



Helicity Asymmetry E for $\gamma p \to \pi^0 p$ from JLAB CLAS g9a/FROST dataset with application of Machine Learning

Chan Kim

The George Washington University

Igor Strakovsky, William Briscoe, Stuart Fegan

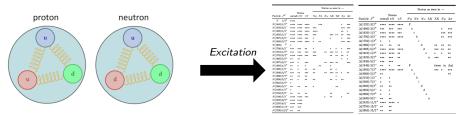
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Overview

- 1 Motivation
- 2 Event Selection
- 3 ML: Target Classification
- 4 ML: Hydrogen Contamination on Carbon
- Helicity Asymmetry E
- 6 Next Steps

Baryon Spectroscopy

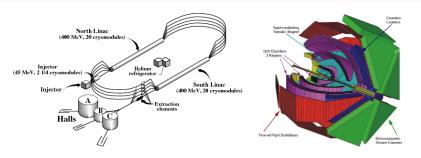
Baryon Spectroscopy is the study of excited nucleon states.



 Different quark models have different degrees of freedom, causing different predictions of resonance states & parameters of resonances (mass, width, etc).



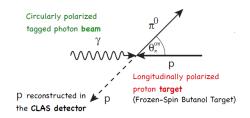
JLab Continuous e^- Beam Accelerator (6 Gev, before upgrade to 12 GeV)



Electron Beam Energy (GeV)	Photon Beam Polarization	# of Events (M)	Observable
1.645	Circular	~1000	E
2.478	Circular	~2000	E
2.751	Linear	\sim 1000	G
3.538	Linear	~2000	G
4.599	Linear	~3000	G

Hall B g9a/FROST run from $12/2007 \sim 2/2008$

CLAS g9a/FROST Experiment



- \circ Bremsstrahlung radiation (gold foil or thin diamond) \to real polarized photon
- \circ Dynamic Nulcear Polarization \to polarized targets
- o g9a/FROST Circularly polarized photons with $E_{\gamma} \approx 0.4-2.4$ GeV and longitudinally polarized proton target
- o 8 observables at fixed $(E_{\gamma}, \theta) \to 4$ helicity amplitudes \to Resonances (PWA)

	UP_T and UP_R	UP_T and P_R	P_T and UP_R	P_T and P_R
UP_B	$\frac{d\sigma}{d\Omega}$	Р	T	$T_{x'}, T_{z'}, L_{x'}, L_{z'}$
LP_B	$-\Sigma$	$O_{x'}, (-T), O_{z'}$	H,(-P),-G	
CP_B		$-C_{x'}, -C_{z'}$	F, - E	

UP, P, LP, CP, B, T, R denote unpolarized, polarized, linearly polarized, circularly polarized, beam, target, and recoil, respectively.

Helicity Asymmetry E

 Double polarization observable E is the helicity asymmetry of the cross section:

$$E=rac{\sigma_{3/2}-\sigma_{1/2}}{\sigma_{3/2}+\sigma_{1/2}}\qquad ext{for }rac{3}{2}\;\&\;rac{1}{2}\; ext{are total helicty states}$$

o $\frac{d\sigma}{d\Omega}$ of polarized beam & polarized target for E (theo. & exp.):

$$\left(rac{d\sigma}{d\Omega}
ight)_{rac{1}{2},rac{3}{2}} = rac{d\sigma_0}{d\Omega}(1\mp (P_zP_\lambda)_{rac{1}{2},rac{3}{2}}E)$$

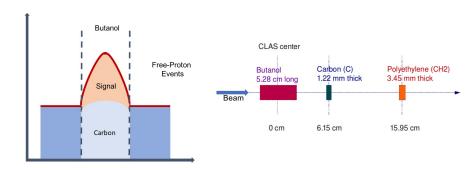
E is measured via:

$$\left[E = \left[\frac{1}{D_f} \right] \left[\frac{1}{P_z P_\lambda} \right] \left[\frac{N_3 - N_{\frac{1}{2}}}{N_{\frac{3}{2}} + N_{\frac{1}{2}}} \right]$$

$$\left(\frac{d\sigma}{d\Omega}\right)_{\frac{1}{2},\frac{3}{2}} = \frac{N_{\frac{1}{2},\frac{3}{2}}}{A \cdot F \cdot \rho \cdot \Delta x_i}$$

 $D_f=$ dilution factor $P_z=$ Polarization of target in \hat{z} $P_\lambda=$ Polarization of beam $N_{\frac{3}{5},\frac{1}{5}}=\#$ of events

Butanol & Carbon Targets

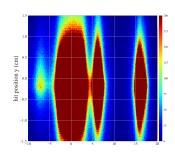


- Butanol target (C_4H_9OH) consists of polarized hydrogen (free-nucleons) & unpolarized carbon and oxygen (bound-nucleons)
- Fermi motion of bound-nucleons \rightarrow negative missing mass M_{π^0}
- Carbon target consists of unpolarized bound-nucleon
- Scale carbon target events & subtract from butanol target events

ML Objectives: Target Selection & Ice on Carbon

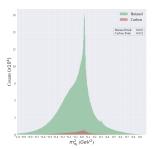
Target Selection

- Events with z-vertex \in [2, 5]cm, uncertain whether γ hit Butanol or Carbon

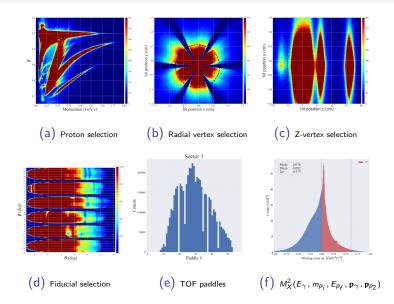


Ice on Carbon

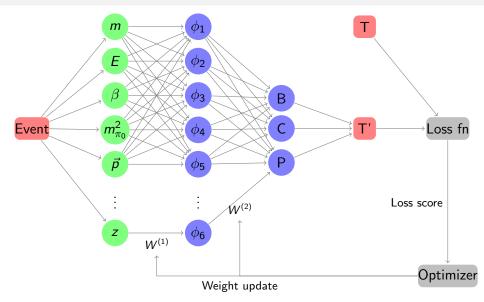
- Carbon events (bound-nucleon) expected to have broader $m_{\pi_0}^2$ peak due to Fermi motion.
- Sharp peak (free-nucleon) observed in the Carbon target region.



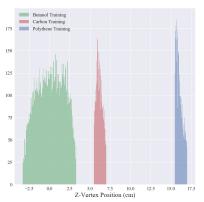
Event Selections



Neural Network Training Flowchart

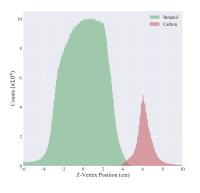


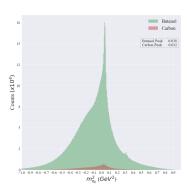
Training Data Selection



- o Randomly select events with z-vertex position in close proximity of each targets
 - Butanol \in [-3.3, 3.3]cm
 - Carbon \in [5.5, 7.0]cm
 - Polythene \in [15.5, 17.0]cm

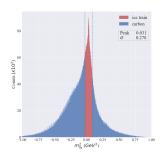
Result on Target Selection



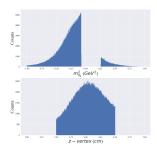


- Classified Carbon events from Butanol in z-vertex ∈ [2.5, 4.5]cm
- o Some Carbon events in Polythene regions & Polythene events in Butanol region.

Training Data for Hydrogen Contamination

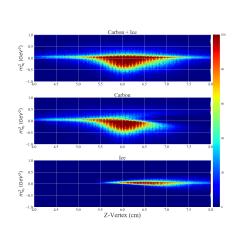


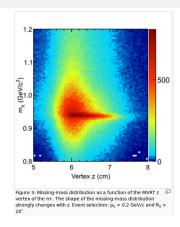
- Tight cut on the $m_{\pi_0}^2$ peak on g9a-Carbon data (or MC sim) as ice
 - Bound-nucleon (fermi p)
 - \rightarrow broader m^2 distribution
 - Sharper peaks from free-nucleon (ice) & Broad background from bound-nucleon (carbon)



- Randomly select events within three criterion:
 - Classified as carbon events in previous target classification distribution
 - Missing mass squared $\notin [-\sigma, \sigma]$
 - Z-vertex position $\in [5.5, 6.5]$

Final Result of ML: ICE vs CARBON

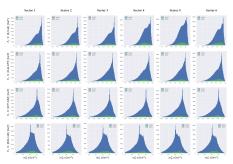




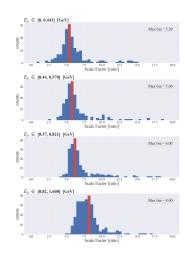
[Result from USC for $\gamma p
ightarrow \pi^+ n$]

- o Classified ice events from Carbon target in z-vertex \in [6.0, 7.5]cm
- It is likely that ice was formed in 20 K heat shield in between Carbon and Polythene targets.

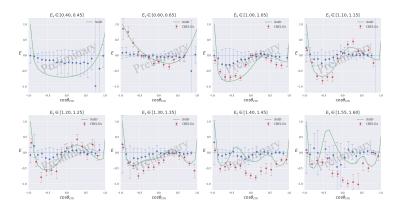
Scale Factor $\left(\frac{N_{C_4H_9OH}}{N_C}\right)$ & Dilution Factor



- Sector dependence only evident in low Energy: $E_{\gamma} \sim [0, 0.45] \, GeV$
- As $E_{\gamma}\uparrow$, more interactions in butanol target than carbon
- o $D_f \Big|_{\text{low lim}} = \frac{\text{free H in butanol}}{\text{total nucleon in butanol}} = \frac{10}{74} \cong 0.135$
- $O_f(E_{\gamma}, \theta_{cm}) = \frac{N_{B,f}}{N_{B,tot}} \cong 1 \frac{s(E_{\gamma}) \times N_C(E_{\gamma}, \theta_{cm})}{N_{B,tot}(E_{\gamma}, \theta_{cm})}$



Preliminary: Helicity Asymmetry E



$$\circ E = \begin{bmatrix} \frac{1}{D_f} \end{bmatrix} \begin{bmatrix} \frac{1}{P_{\gamma}P_{T}} \end{bmatrix} \begin{bmatrix} \frac{N_{\frac{3}{2}} - N_{\frac{1}{2}}}{N_{\frac{3}{2}} + N_{\frac{1}{2}}} \end{bmatrix}$$

- Result of ~ 30% of JLab CLAS g9a experiment data
- Measured E comparison to SAID Partial Wave Analysis predictions

Next Steps

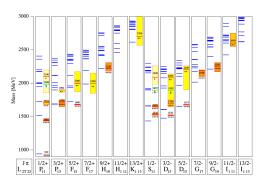
- Process all g9a data for full statistics
- Quantify uncertainties in neural network training
 - Bayesian Neural Network probability distribution to weights and biases while training
 - Compute purity of the training data used for uncertainty.
- Energy loss reconstruction
- Systematic Error studies
- $\,\circ\,$ Measured E into SAID database \to new pole positions for resonances

Acknowledgements

This work was performed with support from US DOE DE-SC001658, The George Washington University.

Backup Slides

Backup: Constituent Quark Models & LQCD Predictions of Non-Strange Baryon Resonances



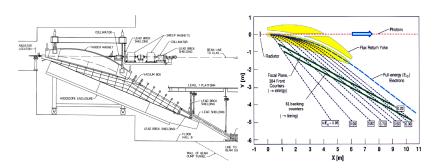
Constituent Quark Model

Lattice QCD

- Constituent Quark Models predicted states: 64 N* & 22 Δ*
- Experimentally confirmed state: 26 N* & 22 Δ*

Backup: Hall B Photon Tagger

- Bremsstrahlung radiation due to slowing of electrons by EM field of radiator (gold foil or thinyo diamond)
- Determine incoming photon energy of $\vec{\gamma}\vec{p} \to \pi^0 p$ by $E_\gamma = E_0 E_e$
- ullet g9a/FROST circularly polarized photons with $E_{\gamma} pprox 0.4 \sim$ 2.4 GeV
- Tagger was built by the GWU, CUA, & ASU nuclear physics group



Backup: Circularly Polarized Photon Beam

Linearly Polarized Electron Beam



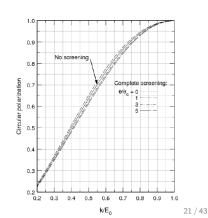
Circularly Polarized Photon Beam

Polarization transfer:

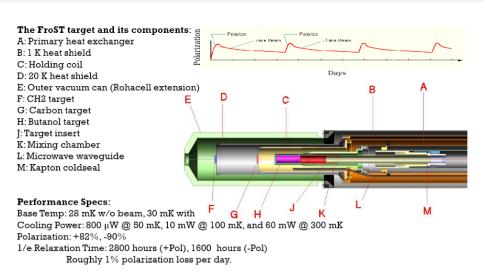
$$P(\gamma) = P(e) \frac{4x - x^2}{4 - 4x + 3x^2}$$

$$x = \frac{k}{E_0} = \frac{\text{photon energy}}{\text{incident electron energy}}$$

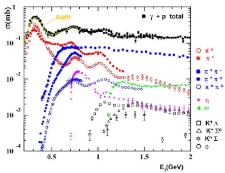
H. Olsen and L.C. Maximon, Phys. Rev. 114, 887 (1959)



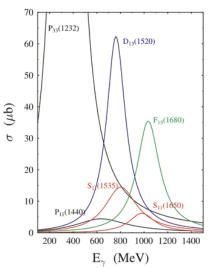
Backup: Frozen Spin Target



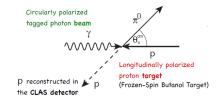
Backup: CLAS g9a/FROST Data



- Select only $\vec{\gamma} \vec{p} \to \pi^0 p$ events
- $\vec{\gamma}\vec{p} \to \pi^0 p$ resonance channels
- Appropriate enegy bins include all resonances (≤ 1500 MeV)



π^0 photoproduction



• From T Matrix to Helicity Amplitudes of $\vec{\gamma}\vec{p} \to \pi^0 p$:

$$\langle \mathbf{q} | m_{s'} | T | \mathbf{k} | m_s | \lambda \rangle = \boxed{\langle m_{s'} | \mathbf{J} | m_s \rangle} \cdot \epsilon_{\lambda}(\mathbf{k})$$

$$H_i(\theta) \equiv \langle \lambda_2 | \mathbf{J} | \lambda_1 \rangle$$

4 Complex Helicity Amplitudes:

$$H_{1}(\theta) = \left\langle +\frac{3}{2} \middle| \mathbf{J} \middle| +\frac{1}{2} \right\rangle \qquad H_{2}(\theta) = \left\langle +\frac{1}{2} \middle| \mathbf{J} \middle| +\frac{1}{2} \right\rangle$$

$$H_{3}(\theta) = \left\langle +\frac{3}{2} \middle| \mathbf{J} \middle| -\frac{1}{2} \right\rangle \qquad H_{4}(\theta) = \left\langle +\frac{1}{2} \middle| \mathbf{J} \middle| -\frac{1}{2} \right\rangle$$

Backup: Complete Experiment - 8 Polarization Observables

- Polarizable: incoming photons, target & recoiling nucleons
- 8 well chosen observables at fixed E_{γ} & angle \rightarrow 4 helicity amplitudes

	UP_T and UP_R	UP_T and P_R	P_T and UP_R	P_T and P_R
UP_B	$\frac{d\sigma}{d\Omega}$	Р	T	$T_{x'}, T_{z'}, L_{x'}, L_{z'}$
LP_B	$-\Sigma$	$O_{x'},(-T),O_{z'}$	H,(-P),-G	
CP_B		$-C_{x'}, -C_{z'}$	F, - E	

UP, P, LP, CP, B, T, R denote unpolarized, polarized, linearly polarized, circularly polarized, beam, target, and recoil, respectively.

Helicity asymmetry E related to other observables via Fierz identities:

$$E^{2} + F^{2} + G^{2} + H^{2} = 1 + P^{2} - \Sigma^{2} - T^{2}$$

 $FG - EH = P - \Sigma T$

Overtraining Limits

Overtraining:

Excess training with only specific training data

Classification succeeds on training data, but fails on actual data

- Must determine adequate classifying variables & size of training data
- Rule of thumb for Decision Tree algorithm:

$$L_D(h) \leq L_S(h) + \sqrt{\frac{(n+1)\log_2(d+3) + \log(2/\delta)}{2m}}$$

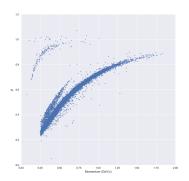
 $L_D(h) = \text{Error of classification on actual data set}$ h = Error of classification on a training data set $\delta = \text{Confidence level of randomly selected training data points}$ n = Number of nodes

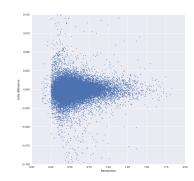
 $L_S(h) = \text{Error of classification on a training data set}$ d = Number of variables

m =Size of training data sets

n & d inversely proportional to L_c

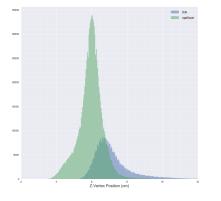
Proton Selection: $\Delta \beta$ Selection

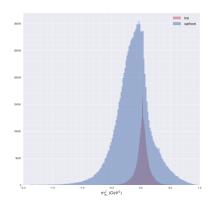




- $\Delta \beta = \beta_{
 m measured} \beta_{
 m p} = \beta_{
 m measured} rac{
 m p}{\sqrt{m_{
 m p}^2 +
 m p^2}}$
- ullet Select events with only 1 positive outgoing particle (for $ec{\gamma}ec{p} o\pi^0 p)$
- \bullet Measure p (via curvature) and β (via SC & TOF) of positive particles
- Select events with $\Delta \beta \approx 0$

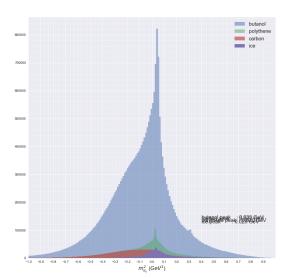
Result on Hydrogen Contamination of Carbon Target



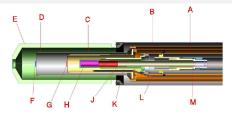


- o Classified ice events from Carbon target in z-vertex ∈ [6.0, 7.5]cm
- It is likely that ice was formed in 20 K heat shield in between Carbon and Polythene targets.

Final Result Target Classification

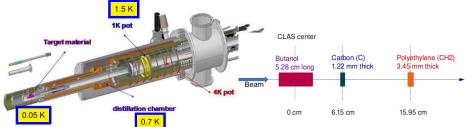


g9a/FROST Target setup



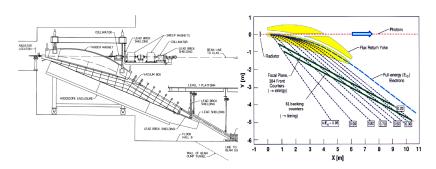
Side view of FROST target with beam entering from the right. (A) Primary head exchanger, (B) I K heat shield, (C) Holding coil, (D) 20 K heat shield, (E) Outer vacuum can, (F) Polyethylene target,(G) Carbon target,(H) Butanol target (J) Target insert, (K) Mixing chamber, (L) Microwave waveguide, and (M) Kapton cold seal.



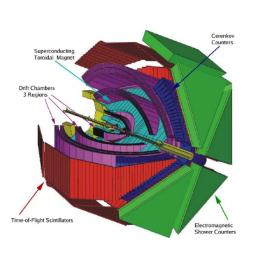


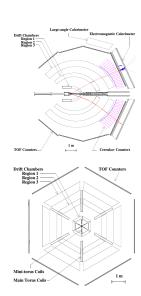
JLab Hall B Photon Tagger

- Bremsstrahlung radiation due to slowing of electrons by EM field of radiator (gold foil or thinyo diamond)
- Determine incoming photon energy of $\vec{\gamma}\vec{p} \to \pi^0 p$ by $E_\gamma = E_0 E_e$
- ullet g9a/FROST circularly polarized photons with $E_{\gamma} pprox 0.4 \sim$ 2.4 GeV
- Tagger was built by the GWU, CUA, & ASU nuclear physics group



CEBAF Large Acceptance Spectrometer





Evidence of Hydrogen Contamination on Carbon

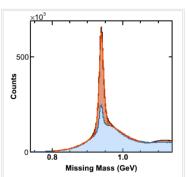
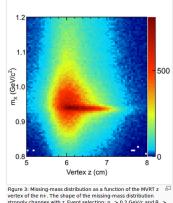


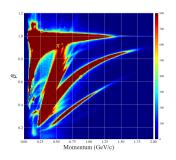
Figure 2: Missing-mass distribution for the π+n channel from -FROST q9a data. W = 1.25 - 1.50 GeV, integrated over all angles. Events in the red histogram are from the butanol target and events in the blue histogram are from the 12C target with z-vertex larger 5.0 cm and smaller than 7.5 cm. The blue histogram is scaled by 5.26. The FROST distribution from the 12C target region show a narrow peak at the mass of then neutron.

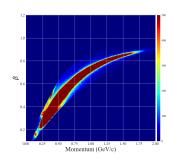


strongly changes with z. Event selection: $p_{m} > 0.2$ GeV/c and $\theta_{m} >$ 20°

- \circ Sharp peak at downstream end of Carbon foil \to ice built up while cooling the target
- o Ice formed on the right side of Carbon target: Z-vertex ∈ [6, 7]cm
- o Plots from [Steffen Strauch]'s Analysis page of FROST Wikipage

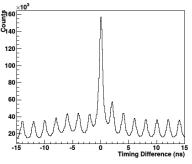
Proton Selection: GPID bank





- $\Delta \beta = \beta_{
 m measured} \beta_{\it p} = \beta_{
 m measured} \frac{\it p}{\sqrt{m_{\it p}^2 + \it p^2}}$
- ullet Select events with only 1 positive outgoing particle (for $ec{\gamma}ec{p} o\pi^0 p)$
- ullet Measure p (via curvature) and eta (via SC & TOF) of positive particles
- Select events with $\Delta \beta \approx 0$

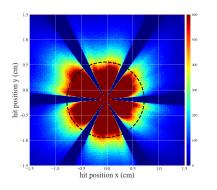
Photon Beam Selection



$$\Delta t = t_{pv} - t_{\gamma v}$$
 = time when p was at event vertex $-$ time when γ was at event vertex

- Readings from SC, DC & TOF system to determine t_{pv} & $t_{\gamma v}$
- JLab e^- beam sent in bunches separated by 2 ns
- ullet Neglect events caused by photons emitted from different e^- bunches
- Select out events with $\Delta t \approx 0$

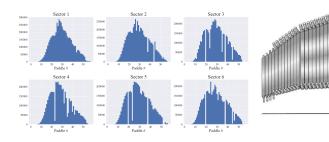
Radial Vertex Selection - Target Cup





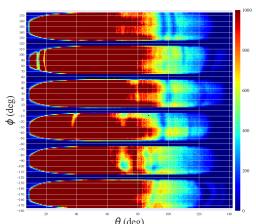
- Removed events outside of target cup (d = 1.5cm)
- He-Bath outer region

Inefficient Time-Of-Flight system paddles



- Events from inefficient scintillator paddles removed
- Sector2 25, Sector3 23, 35, Sector4 23 and etc

Fiducial Selection - Inactive CLAS regions



 θ (deg) lnactive regions of detector - coil of torus magnet, beamline holes, etc

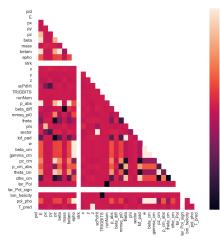
$$\begin{array}{lll} \circ & \theta < 7, \ -180 < \phi < -175, \ -125 < \phi < -115, \ -65 < \phi < -55 \\ -5 < \phi < 5, \ 55 < \phi < 65, \ 115 < \phi < 125, \ 175 < \phi < 180 \\ \end{array}$$

Neural Network Model Setup

- Two fully-connected (dense) neural layers
 - 1 Dense layer with 15 nodes 15 parameters:
 - E, β , β_{diff} , β_m E_{γ} , m, $m_{\pi_0}^2$, pid,|p|, p_x , p_y , p_z , x, y, and z.
 - Too many parameters + insufficient train data \to Too specific training \to Overfitting (fail)
 - 2 Dense layer with 3 nodes one for each target
 - For each event, this layer returns an array of 3 probability scores (butanol, carbon, or polythene) that sum to 1
- o Optimizer used: AdamOptimizer
- Loss function used Sparse categorical cross entropy:
 - $H_{y'}(y) = -\sum_i y_i' \log(y_i)$,where y_i is the predicted target and y_i' is the true target
- Python and Tensorflow

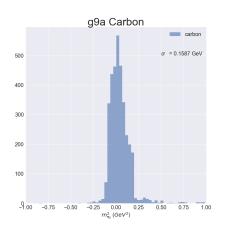
0.4

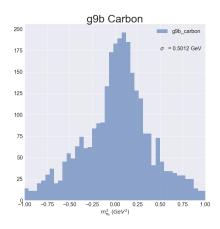
Choosing Classifying Parameters



- \circ Choose 10 \sim 15 adequately correlated parameters to avoid overfitting and underfitting
- \circ Higher correlation \rightarrow lesser contribution to classification
 - \circ Lower correlation \to biased training \to overfitting

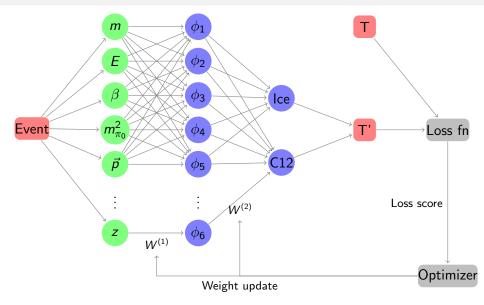
Training Data for Carbon from g9b experiment



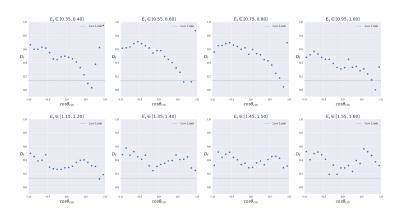


- \circ g9b-carbon $m_{\pi_0}^2$ peak broader than g9a/Carbon o No ice on g9b
- o During g9b, Carbon target was moved further in downstream.
- o Shifted Z-vertex of g9b-Carbon events to use as training events for g9a [F. Klein].
- \circ Failed (under investigation) \to Different training data for carbon used

Neural Network Training Flowchart: ICE vs CARBON



Dilution Factor



$$\begin{array}{l} \circ \;\; D_f(E_{\gamma},\theta_{\mathit{cm}}) = \frac{N_{\mathit{B},f}}{N_{\mathit{B},tot}} = \frac{N_{\mathit{B},tot} - N_{\mathit{B},b}}{N_{\mathit{B},tot}} \cong 1 - \frac{s(E_{\gamma}) \times N_{\mathit{C}}(E_{\gamma},\theta_{\mathit{cm}})}{N_{\mathit{B},tot}(E_{\gamma},\theta_{\mathit{cm}})} \\ \circ \;\; D_f\big|_{\mathsf{low \; lim}} = \frac{\mathsf{free \; H \; in \; butanol}}{\mathsf{total \; nucleon \; in \; butanol}} = \frac{10}{74} \cong 0.135 \\ \end{array}$$