

Analysis Tools for MesonEx at CLAS12 Derek Glazier University of Glasgow

## Hadron 2015 Newport News, VA



## MesonEx

(also Carlos talk)

HASPECT

Data Handling Software Longitudinal Plots

Likelihood Analysis

## Jefferson Lab at 12 GeV



## MesonEx - Spectroscopy with CLAS12

## Primary Physics Goal for CLAS12 e- beam

- Nucleon Structure(not discussed here)
- But high potential for Meson spectro. (see also Hall D GLUEX)

•Strategy :

- High Intensity electron Beam
  - Tag quasi-real photons
- Large Acceptance Magnetic Spectrometer
- Many final states
- Linearly polarised photons
- Amplitude analysis sensitive to small contributions
  - Close interplay of exp theory

## HASPECT

International collaboration preparing for MesonEx at CLAS12

Implement reliable amplitudes
 Revisit techniques from earlier efforts
 Accessible to all
 Common Tools e.g IU AMPTOOLS
 Work closely with JPAC

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Frequent meetings and Workshops
     HASPECT weekly meetings with experimentalists and JPAC
     HASPECT weeks with guests from other projects
     ATHOS Amplitude Analysis Workshops
                                                              Theoretical support:
                                                               A.Szczepaniak (IU/JPAC), V.Mathieu (IU),
      . . .
                                                               E.Santopinto (INFN-GE), A.Vassallo (GE),
                                                              I.Ferretti (UMAS)
Apply/develop with existing CLAS data
                                                               Experimental Analysis:
\gamma p \rightarrow N \pi \pi \quad \gamma p \rightarrow N K K \quad \gamma p \rightarrow N \eta
                                                               M.Battagleiri, R.deVita, A.Celentano,
\pi\gamma p \rightarrow N \omega \gamma p \rightarrow N \pi \pi \pi \gamma p \rightarrow N \eta \pi \pi
                                                               S.Fegan (INFN-GE), A. Filippi (INFN-TO),
                                                               D.Glazier(Glasgow), S.Hughes (Edinburgh),
\gamma p \rightarrow N \pi^+\pi^-\pi^-K^+
                                                               K.Hicks (OhioU), S.Lombardo (Cornell),
\gamma p \rightarrow N \phi \pi \gamma p \rightarrow N \phi \eta \gamma p \rightarrow N....
                                                               A.Rizzo (RomaTV), I Stankovich (Edinburgh),
                                                               L.Zana (Edinburgh)
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## Use of Software in hadron physics



## Use of Software in hadron physics



## HASPECT Event Reconstruction

Provide code to handle routine tasks allowing procedures to become standardised Input/Output/Interfacing Histogramming Particle/reaction identification Event weighting

Maintain normal ROOT flexability for users

Users shift to physics and systematic studies

Promote full potential of ROOT Based on TSelector Tree analysis class Use of TEntryList class to prevent duplicating data ROOT system takes care of compilation and configuration Parallel ROOT Facility (PROOF) Statistical Analysis Packages (RooFit/Stats) https://github.com/HASPECT/Events



## Event Weighting : Qvalue

Developed Mike Williams (CMU) for CLAS analysis Look for N nearest events in kinematic space Fit discriminatory variable for signal and back. e.g. missing mass Qval = S/(S+B) Qval can then be used to weight events HASPECT

\*Selector class inherits additional Qvalue class \*Use RooFit event-by-event maximum likelihood \*Near. Neigh. saved \*Limit NN search with TEntryList



## Event Weighting : sWeights

M. Pivk,F.R. Le Diberder,Nucl.Inst.Meth.A 555, 356-369, 2005

Given discriminatory PDF for signal and background calculates weight :  $N_s = N_s$  N = Number of species

$$_{s}\mathcal{P}_{n}(y_{e}) = \frac{\sum_{j=1}^{N_{s}} \mathbf{V}_{nj} \mathbf{f}_{j}(y_{e})}{\sum_{k=1}^{N_{s}} N_{k} \mathbf{f}_{k}(y_{e})}$$

Part of RooStats(used here) Can include multiple signal and background species



N<sub>k</sub> = Yield for species k
V = covariance matrix
Can fit multidimensional
discriminatory PDF
MKO distribution fit

 $f_{\mu}$  = PDF for species k









sWeights correctly reproduces M2 signal and background shape and uncertainties

## Example CLAS Analysis

 $\gamma p \rightarrow f1(1285) p$  $\rightarrow \pi^{+}\pi^{-}(\eta)p$ 



#### Van Hove Plots (Longitudinal)





CLAS g11 dataset



Example 3-3.8GeV  $\gamma p \rightarrow K+K-p$ 

$$p_{K^{+}L} = \sqrt{\frac{2}{3}}q\sin\omega,$$

$$p_{K^{-}L} = \sqrt{\frac{2}{3}}q\sin\left(\frac{2}{3}\pi + \omega\right),$$

$$p_{PL} = \sqrt{\frac{2}{3}}q\sin\left(\frac{4}{3}\pi + \omega\right).$$

## Example $\gamma p \rightarrow K+K-p$ at around 3-3.8GeV







## Monte-Carlo Sampling of likelihood

Minuit Maximum Likelihood Single solution

We often have local maxima How to choose initial parameters? How to judge goodness of fit? Implement Occam's Razor?

#### MCMC

Samples full likelihood Not very efficient sampling Only finds unimodal solution Difficult to calculate evidence

Nested Sampling for General Bayesian Computation -

• J. Skilling, 2006, International Society for Bayesian Analysis More efficient sampling model selection via Bayes factor Intrinsically calculates evidence +Occams Razer

MultiNest

F.Feroz and M.P. Hobson, 2007, *Cambridge*, arXiv:0704.3704v3 Finds many maxima and the evidence for each Bayesian Statistics is used for parameter estimation and hypothesis testing.

INPUT OUTPUT Likelihood x Prior = Evidence x Posterior  $P(D|\theta,H) \times P(\theta|H) = P(D|H) \times P(\theta|D,H)$  $L(\theta) \times \pi(\theta)d\theta = Z \times p(\theta)d\theta$ 

Where  $\mathbf{D}$  is the data set,  $\mathbf{0}$  is a parameter vector, H is a model and

Evidence = 
$$Z = \int L(\theta) \pi(\theta) d\theta$$

Likelihood integrated over the prior distribut.

## Nested Sampling estimates evidence and finds likelihood maxima.

Define accumulated mass

$$X = \int_{L(\theta) > \lambda} \pi(\theta) d\theta$$

Evidence can be evaluated as a 1D integral of likelihood over the prior accumulated mass

$$Z = \int L(X) \, dX$$

N "live points" maintained and one with lowest likelihood is replaced



# Nested Sampling procedure involves points exploring the likelihood.



Note that only one new point needs to be calculated at each iteration, the N points at iteration *i* are the active live points.

Rejected points are kept to give the posterior distribution (All samples)





#### MultiNest Dalitz Fit, Posterior and Live Points

R12 Real V R12 Imag. R12 Real V R13 Real **FitPars** FitPars 100 Entries 1100 Entries 1100 0.1089 Mean x Mean x 0.1089 80 -0.06577 Mean y Mean y 1.81 RMS x RMS y RMS x 56.72 56.72 60 RMS y 5.675 56.48 20 -20 -40 -60 -80 -10-100 -100 80 80 -60 60 100 -80 -60 60 100 20 -20 Live R12 Real V R12 Imag. Live R12 Real V R13 Real FitLive FitLive 100 Entries 1000 Entries 1000 -0.08279 -0.08279 Mean x Mean x 80 -0.061 2.273 Mean y Mean y 53.86 RMS x RMS y 53.86 RMS x 60 RMS y 5.697 52.4 20 -20 -40 -60

-80

-100

-80

-60

-20

20

60

40

80

100

http://www.nuclear.gla.ac.uk/~dglazier/multinest/Dalitz1.mov

60

80

100

-10-100

-80

-60

-40

-20

0

20

40

#### MultiNest Dalitz Fit, Posterior and Live Points

R12 Real V R12 Imag.



Live R12 Real V R12 Imag.





Live R12 Real V R13 Real



## MultiNest Dalitz Fit - 2 solutions, Evolution with iteration number



CPU time for 100 Live Points = 0.7s

#### Start with a simple case $\pi N \rightarrow \pi N$

#### $\pi^+ p \, d\sigma/d\Omega$ from SAID multipoles Lmax=2 Helicity Amplitudes and partial waves $g(z) = \frac{1}{k} \Sigma_L \left[ (L+1) T_L^+ + L T_L^- \right] P_L(z)$ 3 2.5 $h(z) = \frac{1}{k} \Sigma_L \left[ T_L^+ - T_L^- \right] \sqrt{1 - z^2} P'_L(z)$ 0.5 $\cos(\theta)$ -0.4 -0.2 -0.6 0 0.2 $\pi^+p$ P from SAID multipoles Lmax=2 Minuit Fit starting with correct parameters 3 D 2 Ρ DataHis Entries 2568864 Mean x 0.07492 Mean y 0.1897 0.6792 RMS x $\rightarrow 0$ RMS y 1.803 $\chi^2$ / ndf 1.013e+04 / 9992 p0 -0.2564 ± 0.0008917 -0.6 -0.4 -0.2 0.2 0.4 0.8 p1 $0.07184 \pm 0.0006189$ -1 -0.8 0 06 p2 $-0.4794 \pm 0.4794$ $\cos(\theta)$ рЗ $0 \pm 0.8415$ p4 $-0.1271 \pm 0.001725$ -2 p5 $0.983 \pm 0.0003224$ p6 $-0.09146 \pm 0.0008278$ p7 $0.009359 \pm 0.0006977$ **p**8 125.7 ± 0 **p**9 1± 0 -0.8 -0.6 -0.4 -0.2 0.2 0.6 0 0.4 0.8 1 -1 $\cos(\theta)$

Use SAID PW

At 200 MeV



#### Use Minuit with Random initial values

#### Physical Value



2M likelihood calculations

#### Fix S1 parameters to true value



#### Relative Partial Waves





Look at P3 phase relative to S1



## Summary

CLAS12 experiment will soon produce mesons through Quasi-real photoproduction

Currently preparing analysis framework to Handle large statistics datasets Analyse many final states Provide alternative methods

Investigated different signal/background separation

Investigated effectiveness of Longitudinal P.S.

Implemented Nested Sampling algorithm into AmpTools
 Investigating its usefulness in Amplitude
 Analysis

Currently implementng amplitudes in collaboration with JPAC and testing on available CLAS data

## Developing Amplitude Analysis



In general greater overlap between different experiments and theorists

## CLAS12 - Forward Tagger

## Detect electrons at small angle to perform quasi-real photo-production experiments.

**Calorimeter:** electron energy/momentum Photon energy (v=E-E') Polarization  $\varepsilon^{-1} \approx 1 + v^2/2EE'$ PbWO<sub>4</sub> crystals with APD/SiPM readout

**Scintillation Hodoscope:** veto for photons Scintillator tiles with WLS readout

**Tracker:** electron angles, polarization plane MicroMegas detectors





$E_{scattered}$	0.5 - 4.5 GeV
$\theta$	$2.5^{o} - 4.5^{o}$
$\phi$	0° - 360°
ν	6.5 - 10.5 GeV
$Q^2$	$0.01 - 0.3 \text{ GeV}^2 \ (< Q^2 > 0.1 \text{ GeV}^2)$
W	3.6 - 4.5 GeV

## CLAS12 Detector Systems

#### Forward Detector

- TORUS Magnet
- Forward silicon vertex
   tracker
- HThresh Cerenkov
   Counter
- LThresh Cerenkov
   Counter
- Forward TOF System
- Preshower calorimeter
- E.M. Calorimeter

#### **Central Detector**

- SOLENOID magnet
- Barrel silicon tracker
- Central TOF

#### Additional Equipment

- Micromegas (CD)
- Neutron detector (CD)
- Forward RICH
- Forward Tagger



Enable e- detection below  $5^{\circ}$ 

Ready for data Summer 2017

## Qvalue with fixed (true) signal width





Qvalue reproduces signal and background shape But uncertainties not correct (need to calculate) And if width not constrained ...



