

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

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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

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IV

J/ψ event selection and resolution

V

Maximum likelihood fit for the extraction of the TCS parameters

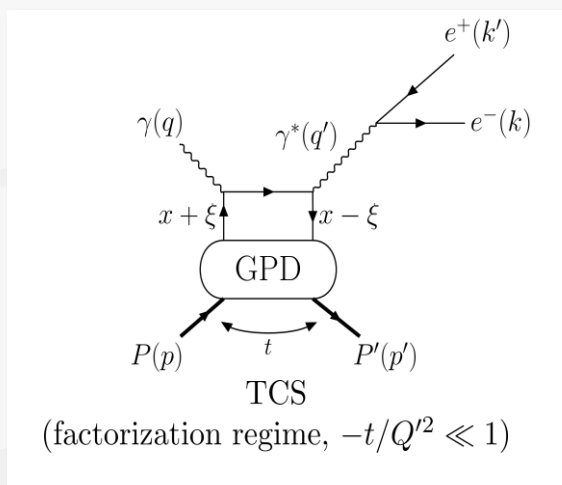
VI

Take-aways and timeline for future work

Motivations for dilepton final state measurement

Timelike Compton Scattering

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

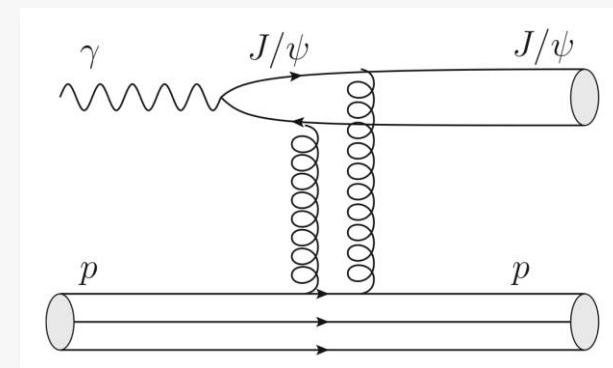


$$\frac{d^4\sigma_{INT}}{dQ'^2 dt d\Omega} \propto \frac{L_0}{L} \left[\cos(\phi) \frac{1 + \cos^2(\theta)}{\sin(\theta)} \text{Re}\mathcal{H} + \dots \right]$$

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

J/ψ photoproduction at threshold

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

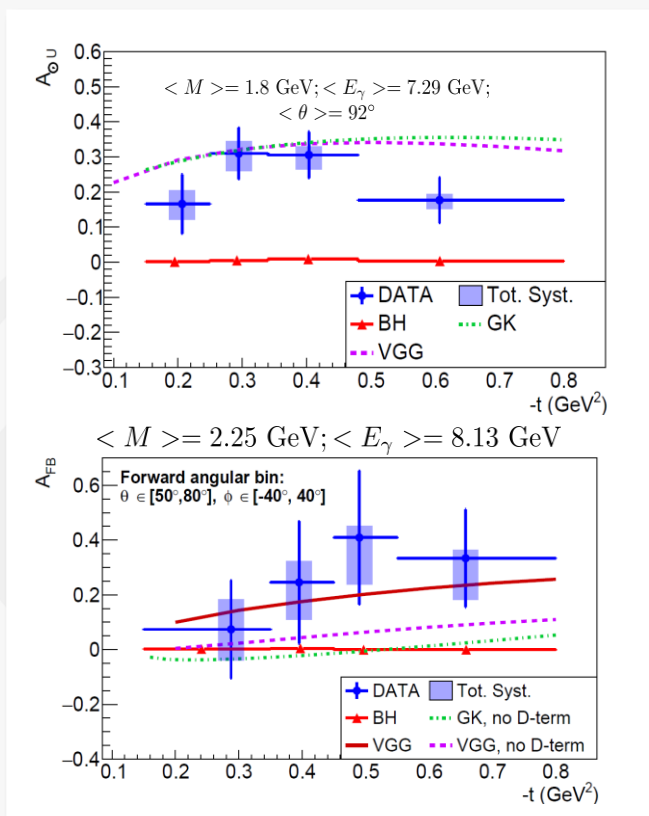


- The t -dependence of the cross-section allow to access gluon Gravitational Form Factors (GFFs), mass radius of the nucleon and gluon GPDs (under 2-gluon exchange assumption and no open-charm contributions)
- Model-dependent limit on the branching ration of the Pc pentaquark.

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

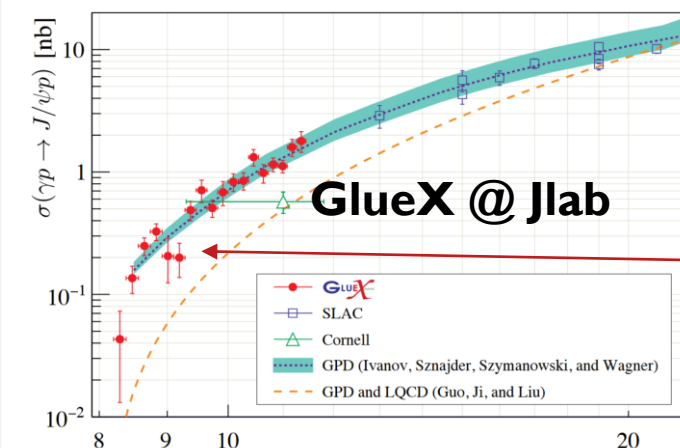


Figure in, Measurement of the J/ψ photoproduction cross section, S. Adhikari et al. (GlueX Collaboration) arXiv:2304.03845

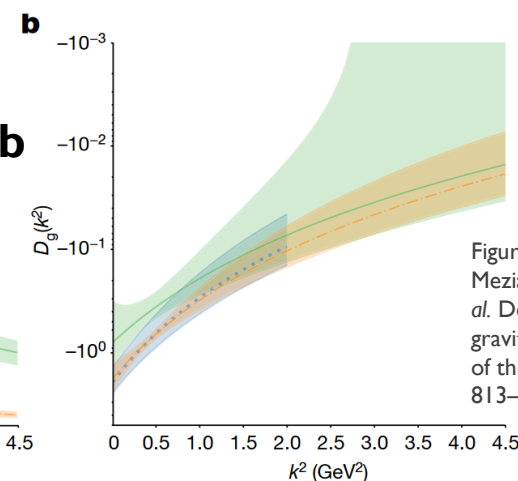
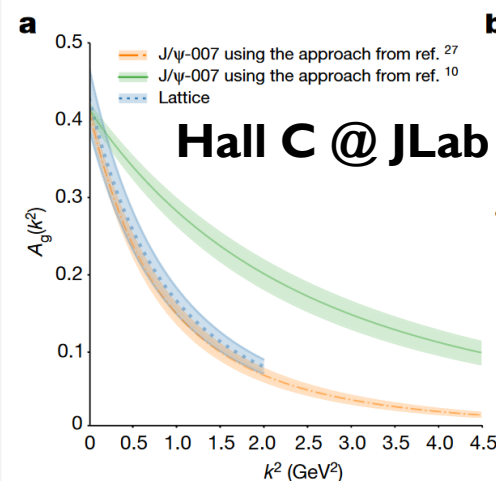
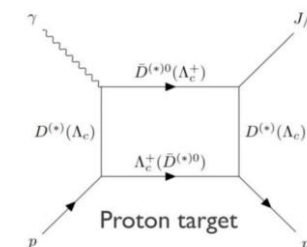
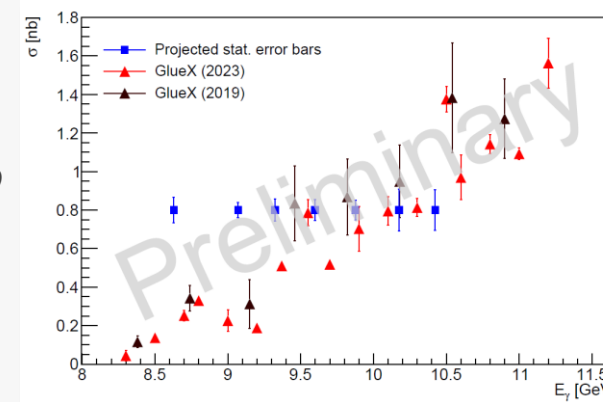


Figure in Duran, B., Meiziani, Z.E., Joosten, S. et al. Determining the gluonic gravitational form factors of the proton. *Nature* 615, 813–816 (2023)

Timelike Compton Scattering

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- Figure 1 is a plot showing the ratio of cross sections, A_{0U} , versus the negative logarithm of the squared invariant mass, $-t$ (in GeV^2). The plot includes data points with error bars for RGA Fall 2018 (blue diamonds) and projected data (grey diamonds). Theoretical curves are shown for VGG (magenta dashed line), BH (red solid line with triangles), and GK (green dotted line). The y-axis ranges from -0.3 to 0.6, and the x-axis ranges from 0.1 to 0.8. The BH curve is flat at 0, while VGG and GK rise from ~0.2 to ~0.35.

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



ID	Name	Start Date	End Date
1	Lepton AI PID	Sep 18, 2023	Dec 08, 2023
2	Data processing	Oct 09, 2023	Jan 02, 2024
4	Energy loss corrections	Sep 18, 2023	Oct 27, 2023
5	Momentum correction	Jan 22, 2024	Apr 09, 2024
9	Fiducial cuts	Jan 02, 2024	Feb 12, 2024
6	Momentum smearing	Jan 22, 2024	Apr 10, 2024
7	Radiative correction validations	Sep 18, 2023	Dec 08, 2023
8	Radiative corrections	Dec 11, 2023	Mar 01, 2024
10	Systematics	Apr 22, 2024	Jun 10, 2024
11	Analysis note writing	Sep 18, 2023	May 31, 2024
12	Article writing	Apr 23, 2024	Jun 18, 2024

23 Objective: summer 2024

General analysis strategy

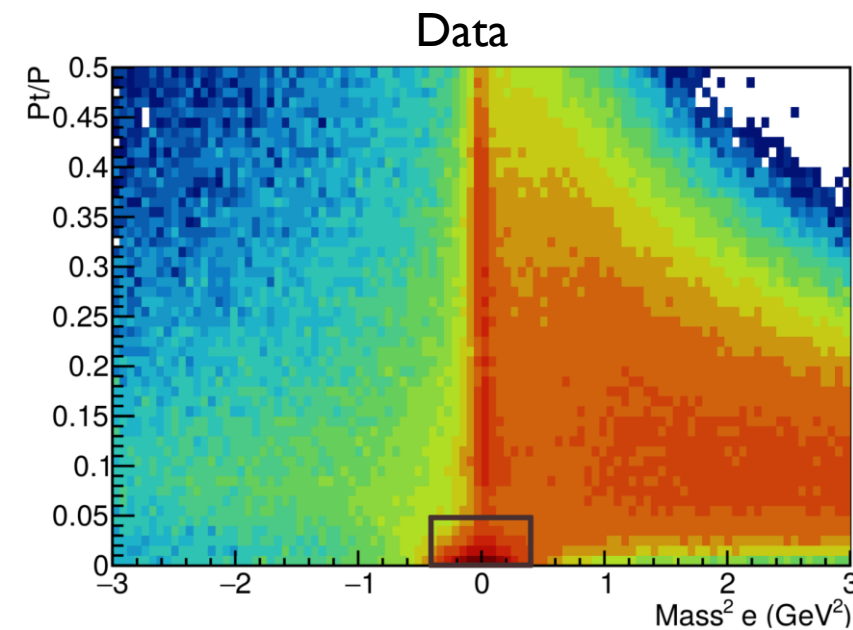
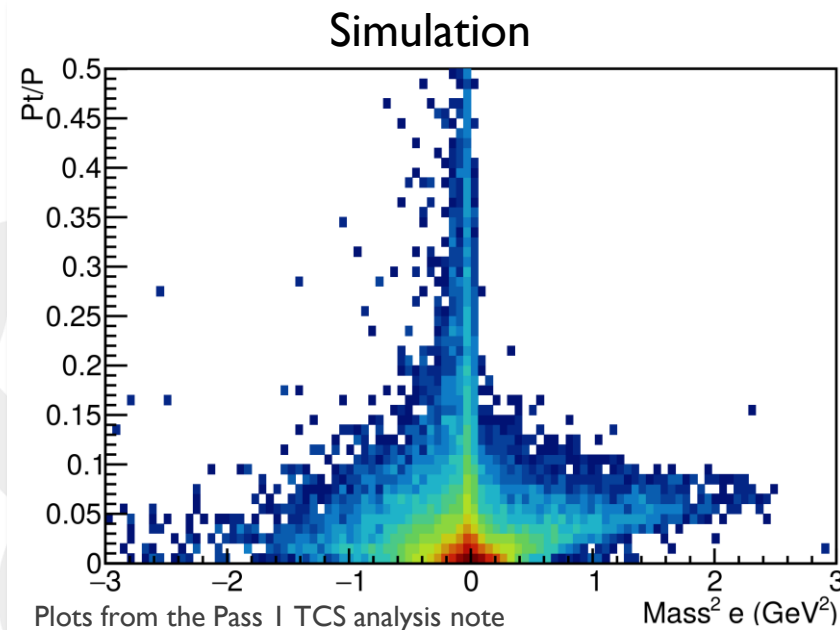
1) CLAS12 PID + Positron NN PID

$$ep \rightarrow (e')\gamma p \rightarrow (X)\overbrace{e^+e^-} p'$$

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

2) $|M_X^2| < 0.4 \text{ GeV}^2$

3) $|\frac{P_{tX}}{P_X}| < 0.05$ or $Q^2 < 0.1 \text{ GeV}^2$



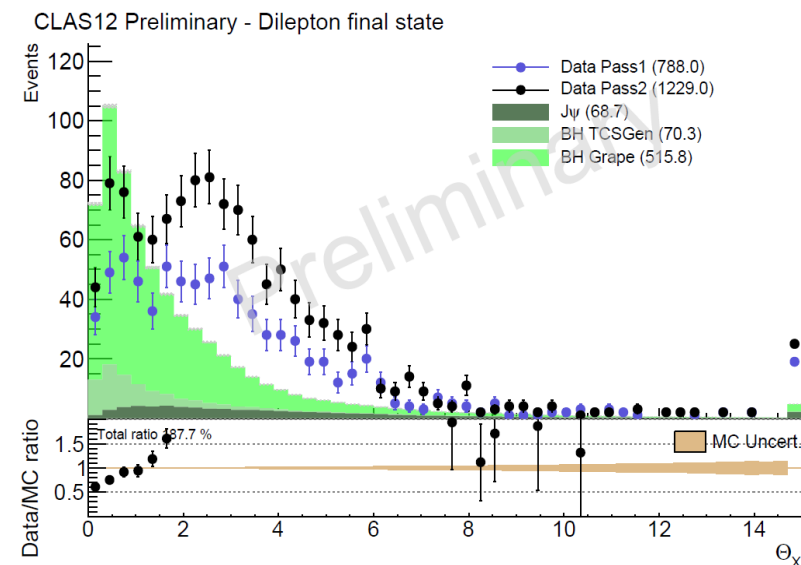
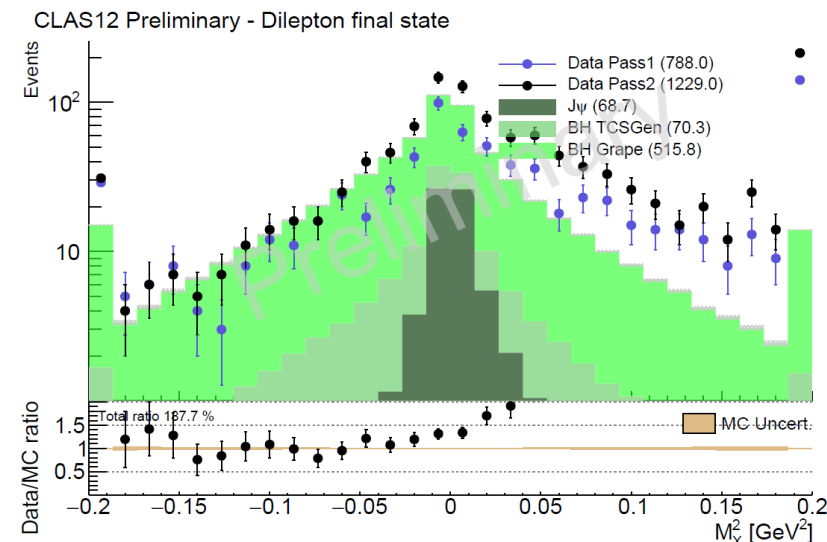
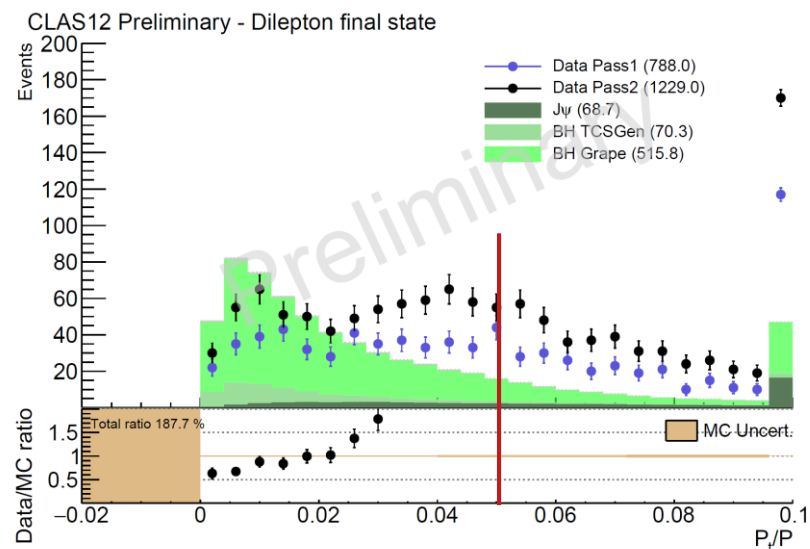
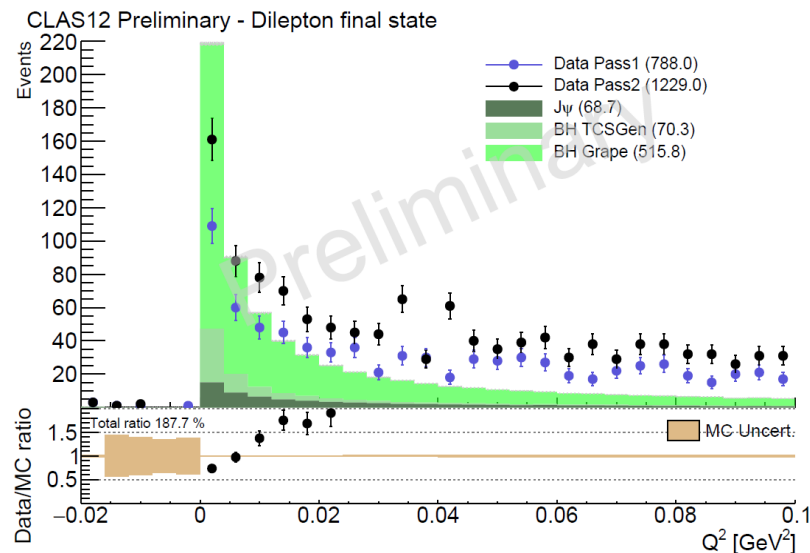
II - Pass 2 data: first look at Spring 2019



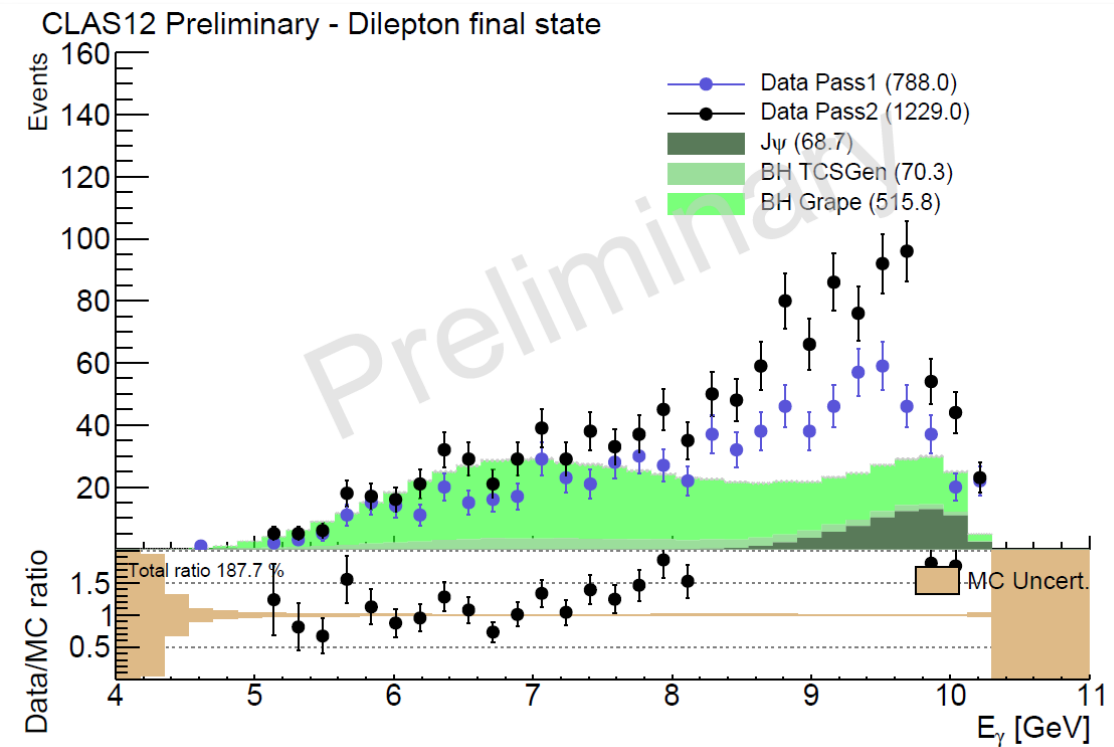
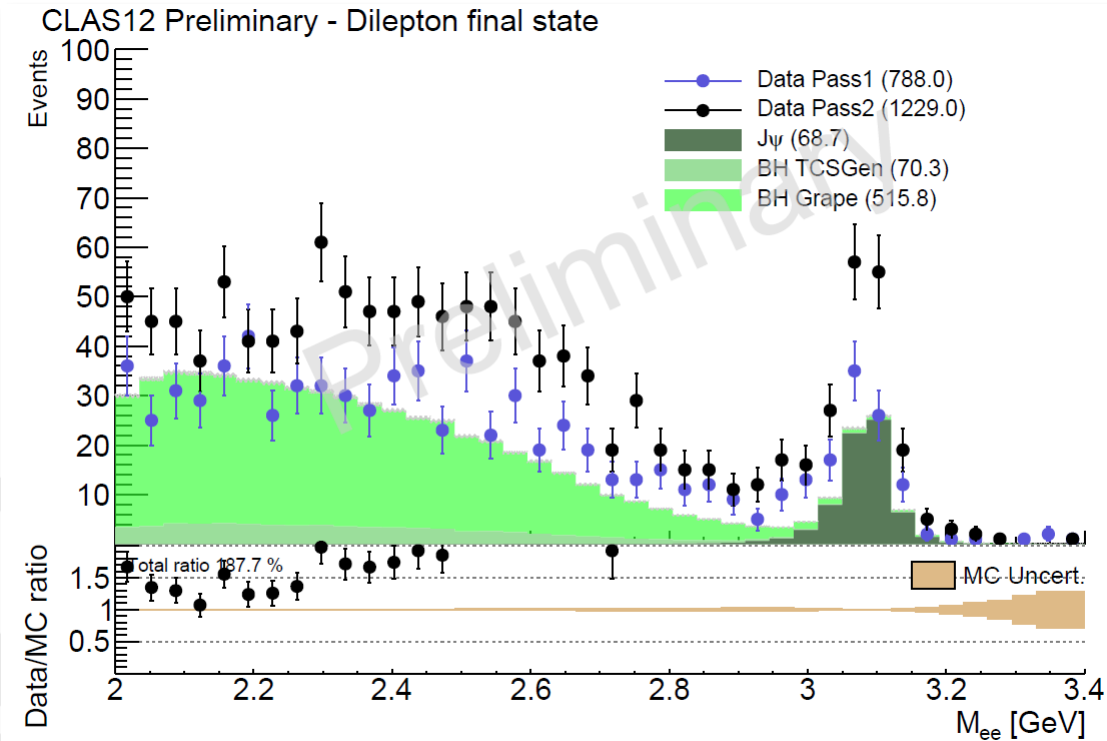
Comparison of data and MC (I)

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

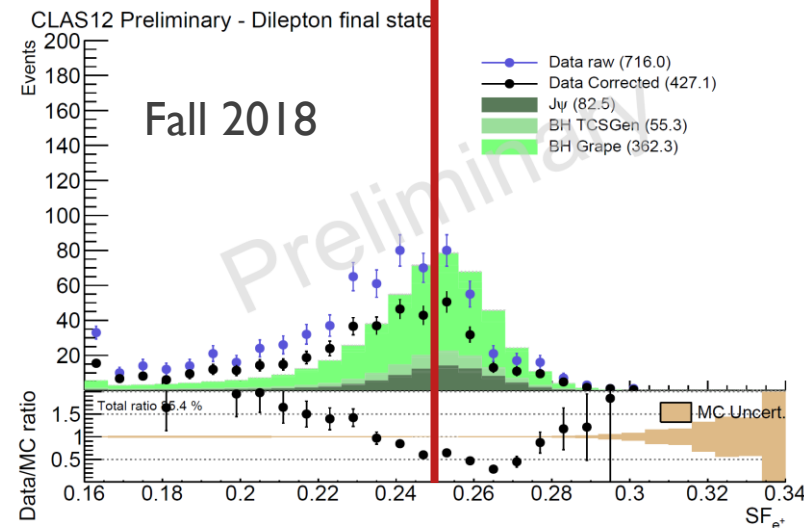
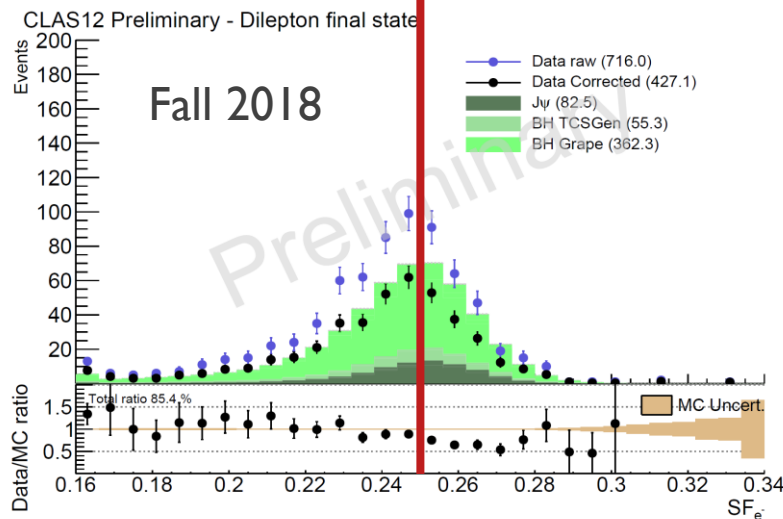
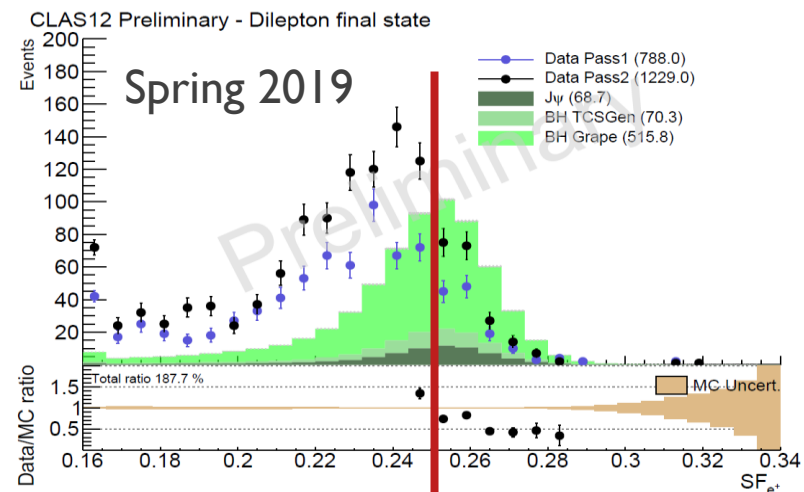
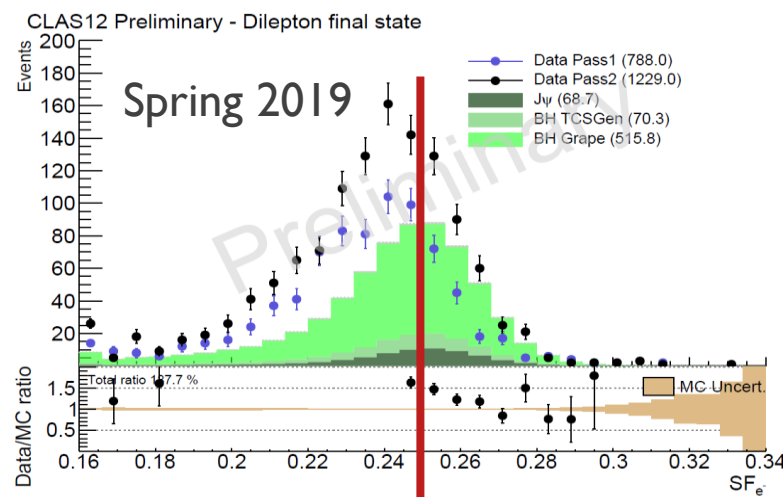


Comparison of data and MC (2)



- Same behavior is seen in Spring 19 and Fall 18 data: the large Q2 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

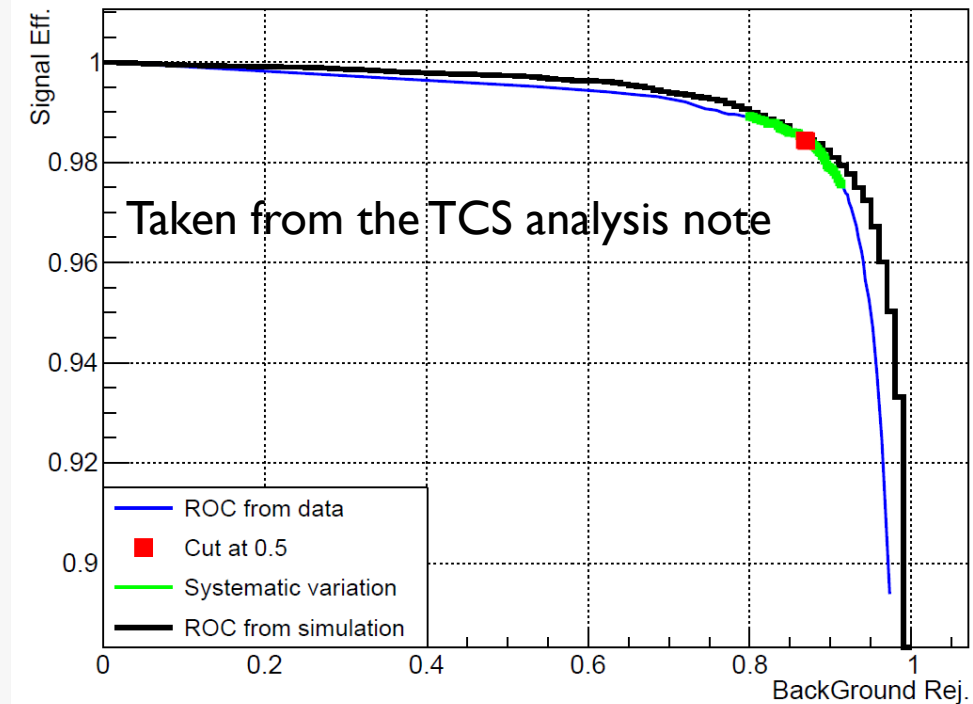


III - Lepton PID using machine learning



Motivations

- $$ep \rightarrow ep\pi^+\pi^- \quad \text{VS} \quad ep \rightarrow epe^+\pi^-$$



Taken from the TCS analysis note

Approach

- 2 (e⁺/e⁻) x 3 (Spring19/Fall18 in/out) = 6 classifiers

- Variables used: P , θ , ϕ , SFs and m2 of PCAL, ECIN, ECOUT

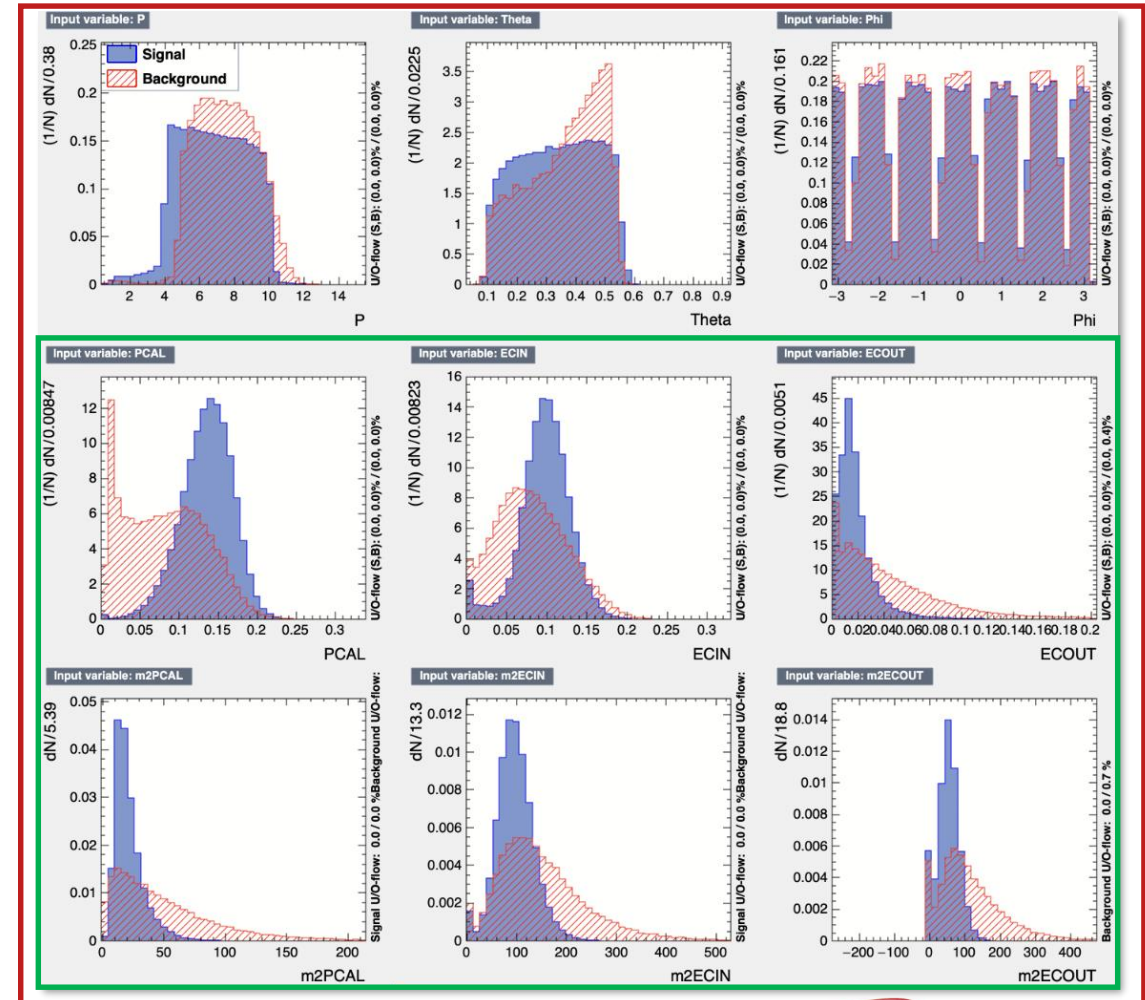
Method tested: NN, BDT

- Signal: flat $e^{+/-}$ distribution, reconstructed as $e^{+/-}$

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

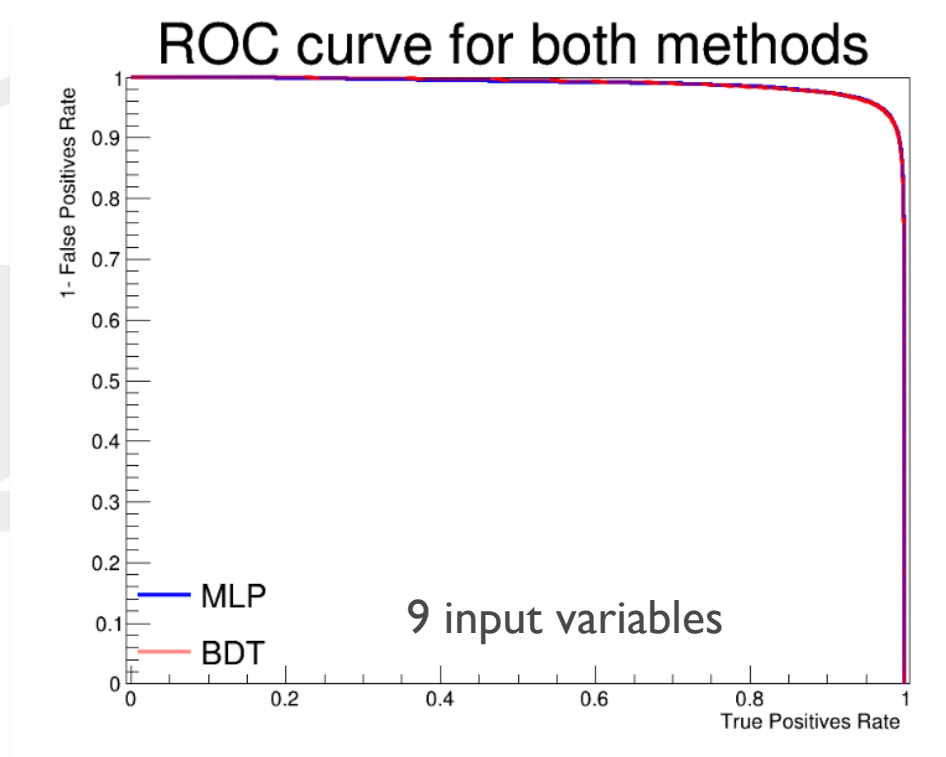
- Only RGA Spring 2019 for now

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



Performances

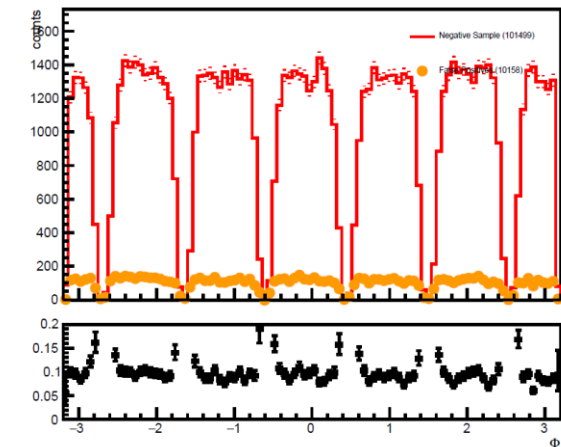
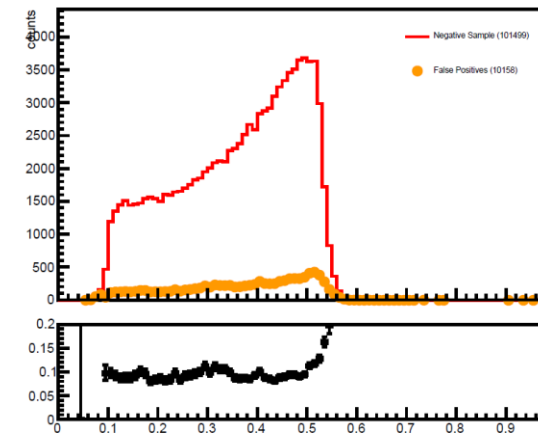
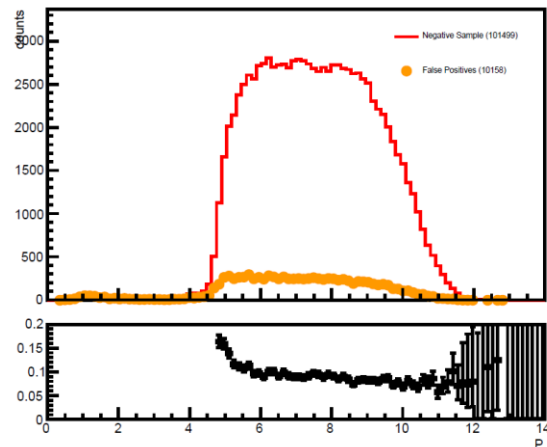
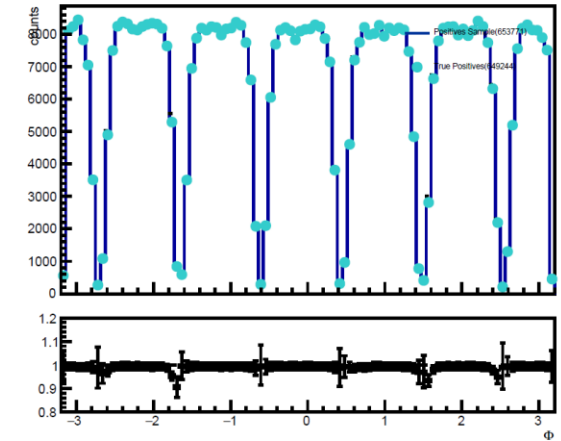
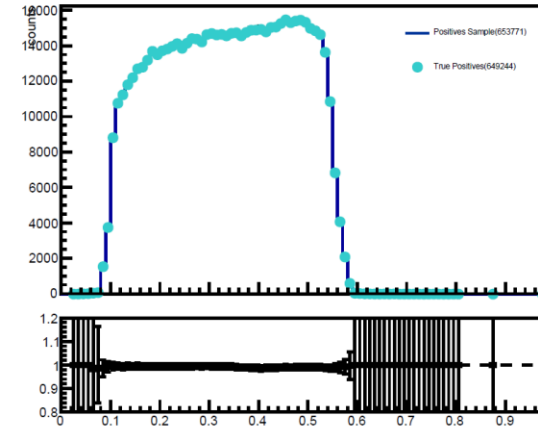
- We tested both 6 and 9 input variables, for 2 methods **NN** and BDT.
- Signal efficiency: 99.4 %
- Background reduction: 10%



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 ⁺	Actual π ⁺
Predicted e ⁺	649244	10158
Predicted π ⁺	4527	91341
Performances	TPR 99.4%	FPR 10%

NN. 9 Variables

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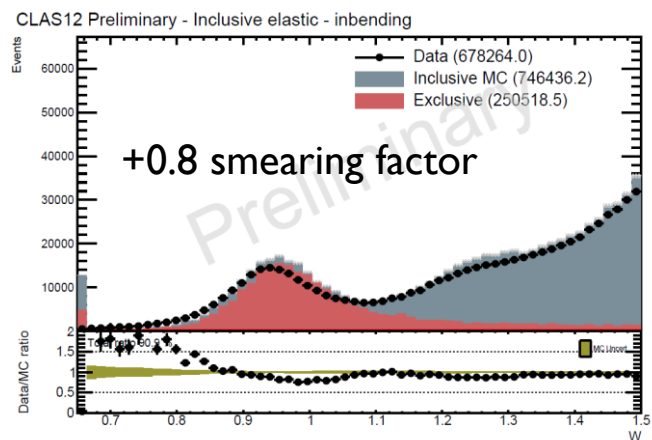
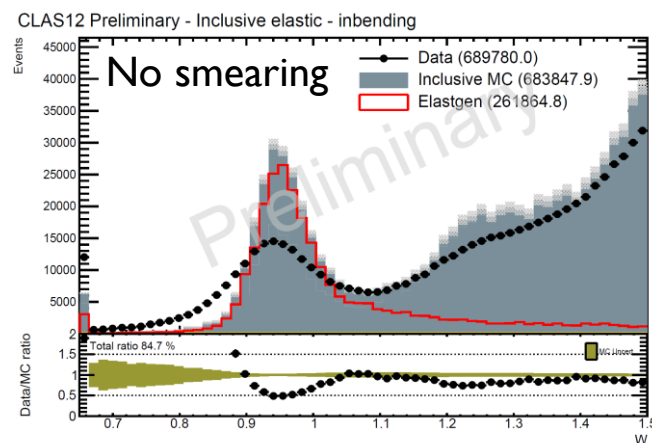
IV - J/ψ event selection, resolution and cross-section



Motivations

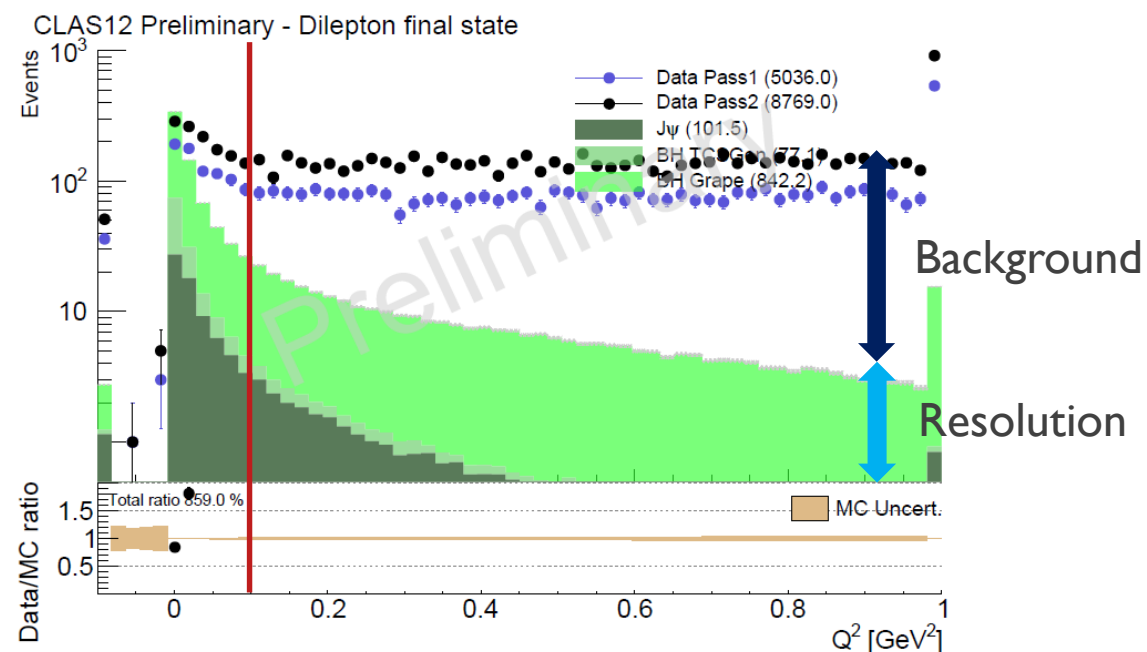
Inclusive elastic events

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



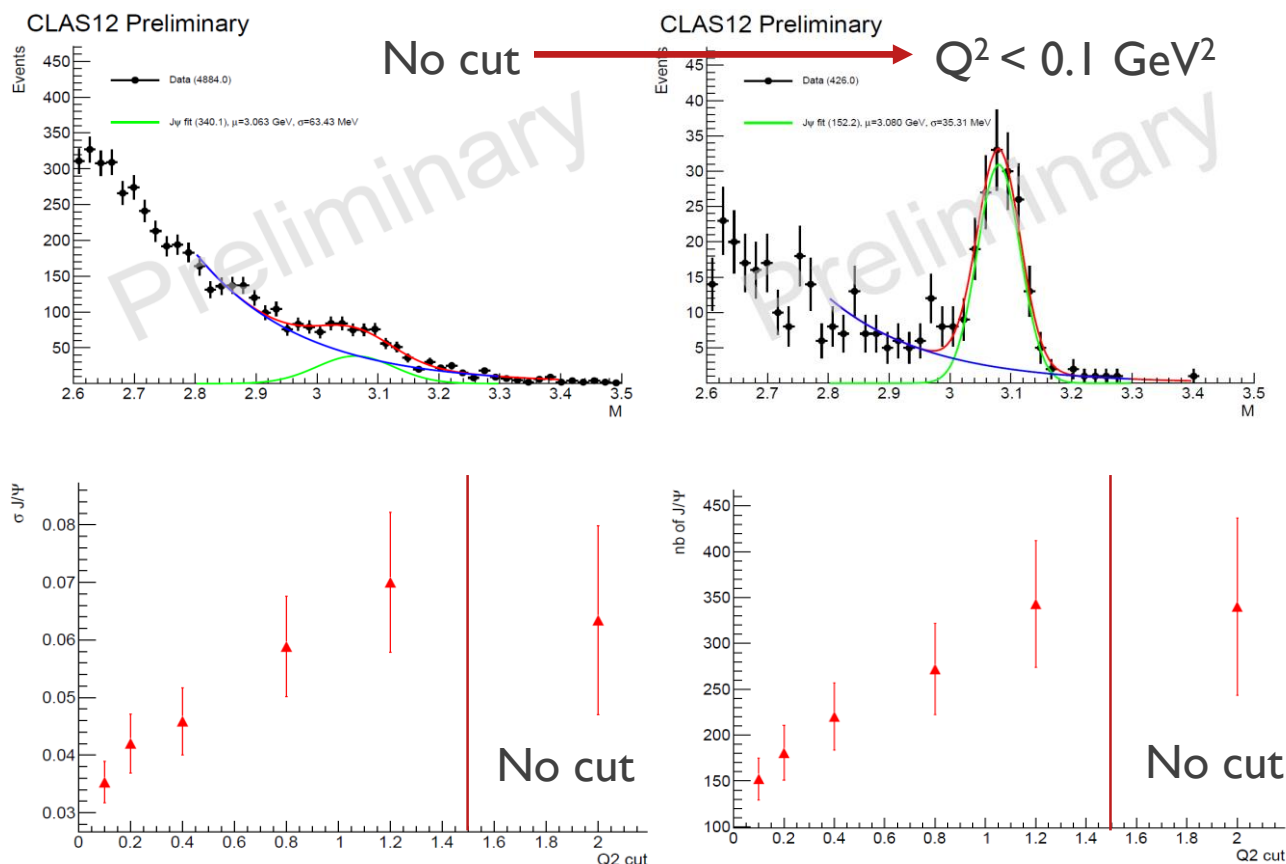
Inclusive elastic events

- 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

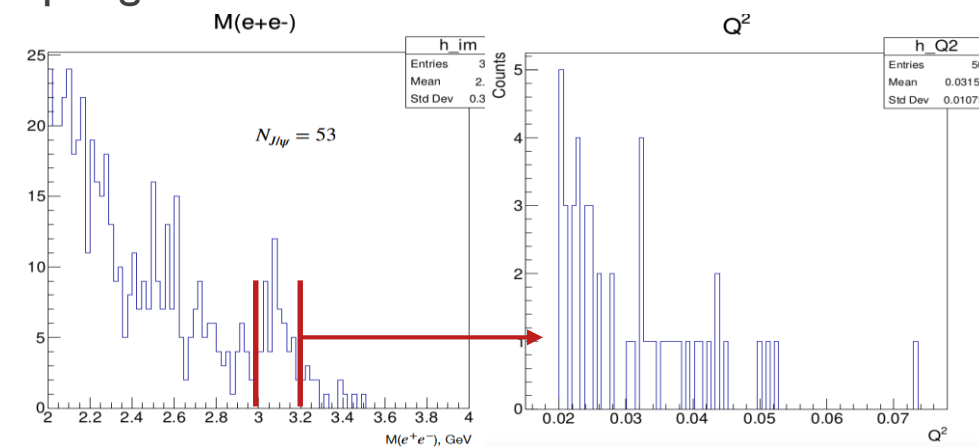


Maximum virtuality of the incoming photon

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

- Using tagged photo-production events, one can measure the virtuality of the incoming photon with only the FT resolution involved

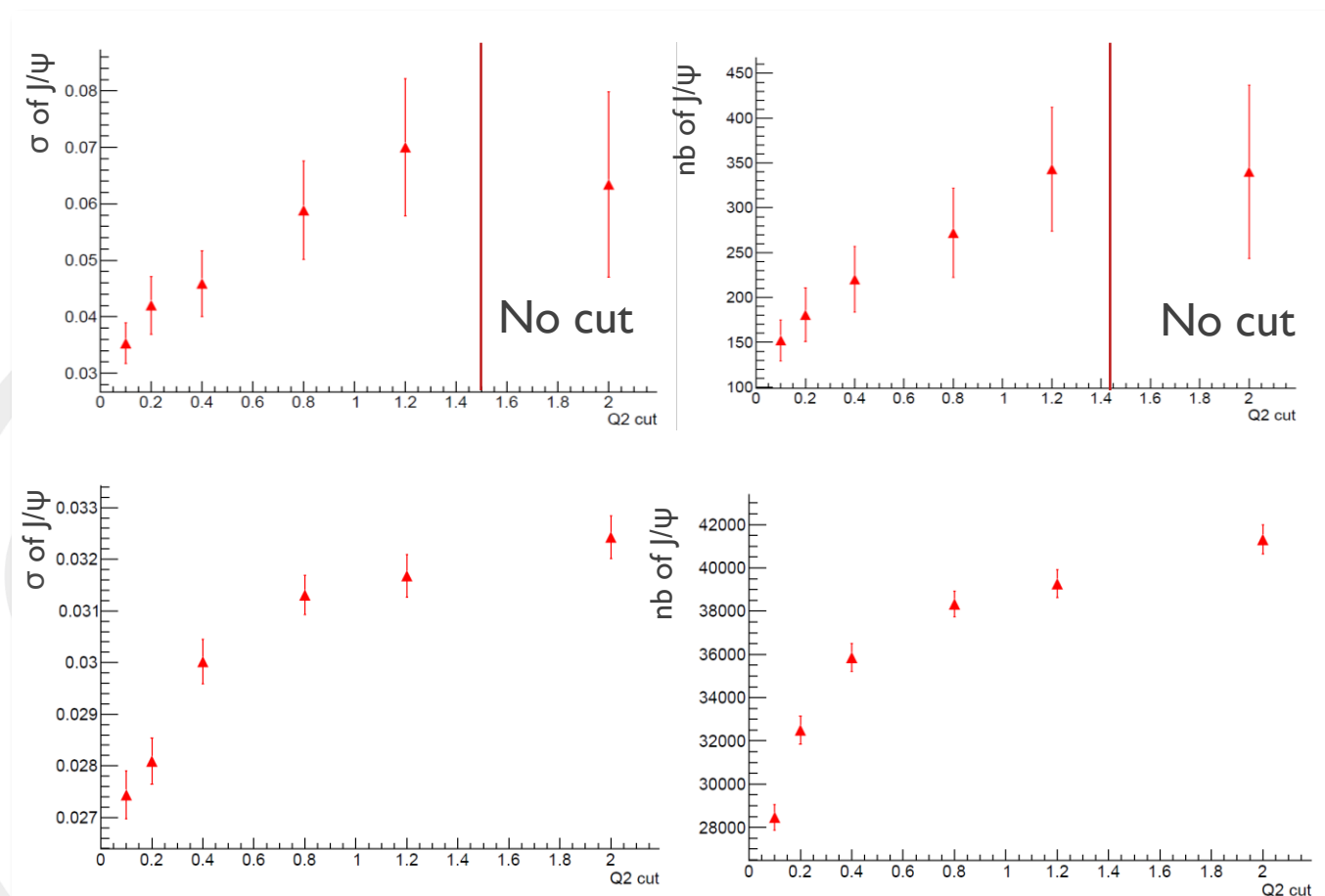
Spring 2019 – Full statistics



Material provided by M.Tenorio Pita

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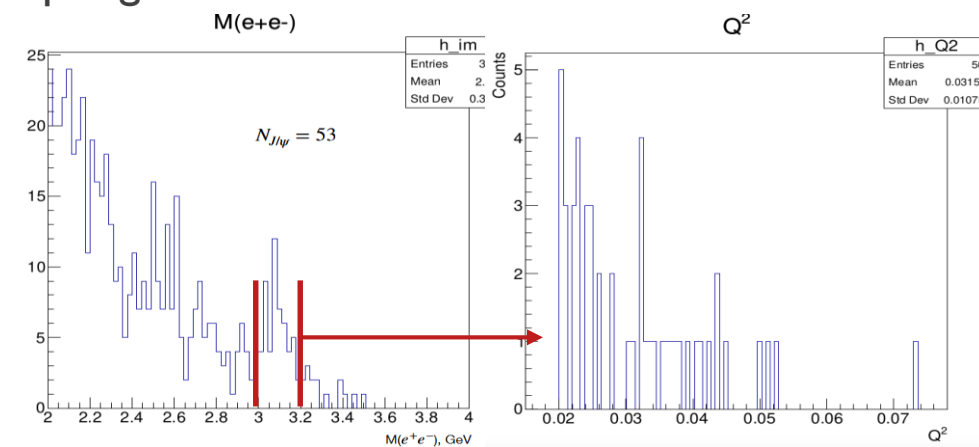


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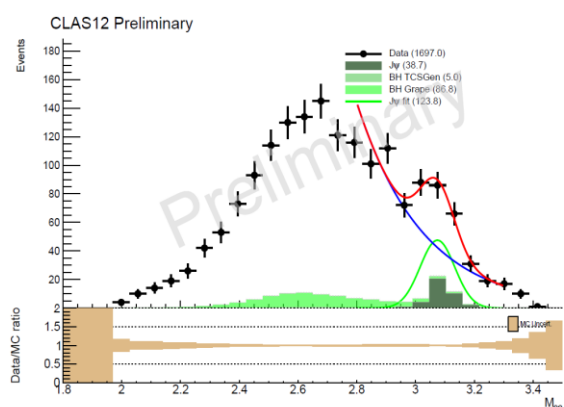
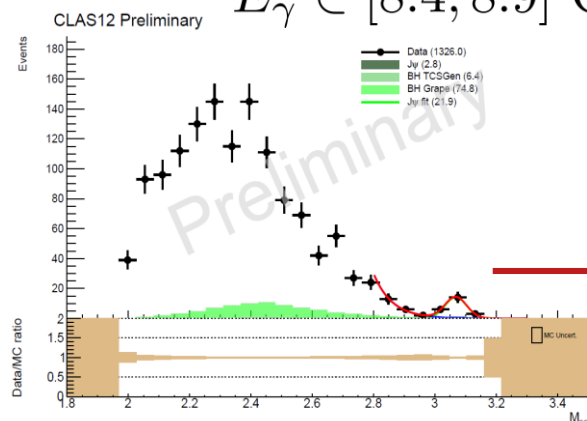
Spring 2019 – Full statistics



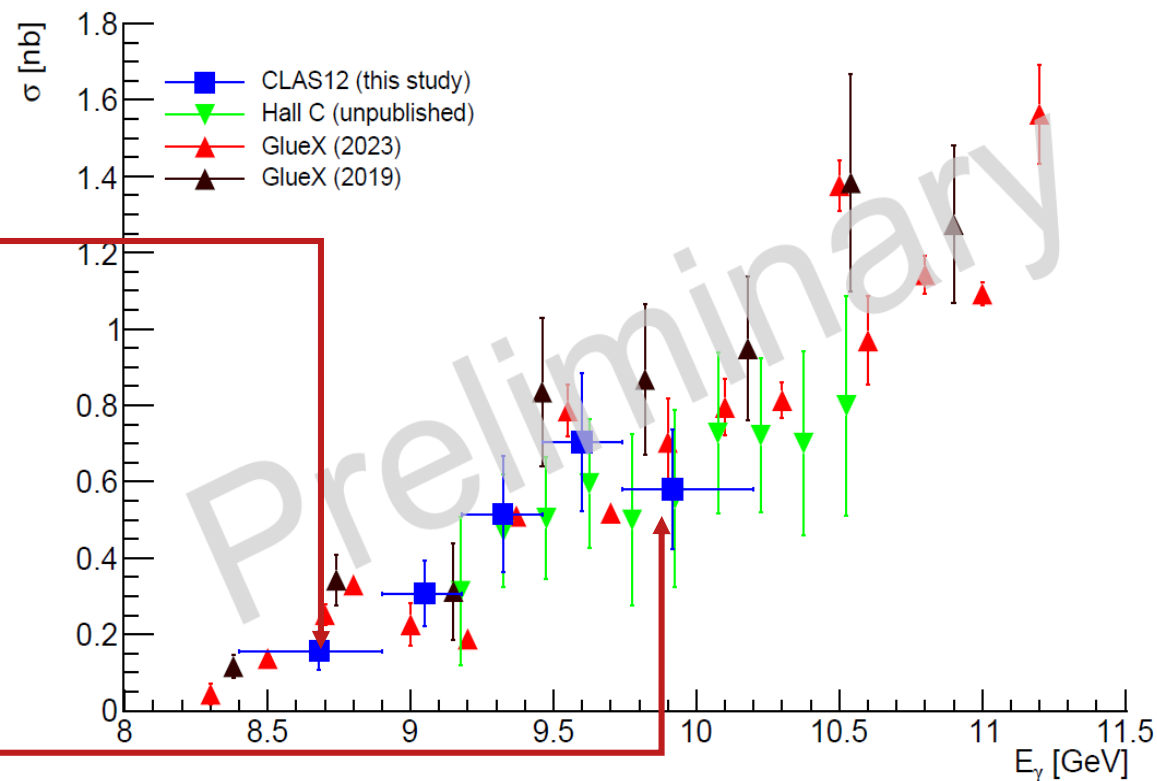
Material provided by M.Tenorio Pita

Effect on the CS extraction

$E_\gamma \in [8.4, 8.9]$ GeV



$E_\gamma \in [9.74, 10.2]$ GeV



- 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



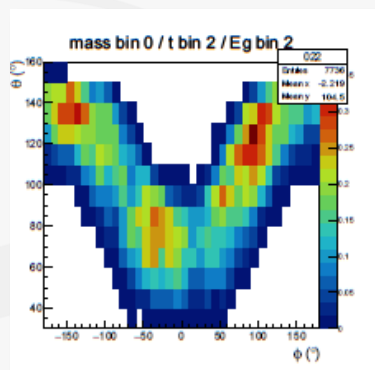
Motivations and formalism

All material provided by D. Glazier

Limitation of the current approach

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

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



... 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

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

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

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,

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

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

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

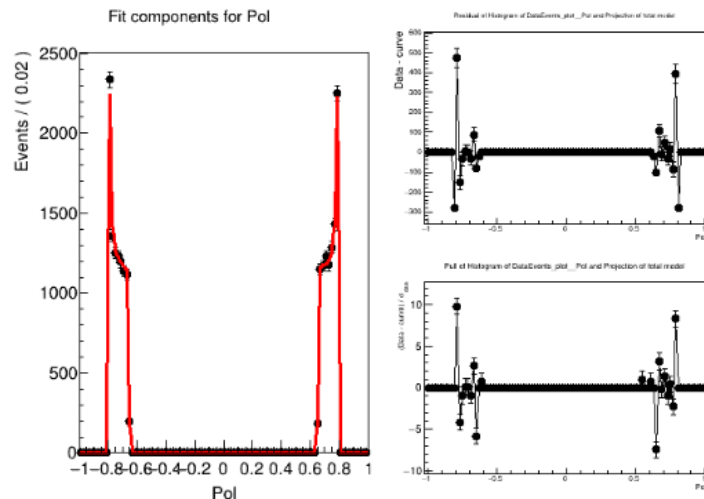
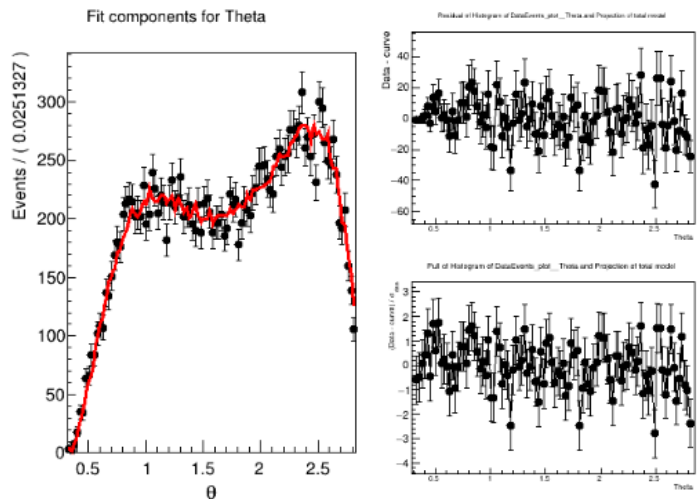
$$\mathcal{L}(\theta_j, Y) = -\ln L(\theta_j, Y) = -\sum_k^N \ln \frac{I(x_{i,k} : \theta_j)}{\sum_s^M I(x_{i,s} : \theta_j)} + Y - N \ln Y$$

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

Reproduce input parameters ?
BH = 0.6 ; ImM = 0.7; TCS = 0.2

BH	5.5324e-01	+/-	1.60e-02
ImM	8.1496e-01	+/-	3.21e-02
TCS	2.5344e-01	+/-	1.72e-02

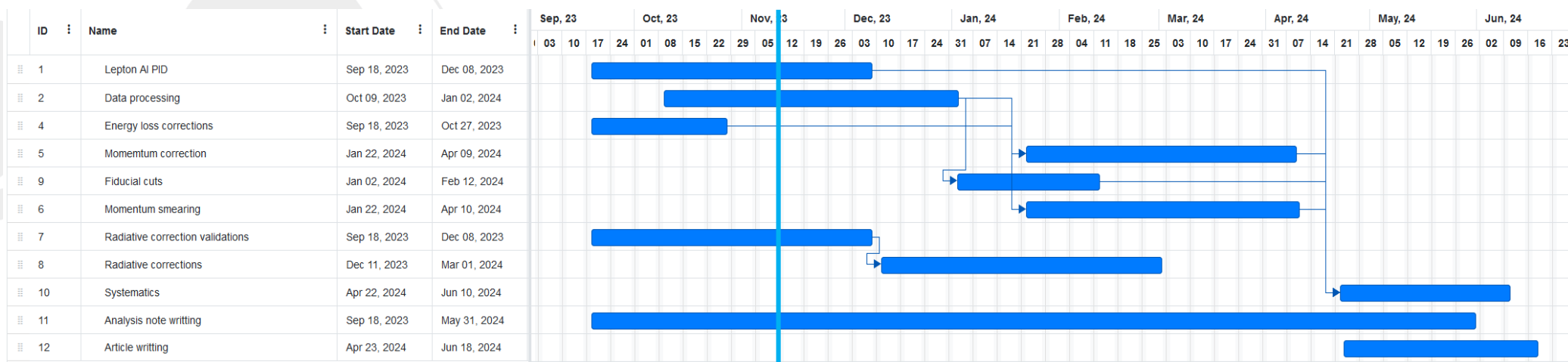
Slight bias from BH to TCS ?



- Method based on brufit
- Tested on MC only
- Fitted function

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), Stepan (J/ψ on RGA).
- Spring 19 Pass 2 dataset looks good, with similar issue than Pass 1 (Resolution and high- Q^2 background).
- AI PID for lepton is well underway and consistent with Pass 1 analysis.
- Maximum likelihood fit method is being developed for TCS observable extraction.



Back-up

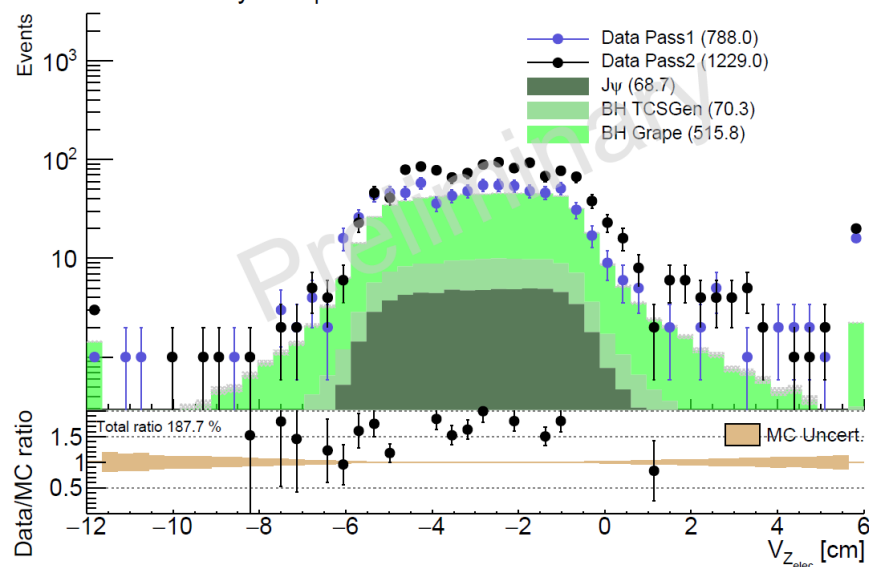


II - Pass 2 data: first look at Spring 2019

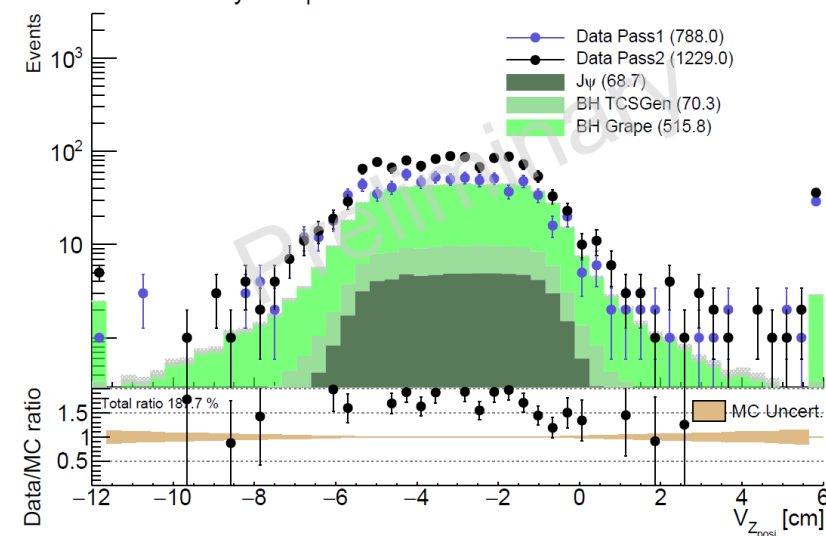


Spring 19 Pass 2: Vertices

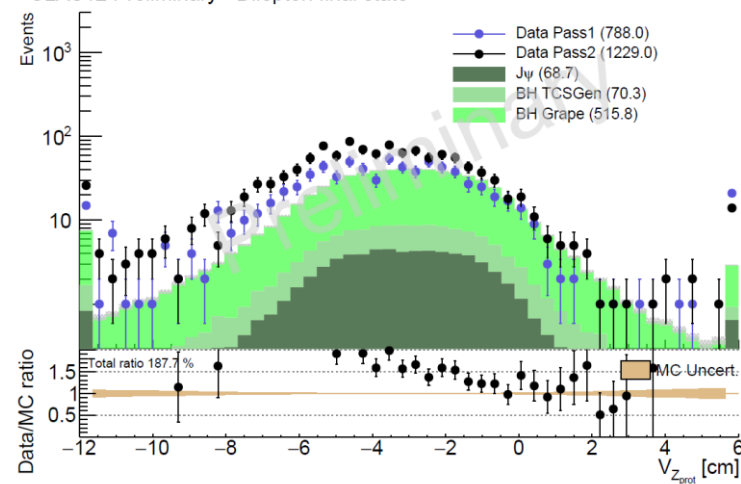
CLAS12 Preliminary - Dilepton final state



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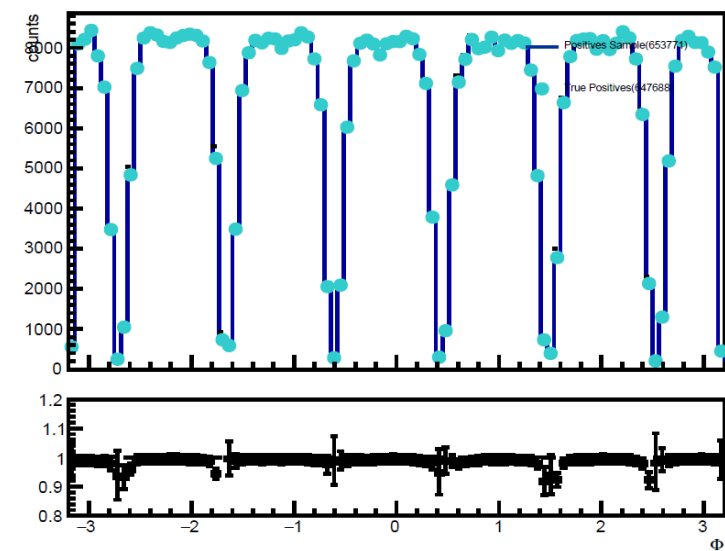
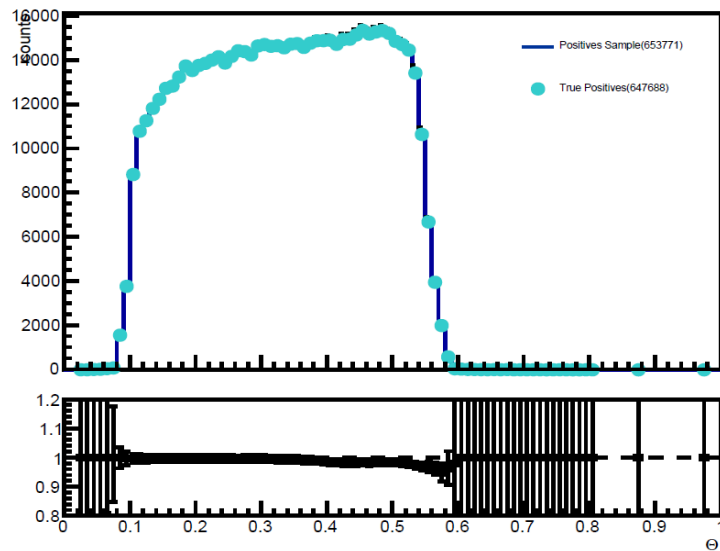
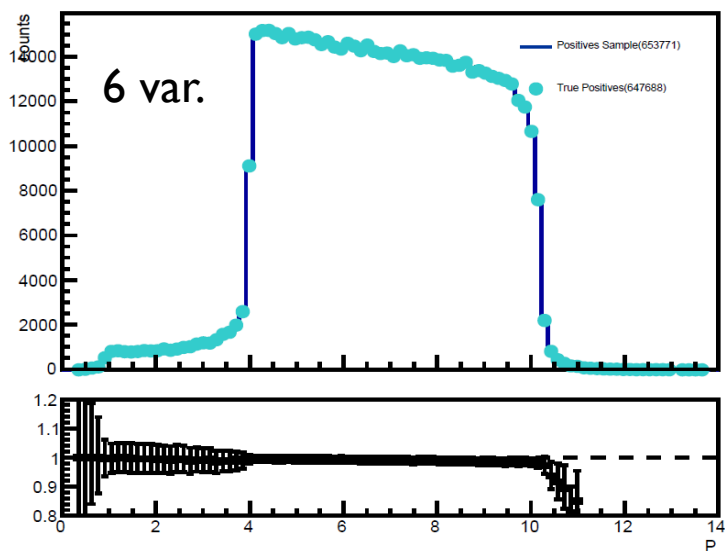
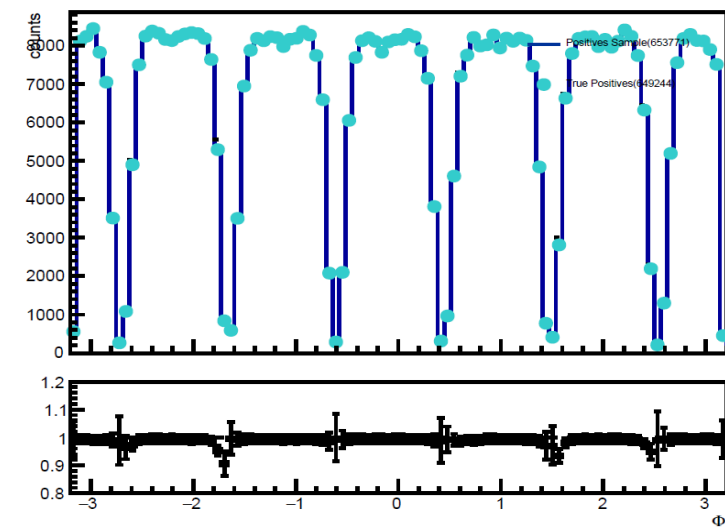
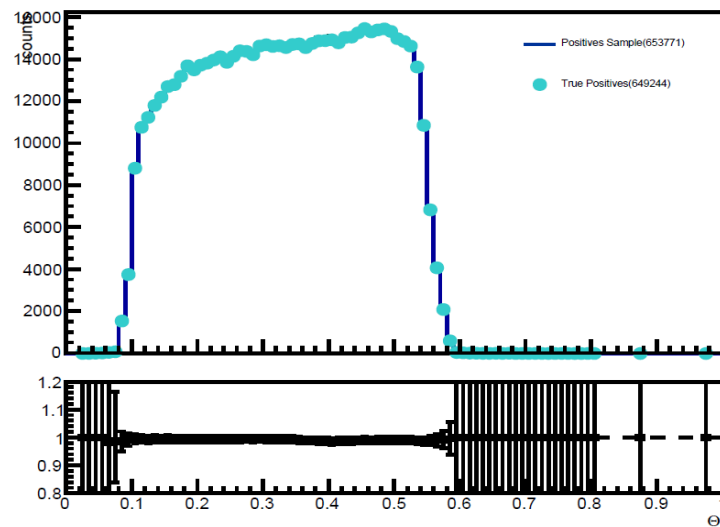
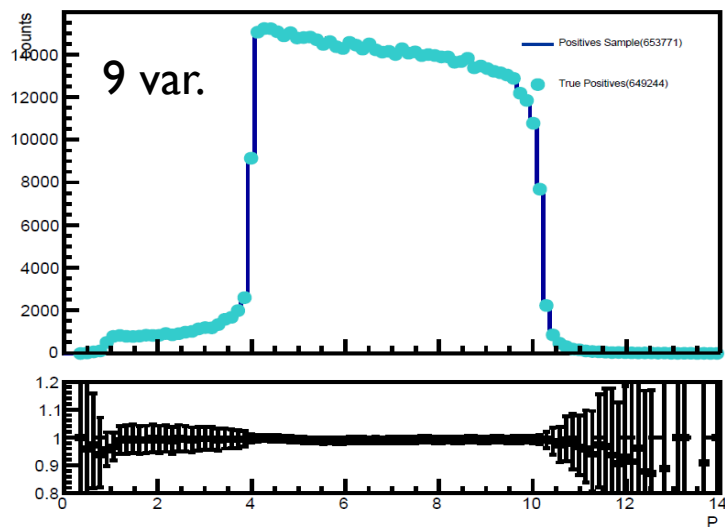
CLAS12 Preliminary - Dilepton final state



III - Lepton PID using machine learning



Validation - Efficiency



Validation - Contamination

