

Constraining DVCS Compton Form Factors Using Lattice QCD calculations

Yuan-Yuan Huang

Oct. 2025

Collaborators: Xu Cao, Taifu Feng, K. Kumericki, Yu Lu



中国科学院近代物理研究所
Institute of Modern Physics, Chinese Academy of Sciences



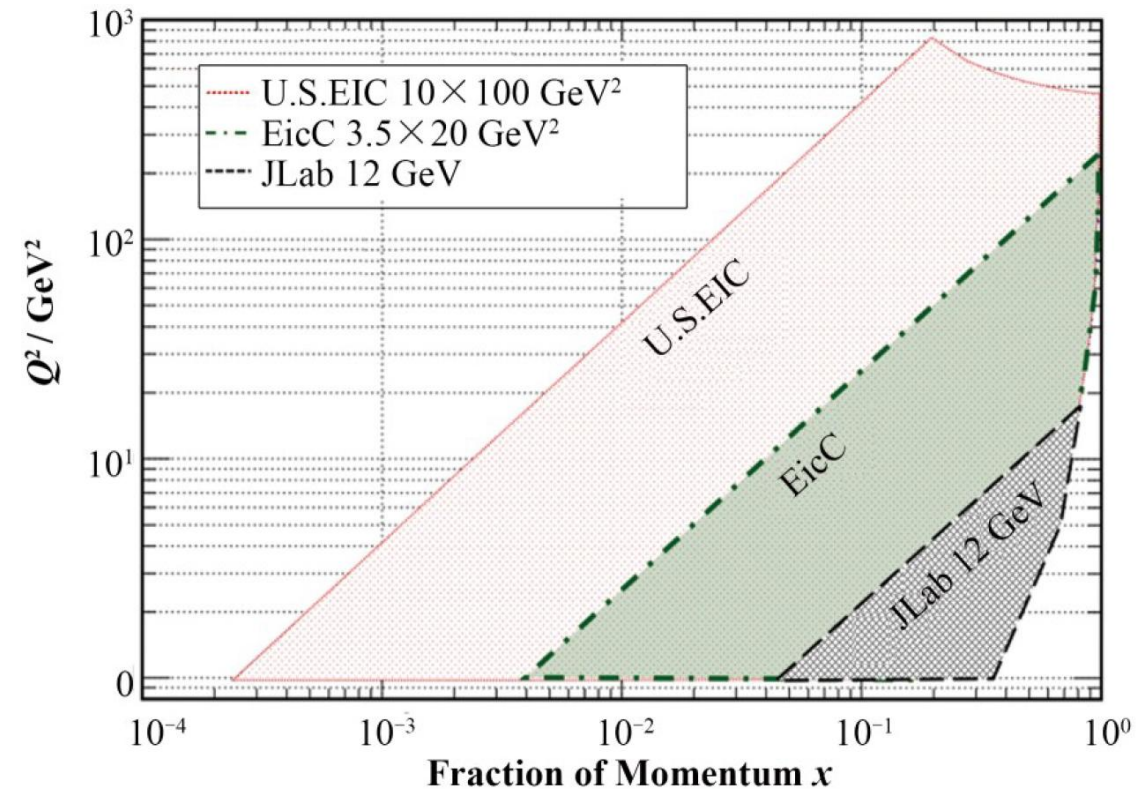
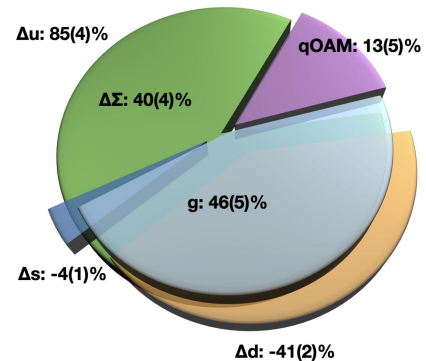
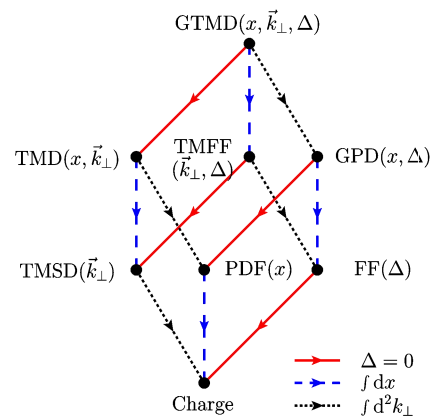
河北大学
HEBEI UNIVERSITY

Contents

- **1. Introduction**
- **2. Neural networks**
- **3. Add LQCD data of the Gravitation Form Factors**
- **4. Summary**

Introduction

- **GPD: General Parton Distributions** (trans. spatial position $b_\perp$ & longi. Momentum):
 - How does proton's spin influence the spatial distribution of partons?
 - probed by the exclusive process
- From 1D to 3D picture of hadron & nuclei
- Origin of the Proton/Meson spin



- 中国科学: 物理学力学天文学, 50: 112005 (2020)
- 核技术, 43(2): 020001 (2020) ;Front. Phys. 16, 64701 (2021)

Introduction

- **GPD(General Parton Distributions):**
- The key theoretical tool for describing the three-dimensional structure of hadrons.
- **CFF(Compton Form Factor):**
- The bridge connecting GPD theory with experimental observables. CFF is the convolution integral of GPDs and QCD Wilson coefficients

$$\begin{aligned}\mathcal{F}^q(\xi, t) &= \mathcal{C}(x, \xi; Q^2) \otimes F^q(x, \xi, t) \\ &= e_q^2 \int_{-1}^{+1} dx \left[\frac{1}{\xi - x - i\epsilon} - \frac{1}{\xi + x - i\epsilon} \right] F^q(x, \xi, t)\end{aligned}$$

$$\begin{aligned}\tilde{\mathcal{F}}^q(\xi, t) &= \mathcal{C}(x, \xi; Q^2) \otimes \tilde{F}^q(x, \xi, t) \\ &= e_q^2 \int_{-1}^{+1} dx \left[\frac{1}{\xi - x - i0} + \frac{1}{\xi + x - i0} \right] \tilde{F}^q(x, \xi, t)\end{aligned}$$

$$\Re \mathcal{F}(\xi, t) = e_q^2 P.V. \int_{-1}^{+1} dx \left[\frac{1}{\xi - x} - \frac{1}{\xi + x} \right] F^q(x, \xi, t)$$

$$\Im \mathcal{F}(\xi, t) = \pi e_q^2 \left(F^q(\xi, \xi, t) - F^q(-\xi, \xi, t) \right)$$

$$\Re \tilde{\mathcal{F}}(\xi, t) = e_q^2 P.V. \int_{-1}^{+1} dx \left[\frac{1}{\xi - x} + \frac{1}{\xi + x} \right] \tilde{F}^q(x, \xi, t)$$

$$\Im \tilde{\mathcal{F}}(\xi, t) = \pi e_q^2 \left(\tilde{F}^q(\xi, \xi, t) + \tilde{F}^q(-\xi, \xi, t) \right)$$

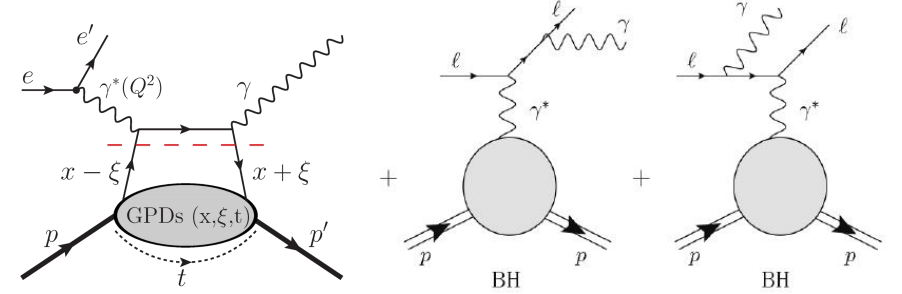
arXiv:2207.10766 [hep-ph].

Introduction

- **DVCS(Deply Virtual Compton Scattering):**
- By measuring CFFs via DVCS, GPDs are indirectly decoded. The Bethe-Heitler process is a background process in DVCS measurements.

$$|T|^2 = |T_{BH} + T_{DVCS}|^2 = |T_{BH}|^2 + |T_{DVCS}|^2 + \mathcal{I}$$

$$\mathcal{I} = T_{BH}^* T_{DVCS} + T_{DVCS}^* T_{BH}$$



- **CFFs measured by the asymmetries:**

$$A_{LU,I}^{\sin \phi} \propto \text{Im} \left[F_1 \mathcal{H} + \xi(F_1 + F_2) \tilde{\mathcal{H}} - \frac{t}{4m^2} F_2 \mathcal{E} \right],$$

$$A_{UL,I}^{\sin \phi} \propto \text{Im} \left[\xi(F_1 + F_2) \left(\mathcal{H} + \frac{\xi}{1+\xi} \mathcal{E} \right) + F_1 \tilde{\mathcal{H}} - \xi \left(\frac{\xi}{1+\xi} F_1 + \frac{t}{4M^2} F_2 \right) \tilde{\mathcal{E}} \right]$$

$$A_{LL,I}^{\cos \phi} \propto \text{Re} \left[\xi(F_1 + F_2) \left(\mathcal{H} + \frac{\xi}{1+\xi} \mathcal{E} \right) + F_1 \tilde{\mathcal{H}} - \xi \left(\frac{\xi}{1+\xi} F_1 + \frac{t}{4M^2} F_2 \right) \tilde{\mathcal{E}} \right]$$

$$A_{UT,I}^{\sin(\phi-\phi_s) \cos \phi} \propto \text{Im} \left[-\frac{t}{4M^2} (F_2 \mathcal{H} - F_1 \mathcal{E}) + \xi^2 \left(F_1 + \frac{t}{4M^2} F_2 \right) (\mathcal{H} + \mathcal{E}) \right. \\ \left. - \xi^2 (F_1 + F_2) \left(\tilde{\mathcal{H}} + \frac{t}{4M^2} \tilde{\mathcal{E}} \right) \right],$$

$$A_{UT,I}^{\cos(\phi-\phi_s) \sin \phi} \propto \text{Im} (F_2 \tilde{\mathcal{H}} - F_1 \xi \tilde{\mathcal{E}}),$$

$$A_{LT,I}^{\sin(\phi-\phi_s) \sin \phi} \propto \text{Re} (F_2 \mathcal{H} - F_1 \mathcal{E}),$$

$$A_{LT,I}^{\cos(\phi-\phi_s) \cos \phi} \propto \text{Re} (F_2 \tilde{\mathcal{H}} - F_1 \xi \tilde{\mathcal{E}}).$$

Introduction



Introduction

- All order dispersion relation with arbitrary subtraction: [Eur.Phys.J.C 85 \(2025\) 1, 105 e-Print: 2410.13518](#)

$$\sum_{j \text{ even}}^n h_j \xi^{-j} = \Re \mathcal{H}(\xi) - \frac{1}{\pi} \int_0^1 \Im \mathcal{H}(x) \left(\frac{x}{\xi} \right)^n \left[\frac{1}{\xi - x} - (-1)^n \frac{1}{\xi + x} \right] dx$$

- for $j=0$: $\Re \mathcal{H}(\xi, t) = \Delta(t) + \frac{1}{\pi} \text{P.V.} \int_0^1 dx \left(\frac{1}{\xi - x} - \frac{1}{\xi + x} \right) \Im \mathcal{H}(x, t)$ $e_0 = - \Delta(t)$ for CFF E

- for $j>0$: $h_{2j} = \frac{2}{\pi} \int_0^1 \Im \mathcal{H}(\xi') (\xi')^{2j-1} d\xi'$ and e_{2j} for CFF E

These higher order subtraction constants are related to generalized form factors

$$\int_{-1}^1 dx x^{m-p_a} H^a(x, \xi, t) = \sum_{j=0}^{\lfloor \frac{m}{2} \rfloor} A_{m;2j}^a(t) (2\xi)^{2j} + \text{mod}(2, m) (2\xi)^{m+1} C_m^a(t),$$

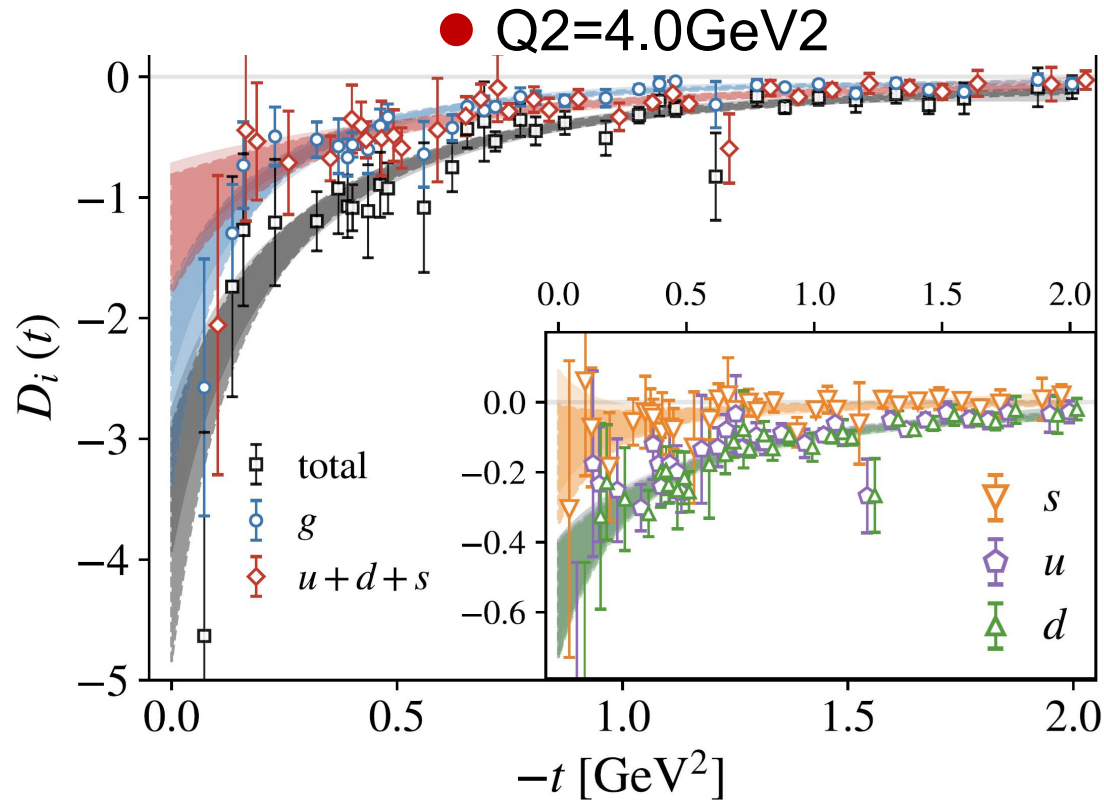
$$\int_{-1}^1 dx x^{m-p_a} E^a(x, \xi, t) = \sum_{j=0}^{\lfloor \frac{m}{2} \rfloor} B_{m;2j}^a(t) (2\xi)^{2j} - \text{mod}(2, m) (2\xi)^{m+1} C_m^a(t),$$

calculable in LQCD, e.g. PRD.77.094502 & 108.014507 with bigger pion mass

Introduction

- PHYSICAL REVIEW LETTERS 132, 251904 (2024)

$$m_\pi \approx 170 \text{ MeV}$$



$$\Delta^q(t) = 2 \int_{-1}^1 \frac{D^q(z, t)}{1 - z} dz = 4 \sum_{k=0}^{+\infty} d_{2k+1}^q$$

$$\bar{D}_1^q(t) = \int_{-1}^1 z D^q(z, t) dz = \frac{4}{5} d_1^q$$

$$h_0^q(t) \simeq 4d_1^q = 5\bar{D}_1^q(t)$$

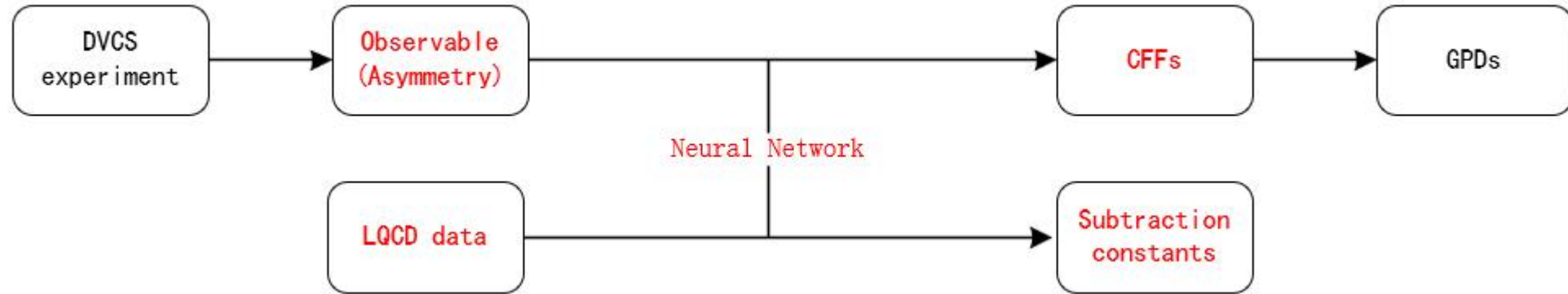
Flavor unseparated subtract const.:

$$\Delta(t) = \sum_q e_q^2 \Delta^q(t)$$

$$\text{Re}\mathcal{H}(\xi, t) = \Delta(t) + \frac{1}{\pi} \text{P.V.} \int_0^1 dx \left(\frac{1}{\xi - x} - \frac{1}{\xi + x} \right) \text{Im}\mathcal{H}(x, t)$$

$$\begin{aligned} & \langle N(\mathbf{p}', s') | \hat{T}^{\mu\nu} | N(\mathbf{p}, s) \rangle \\ &= \frac{1}{m} \bar{u}(\mathbf{p}', s') \left[P^\mu P^\nu A(t) + i P^{\{\mu} \sigma^{\nu\} \rho} \Delta_\rho J(t) + \frac{1}{4} (\Delta^\mu \Delta^\nu - g^{\mu\nu} \Delta^2) D(t) \right] u(\mathbf{p}, s), \end{aligned}$$

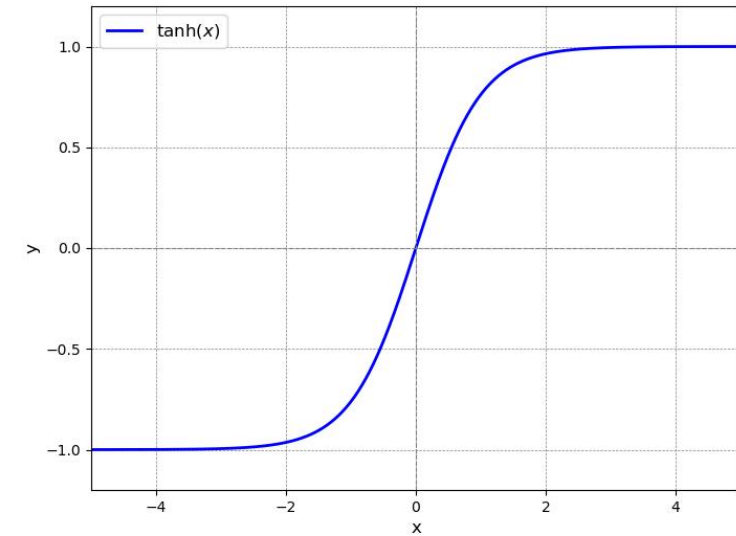
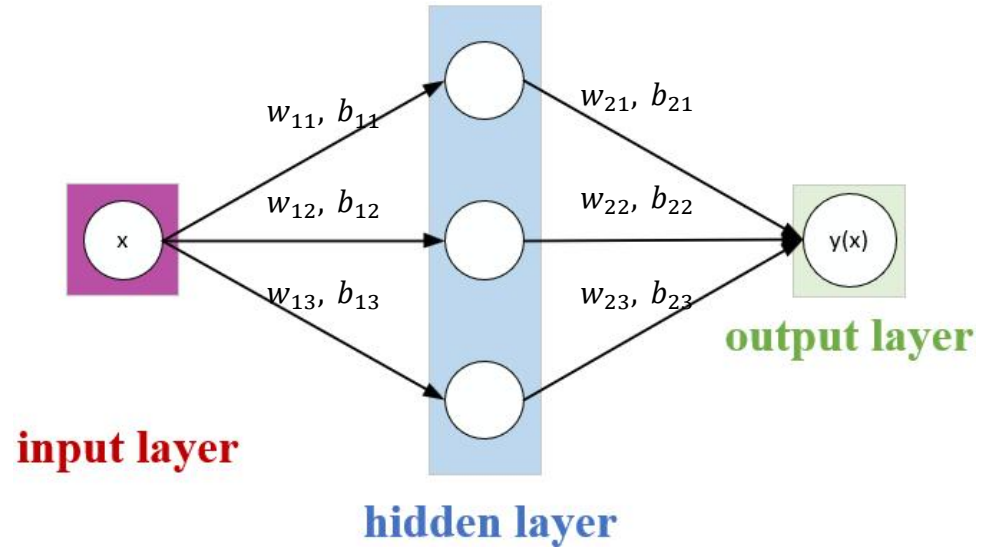
Introduction



Neural network

- Perceptron and Activation function

- The simplest neural network

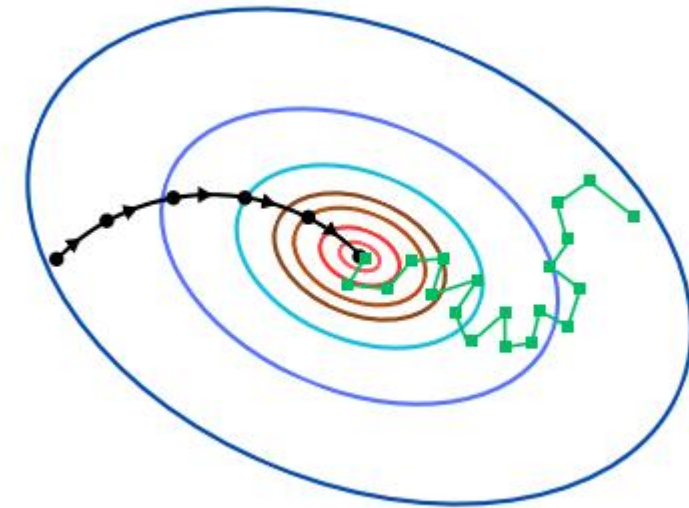
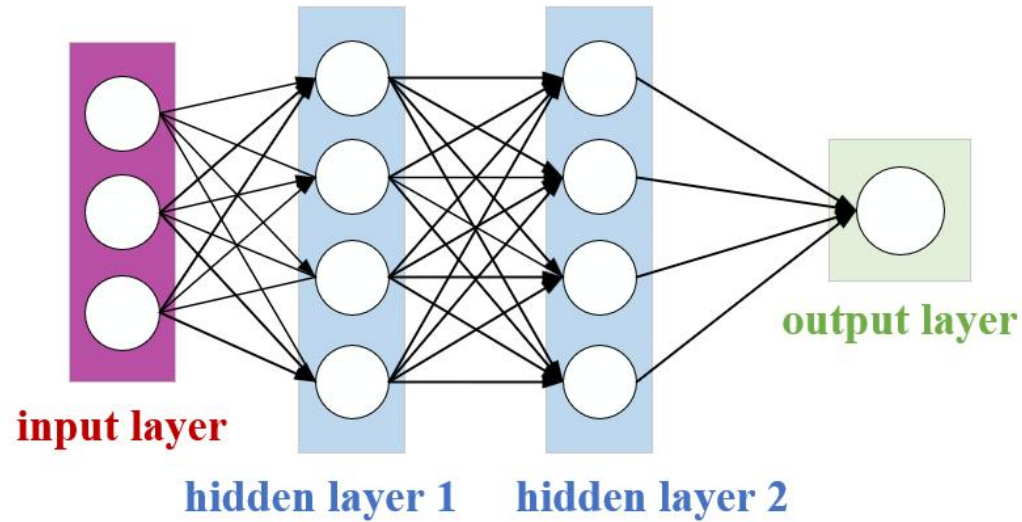


$$y(x) = \sum w_{2i} \tanh(w_{1i}x + b_{1i}) + b_{2i}$$

Neural network

- **Multilayer Perceptron and loss function**

- The simplest neural network



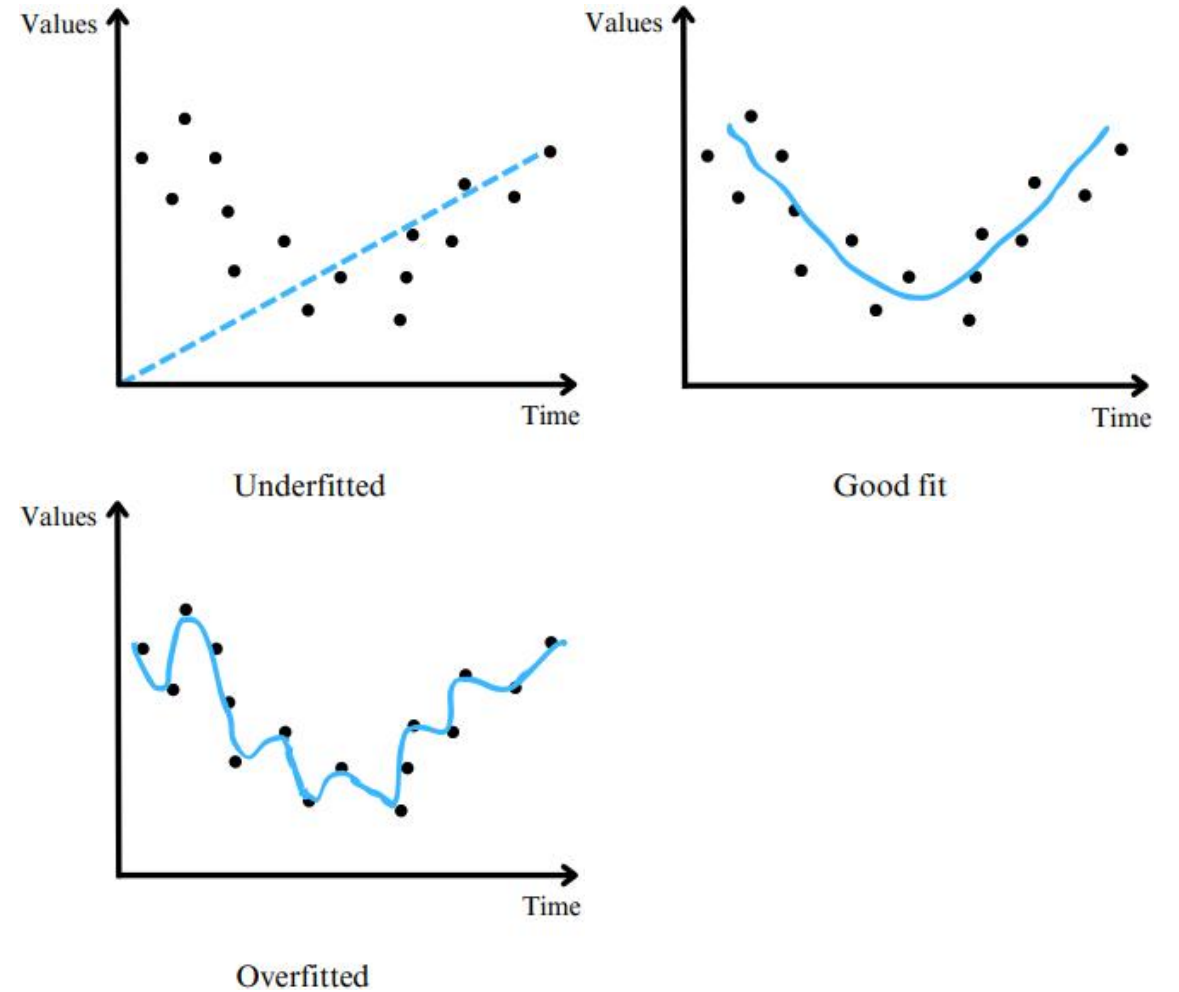
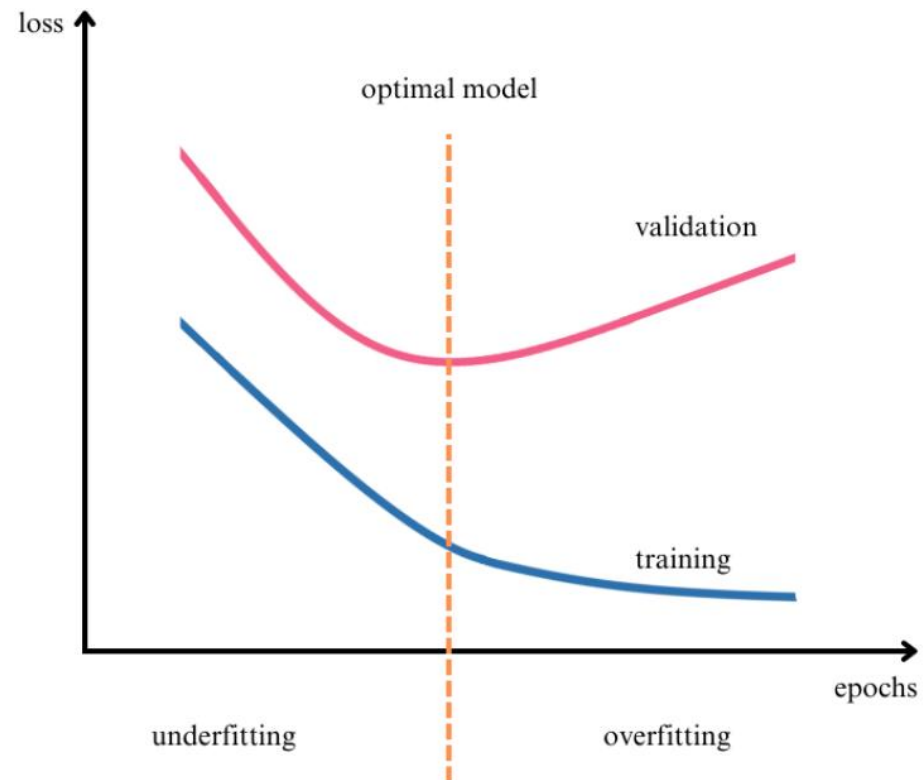
$$y(x) = (W_i((W_{i-1}(\dots(W_1(x) + B_1) + \dots) + B_{i-1}) + B_i)$$

$$Loss = (y(x) - y_{obs})^2$$

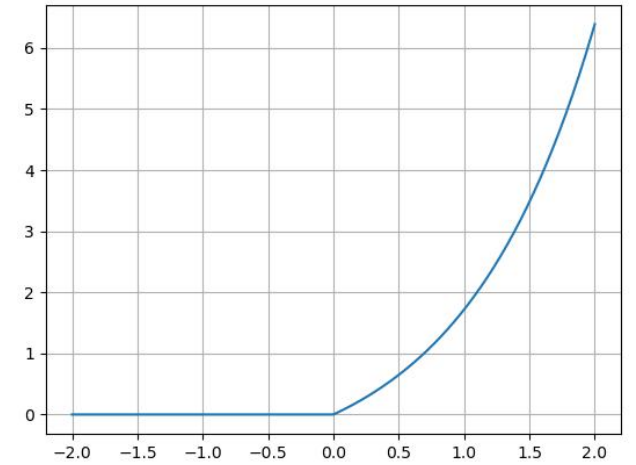
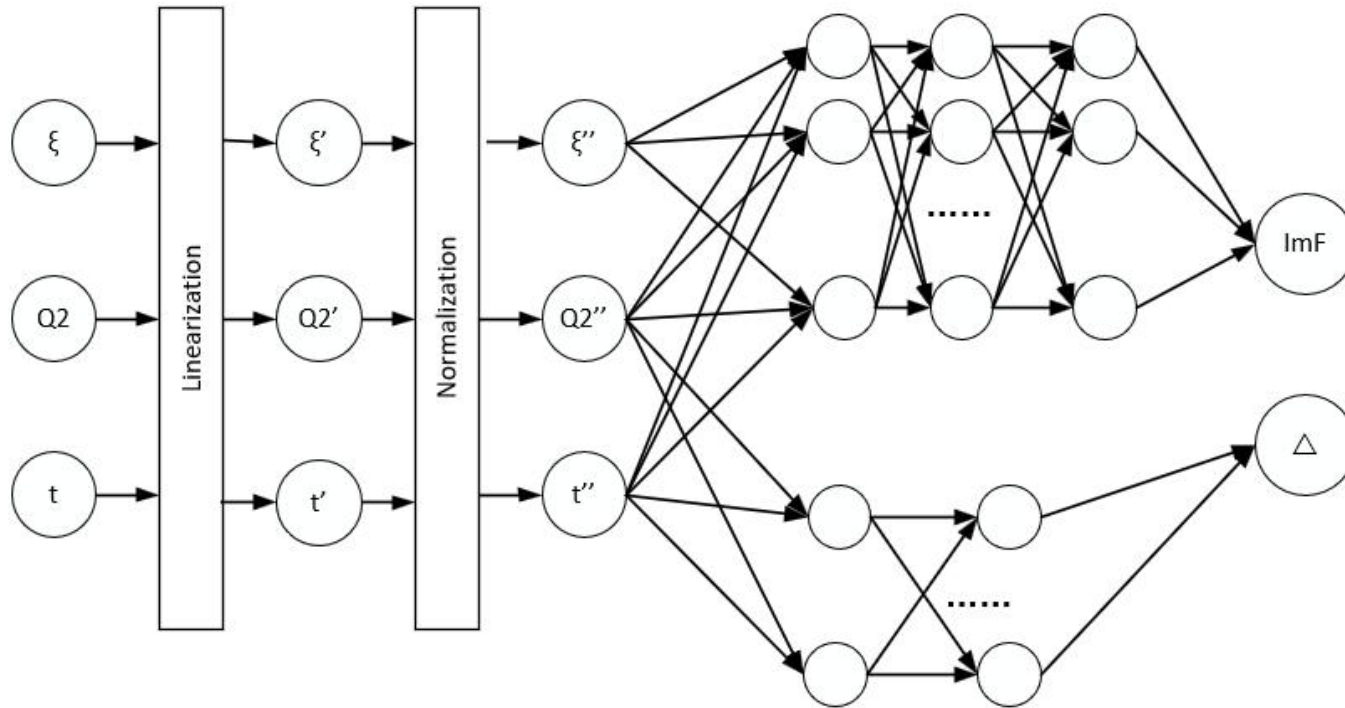
$$w' = w - \frac{\nabla Loss}{\nabla w}$$

Neural network

● Early stopping



Neural network



$$f(x) = \begin{cases} e^x - 1, & x > 0 \\ 0, & x \leq 0 \end{cases}$$

$$\mathcal{L}_{\theta}^{\chi^2}(\mathbf{x}, \sigma) = \mathcal{L}_{\theta}^{\chi^2}(\mathbf{x}, \sigma_{\text{DVCS}}) + \lambda \mathcal{L}_{\theta}^{\chi^2}(\mathbf{x}, \sigma_{\text{LQCD}})$$

Add LQCD data of the Gravitation Form Factors

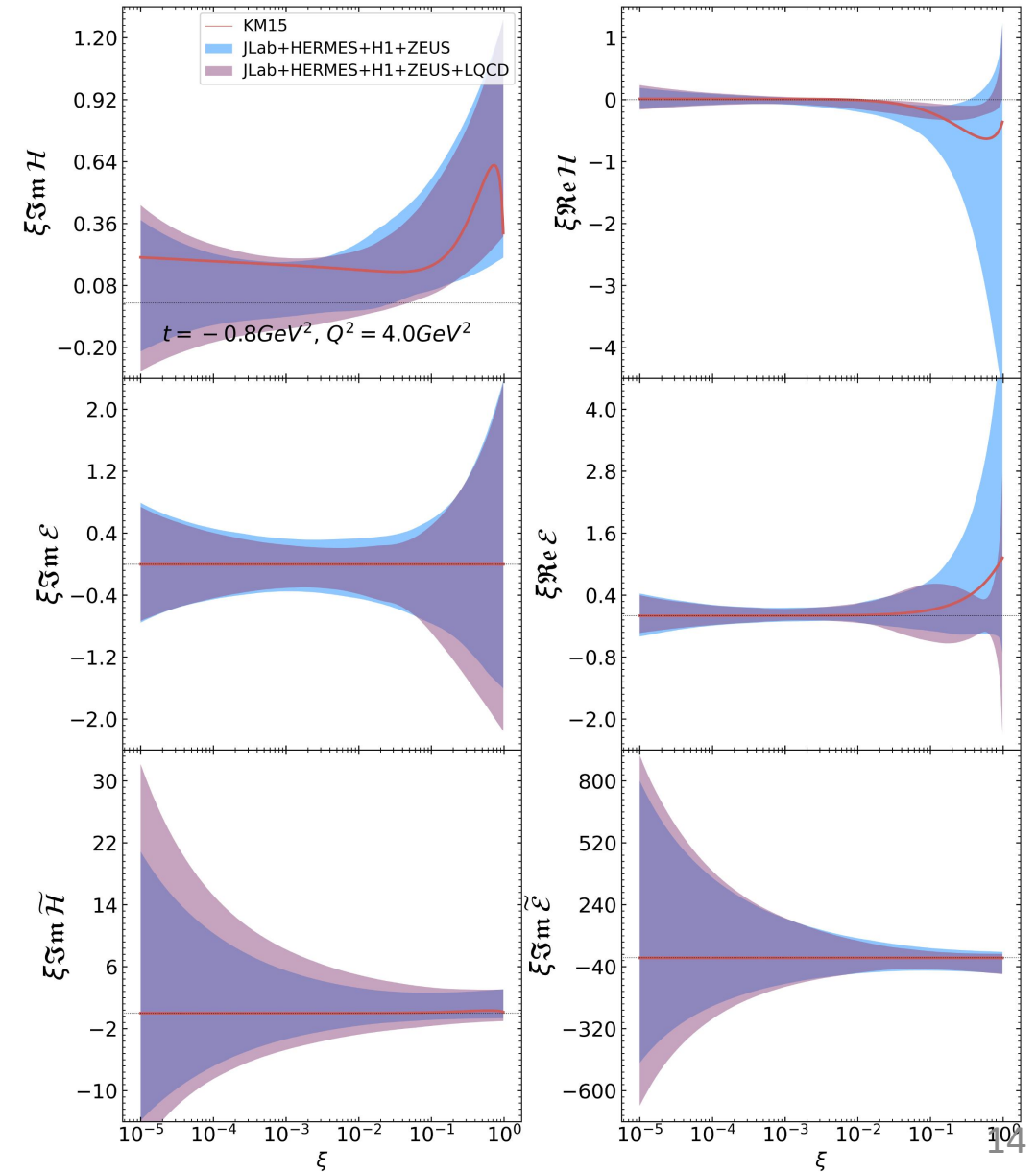
-
- The blue part fit DVCS data without LQCD data

Consistent with the result of PRL 125, 232005 (2020)

- The red part fit DVCS data with LQCD data

Effectively constrains the real parts of H and E

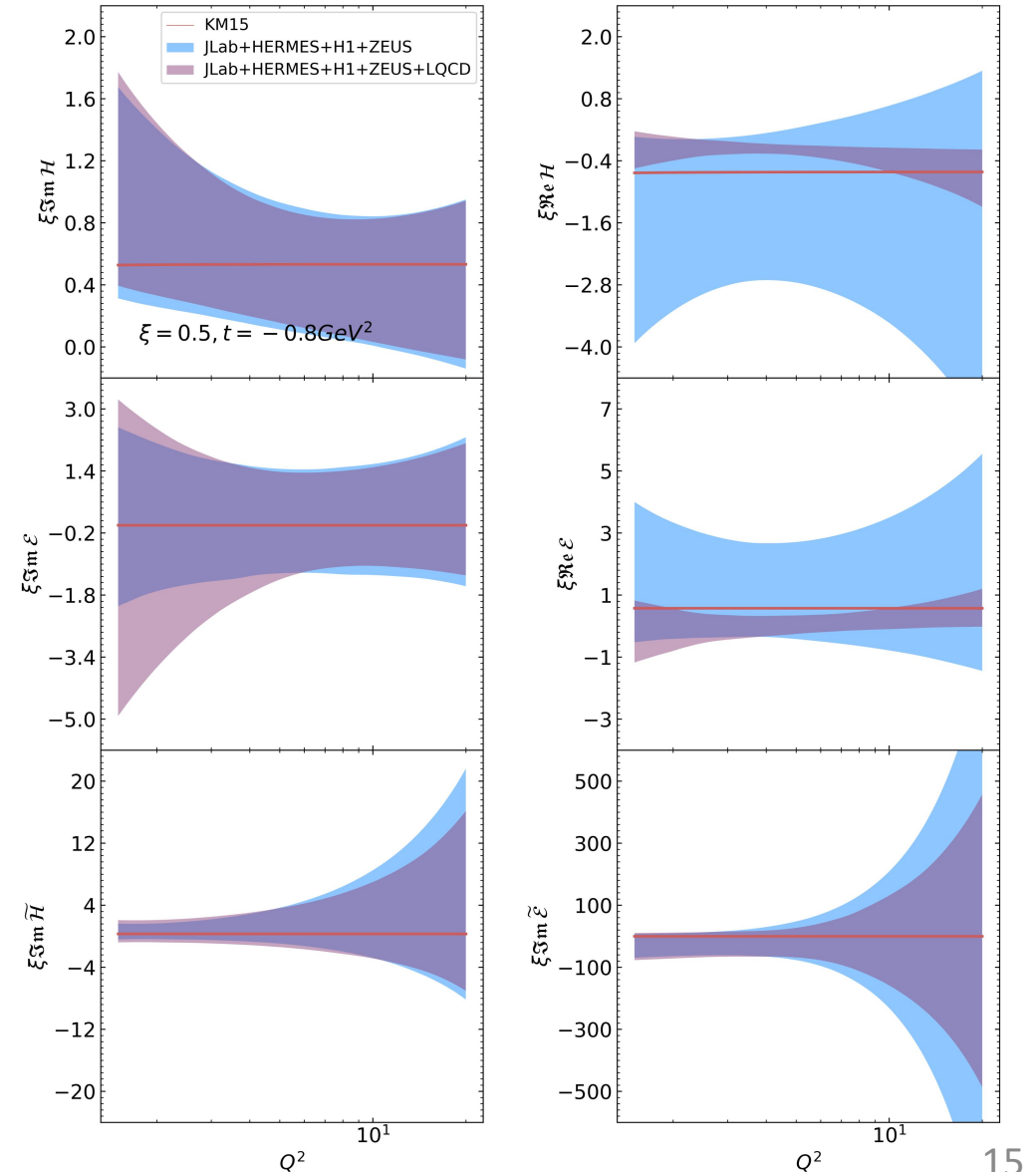
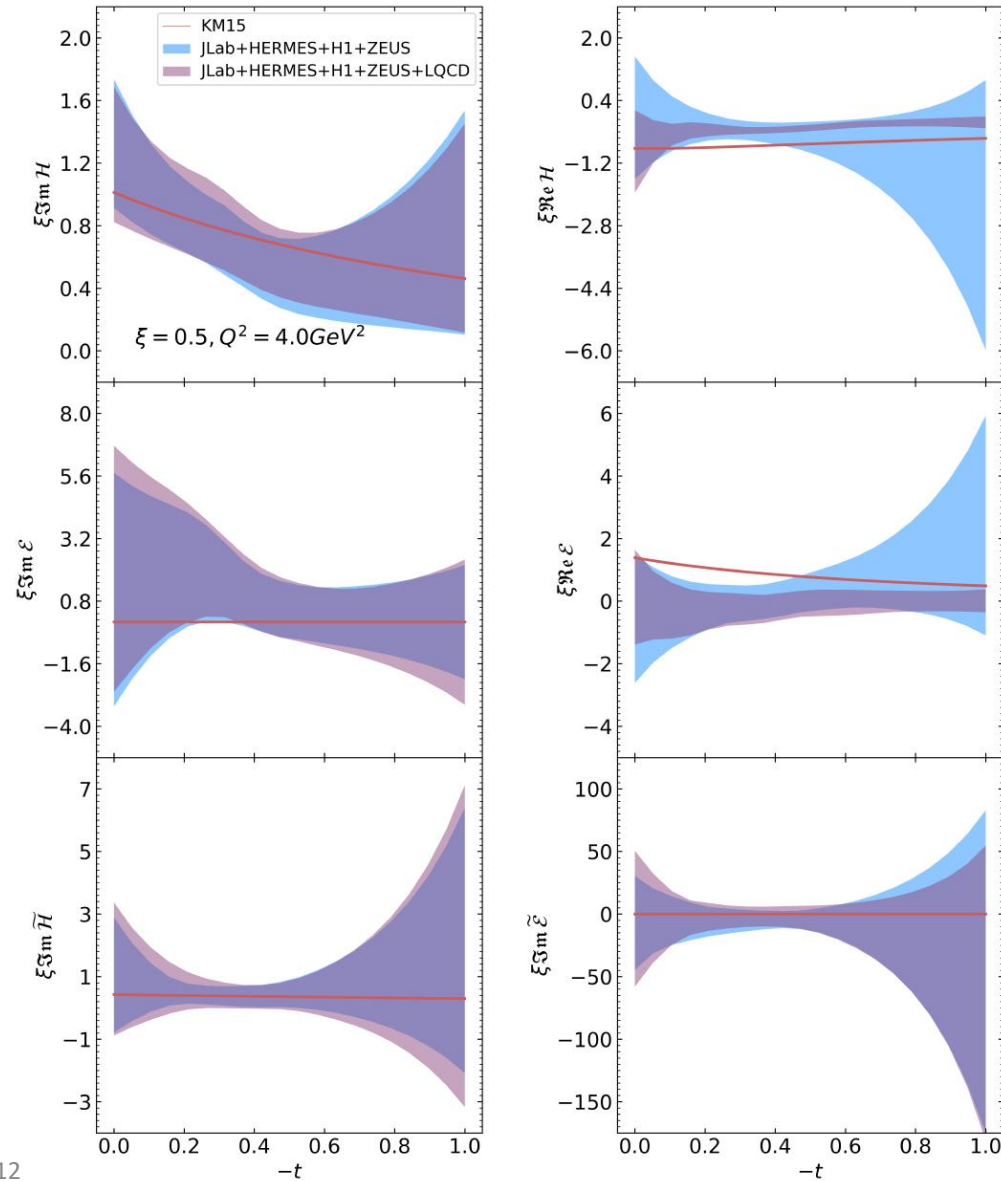
$-t = 0.8\text{GeV}^2, Q^2 = 4.0\text{GeV}^2$



Add LQCD data of the Gravitation Form Factors

$\xi = 0.5, Q^2 = 4.0 \text{ GeV}^2$

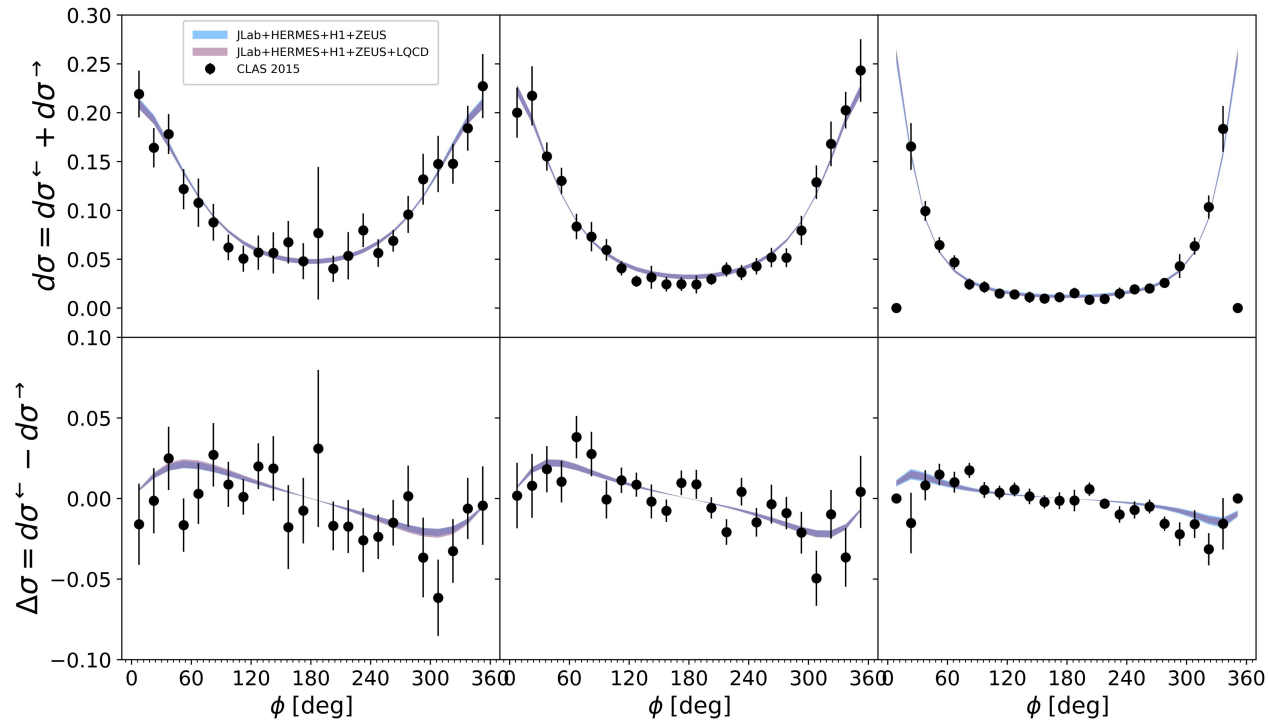
$\xi = 0.5, -t = 0.8 \text{ GeV}^2$



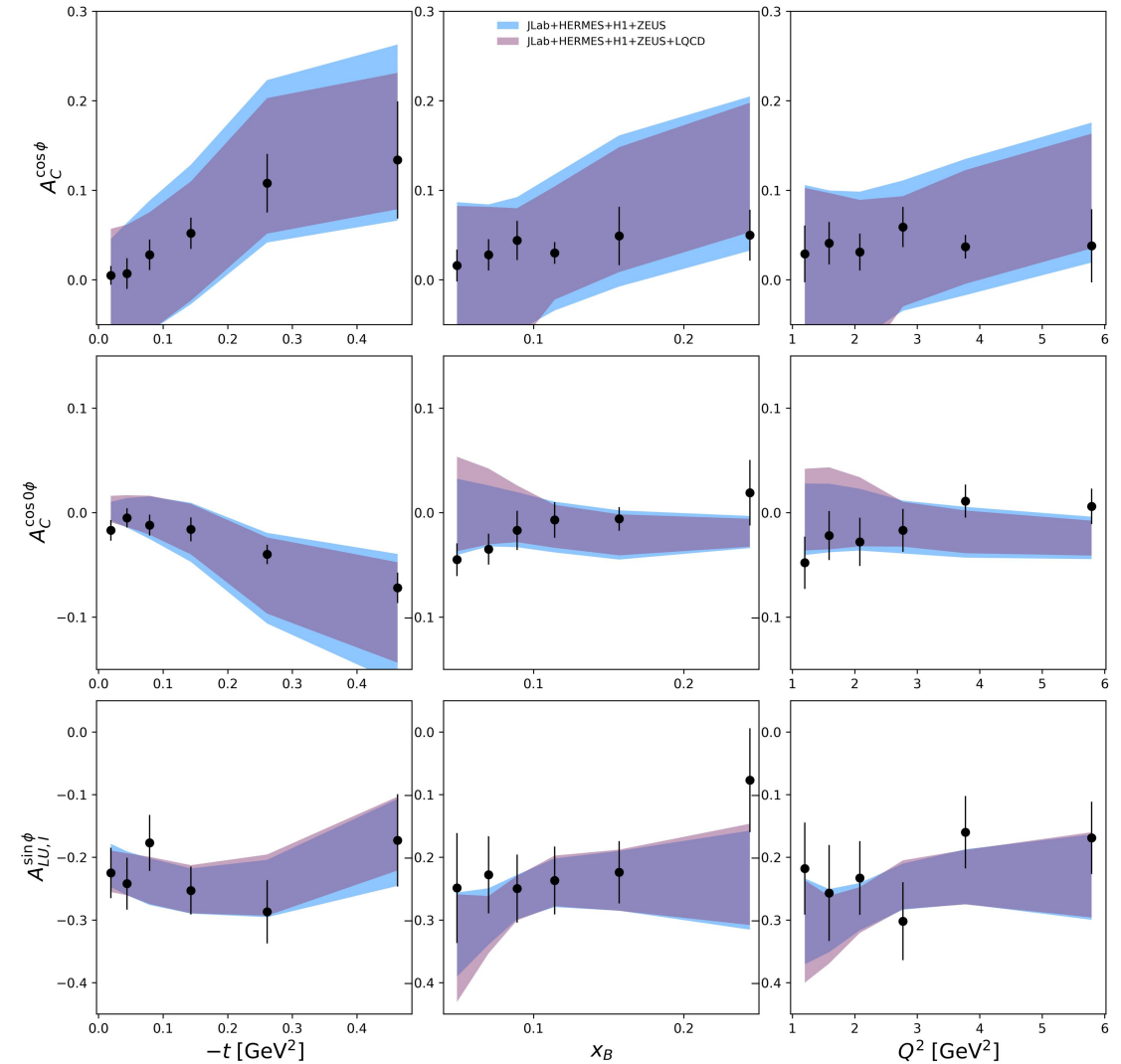
Add LQCD data of the Gravitation Form Factors

● CLAS2015

HERMES2012



χ^2_{DVCS}	$\chi^2_{DVCS+LQCD}$
0.87	1.27

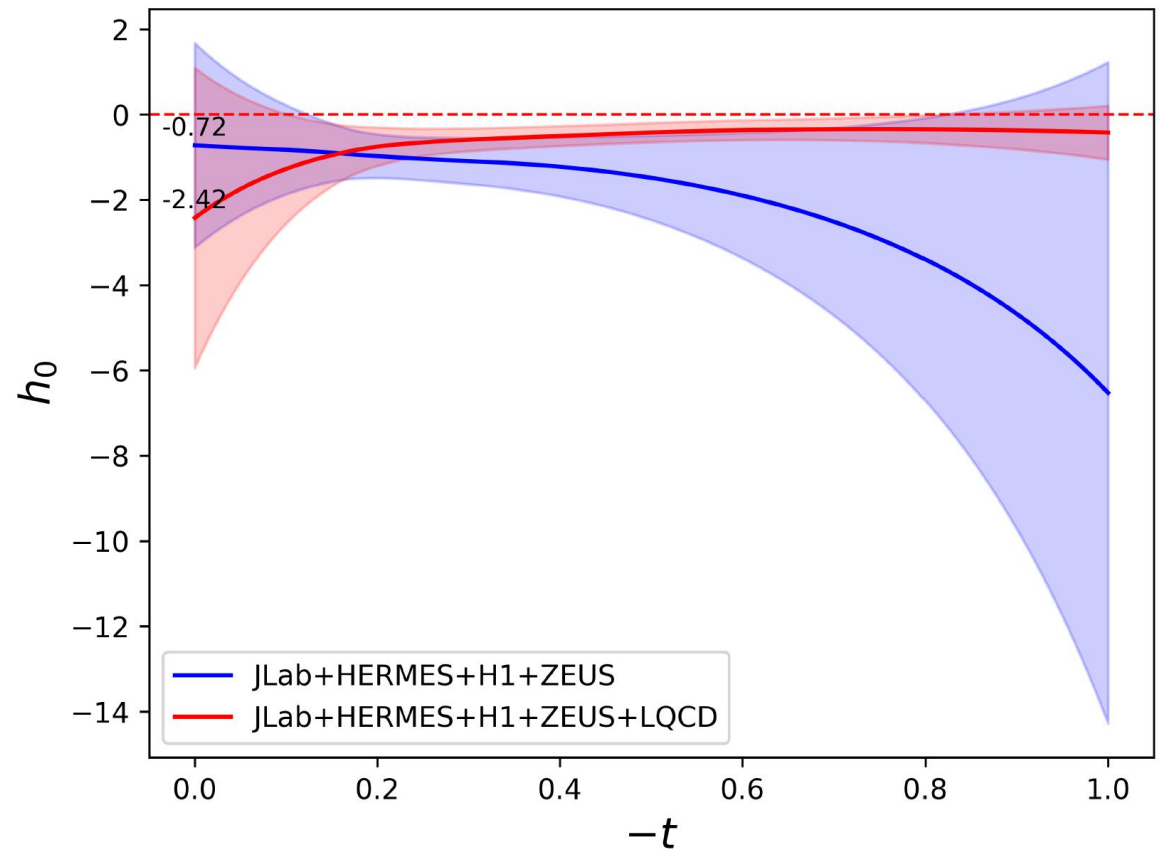
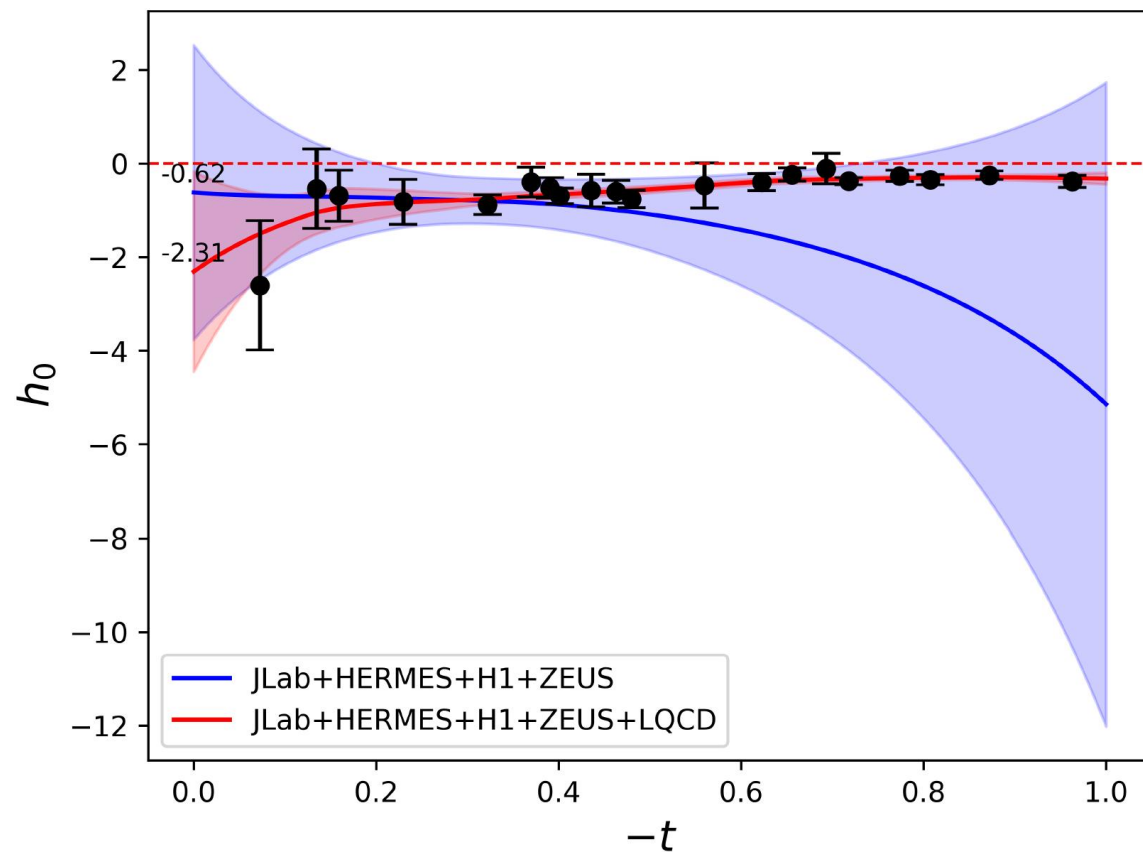


Add LQCD data of the Gravitation Form Factors

● for $j=0$: $\text{Re}\mathcal{H}(\xi, t) = \Delta(t) + \frac{1}{\pi} \text{P.V.} \int_0^1 dx \left(\frac{1}{\xi-x} - \frac{1}{\xi+x} \right) \text{Im}\mathcal{H}(x, t)$ $e_0 = -\Delta(t)$ for CFF E

● **$Q^2 = 4.0 \text{ GeV}^2$**

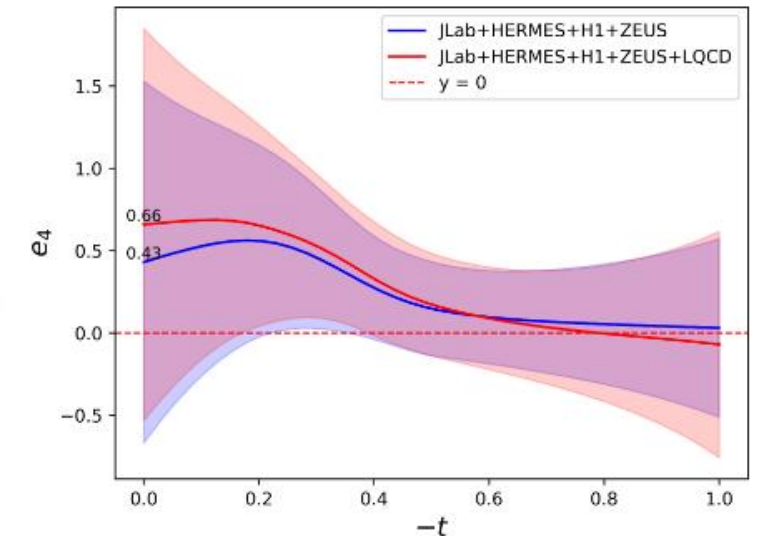
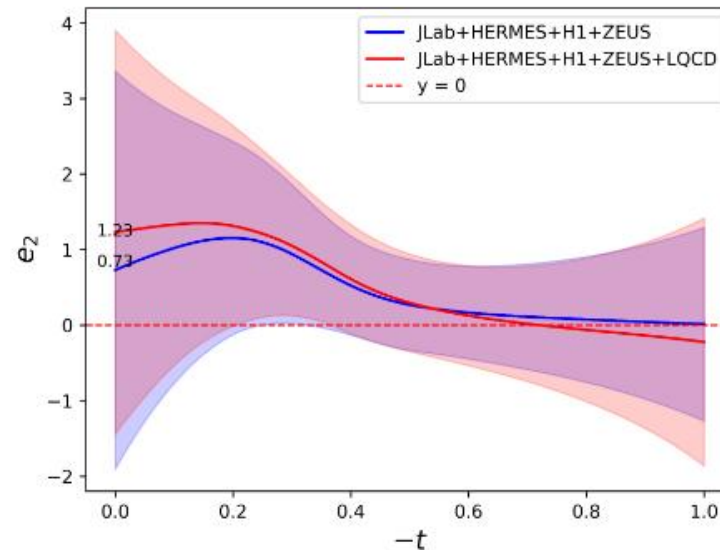
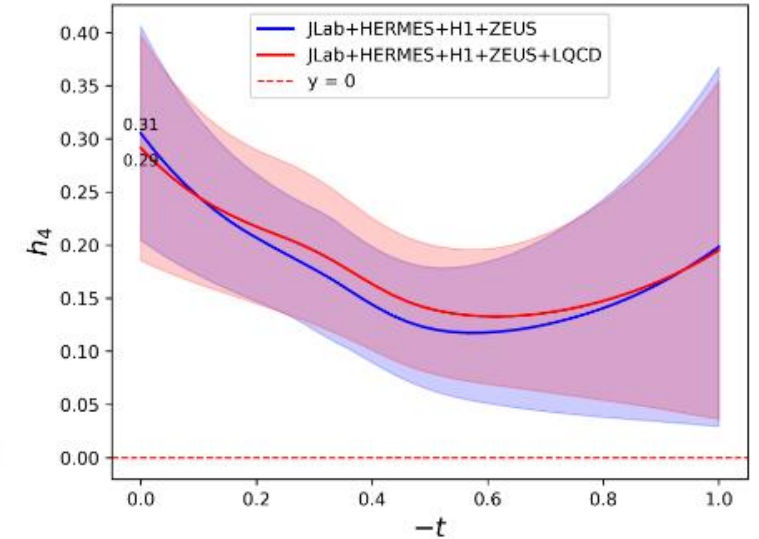
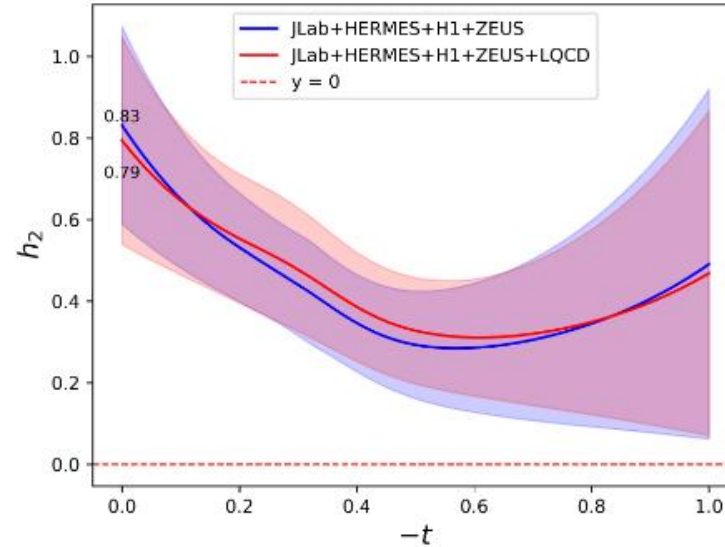
$Q^2 = 2.0 \text{ GeV}^2$



Add LQCD data of the Gravitation Form Factors

- for $j>0$: $h_{2j} = \frac{2}{\pi} \int_0^1 \Im \mathcal{H}(\xi') (\xi')^{2j-1} d\xi'$ and e_{2j} for CFF E

● **$Q^2 = 4.0 \text{ GeV}^2$**



Summary

●1. For CFF

The inclusion of LQCD data has shown a significant constraint on both $\text{Re}H$ and $\text{Re}E$.

●2. For observables

The chi-square values of the observables exhibit strong consistency before and after incorporating LQCD data.

●3. For subtraction constants

The inclusion of LQCD data provides a good constraint on L_0 and demonstrates good consistency in higher-order subtraction constants.

Thanks