

ANN Input

IO startup

D. Keller

The Real Fit Function

$$\sigma_{UU} = \sigma_{UU}^{DVCS} + \sigma_{UU}^{BH} + \sigma_{UU}^I$$

Fitting parameter

$$\sigma_{UU}^{BH} = \frac{\Gamma}{t} \left[A_{UU}^{BH} (\underline{F_1^2} + \tau \underline{F_2^2}) + B_{UU}^{BH} \tau \underline{G_M^2}(t) \right]$$

Elastic Form Factors

$$F_1 = \frac{2.7928}{(1 - 1.40716 t)^2} - F_2 \quad F_2 = \frac{1.7928}{(1 - 1.40716 t)^2 \left(1 - \frac{t}{4M^2}\right)}$$

$$A_{UU}^{BH} = \frac{16 M^2}{t(kq')(k'q')} \left[4\tau \left((kP)^2 + (k'P)^2 \right) - (\tau + 1) \left((k\Delta)^2 + (k'\Delta)^2 \right) \right]$$

$$B_{UU}^{BH} = \frac{32 M^2}{t(kq')(k'q')} \left[(k\Delta)^2 + (k'\Delta)^2 \right]$$

$$\sigma_{UU}^I = \frac{\Gamma}{Q^2(-t)} \left[A_{UU}^I (F_1 \underline{\Re\mathcal{H}} + \tau F_2 \underline{\Re\mathcal{E}}) + B_{UU}^I G_M (\underline{\Re\mathcal{H}} + \underline{\Re\mathcal{E}}) + C_{UU}^I G_M \underline{\Re\tilde{\mathcal{H}}} \right]$$

Compton Form Factors (CFFs)
Fitting parameters

-small

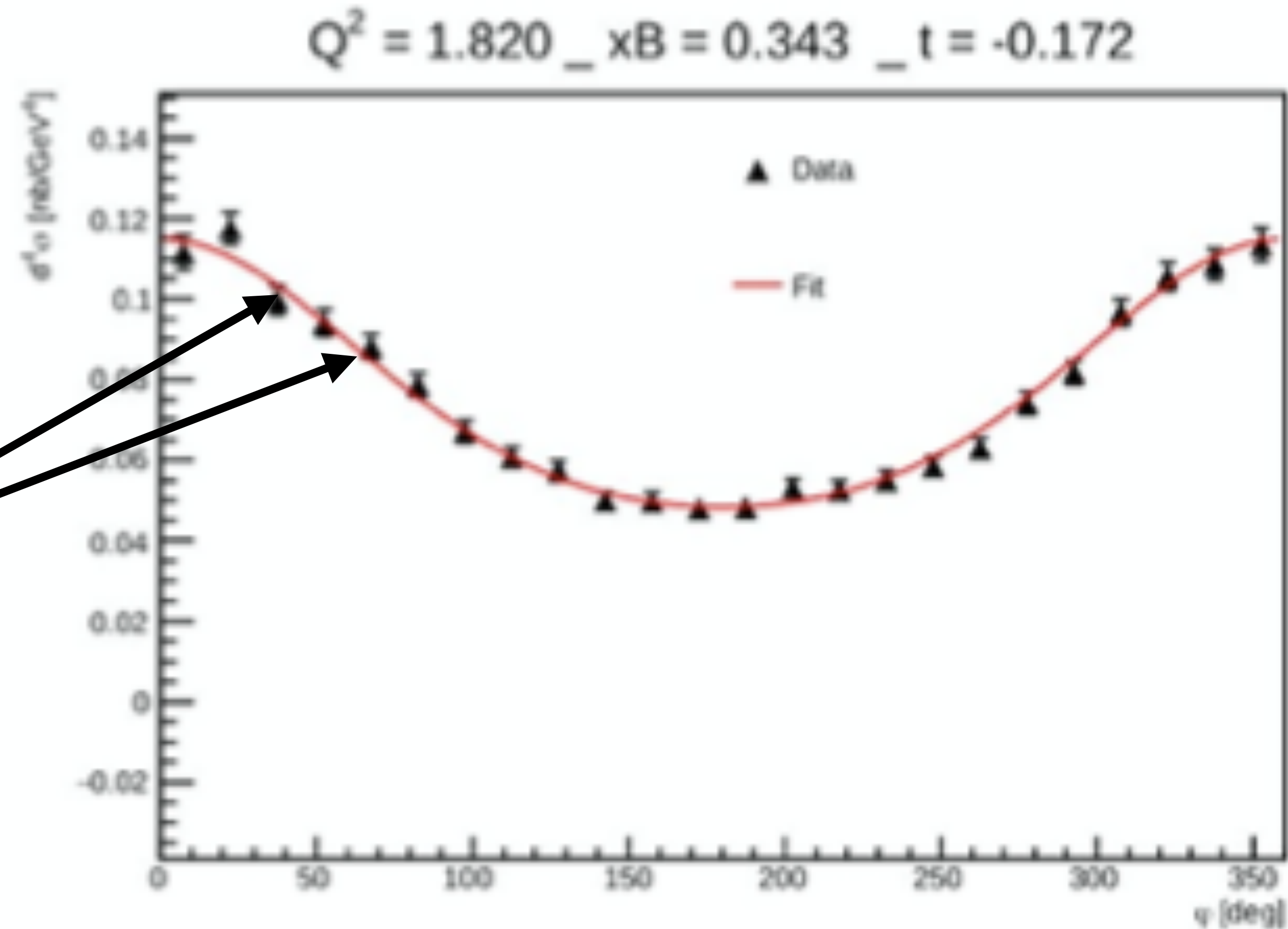
$$A_{UU}^I = -\frac{4}{(kq')(k'q')} \left\{ (Q^2 + t) \left[2((kP) + (k'P))(kk)_T + (Pq)(kq')_T + 2(k'P)(kq') \right. \right. \\ \left. \left. - 2(kP)(k'q') + (k'q')(kP)_T + (kq')(k'P)_T - 2(kk')(kP)_T \right] \right. \\ \left. + (Q^2 - t + 4(k\Delta)) \left[(Pq)((kk')_T + (kq')_T - 2(kk')) \right] \right. \\ \left. + 2(kk')(Pq)_T - (k'q')(kP)_T - (kq')(k'P)_T \right\} \cos \phi$$

$$B_{UU}^I = \frac{2\xi}{(kq')(k'q')} \left\{ (Q^2 + t) \left[2(kk)_T((k\Delta) + (k'\Delta)) + (kq')_T((q\Delta) - (kq') - (k'q')) \right. \right. \\ \left. \left. + 2(kk') \right] + 2(kq')(k'\Delta) - 2(k'q')(k\Delta) \right] + (Q^2 - t + 4(k\Delta)) \left[((kk)_T \right. \\ \left. - 2(kk')(q'\Delta) - (kk')\Delta_T^2 - 2(k\Delta)_T(kq') \right] \right\} \cos \phi$$

The Cross Section

$$\frac{A\cos^2(\phi) + B\cos(\phi) + C}{D\cos^2(\phi) + E\cos(\phi) + F} \cos(\phi)$$

- See github: BHDVCS.py



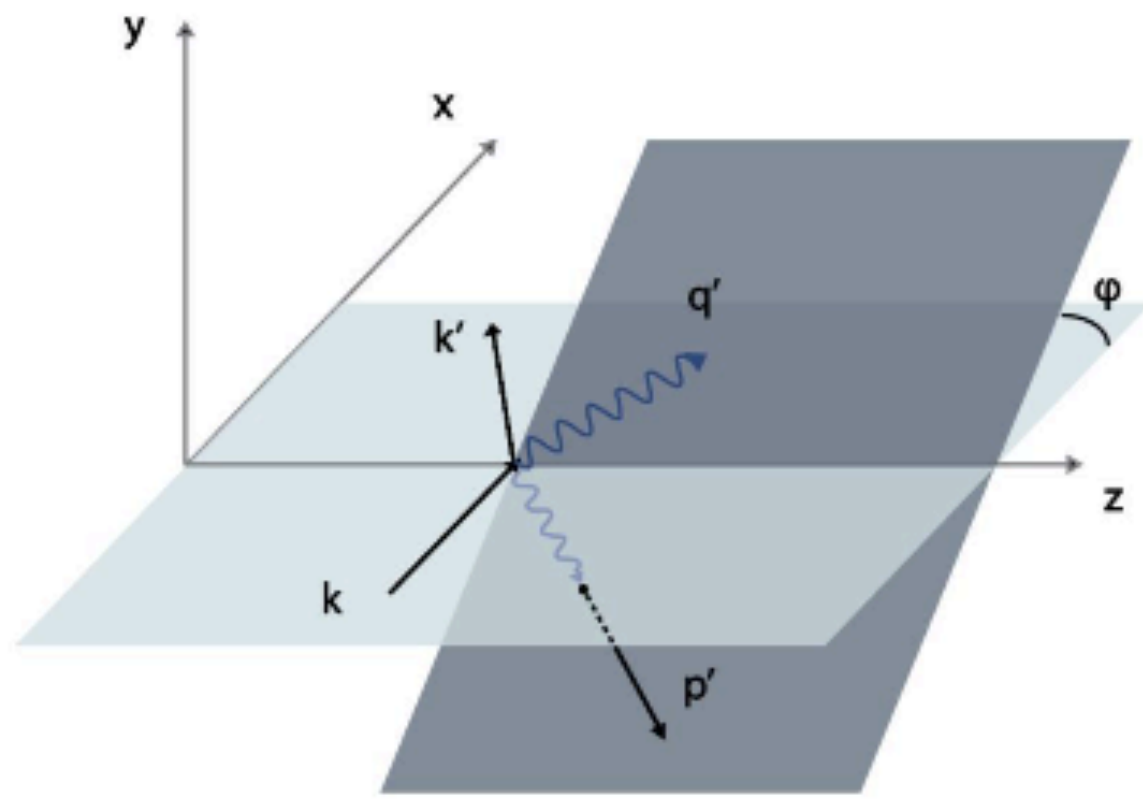
- F(phi_x)

- errF: error at each F(phi_x)

Inputs

Kinematic settings

$$ep \rightarrow e'p'\gamma$$



4-momentum vectors

- k incoming electron
- k' outgoing electron
- q virtual exchange photon
- q' outgoing photon
- p' outgoing proton

$$d\sigma(k, x_B, t, Q^2, \phi)$$

- k Energy of the incoming electron
- Q^2 Electron squared momentum transfer
- t Squared momentum transfer to the proton
- x_B Bjorken variable

$$x_B = \frac{Q^2}{2(pq)}$$

Determines the momentum fraction of the quark or gluon on which the photon scatters.

- ◆◆ Azimuthal angle between the hadron plane formed by the outgoing proton and photon and the lepton plane

Elastic Form Factors

$$F_1 = \frac{2.7928}{(1 - 1.40716 t)^2} - F_2 \quad F_2 = \frac{1.7928}{(1 - 1.40716 t)^2 \left(1 - \frac{t}{4M^2}\right)}$$

- k
- Q^2
- t
- x_b
- ϕ_x
- F and $errF$
- F_1 and F_2

In/Out

Starting point

- Use k , QQ , x_b , t as inputs to $F(\text{phi}_x)$ for several phi_x points
- errF is the Gaussian error in F at those several phi_x points
- So first start with fixed kinematics to produce a fit: a minimization in errF using $F(k, QQ, x_b, t, \text{phi}_x)$ for fixed k , QQ , x_b , t over 0-360 in phi_x
- $F1$ and $F2$ are simply calculated values that go into the function and are the same for fixed kinematics
- After this step compare your results to a χ^2 minimization fit of the same thing

The Function

Branch: master ▾ [ptgroupneuralnet / Compton_FF_Code / BHDVCS.py](#) / <> Jump to ▾ Find file Copy path

 [tvallar](#) Added untracked files e5999e5 on Sep 29, 2019

[1 contributor](#)

235 lines (191 sloc) | 12.5 KB Raw Blame History   

```
1
2 import numpy as np
3 import Lorentz_Vector as lv
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5 class BHDVCS(object):
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207     def TotalUUXS(self, angle, par):
208         phi = angle[0]
209         # Set QQ, xB, t and k and calculate 4-vector products
210         #print(par)
211         self.SetKinematics( par[0], par[1], par[2], par[3] )
212         self.Set4VectorsPhiDep(phi)
213         self.Set4VectorProducts(phi)
214
215         xsbhUU = self.GetBHUUXs(par[4], par[5])
216         xsiiUU = self.GetIIUUXs(phi, par[4], par[5], par[6], par[7], par[8])
217
218         tot_sigma_uu = xsbhUU + xsiiUU + par[9] # Constant added to account for DVCS contribution
219         #print(xsbhUU, " ", xsiiUU, " ", tot_sigma_uu)
220         return tot_sigma_uu
221
```


Before first steps

Sanity Check

- Just do a out of the box χ^2 minimization fit
- Lets choose a single fix kinematics for Liliet to fit
- Compare results and make sure we are all on the same page
- Use resulting error and the true CFF for those points to check results
- Then use Teddys utilities (Numpy, TensorFlow) and check the same
- Then: see if you can improve on Teddys and/or make your own

Start with first fixed setting in DVCS_cross_fixed.csv

After First Steps

Now things get interesting

- Then try additional kinematics
- Full scope of data
- Use ANN (or any ML) to train on dynamic set
- Impose additional linear constraints and use simultaneous fitting
- Multivariate Optimization
- Use real experimental data
- — —> Then next phase

Additional Info

- We can meet weekly (same meeting same time)
- Everyone should have a Github or something
- Help each other and work together
- make sure your hours reflect productivity
- Project site: confluence.its.virginia.edu/display/twist/ANN+Fitting+Project
- Liliet: lc2fc@virginia.edu
- me: dustin@virginia.edu