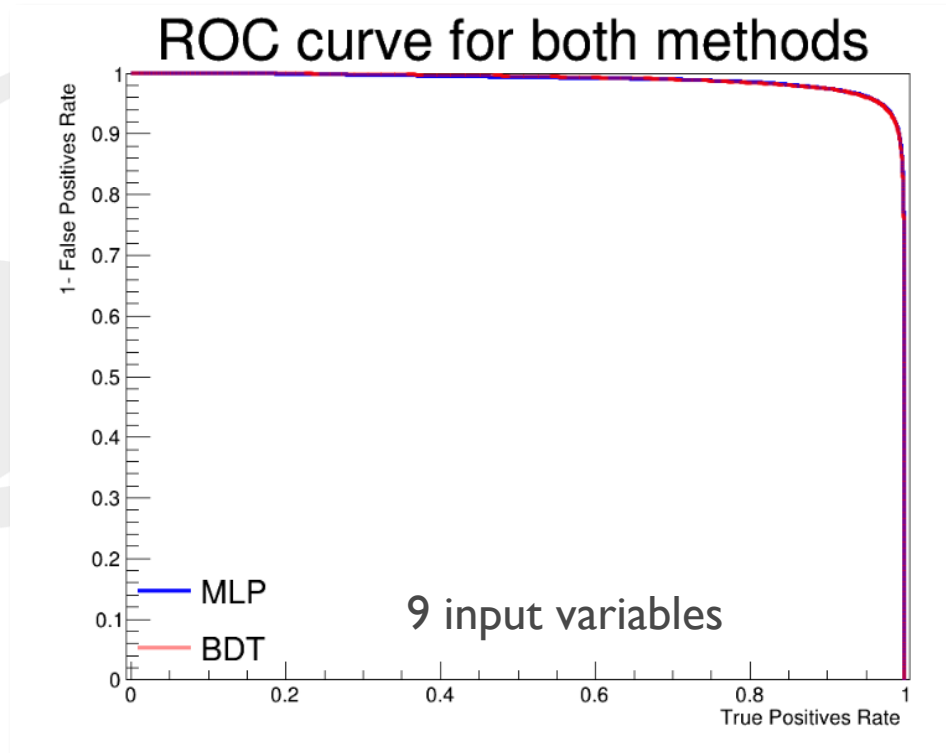
- Only RGA Spring 2019 for now

Input variables for signal (blue) and background (red)



Performances

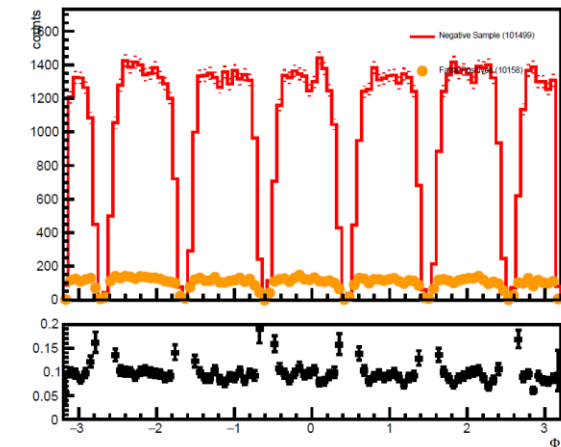
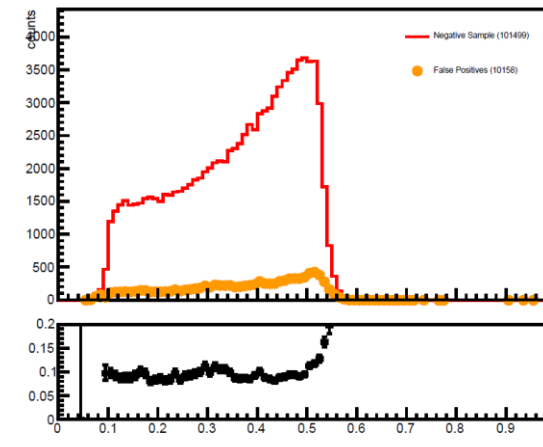
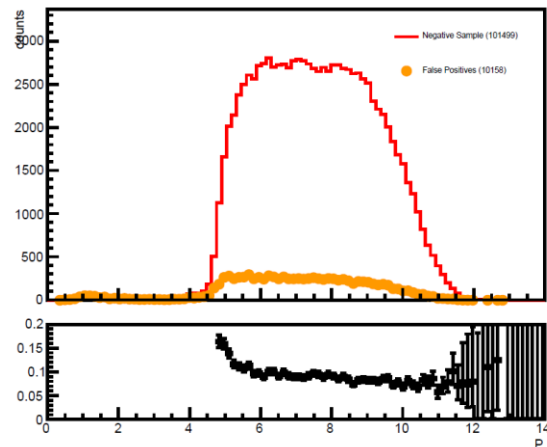
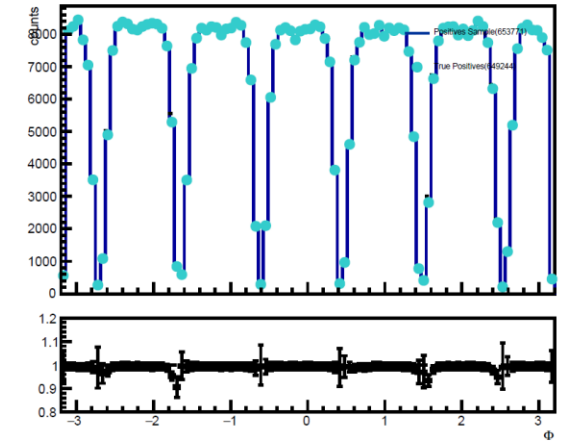
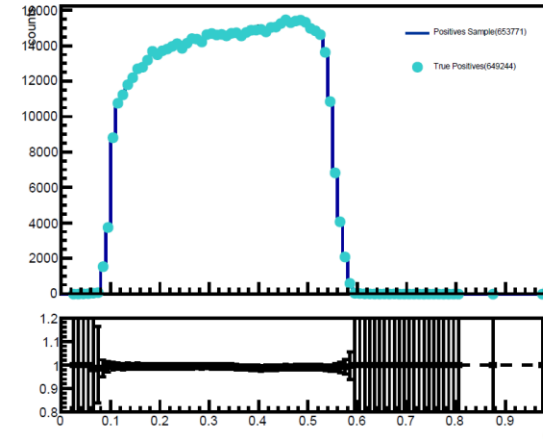
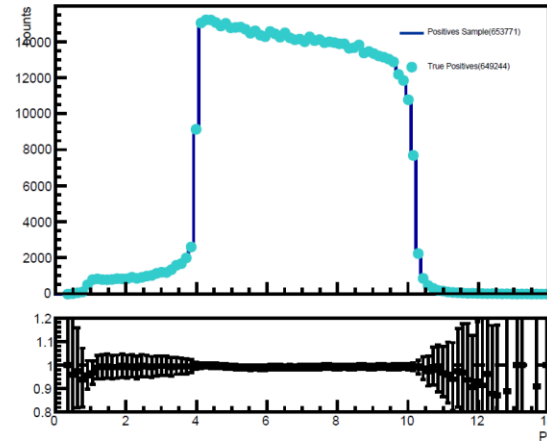
- We tested both 6 and 9 input variables, for 2 methods **NN** and BDT.
- Signal efficiency: 99.4 %
- Background rate: 10%



NN 6 var.	Actual e+ (653771)	Actual π^+ (101499)
Predicted e+	647688	12805
Predicted π^+	6083	88694
	TPR 99.1 %	FPR 12.6 %
NN 9 var.	Actual e+ (653771)	Actual π^+ (101499)
Predicted e+	649244	10158
Predicted π^+	4527	91341
Performances	TPR 99.4%	FPR 10%

Validation

NN. 9 Variables



- Signal efficiency and background reduction as a function of particle kinematics
- Done on separate samples

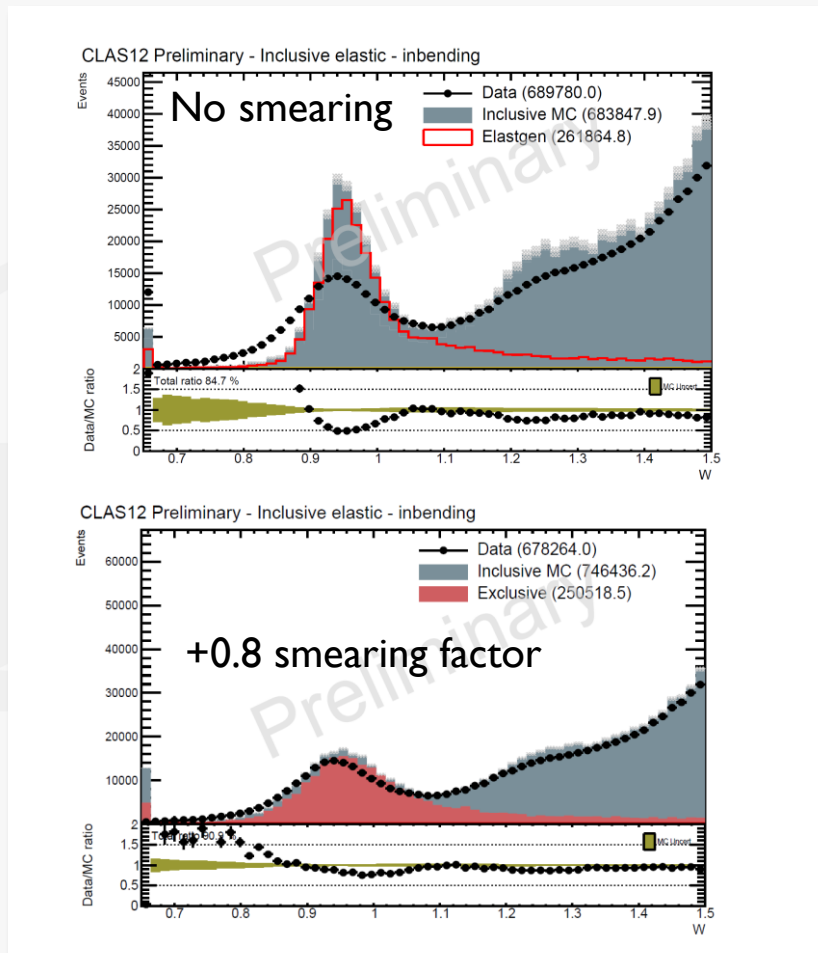
IV - J/ψ event selection, resolution and cross-section



Motivations

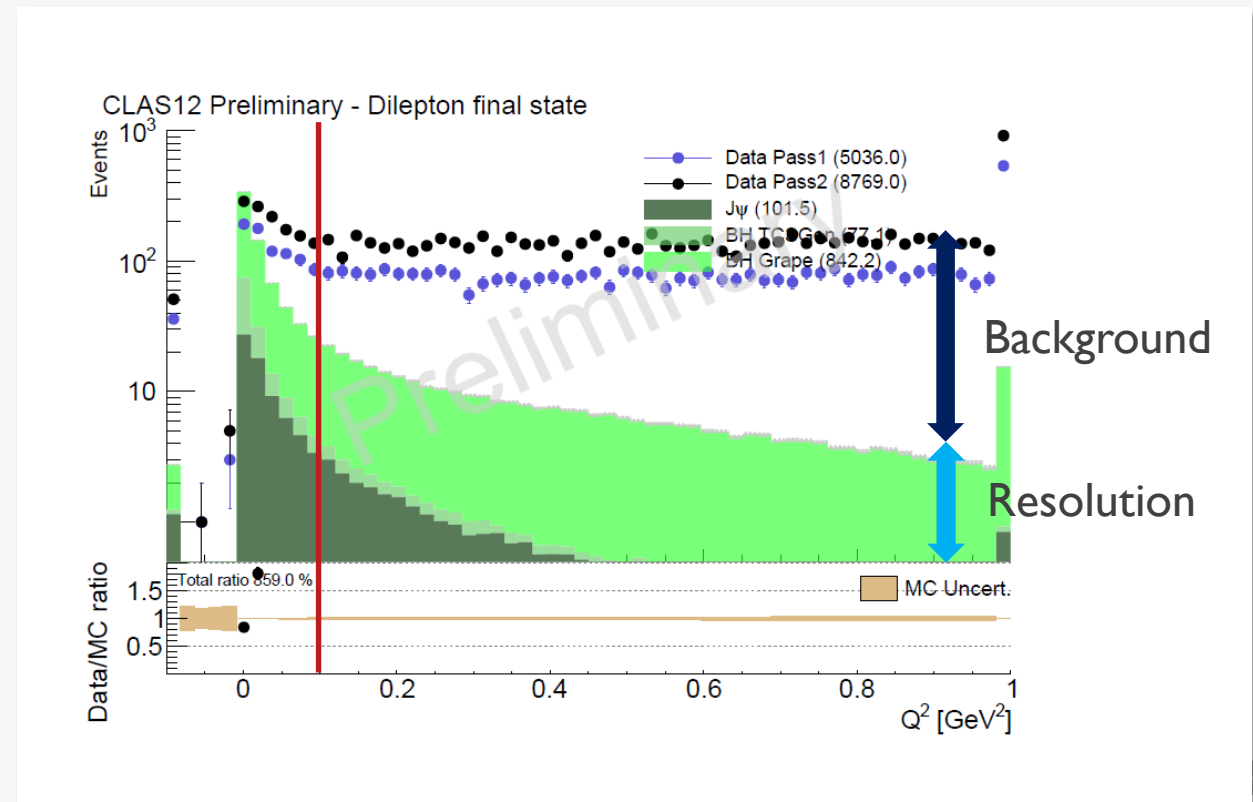
Inclusive elastic events

- In Pass I data, the smearing of the MC is key to understand the elastic peak resolution



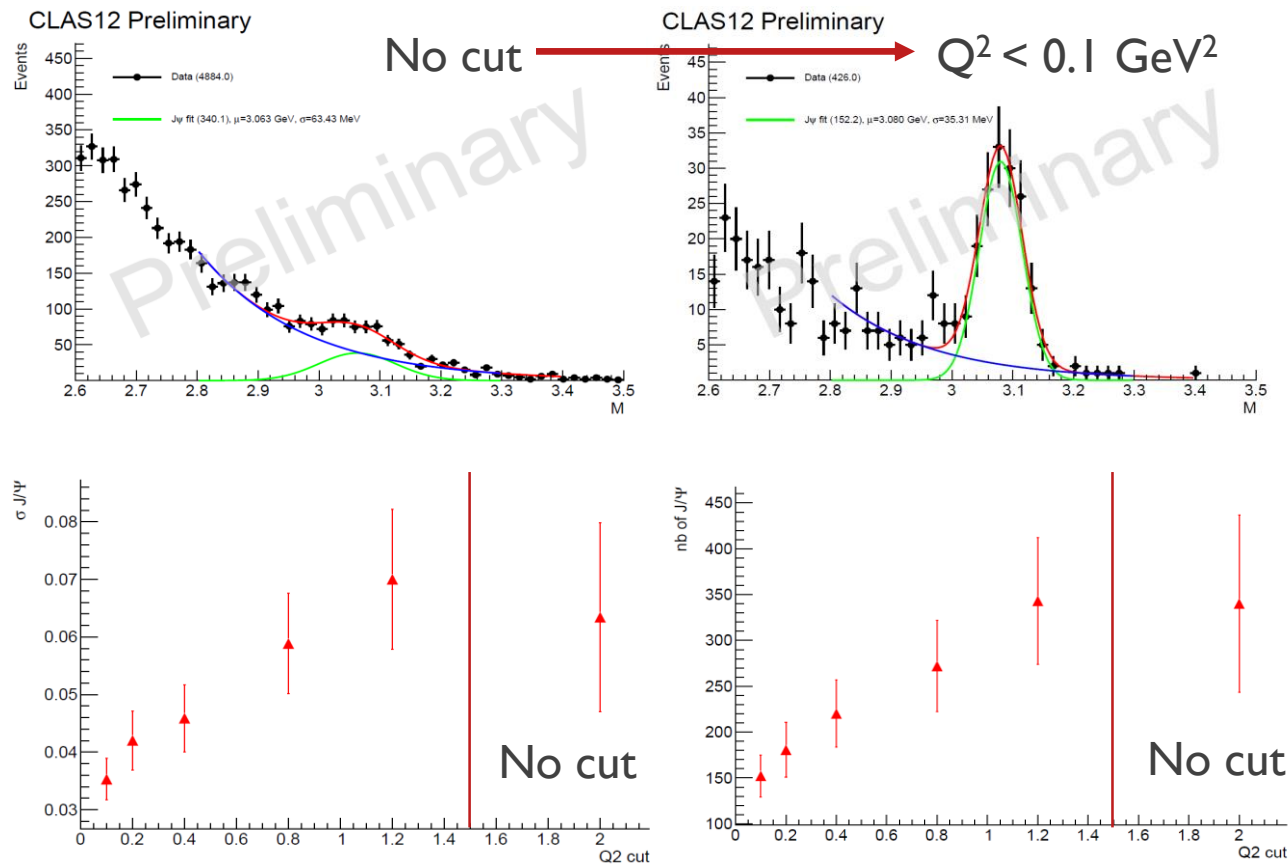
Inclusive elastic events

- Although photo-production events are generated ($Q^2=0$ GeV), the reconstructed virtuality of the incoming photon is large
- If the data resolution is not well reproduced by MC, the tail will be mis-reproduced and thus the extracted efficiency



Consequence for the number of J/ψ

- The J/ψ photoproduction yield should depend on the Q^2 cut similarly in data and simulation

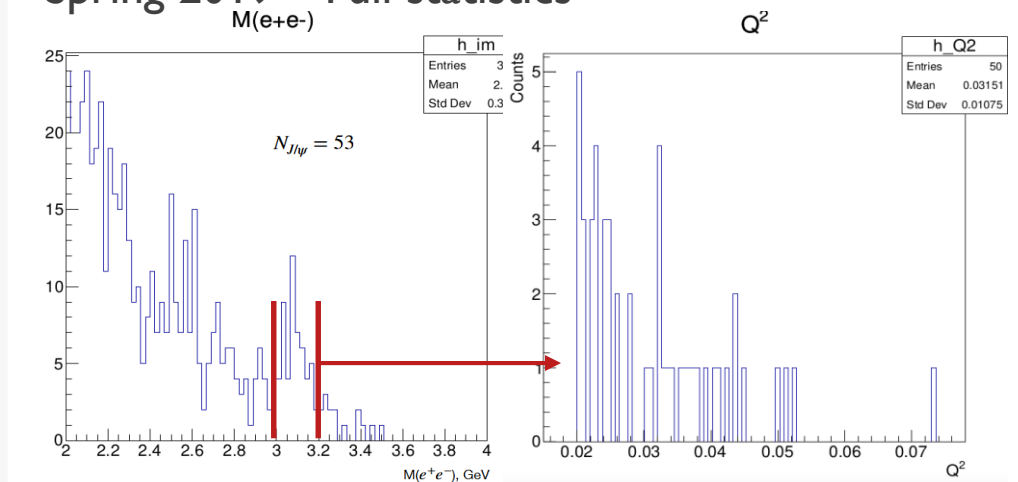


Maximum virtuality of the incoming photon

$$ep \rightarrow e' J/\psi p' \rightarrow e' l^+ l^- (X)$$

- Using tagged photo-production events, one can measure the virtuality of the incoming photon with only the FT resolution involved
- The large Q^2 events in the untagged case are solely due to FD resolution effects

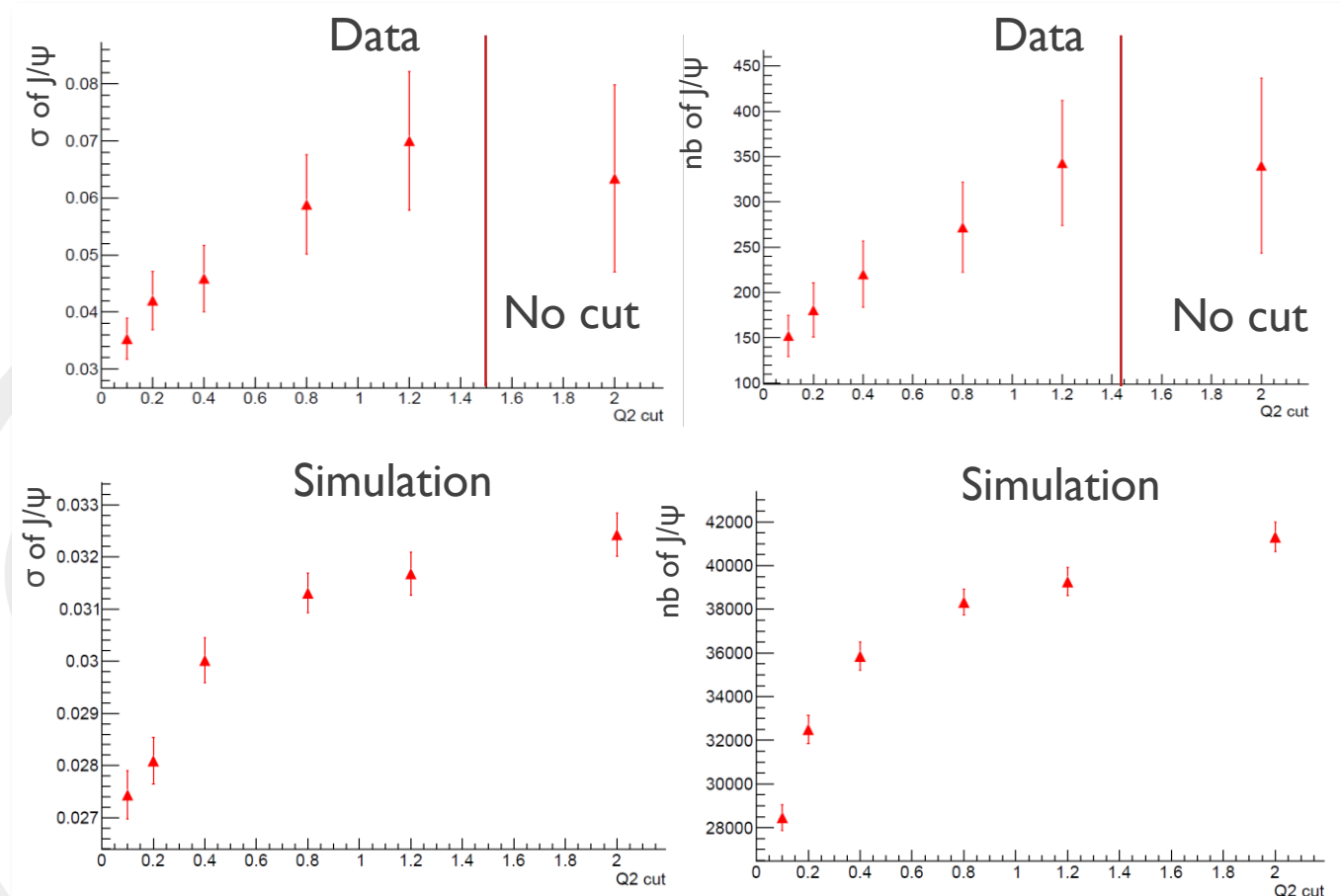
Spring 2019 – Full statistics



Material provided by M.Tenorio Pita

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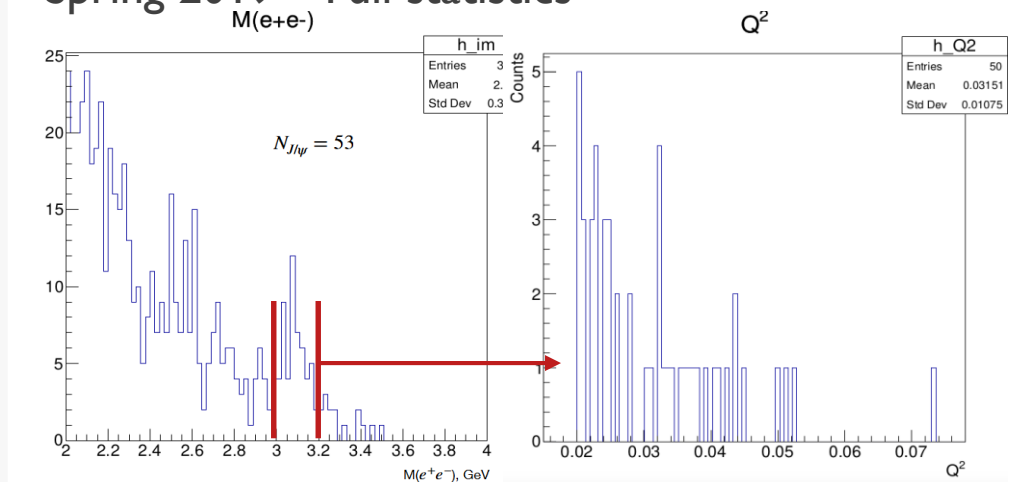


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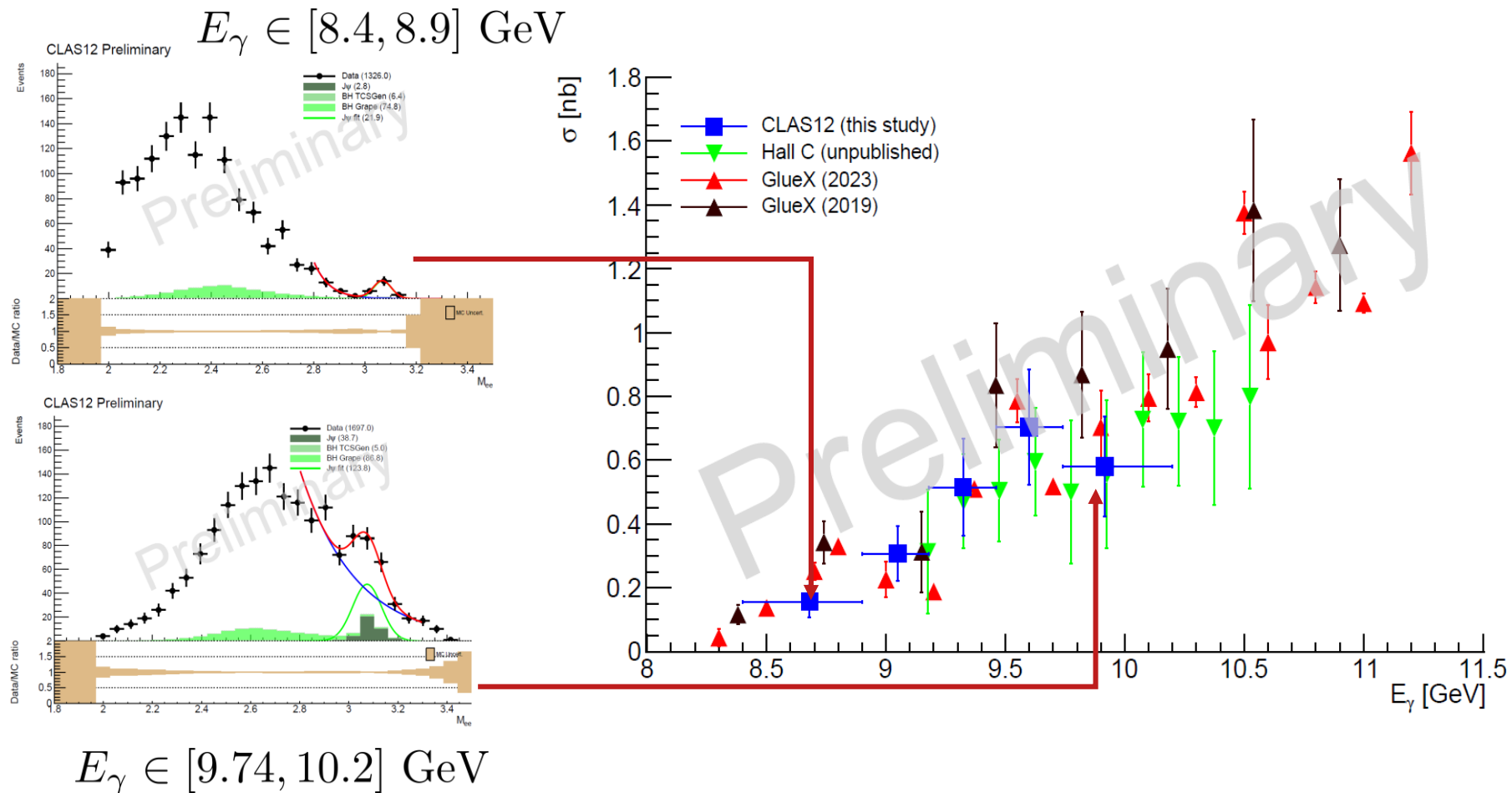
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Spring 2019 – Full statistics



Material provided by M. Tenorio Pita

Effect on the CS extraction



- Acceptance calculated using J/ψ photoproduction MC events and no Q^2 cut
- No cross-normalization with BH
- Fit using gaussian + exponential

V-TCS observable extraction: maximum likelihood approach



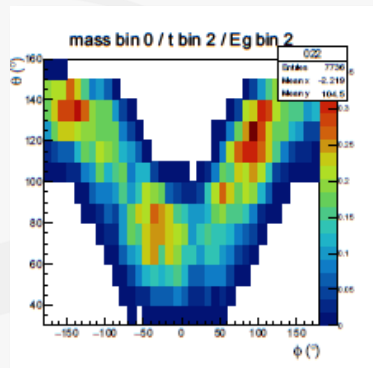
Motivations and formalism

All material provided by D. Glazier

Limitation of the current approach

- Both non-trivial angle dependence and non-trivial angular integration...

$$\frac{d^4\sigma_{INT}}{dQ'^2 dt d\Omega} = \frac{d^4\sigma_{INT} |_{\text{unpol.}}}{dQ'^2 dt d\Omega} - \nu \cdot A \frac{L_0}{L} \left[\sin(\phi) \frac{1 + \cos^2(\theta)}{\sin(\theta)} \text{Im}\mathcal{H} + \dots \right]$$



... makes the naive fitting procedure not straight forward to interpret

- What about the pure TCS contribution ?

$$\sigma(\gamma p \rightarrow e^+ e^- p) = \sigma_{BH} + \sigma_{INT} + \sigma_{TCS}$$

Maximum likelihood fit

$$I(\theta, \phi, hP) = \sigma_{BH} + \sigma_{TCS} + \sigma_{INT}$$

$$I(\theta, \phi, hP) = B \frac{1 + \cos^2(\theta)}{\sin^2(\theta)} + T(1 + \cos^2(\theta)) + A \frac{1 + \cos^2(\theta)}{\sin(\theta)} (\text{Re}M \cos(\phi) - hP \cdot \text{Im}M \sin(\phi))$$

If our data distribution, f , depends on an acceptance function $\eta(x_i)$ and a physics model $I(x_i : \theta_j)$:

$$f(x_i : \theta_j) = I(x_i : \theta_j) \cdot \eta(x_i)$$

Then we can approximate p by summing over M accepted Monte-Carlo events,

$$p(x_i : \theta_j) = \frac{I(x_i : \theta_j) \eta(x_i)}{\sum_s^M I(x_{i,s} : \theta_j)}$$

$$L(\theta_j, Y) = \prod_k^N p(x_{i,k} : \theta_j) e^{-Y} \frac{Y^N}{N!}$$

$$-\ln L(\theta_j, Y) = -\sum_k^N \ln \left[\frac{I(x_{i,k} : \theta_j)}{\sum_s^M I(x_{i,s} : \theta_j)} \right] + Y - N \ln Y - \sum_k^N \ln \eta(x_{i,k})$$

$$\mathcal{L}(\theta_j, Y) = -\ln L(\theta_j, Y) = -\sum_k^N \ln \frac{I(x_{i,k} : \theta_j)}{\sum_s^M I(x_{i,s} : \theta_j)} + Y - N \ln Y$$

Current status of brufit for TCS

Brufit tutorial: <https://indico.jlab.org/event/343/contributions/5450/attachments/4585/5691/GlazierBruFit>

Maximum Likelihood Approach Gen with truth

Reproduce input parameters ?

BH = 0.6 ; ImM = 0.8 ; TCS = 0.2

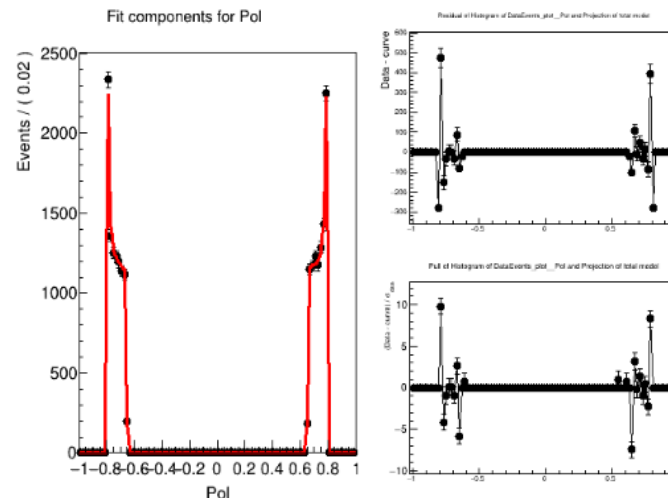
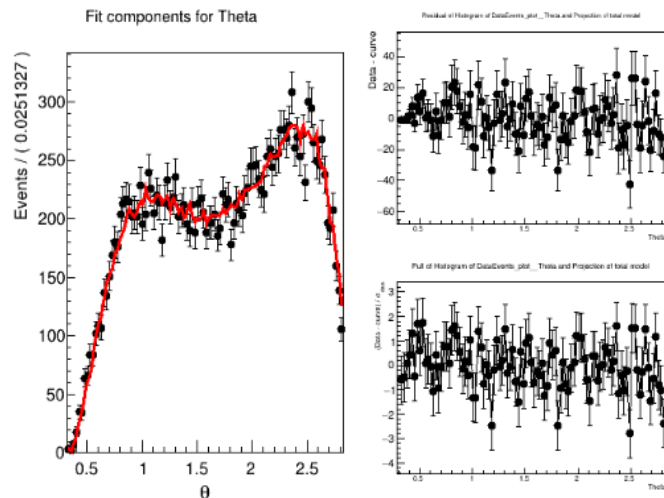
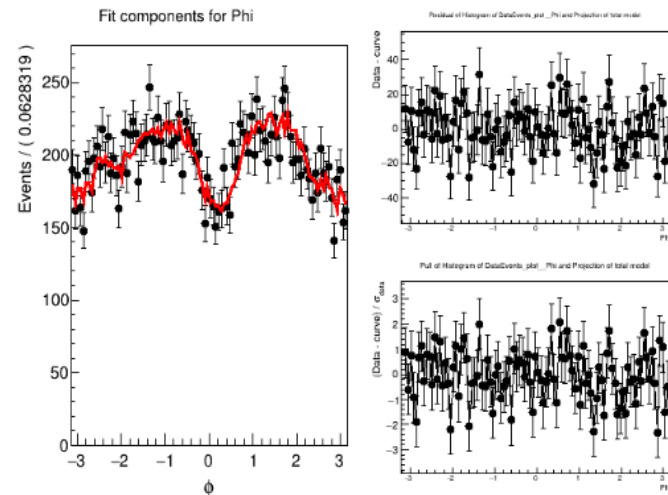
Results

BH 5.5324e-01 +/- 1.60e-02

ImM 8.1496e-01 +/- 3.21e-02

TCS 2.5344e-01 +/- 1.72e-02

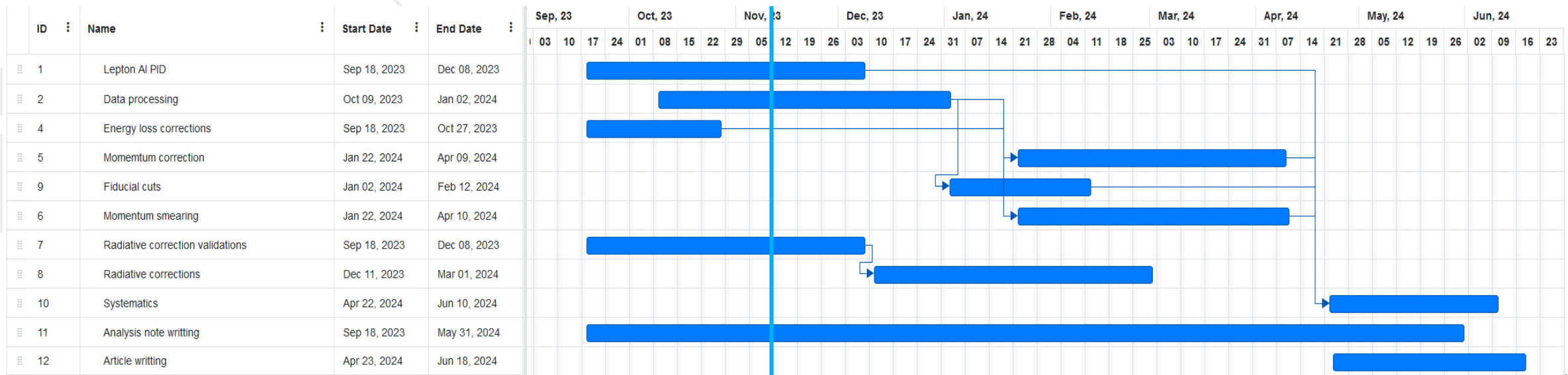
Slight bias from BH to TCS ?



- Method based on brufit https://github.com/dglazier/brufit/tree/R6_28/tutorials/TCS
- Tested on MC event passed through GEMC (geometrical acceptance + resolution effects)

Summary and outlook

- We have established a plan to reach both a new TCS and a first J/ψ publication on RGA.
- The work force matches the need: Derek (brufit for TCS), Kayleigh (TCS on RGC), Mariana (J/ψ on RGA), Pierre (TCS and J/ψ on RGA), Richard (J/ψ on RGA and RGB), Rafo (Simulation)
- Spring 19 Pass 2 dataset looks good, with similar issue than Pass I (Resolution and high- Q^2 background).
- AI PID for lepton is well underway and consistent with Pass I analysis.
- Maximum likelihood fit method is being developed for TCS observable extraction.



Back-up

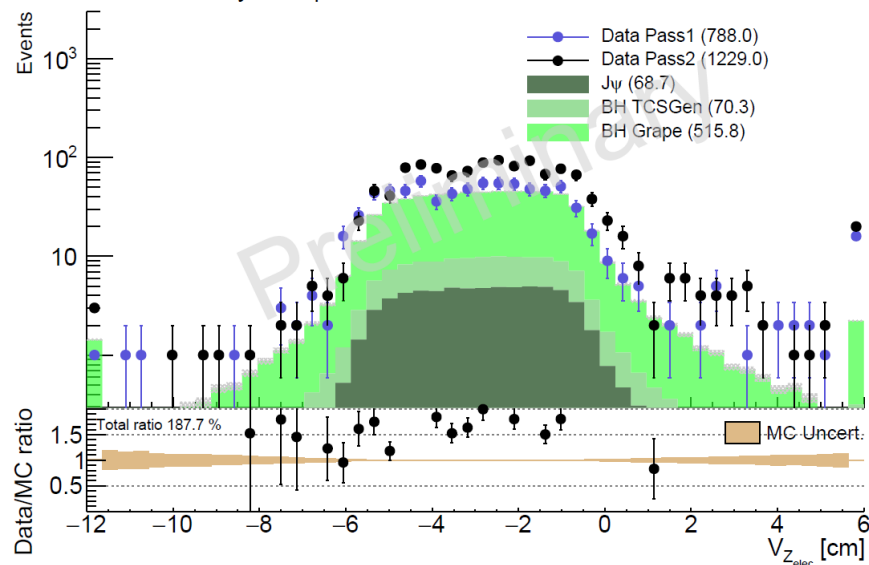


II - Pass 2 data: first look at Spring 2019

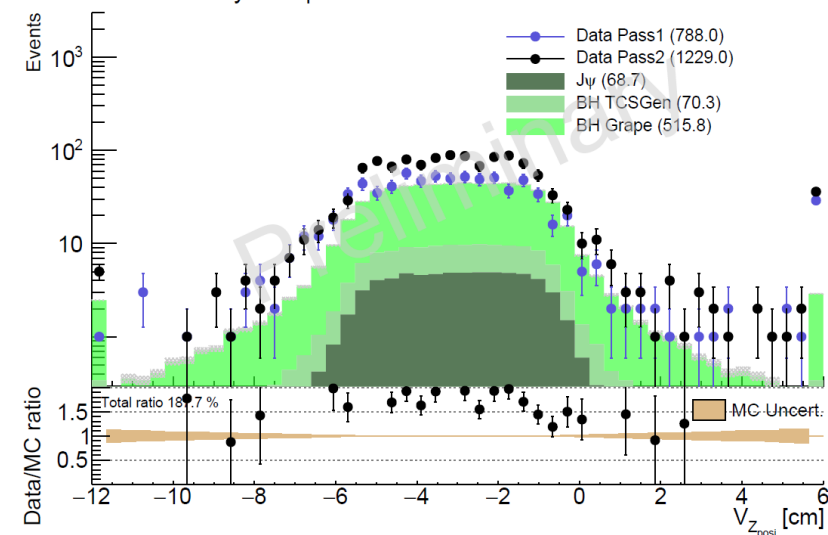


Spring 19 Pass 2: Vertices

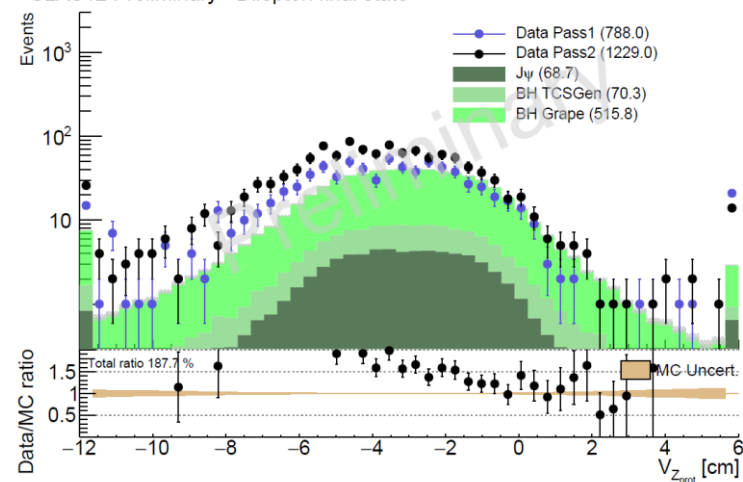
CLAS12 Preliminary - Dilepton final state



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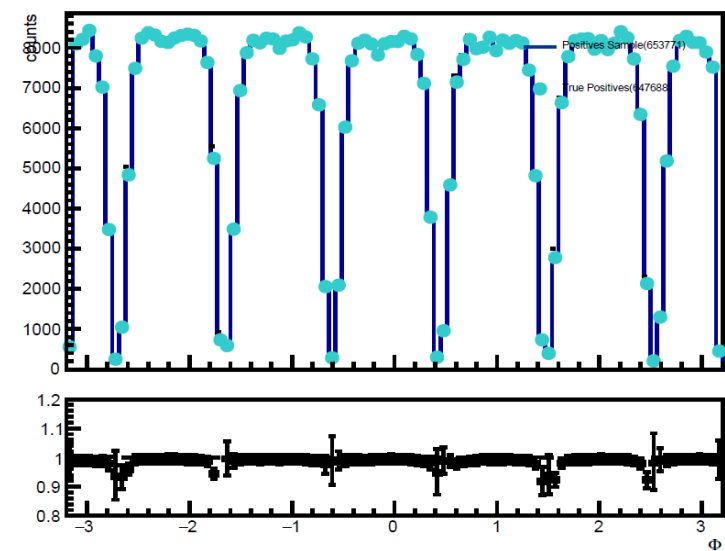
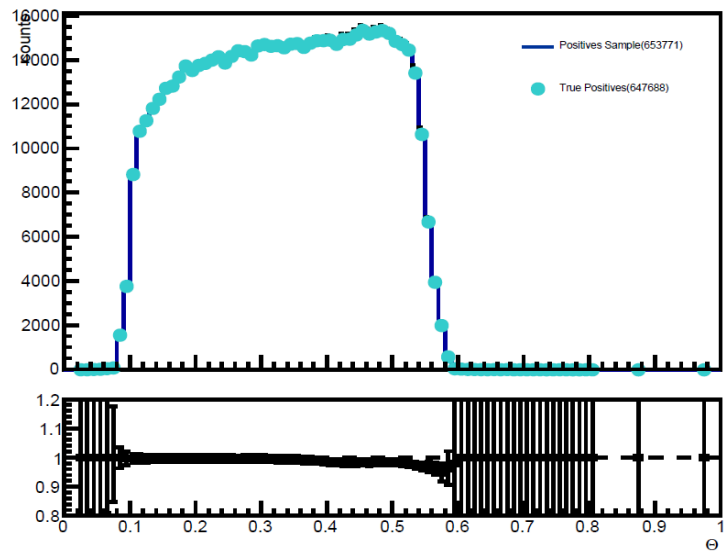
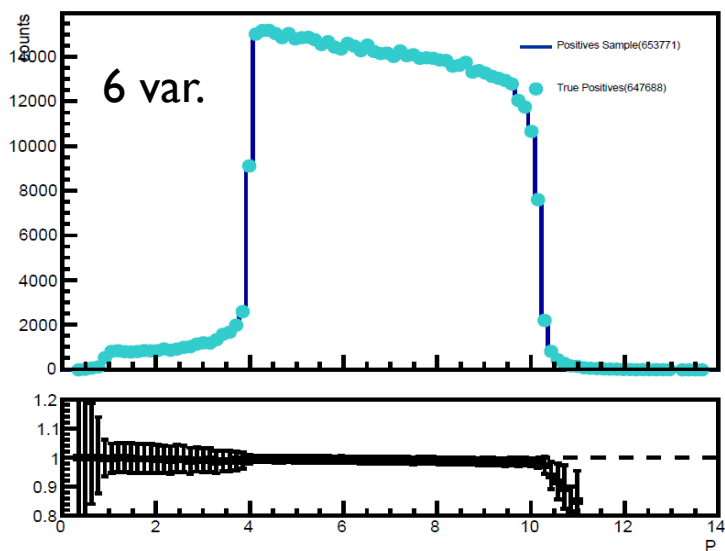
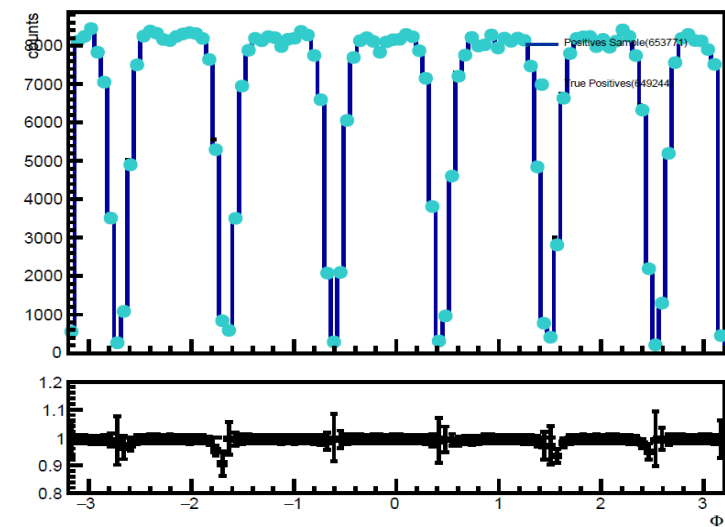
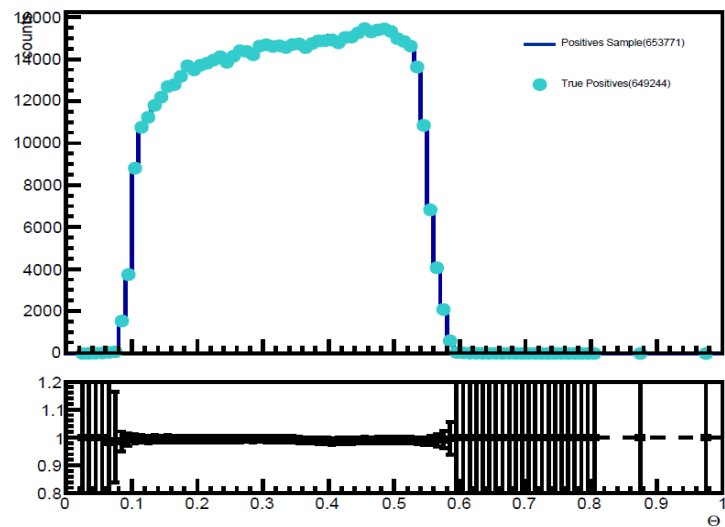
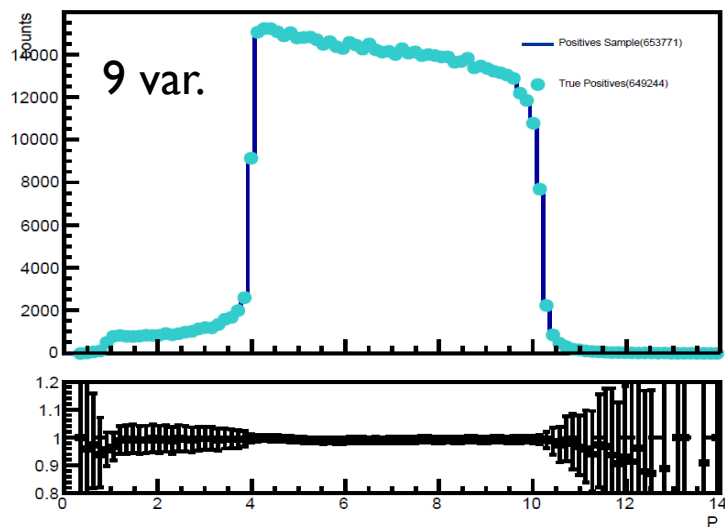
CLAS12 Preliminary - Dilepton final state



III - Lepton PID using machine learning



Validation - Efficiency



Validation - Contamination

