

# **Validating the GPD extraction framework**

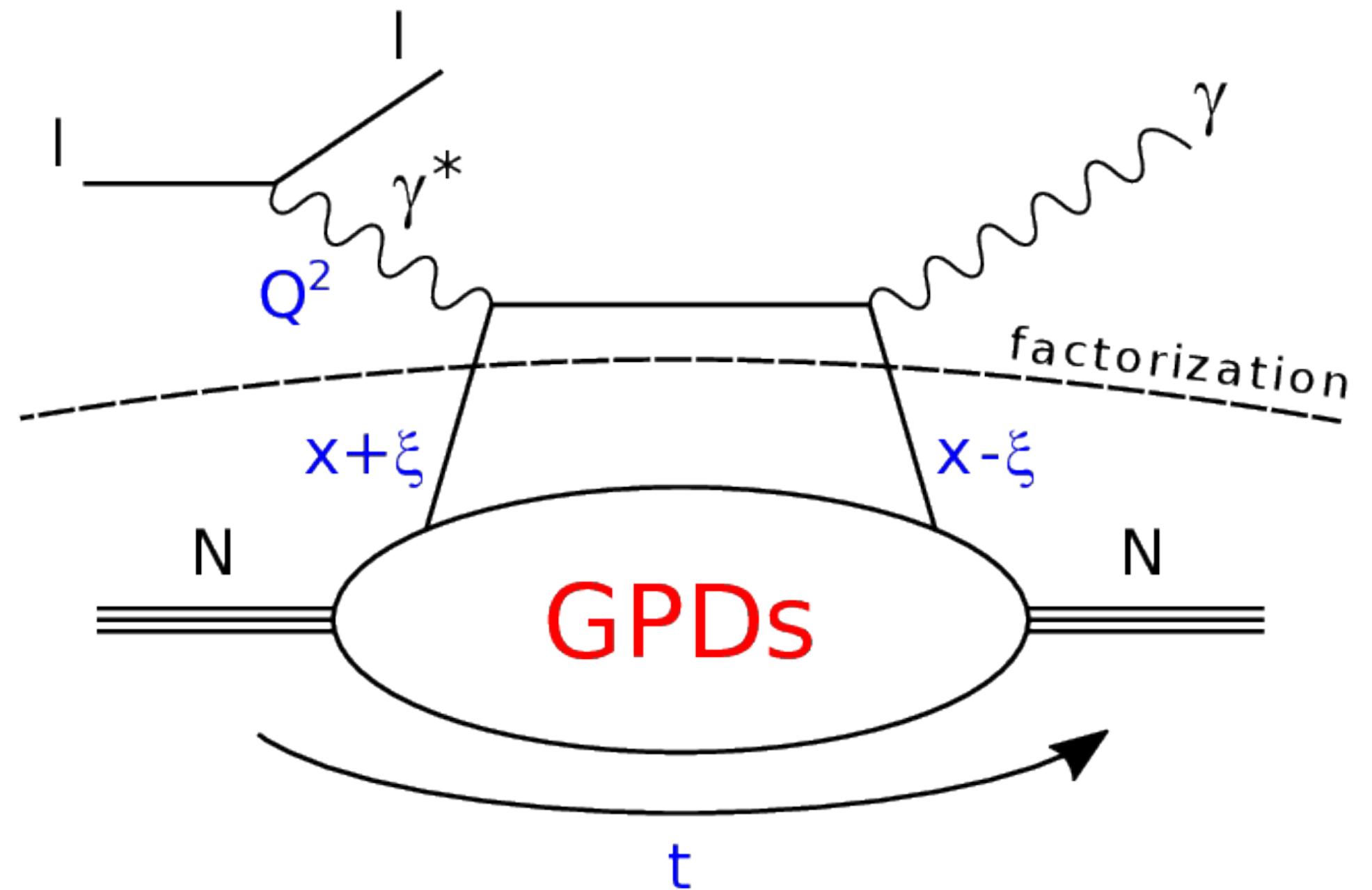


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Opportunities with JLab20+ upgrade workshop, Trento, Italy, September 29th, 2022

# **1. Introduction**

## Deeply Virtual Compton Scattering (DVCS)



*factorisation for  $|t|/Q^2 \ll 1$*

Chiral-even GPDs:  
(helicity of parton conserved)

$H^{q,g}(x, \xi, t)$	$E^{q,g}(x, \xi, t)$	for sum over parton helicities
$\tilde{H}^{q,g}(x, \xi, t)$	$\tilde{E}^{q,g}(x, \xi, t)$	for difference over parton helicities
nucleon helicity conserved	nucleon helicity changed	

## Reduction to PDF:

$$H(x, \xi = 0, t = 0) \equiv q(x)$$

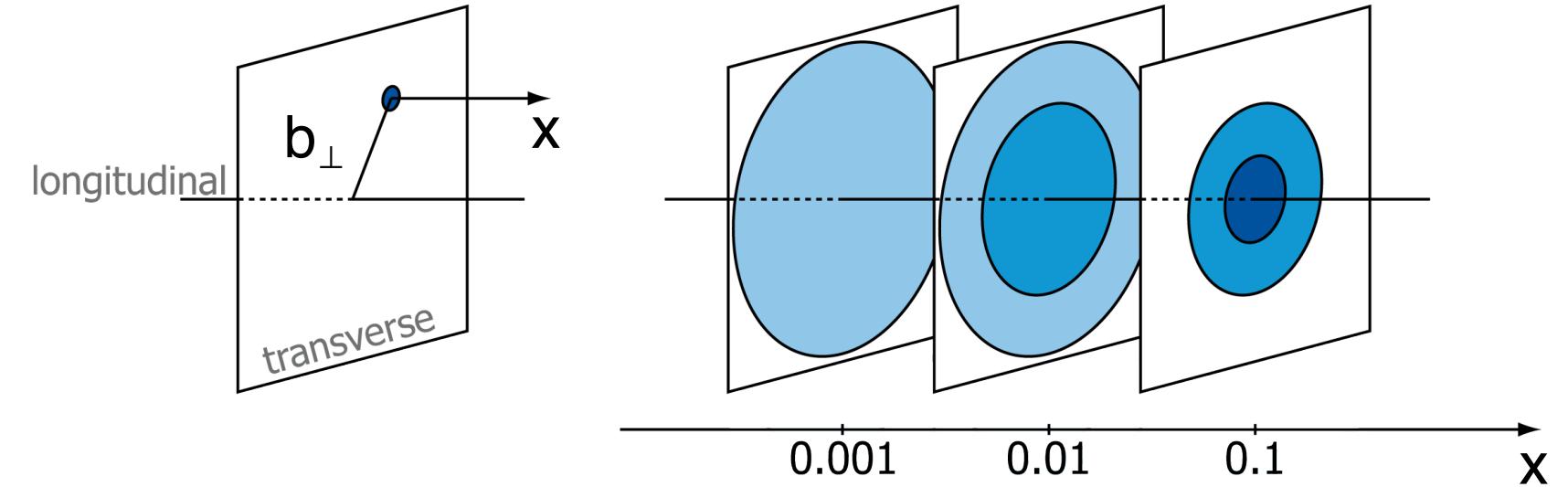
## Polynomiality - non-trivial consequence of Lorentz invariance:

$$\mathcal{A}_n(\xi, t) = \int_{-1}^1 dx x^n H(x, \xi, t) = \sum_{\substack{j=0 \\ \text{even}}}^n \xi^j A_{n,j}(t) + \text{mod}(n, 2) \xi^{n+1} A_{n,n+1}(t)$$

## Positivity bounds - positivity of norm in Hilbert space, e.g.:

$$|H(x, \xi, t)| \leq \sqrt{q \left( \frac{x + \xi}{1 + \xi} \right) q \left( \frac{x - \xi}{1 - \xi} \right) \frac{1}{1 - \xi^2}}$$

## Nucleon tomography:



$$q(x, \mathbf{b}_\perp) = \int \frac{d^2 \Delta}{4\pi^2} e^{-i\mathbf{b}_\perp \cdot \Delta} H^q(x, 0, t = -\Delta^2)$$

**Energy momentum tensor in terms of form factors  
(OAM and mechanical forces):**

$$\begin{aligned} \langle p', s' | \hat{T}^{\mu\nu} | p, s \rangle &= \bar{u}(p', s') \left[ \frac{P^\mu P^\nu}{M} A(t) + \frac{\Delta^\mu \Delta^\nu - \eta^{\mu\nu} \Delta^2}{M} C(t) + M \eta^{\mu\nu} \bar{C}(t) + \right. \\ &\quad \left. \frac{P^\mu i \sigma^{\nu\lambda} \Delta_\lambda}{4M} [A(t) + B(t) + D(t)] + \frac{P^\nu i \sigma^{\mu\lambda} \Delta_\lambda}{4M} [A(t) + B(t) - D(t)] \right] u(p, s) \end{aligned}$$

$$T^{\mu\nu} = \begin{bmatrix} T^{00} & & & \\ T^{01} & T^{02} & T^{03} & \\ T^{10} & T^{11} & T^{12} & T^{13} \\ T^{20} & T^{21} & T^{22} & T^{23} \\ T^{30} & T^{31} & T^{32} & T^{33} \end{bmatrix}$$

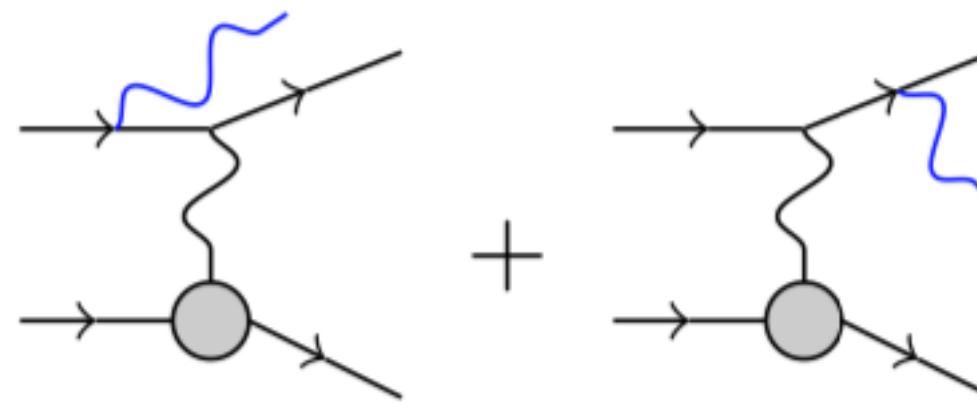
Energy density      Momentum density  
Energy flux      Momentum flux  
Shear stress      Normal stress

## **2. Phenomenology at level of DVCS amplitudes (Compton form factors)**

Cross-section for single photon production ( $l + N \rightarrow l + N + \gamma$ ):

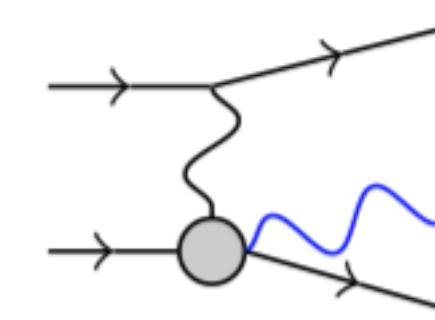
$$\sigma \propto |\mathcal{A}|^2 = |\mathcal{A}_{BH} + \mathcal{A}_{DVCS}|^2 = |\mathcal{A}_{BH}|^2 + |\mathcal{A}_{DVCS}|^2 + \mathcal{J}$$

*Bethe-Heitler process*



*calculable within QED  
parametrised by elastic FFs*

*DVCS*



*calculable within QCD  
parametrised by CFFs*

**For more details and formulae  
see e.g.:  
A. V. Belitsky et al.  
NPB 878 (2014) 214**

$$\text{Im}\mathcal{H}(\xi, t) \stackrel{\text{LO}}{=} \pi \sum_q e_q^2 H^{q(+)}(\xi, \xi, t)$$

$$\text{Re}\mathcal{H}(\xi, t) = \text{PV} \int_0^1 \frac{d\xi'}{\pi} \text{Im}\mathcal{H}(\xi', t) \left( \frac{1}{\xi - \xi'} - \frac{1}{\xi + \xi'} \right) + C_H(t)$$

$$G = \{H, E, \widetilde{H}, \widetilde{E}\}$$

H. Moutarde, PS, J. Wagner,  
Eur. Phys. J. C 78 (2018) 11, 890

$$G^q(x, 0, t) = \text{pdf}_G^q(x) \exp(f_G^q(x)t)$$

$$f_G^q(x) = A_G^q \log(1/x) + B_G^q(1-x)^2 + C_G^q(1-x)x$$

- reduction to PDFs and correspondence to EFFs
- modify "classical"  $\log(1/x)$  term by  $B_G^q(1-x)^2$  in low- $x$  and by  $C_G^q(1-x)x$  in high- $x$  regions
- polynomials found in analysis of EFF data → good description of data
- allow to use the analytic regularisation prescription
- finite proton size at  $x \rightarrow 1$

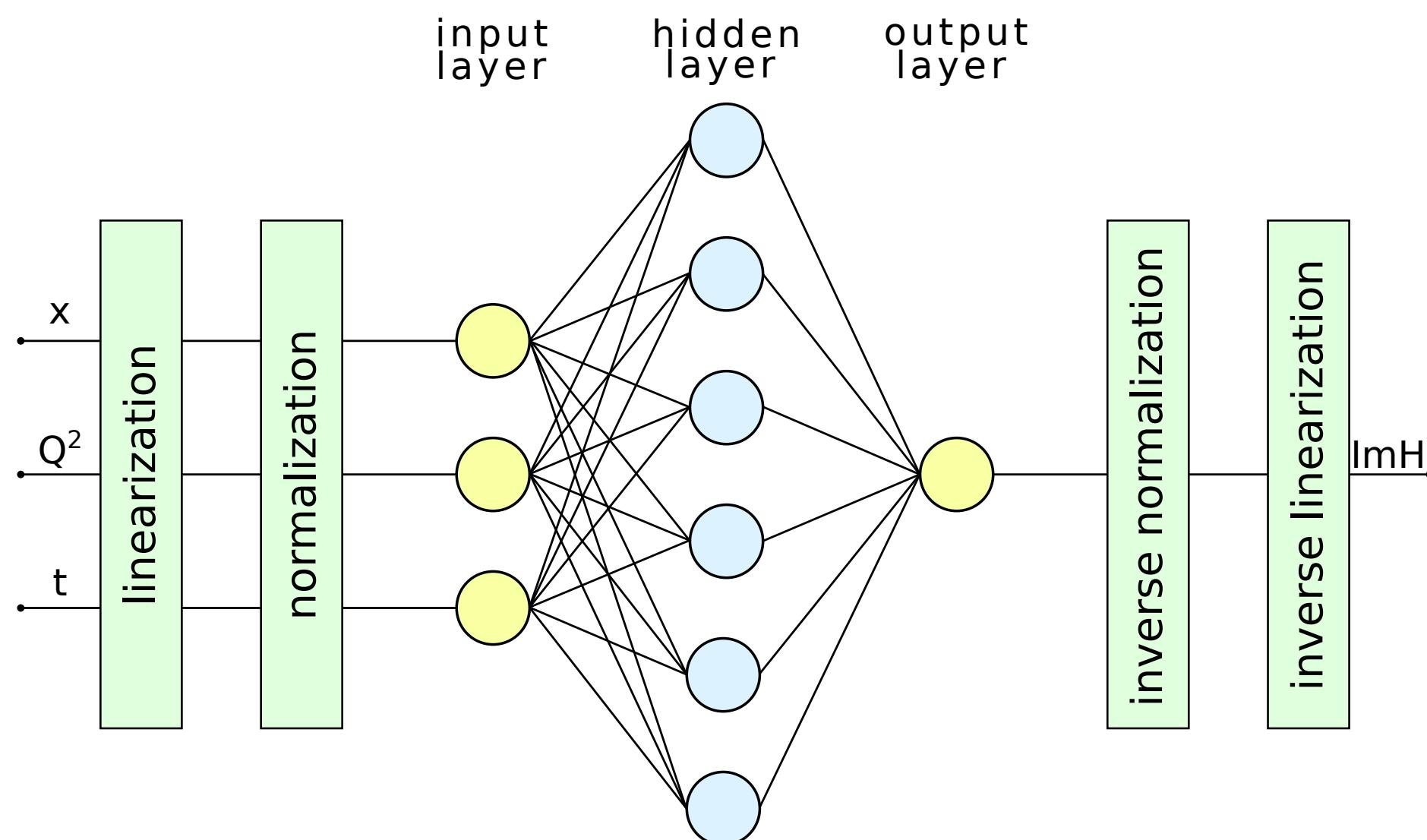
$$G^q(x, x, t) = G^q(x, 0, t) g_G^q(x, x, t) \quad g_G^q(x, x, t) = \frac{a_G^q}{(1-x^2)^2} (1 + t(1-x)(b_G^q + c_G^q \log(1+x)))$$

- at  $x \rightarrow 0$  constant skewness effect
- at  $x \rightarrow 1$  reproduce power behaviour predicted for GPDs in Phys. Rev. D69, 051501 (2004)
- t-dependence similar to DD-models with  $(1-x)$  to avoid any t-dep. at  $x = 1$

$$C_G^q(t) = 2 \int_{(0)}^1 \left( G^{q(+)}(x, x, t) - G^{q(+)}(x, 0, t) \right) \frac{1}{x} dx$$

- subtraction constant as analytic continuation of Mellin moments to  $j = -1$

## Features of analysis:



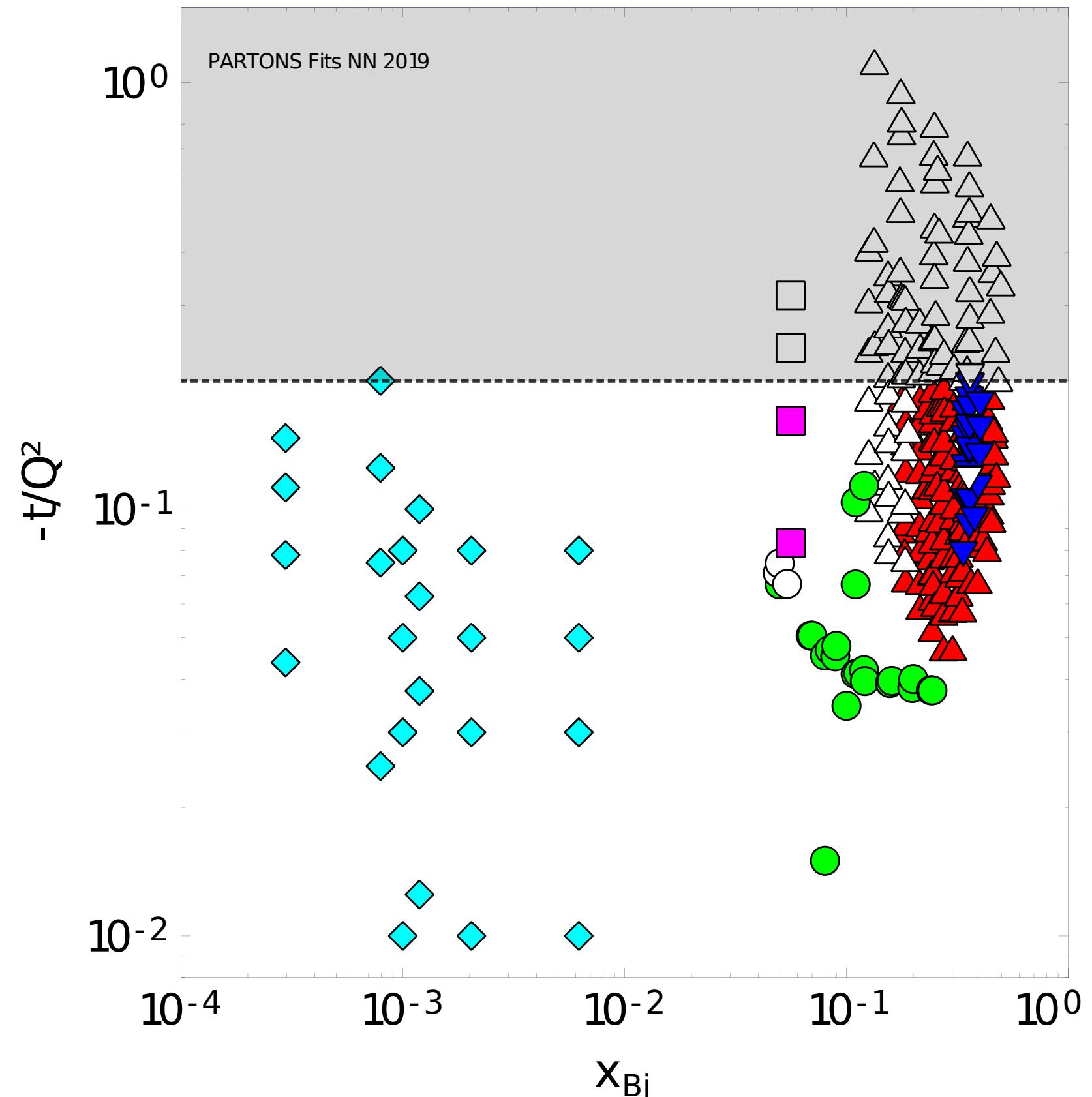
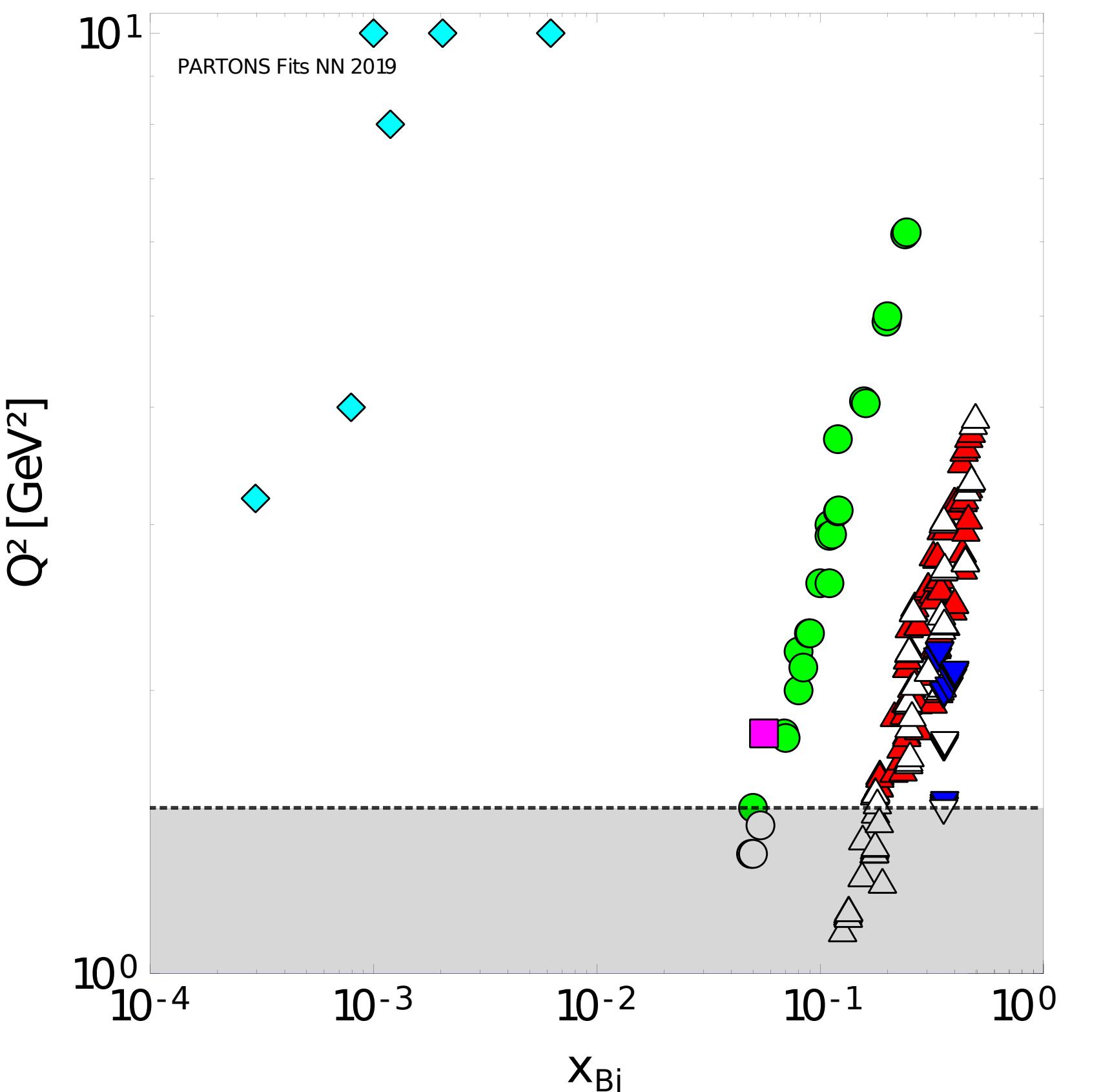
- Independent artificial neural network for each CFF and Re/Im parts
- Functions of  $x_B$ ,  $Q^2$  and  $t$
- Network size determined using benchmark sample
- No power-behaviour pre-factors
- Trained with genetic algorithm
- Regularisation method based on early stopping criterion
- Replica method for propagation of experimental uncertainties

Kinematic cuts  
 used in our recent analyses:

$$Q^2 > 1.5 \text{ GeV}^2$$

$$-t/Q^2 < 0.2$$

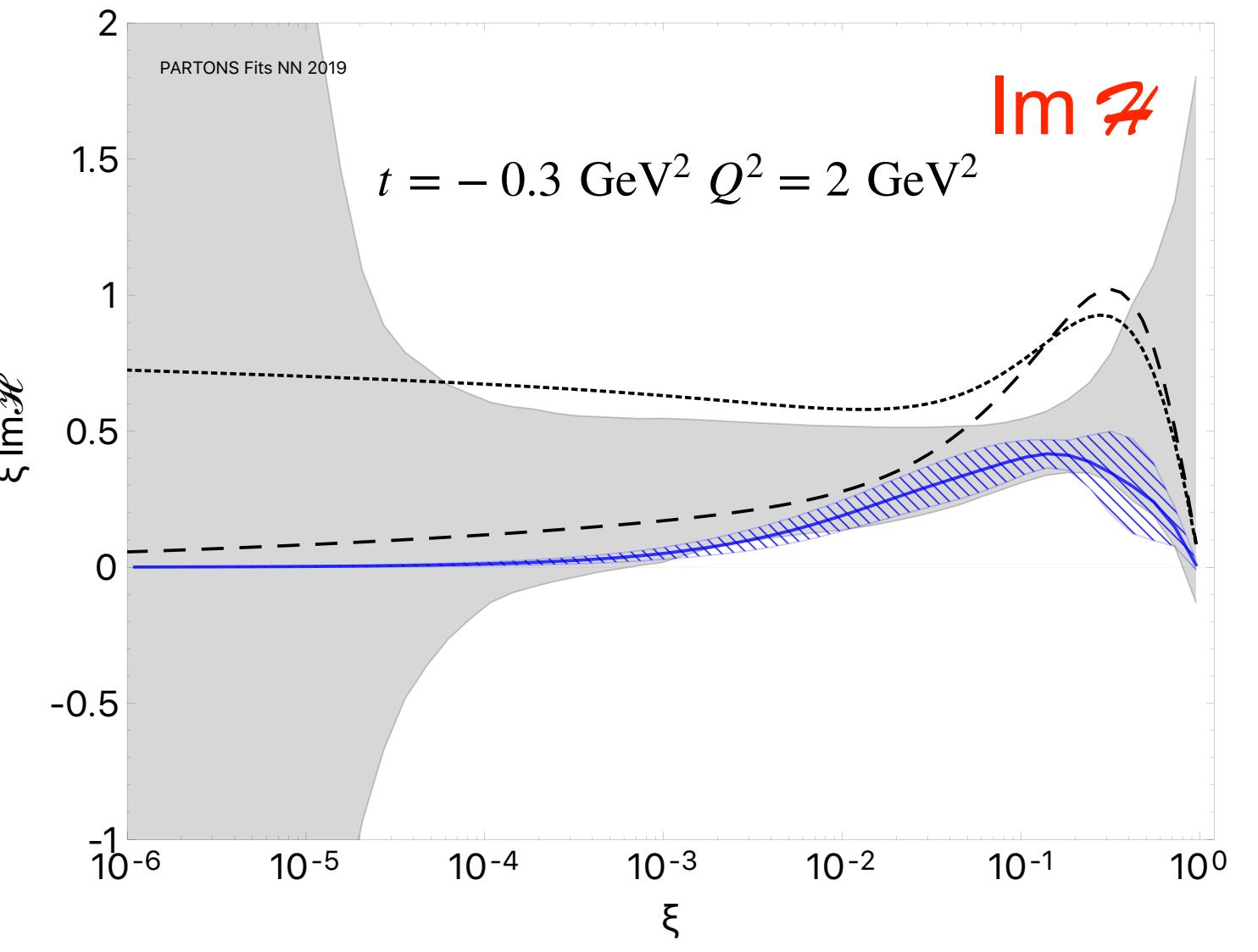
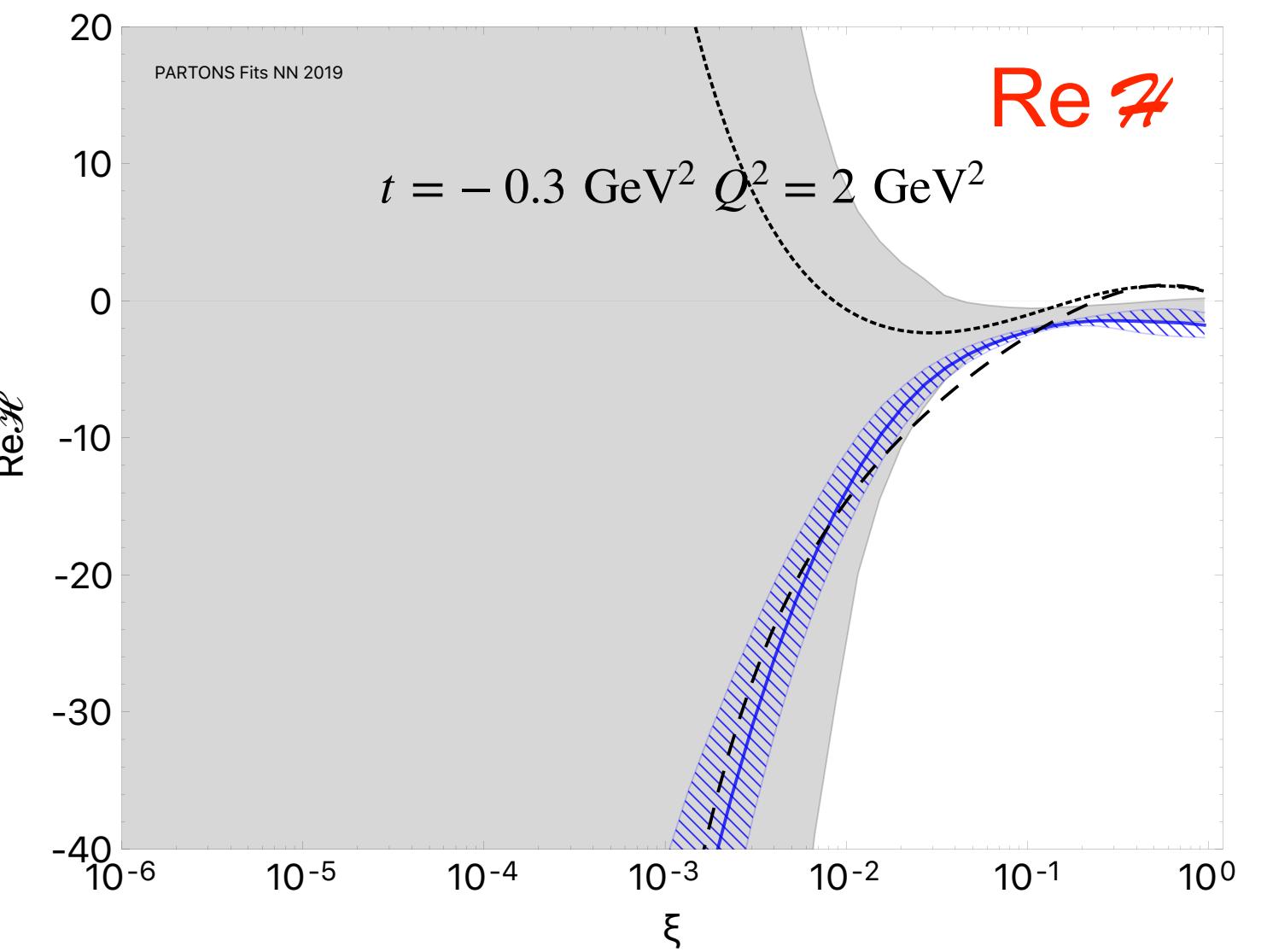
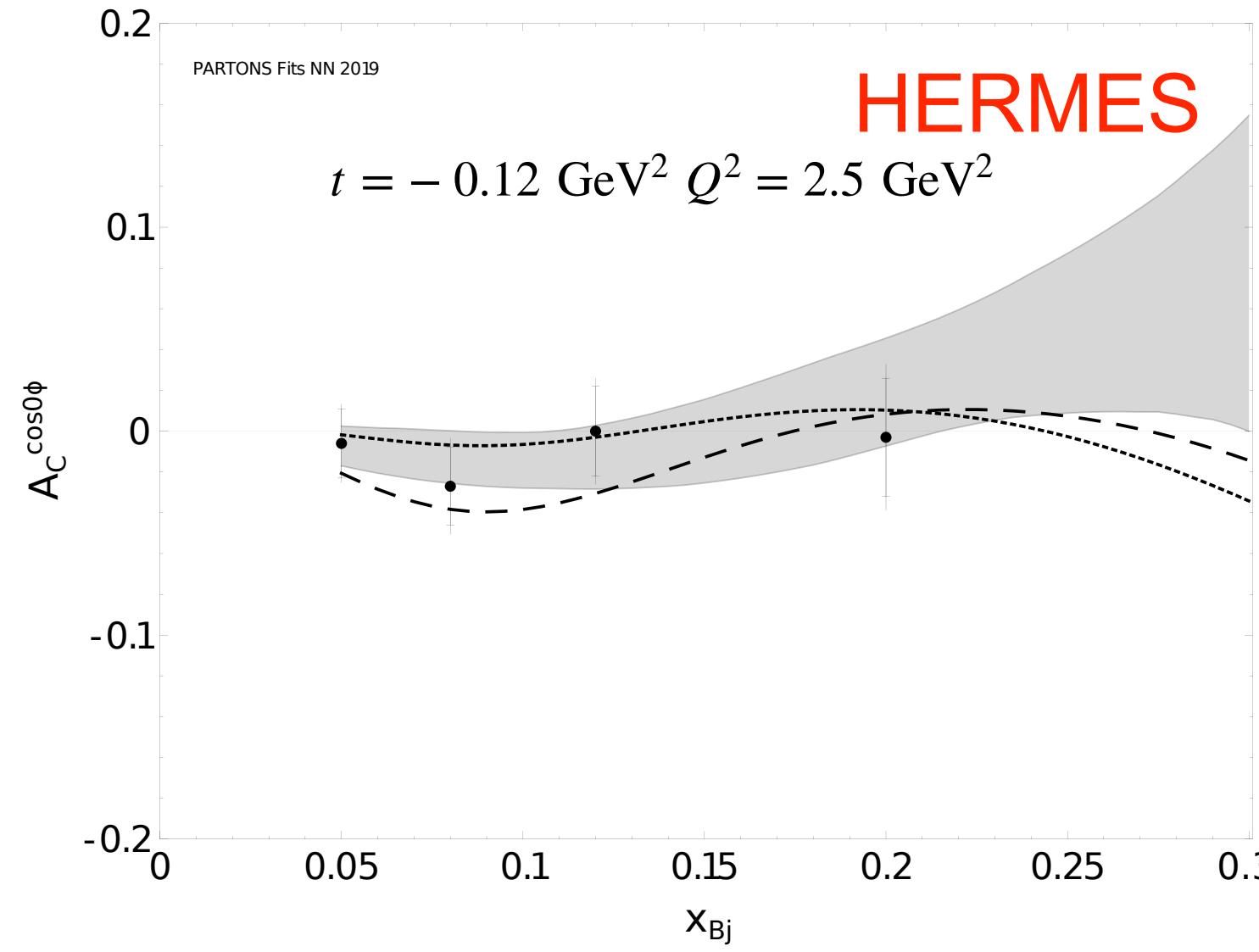
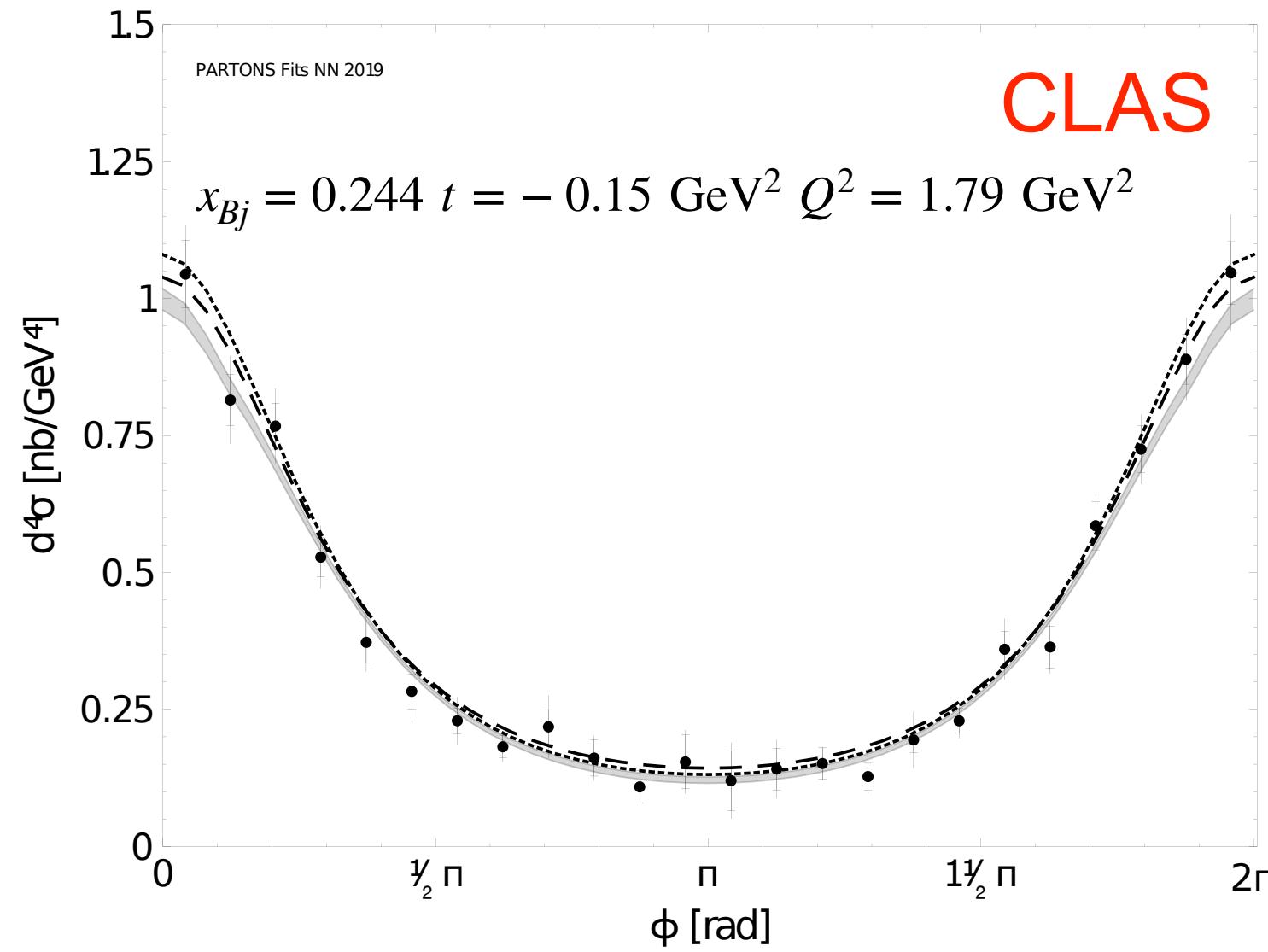
- ▼ HALL A
- ▲ CLAS
- HERMES
- COMPASS
- ◆ H1 and ZEUS



Note: both analytic and non-analytic Ansätze use specific PDF parametrisations  
 analytic Ansatz is also fitted to elastic FF data

# Results

H. Moutarde, PS, J. Wagner,  
Eur. Phys. J. C 78 (2018) 11, 890



Non-parametric

Parametric

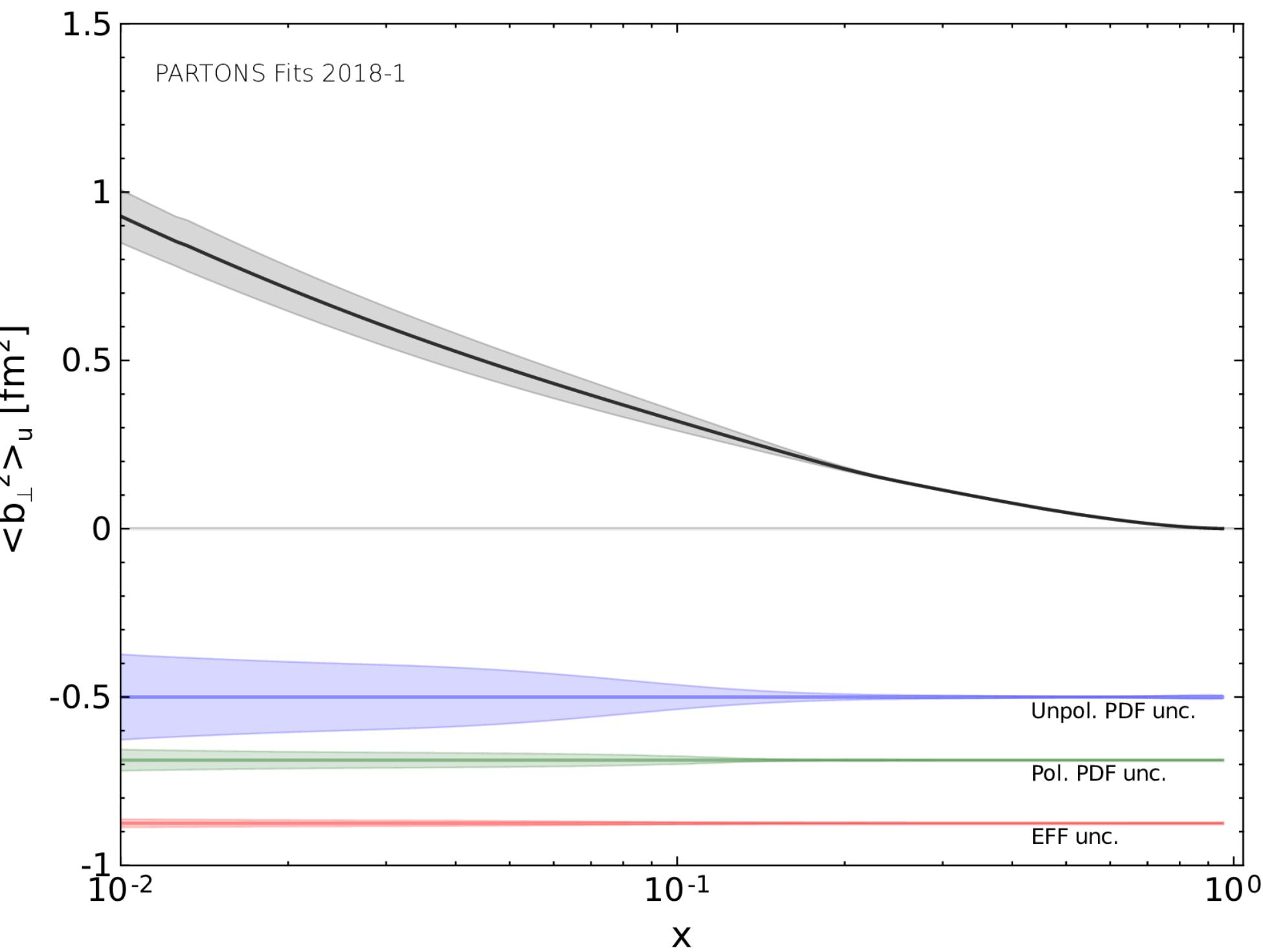
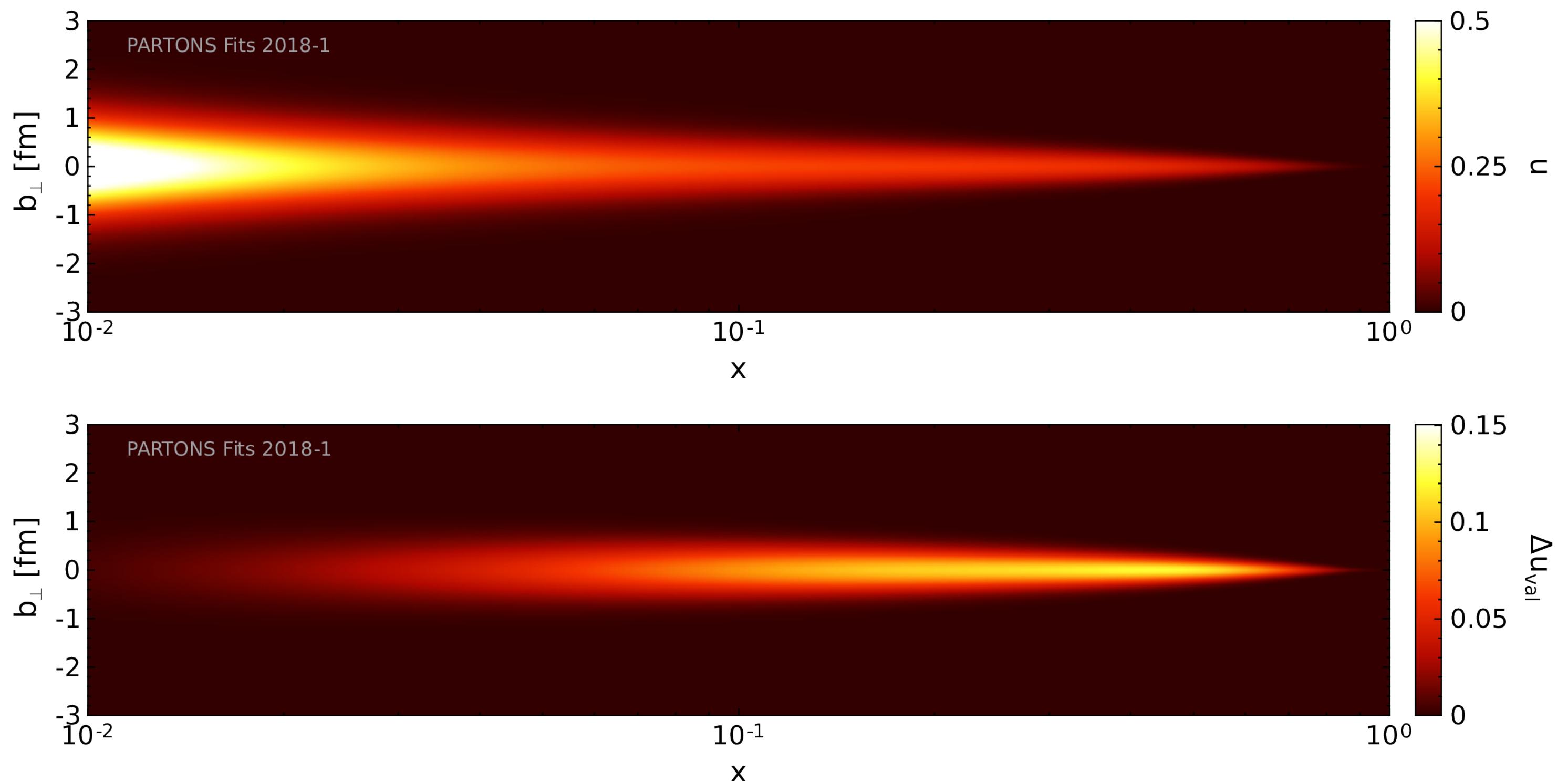
VGG

GK

LO evaluation

$$\xi \approx x_{Bj}/(2 - x_{Bj})$$

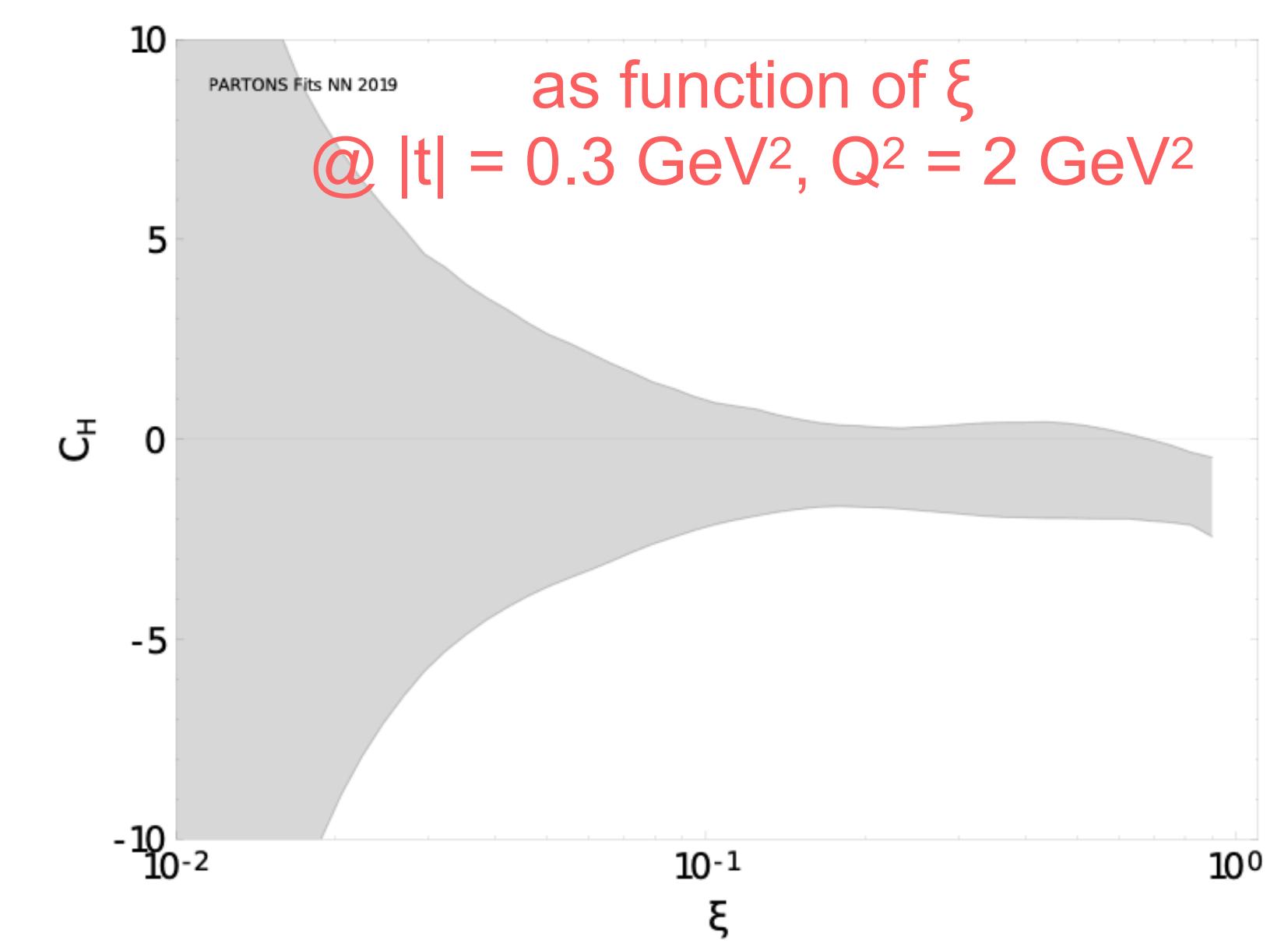
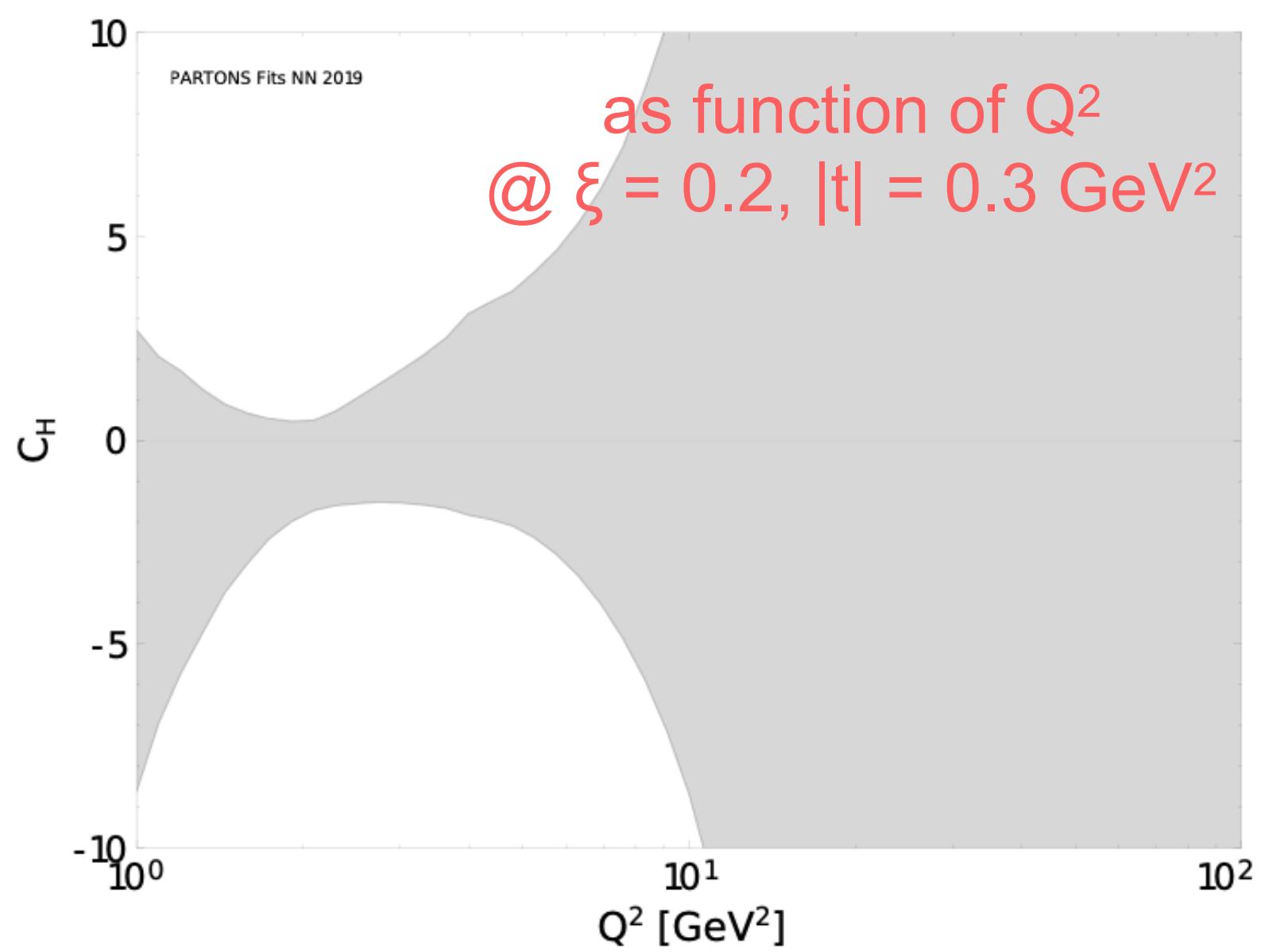
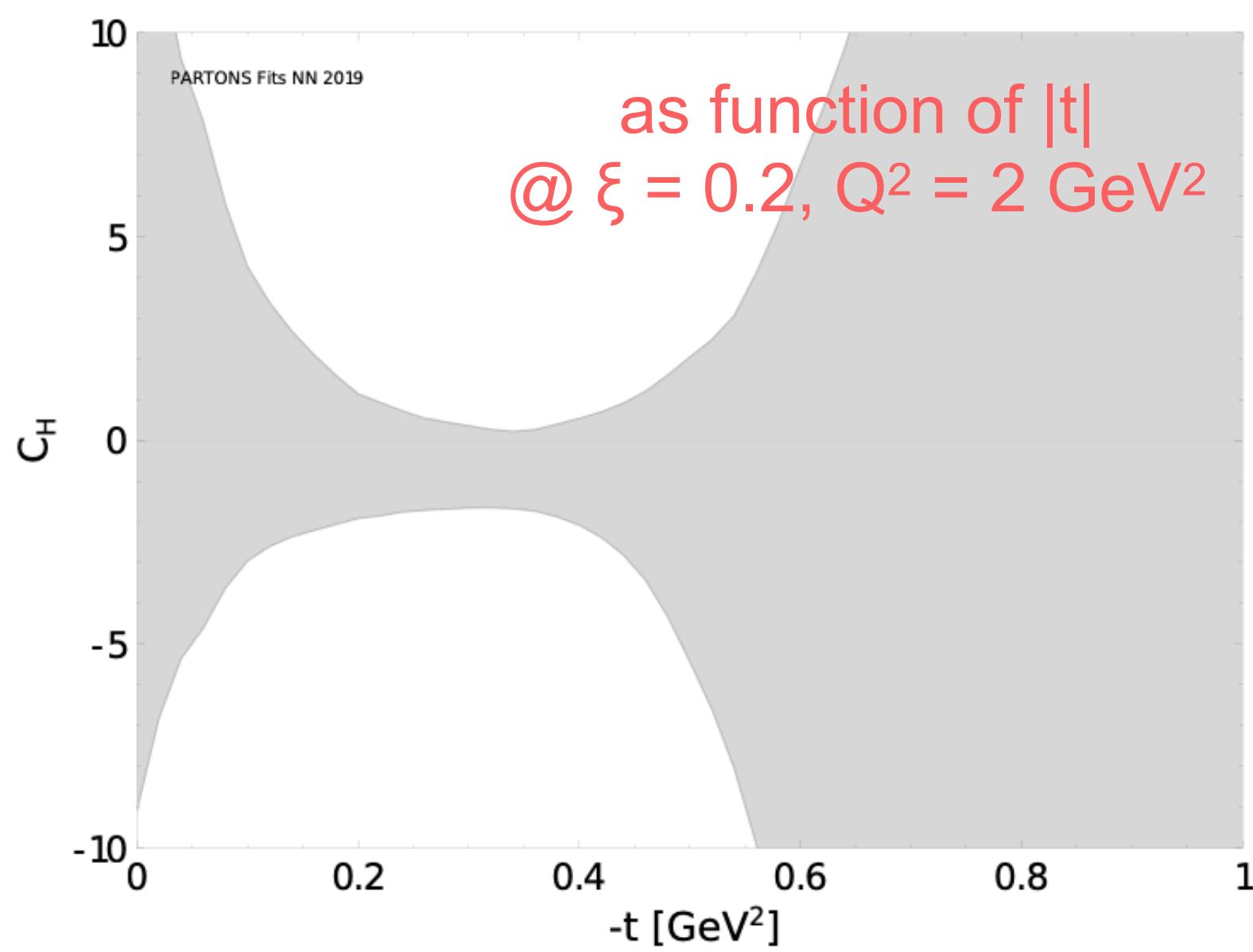
## Parametric Ansatz allows us to access nucleon tomography



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

## Subtraction constant extracted using dispersion relation

$$\mathcal{C}_H(t, Q^2) = \operatorname{Re} \mathcal{H}(\xi, t, Q^2) - \frac{1}{\pi} \int_0^1 d\xi' \operatorname{Im} \mathcal{H}(\xi', t, Q^2) \left( \frac{1}{\xi - \xi'} - \frac{1}{\xi + \xi'} \right)$$



PARTONS ANN

## Non-parametric Ansatz allows us to access EMT FF C

Dispersion relation:

$$\mathcal{C}_H(t, Q^2) = \operatorname{Re} \mathcal{H}(\xi, t, Q^2) - \frac{1}{\pi} \int_0^1 d\xi' \operatorname{Im} \mathcal{H}(\xi', t, Q^2) \left( \frac{1}{\xi - \xi'} - \frac{1}{\xi + \xi'} \right)$$

Relation between subtraction constant and D-term ( $z=x/\xi$ ):

$$\mathcal{C}_H(t, Q^2) \stackrel{\text{LO}}{=} 2 \sum_q e_q^2 \int_{-1}^1 dz \frac{D_{\text{term}}^q(z, t, \mu_F^2 \equiv Q^2)}{1 - z}$$

Decomposition into Gegenbauer polynomials:

$$D_{\text{term}}^q(z, t, \mu_F^2) = (1 - z^2) \sum_{\text{odd } n} d_n^q(t, \mu_F^2) C_n^{3/2}(z)$$

Finally:

$$\mathcal{C}_H(t, Q^2) \stackrel{\text{LO}}{=} 4 \sum_q e_q^2 \sum_{\text{odd } n} d_n^q(t, \mu_F^2 \equiv Q^2)$$

Connection to EMT FF:

$$d_1^q(t, \mu_F^2) = 5C_q(t, \mu_F^2)$$

## Master formula:

$$\text{Re} \mathcal{H}(\xi, t, Q^2) - \frac{1}{\pi} \oint_0^1 d\xi' \text{Im} \mathcal{H}(\xi, t, Q^2) \left( \frac{1}{\xi - \xi'} - \frac{1}{\xi + \xi'} \right)^{\text{LO}} = 4 \sum_q e_q^2 \sum_{\text{odd } n} d_n^q(t, \mu_F^2 \equiv Q^2)$$

## Extraction of subtraction constant from DVCS data requires:

- integral over  $\xi$  (alternatively:  $x_{Bj}$  or  $v$ ) between  $\epsilon$  and 1
  - good knowledge of both Re and Im parts of CFF  $H$
- $\epsilon = 10^{-6}$

## Model assumptions to extract EMT FF C from subtraction constant:

- truncation to d1

$$C_H(t, Q^2) = 4 \sum_q e_q^2 d_1^q(t, \mu_F^2 \equiv Q^2)$$

- symmetry of light quark contributions

$$d_1^u(t, \mu_F^2) = d_1^d(t, \mu_F^2) = d_1^s(t, \mu_F^2) \equiv d_1^{uds}(t, \mu_F^2)$$

- sensitivity to gluon contribution via evolution

$$d_1^G(t, \mu_{F,0}^2) = 0 \quad \mu_{F,0}^2 = 0.1 \text{ GeV}^2$$

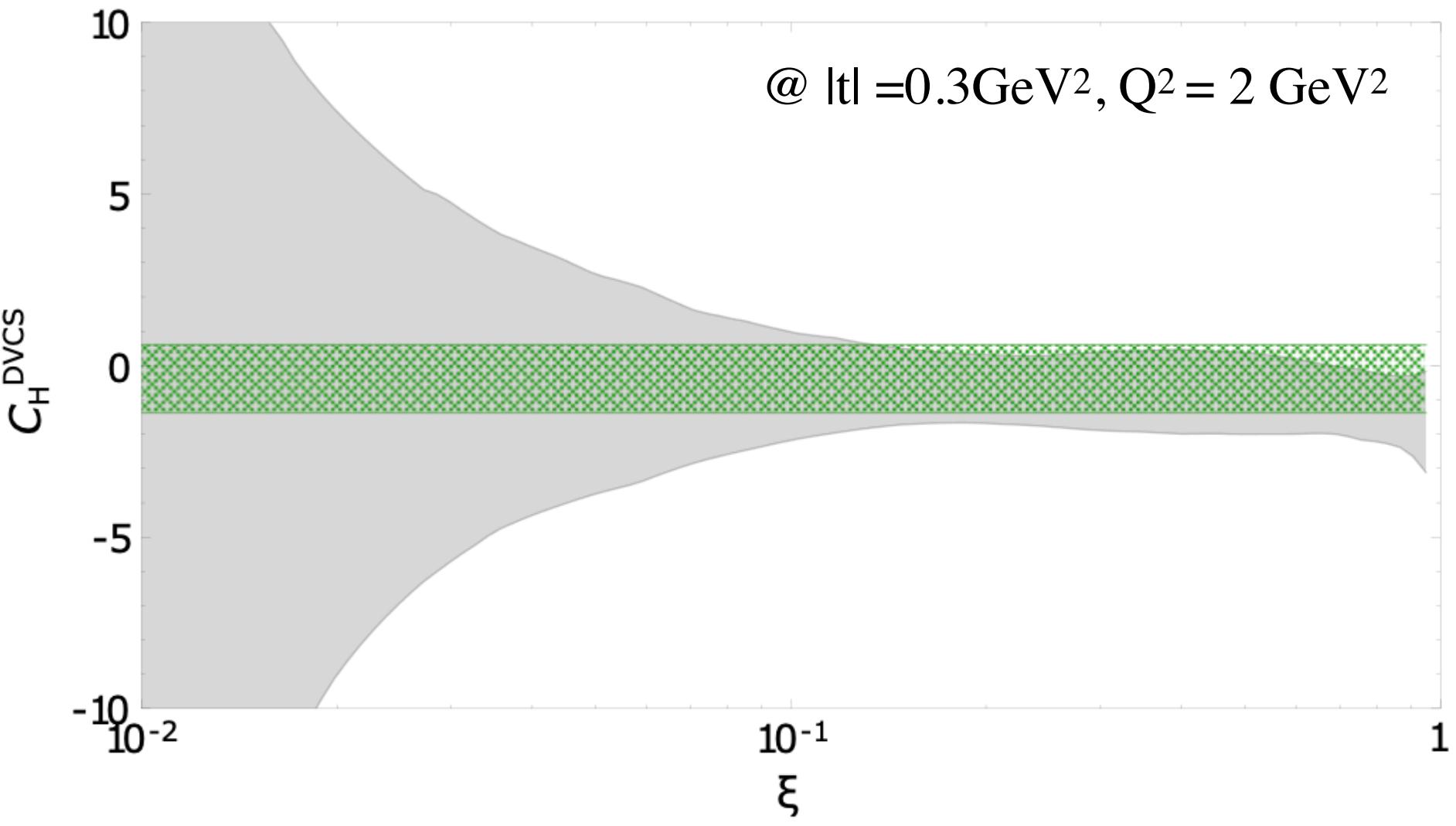
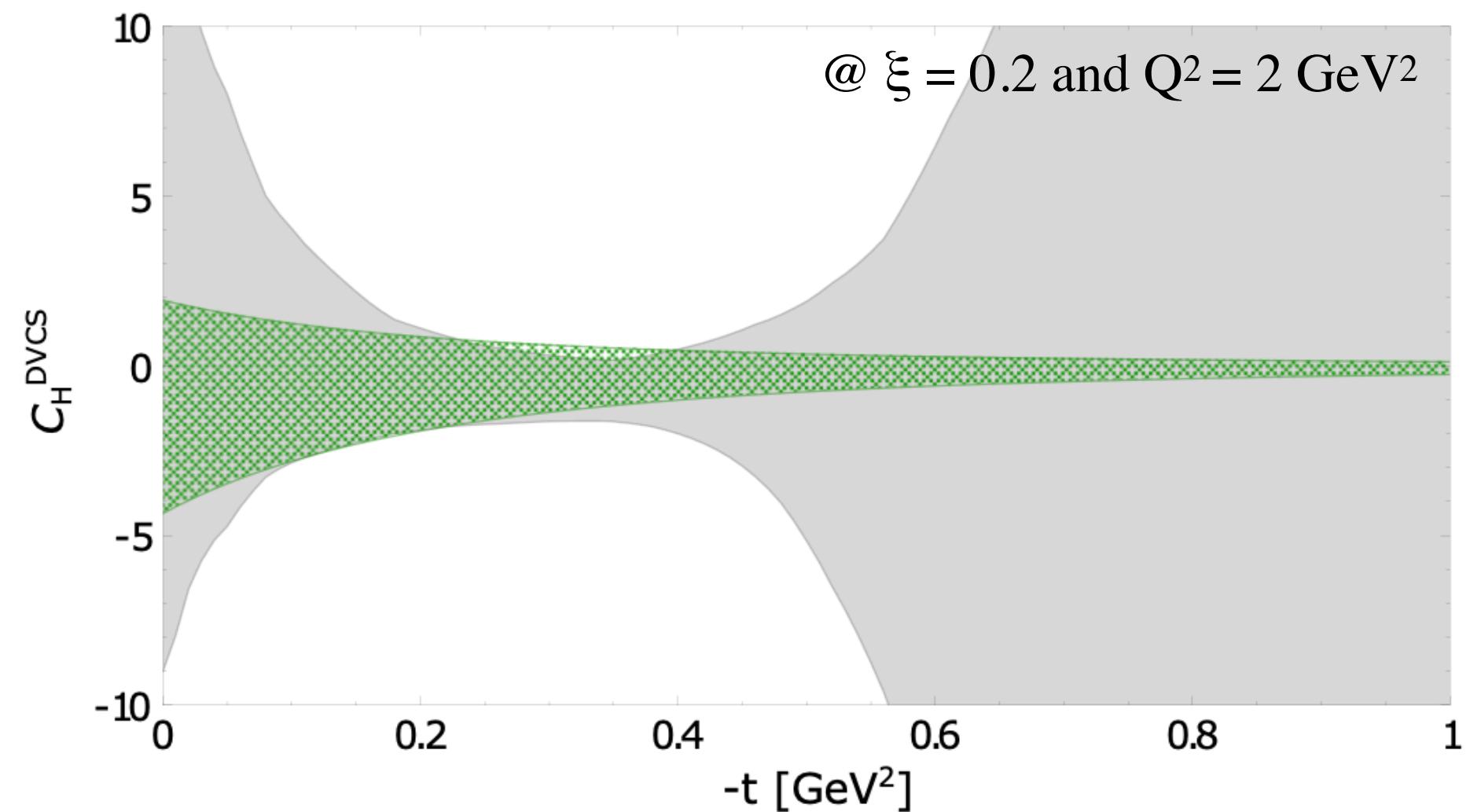
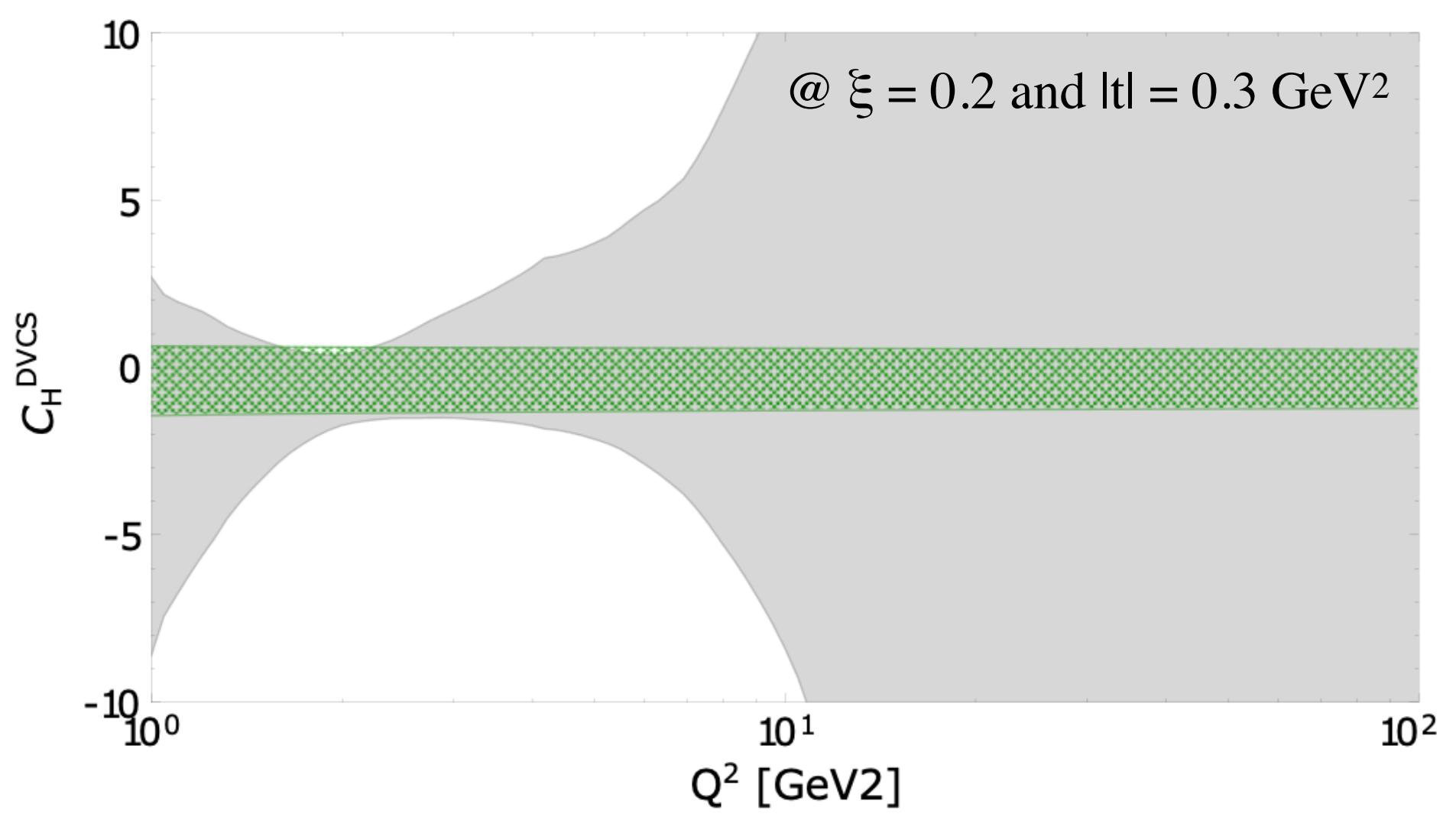
- tripole Ansatz for t-dependence

$$d_1^{uds}(t, \mu_F^2) = d_1^{uds}(\mu_F^2) \left( 1 - \frac{t}{\Lambda^2} \right)^{-\alpha} \quad \alpha = 3 \quad \Lambda = 0.8 \text{ GeV}$$

- Subtraction constant:

ANN analysis

Model dependent extraction

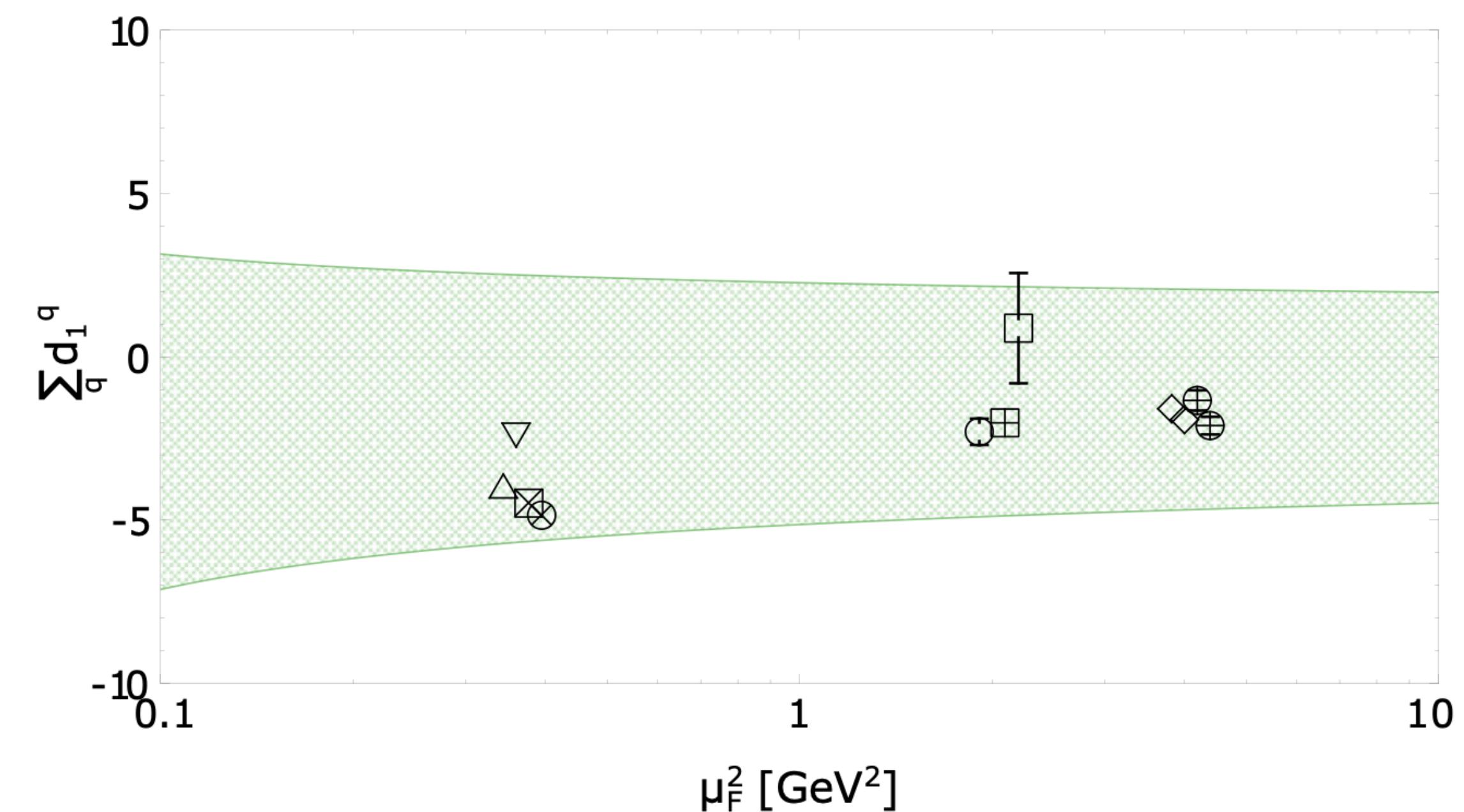


- Obtained values

Parameter	Value
$d_1^{uds}(\mu_F^2)$	$-0.5 \pm 1.2$
$d_1^c(\mu_F^2)$	$-0.0020 \pm 0.0053$
$d_1^g(\mu_F^2)$	$-0.6 \pm 1.6$

@ $\mu_F^2 = 2 \text{ GeV}^2$

- Comparison with other extractions and theory



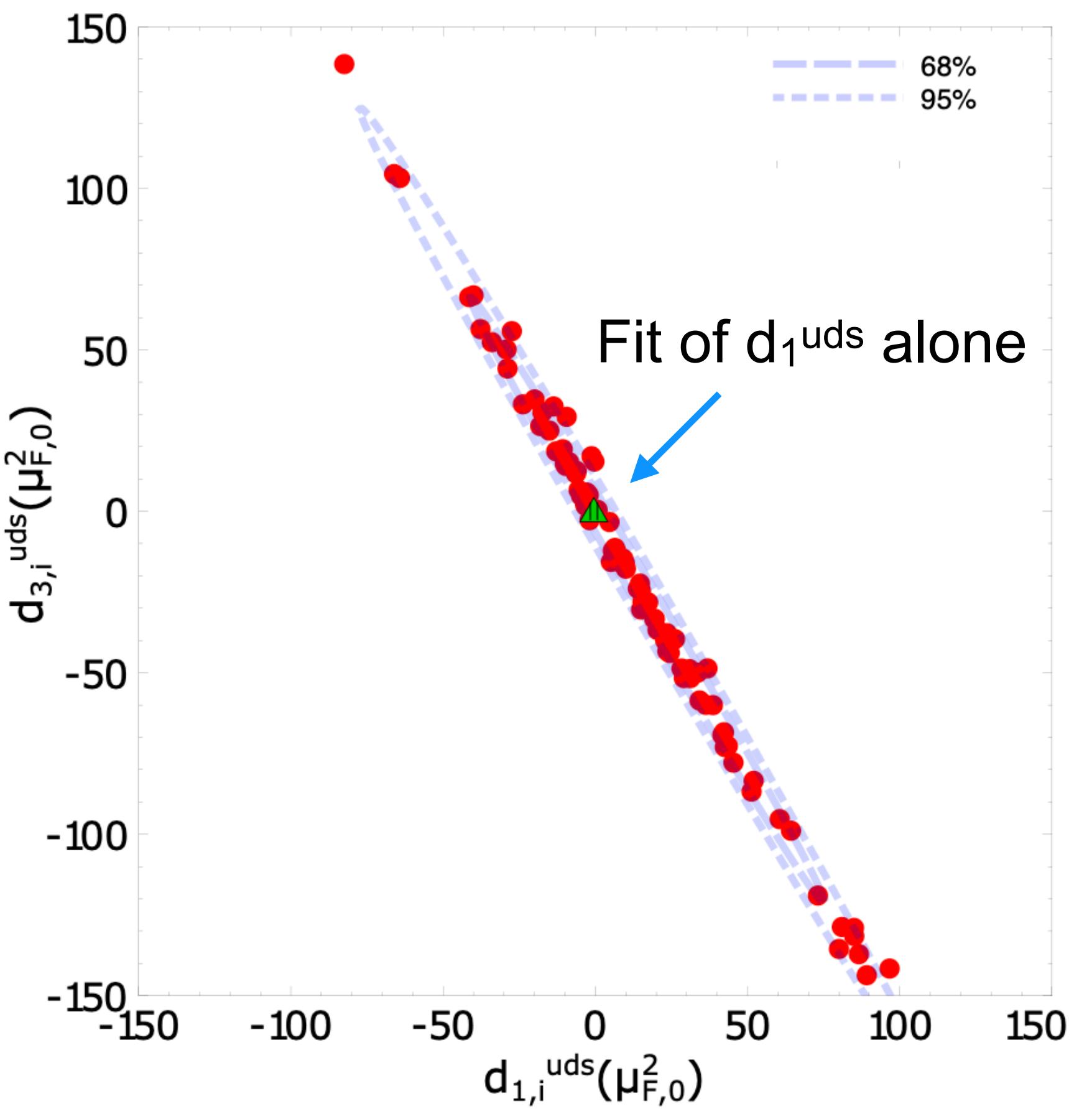
No.	Marker	$\sum_q d_1^q(\mu_F^2)$	$\mu_F^2$ in $\text{GeV}^2$	# of flavours	Type
1	○	$-2.30 \pm 0.16 \pm 0.37$	2.0	3	from experimental data
2	□	$0.88 \pm 1.69$	2.2	2	from experimental data
3	◊	$-1.59$	4	2	$t$ -channel saturated model
		$-1.92$	4	2	$t$ -channel saturated model
4	△	-4	0.36	3	$\chi$ QSM
5	▽	-2.35	0.36	2	$\chi$ QSM
6	⊗	-4.48	0.36	2	Skyrme model
7	田	-2.02	2	3	LFWF model
8	⊗	-4.85	0.36	2	$\chi$ QSM
9	⊕	$-1.34 \pm 0.31$	4	2	lattice QCD ( $\overline{\text{MS}}$ )
		$-2.11 \pm 0.27$	4	2	lattice QCD ( $\overline{\text{MS}}$ )

- Alternative fit with  $d_1$  and  $d_3$  extracted together

$$\begin{aligned} d_1^{uds}(\mu_F^2) & 11 \pm 25 \\ d_3^{uds}(\mu_F^2) & -11 \pm 26 \end{aligned}$$

@ $\mu_F^2 = 2 \text{ GeV}^2$

- Correlation between  $d_1^{uds}$  and  $d_3^{uds}$  →



### **3. Phenomenology at level of GPDs**

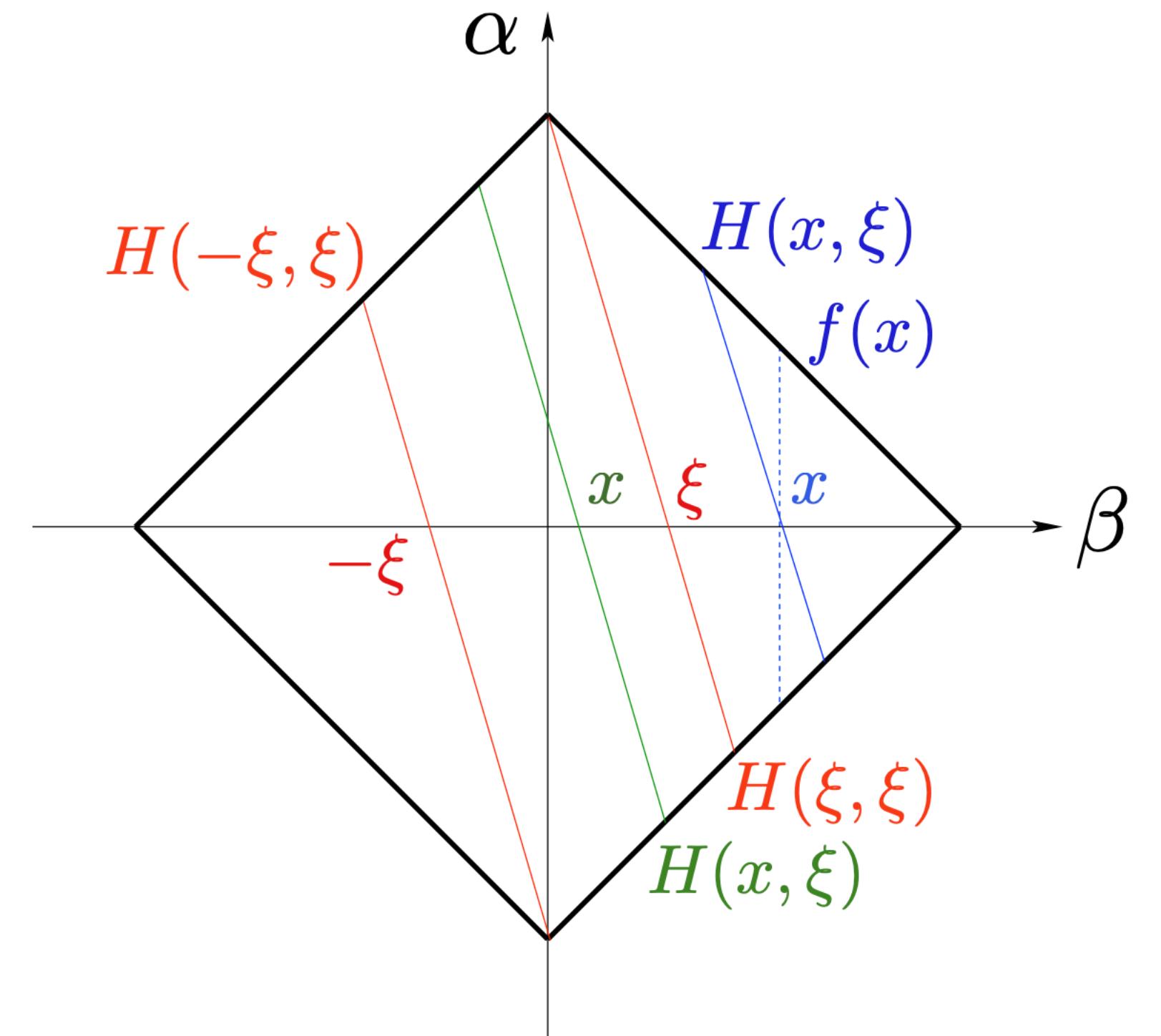
**Double distribution:**

$$H(x, \xi, t) = \int d\Omega F(\beta, \alpha, t)$$

**where:**

$$d\Omega = d\beta d\alpha \delta(x - \beta - \alpha\xi)$$

$$|\alpha| + |\beta| \leq 1$$



from PRD83, 076006, 2011

We also consider non-parametric GPD modelling in  $(x, \xi)$ -space, see our paper  
 The drawback of this modelling is that one can not keep PDF singularity for only  $x=0$  and  $\xi=0$

**Double distribution:**

$$(1 - x^2)F_C(\beta, \alpha) + (x^2 - \xi^2)F_S(\beta, \alpha) + \xi F_D(\beta, \alpha)$$

**Classical term:**

$$F_C(\beta, \alpha) = f(\beta)h_C(\beta, \alpha) \frac{1}{1 - \beta^2}$$

$$f(\beta) = \text{sgn}(\beta)q(|\beta|)$$

$$h_C(\beta, \alpha) = \frac{\text{ANN}_C(|\beta|, \alpha)}{\int_{-1+|\beta|}^{1-|\beta|} d\alpha \text{ANN}_C(|\beta|, \alpha)}$$

**Shadow term:**

$$F_S(\beta, \alpha) = f(\beta)h_S(\beta, \alpha)$$

$$f(\beta) = \text{sgn}(\beta)q(|\beta|)$$

$$h_S(\beta, \alpha)/N_S = \frac{\text{ANN}_S(|\beta|, \alpha)}{\int_{-1+|\beta|}^{1-|\beta|} d\alpha \text{ANN}_S(|\beta|, \alpha)} - \frac{\text{ANN}_{S'}(|\beta|, \alpha)}{\int_{-1+|\beta|}^{1-|\beta|} d\alpha \text{ANN}_{S'}(|\beta|, \alpha)}.$$

$$\text{ANN}_{S'}(|\beta|, \alpha) \equiv \text{ANN}_C(|\beta|, \alpha)$$

**D-term:**

$$F_D(\beta, \alpha) = \delta(\beta)D(\alpha)$$

$$D(\alpha) = (1 - \alpha^2) \sum_{\substack{i=1 \\ \text{odd}}} d_i C_i^{3/2}(\alpha)$$

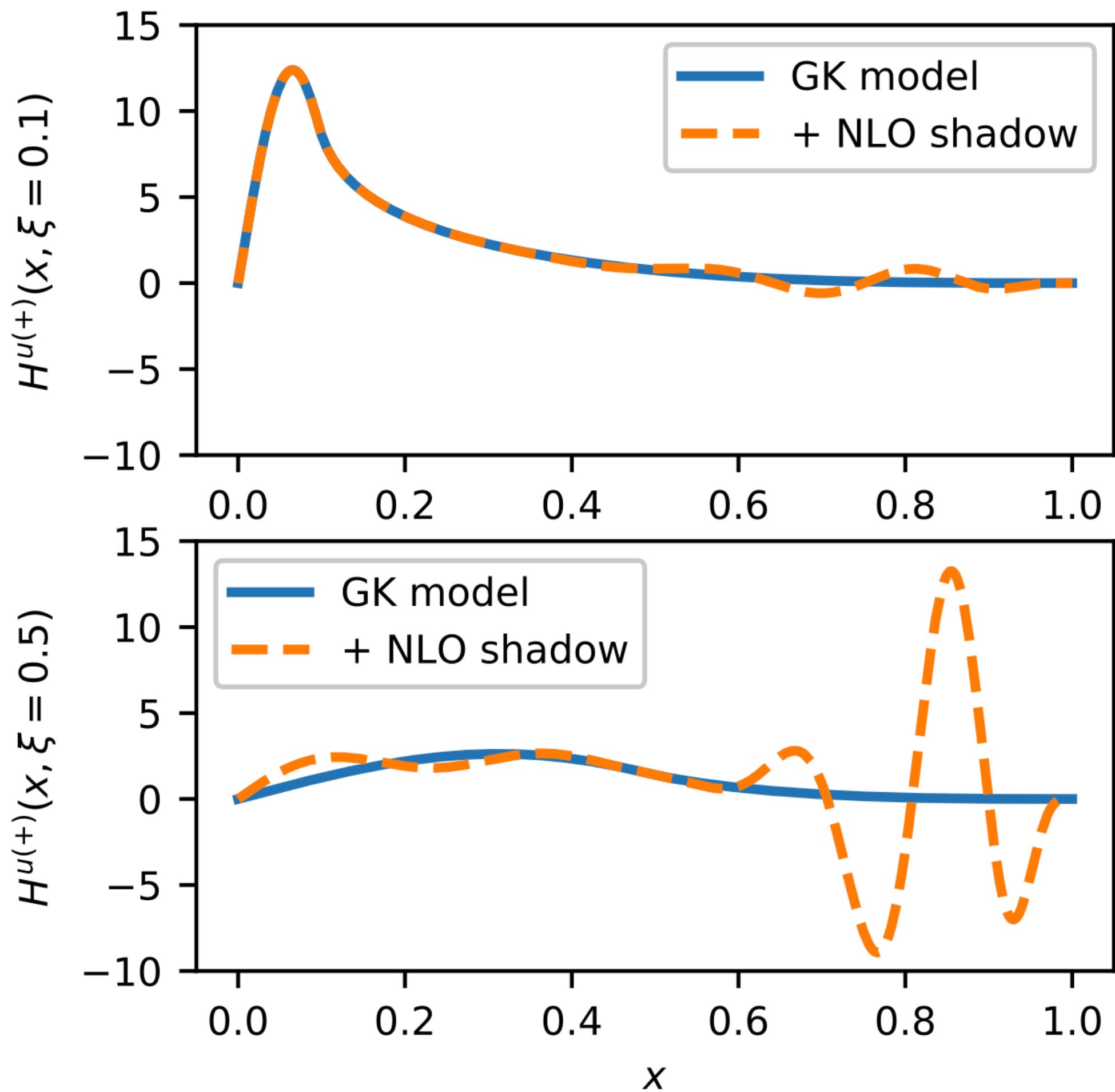
Shadow term is closely related to the so-called shadow GPDs

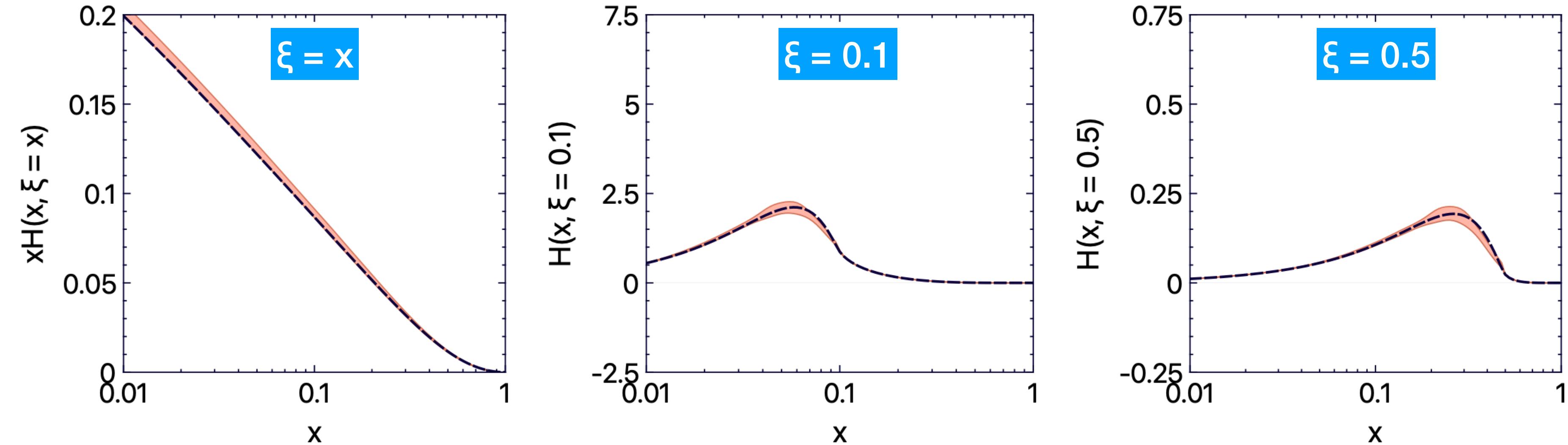
Shadow GPDs have considerable size and:

- at the initial scale do not contribute to both PDFs and CFFs
- at some other scale they contribute negligibly

making the deconvolution of CFFs ill-posed

We found such GPDs for both LO and NLO





## Conditions:

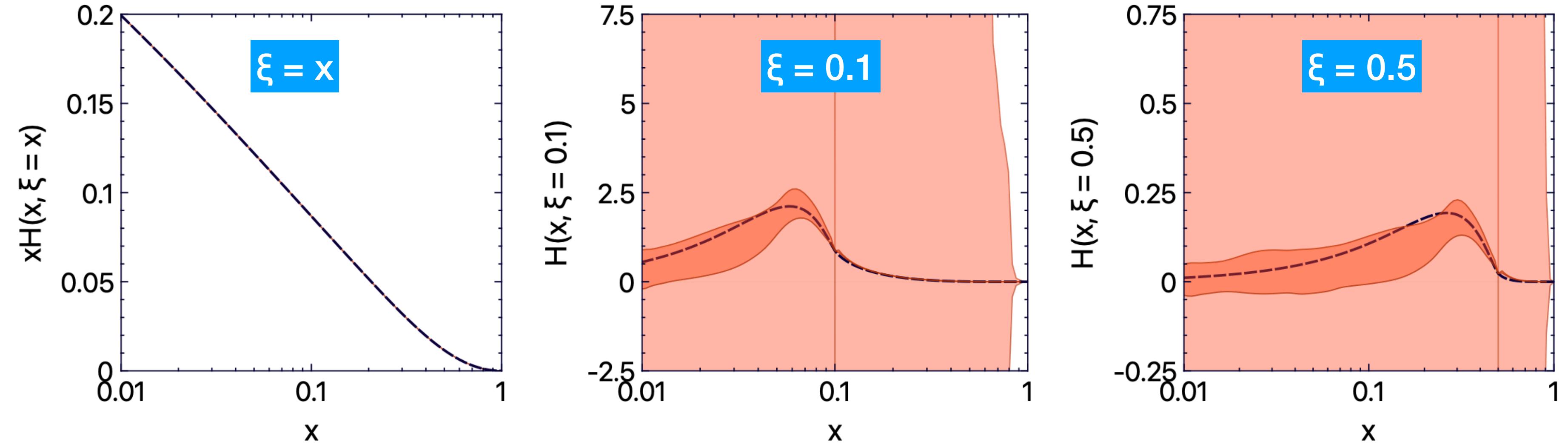
- Input: 400  $x \neq \xi$  points generated with GK model
- Positivity not forced

## Technical detail of the analysis:

- Minimisation with genetic algorithm
- Replication for estimation of model uncertainties
- “Local” detection of outliers
- Dropout algorithm for regularisation

----- GK

ANN model  
68% CL  
 $F_c + F_s + F_D$

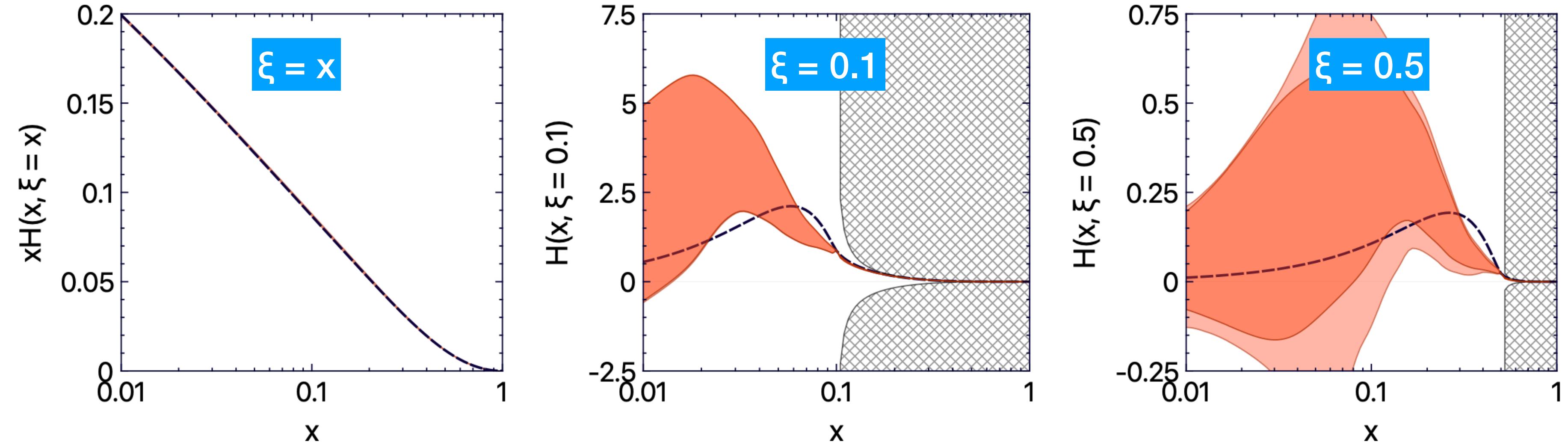


----- GK

## Conditions:

- Input: 200  $x = \xi$  points generated with GK model
- Positivity not forced

ANN model  
68% CL  
 $F_c$   
ANN model  
68% CL  
 $F_c + F_s$



----- GK

## Conditions:

- Input: 200  $x = \xi$  points generated with GK model
- Positivity **forced**

Excluded  
by positivity

ANN model  
68% CL  
 $F_c$

ANN model  
68% CL  
 $F_c + F_s$

## **4. New sources of GPD information**

## Relation between DVCS and TCS CFFs:

for more details see:

Mueller, Pire, Szymanowski, Wagner  
*Phys. Rev. D86, 031502 (2012)*

$$T \mathcal{H} \stackrel{\text{LO}}{=} s \mathcal{H}^*$$

$$T \widetilde{\mathcal{H}} \stackrel{\text{LO}}{=} -s \widetilde{\mathcal{H}}^*$$

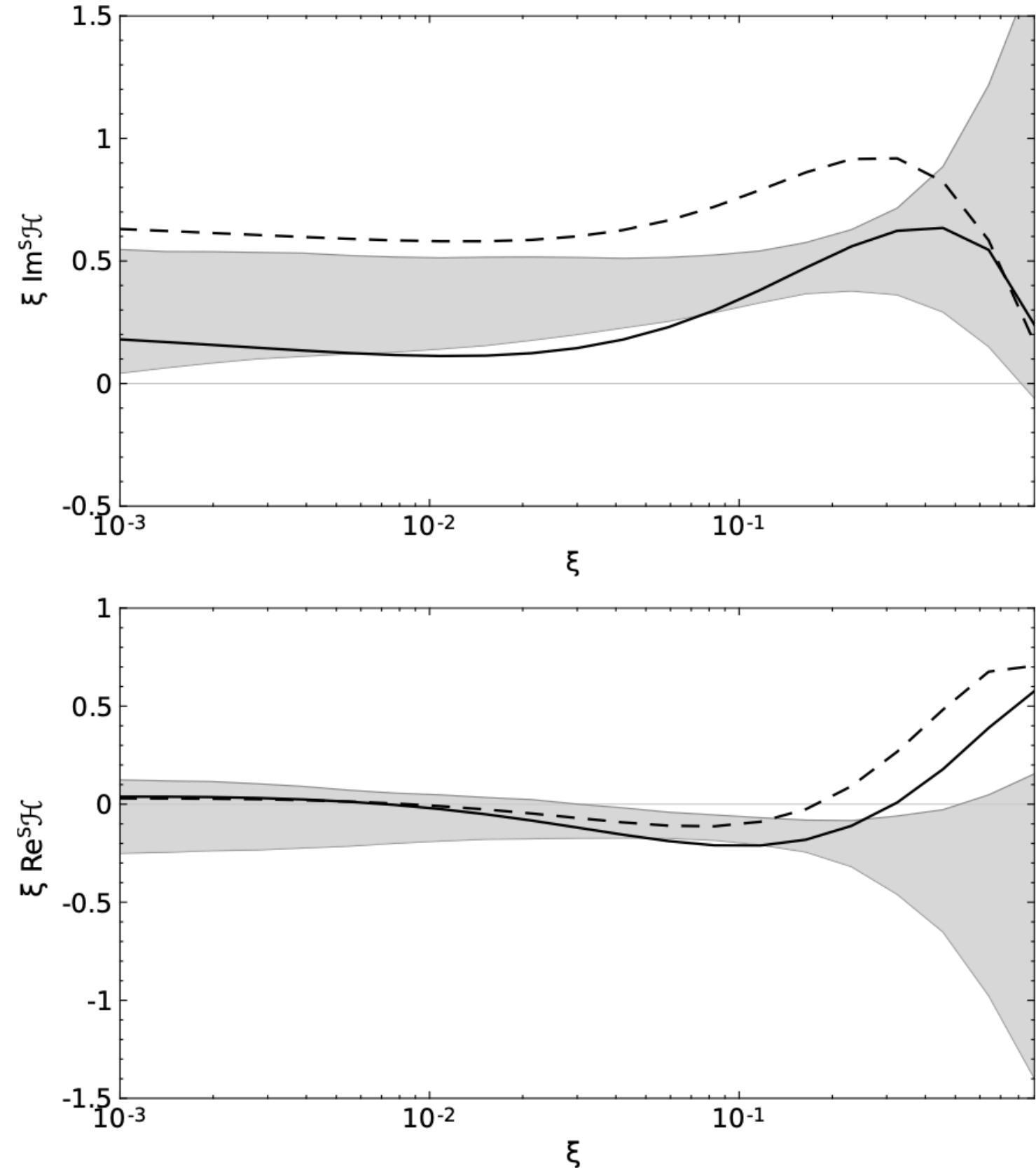
$$T \mathcal{H} \stackrel{\text{NLO}}{=} s \mathcal{H}^* - i\pi \mathcal{Q}^2 \frac{\partial}{\partial \mathcal{Q}^2} s \mathcal{H}^*$$

$$T \widetilde{\mathcal{H}} \stackrel{\text{NLO}}{=} -s \widetilde{\mathcal{H}}^* + i\pi \mathcal{Q}^2 \frac{\partial}{\partial \mathcal{Q}^2} s \widetilde{\mathcal{H}}^*.$$

## Combined study of DVCS and TCS:

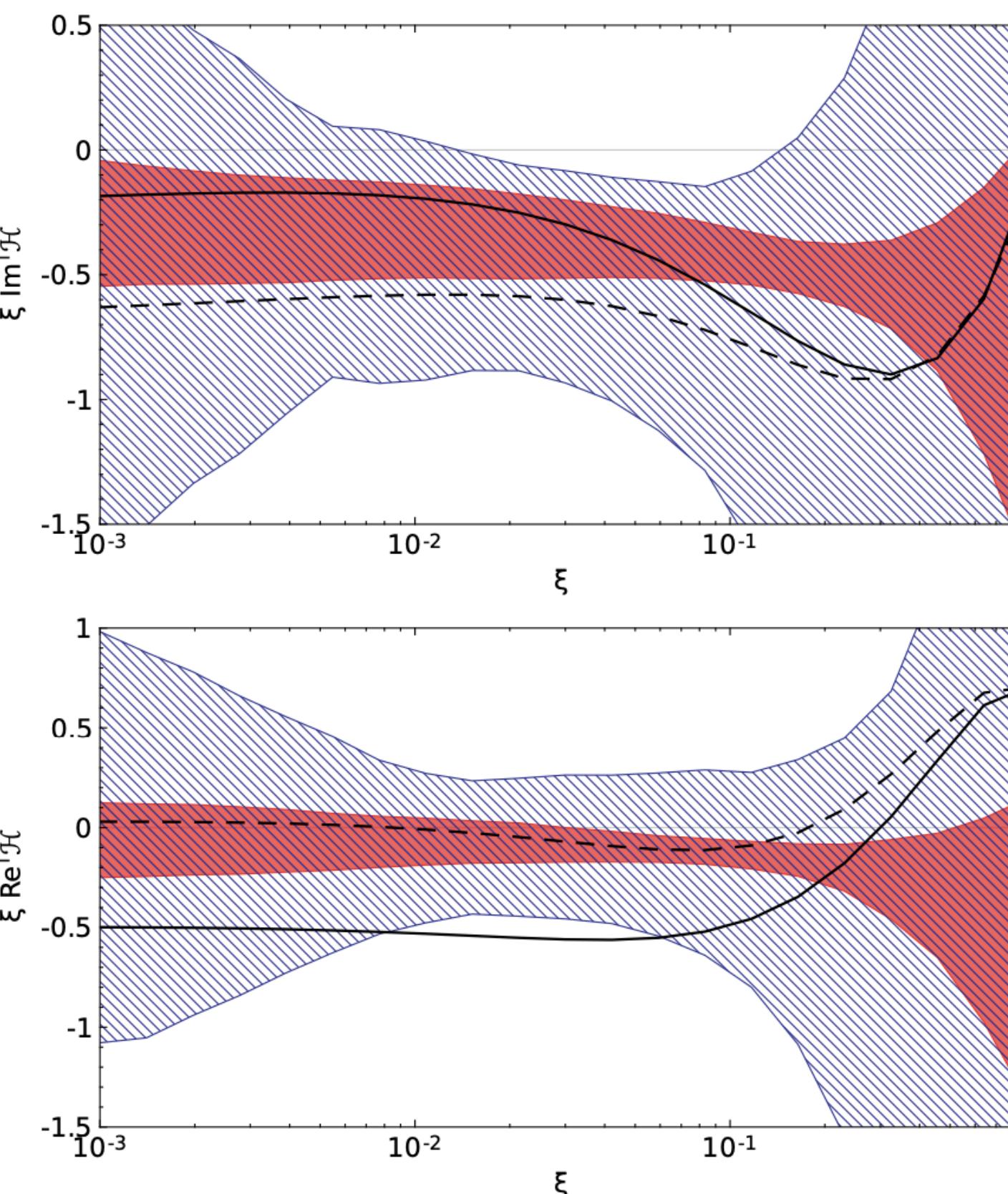
- source of GPD information
- useful to prove universality of GPDs
- allows to assess impact of NLO corrections
- constrain Q2-dep. of CFFs

DVCS CFF (Non-parametric):



DVCS

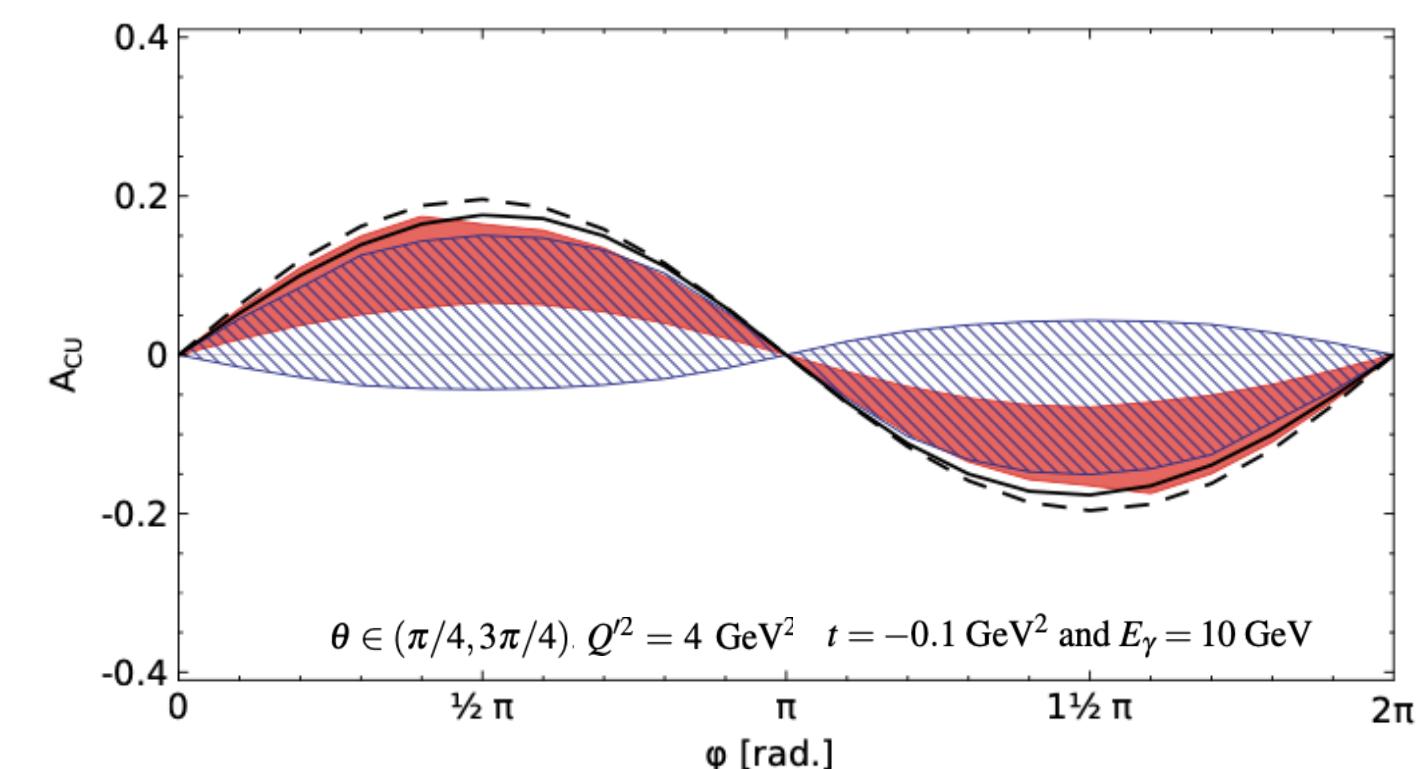
TCS CFF:



TCS from DVCS (LO)

TCS from DVCS (NLO)

TCS circular beam asymmetry:



..... GK model (LO)

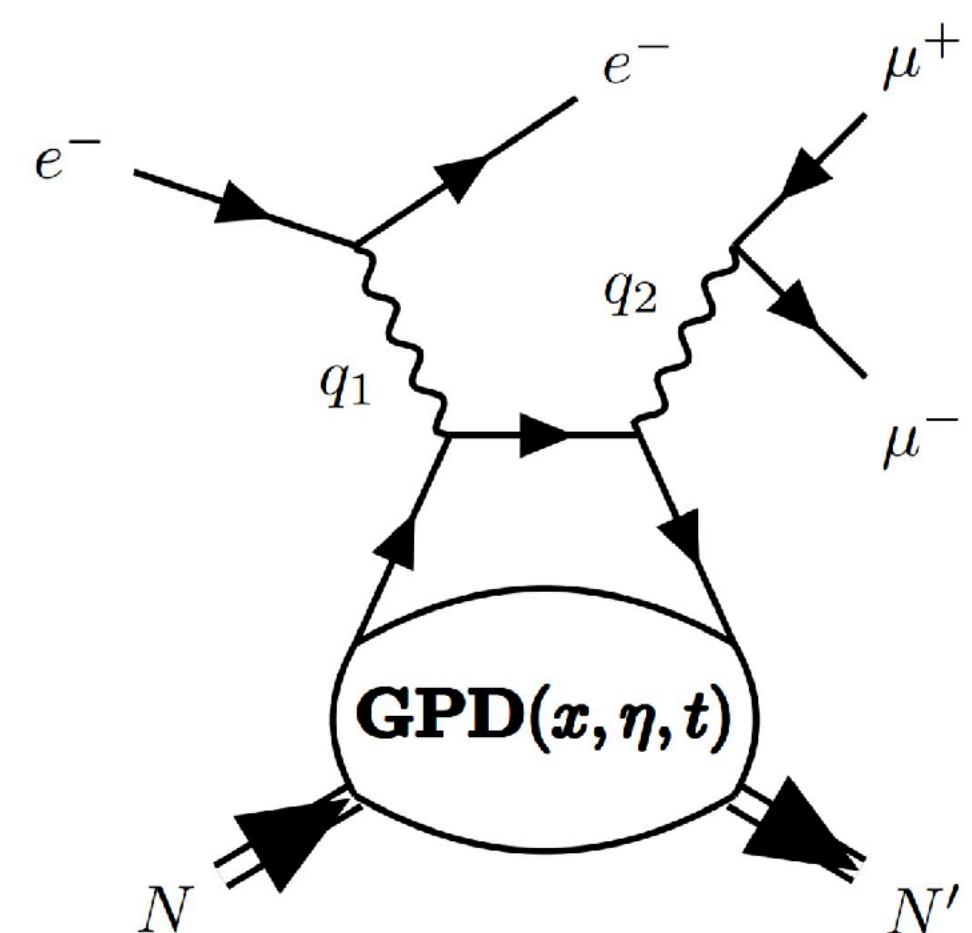
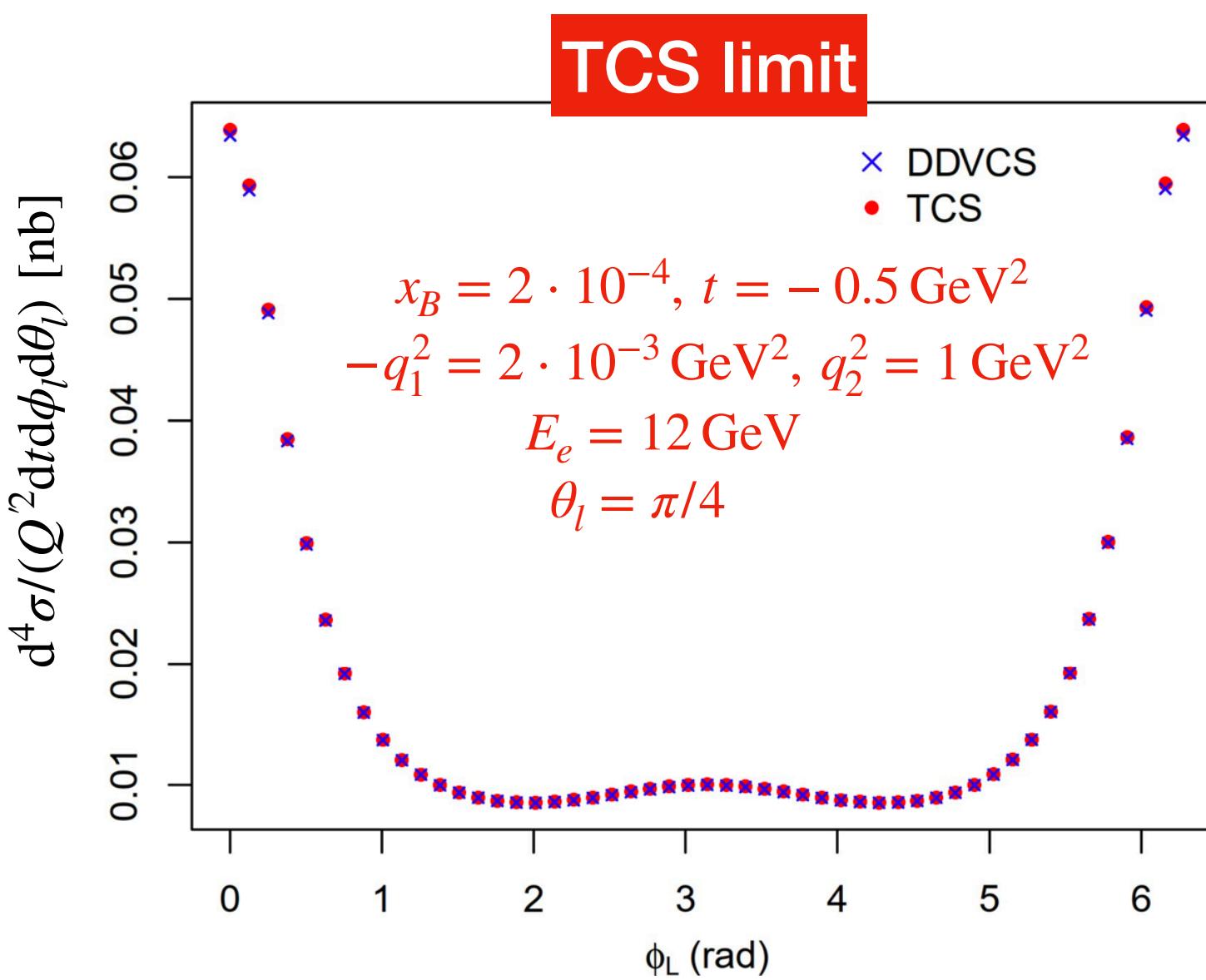
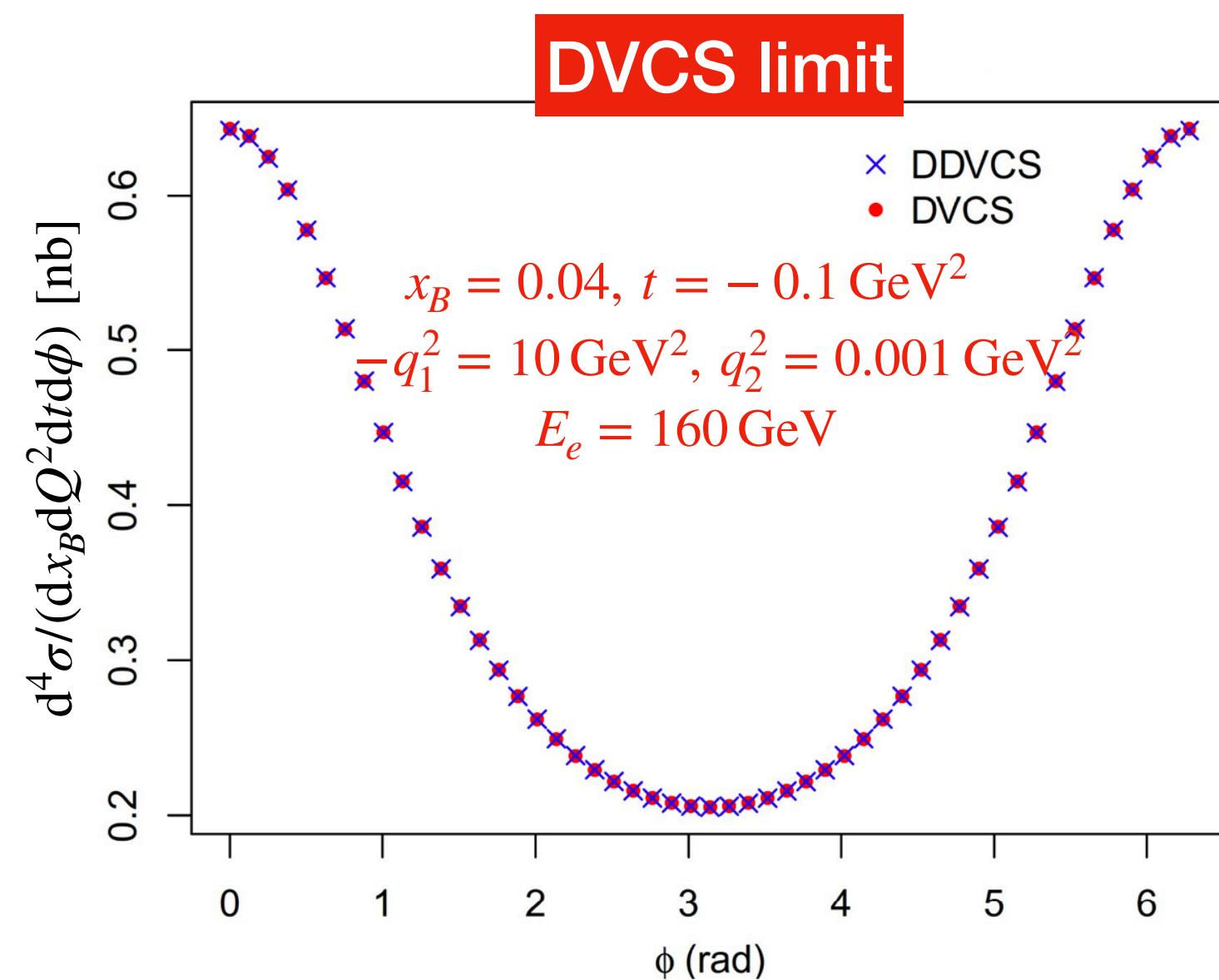
— GK model (NLO)

- The process allows to probe GPDs outside  $x=\xi$  line, but is much more challenging experimentally

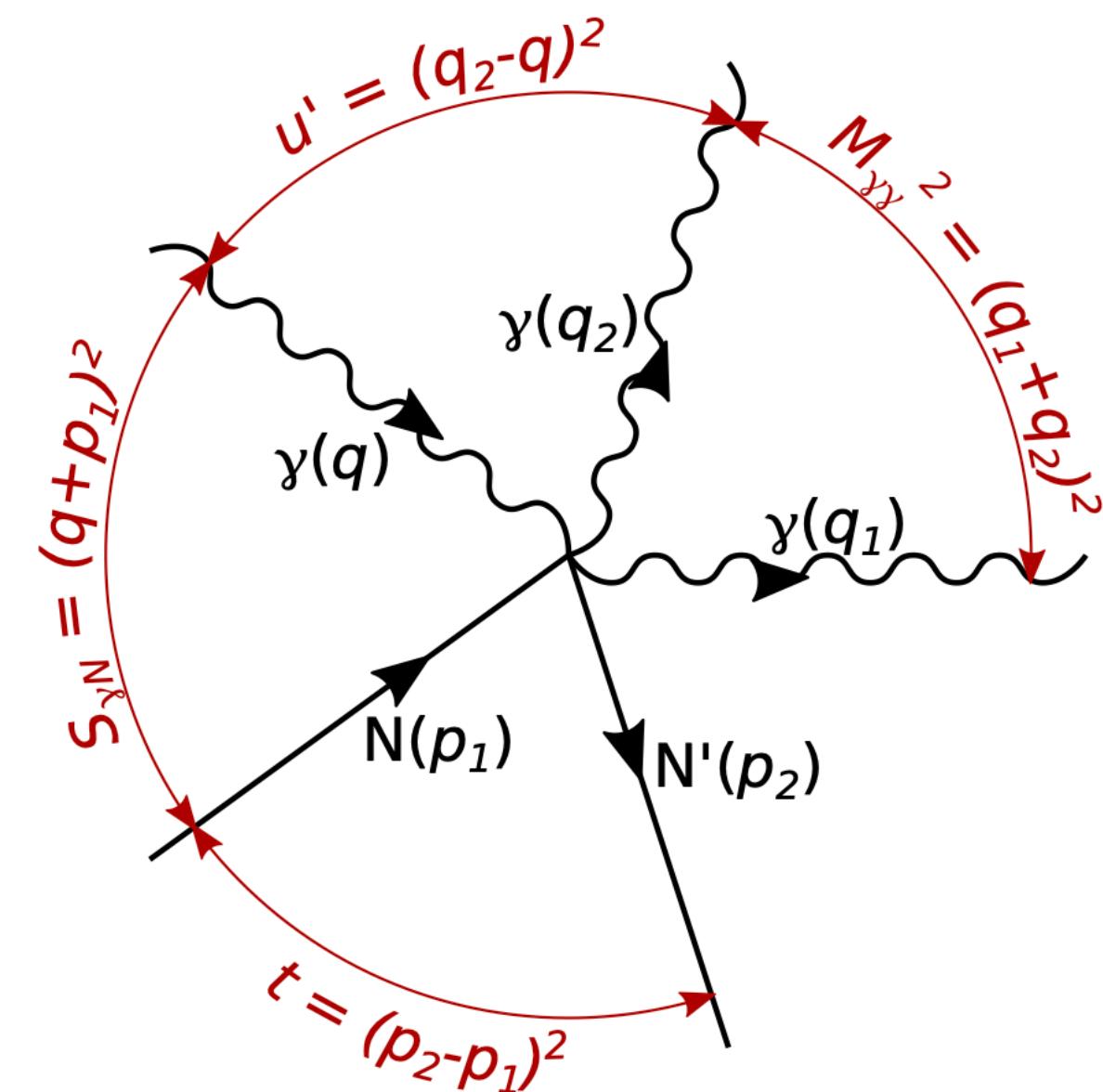
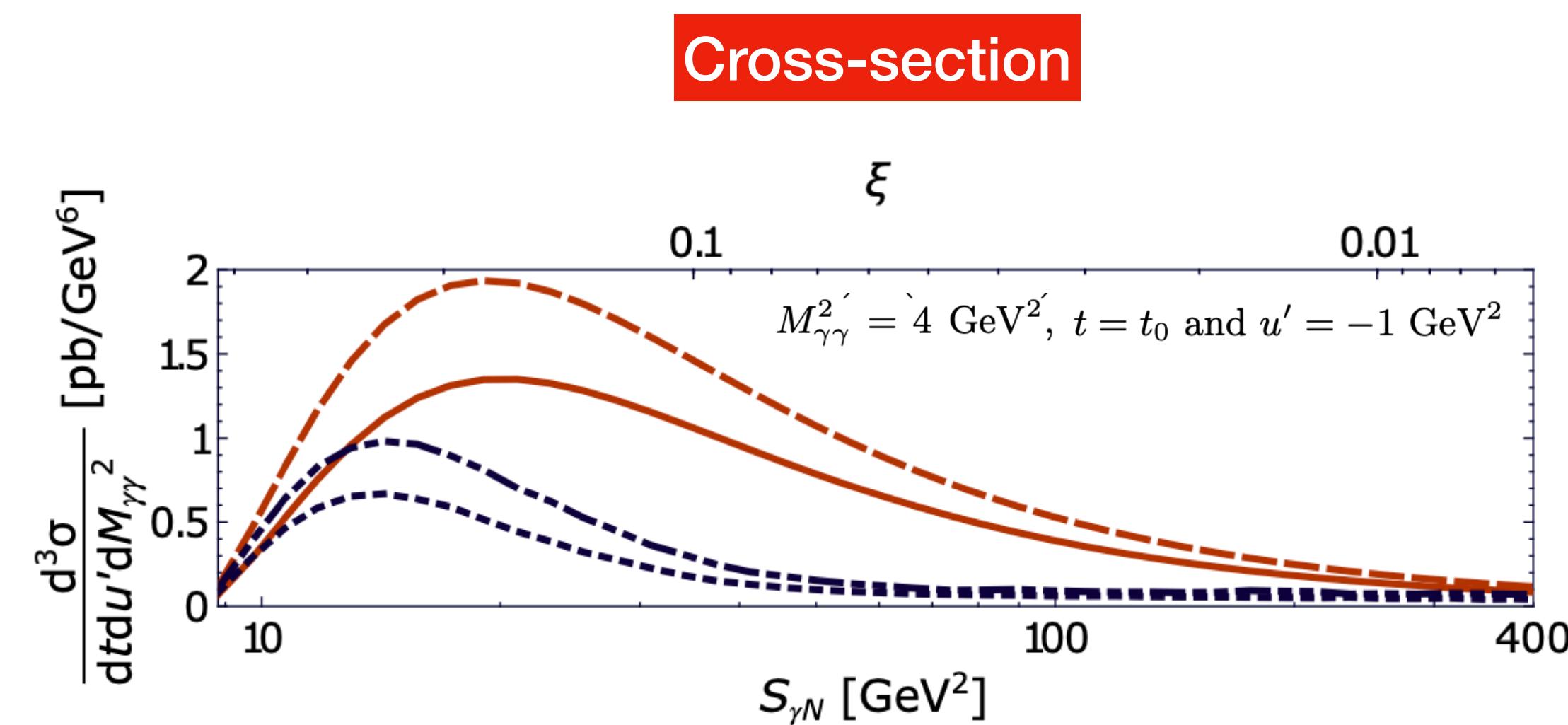
$$\mathcal{A}_{\text{DDVCS}} \stackrel{\text{LO}}{\sim} \int_{-1}^1 dx \frac{1}{x - \xi + i0} \text{GPD}(x, \eta, t)$$

- We are revisiting DDVCS for phenomenological studies, i.e. we reevaluate DDVCS and related BH amplitudes using Kleiss-Stirling technique
- We plan to release obtained formulae in PARTONS and EpIC MC generator

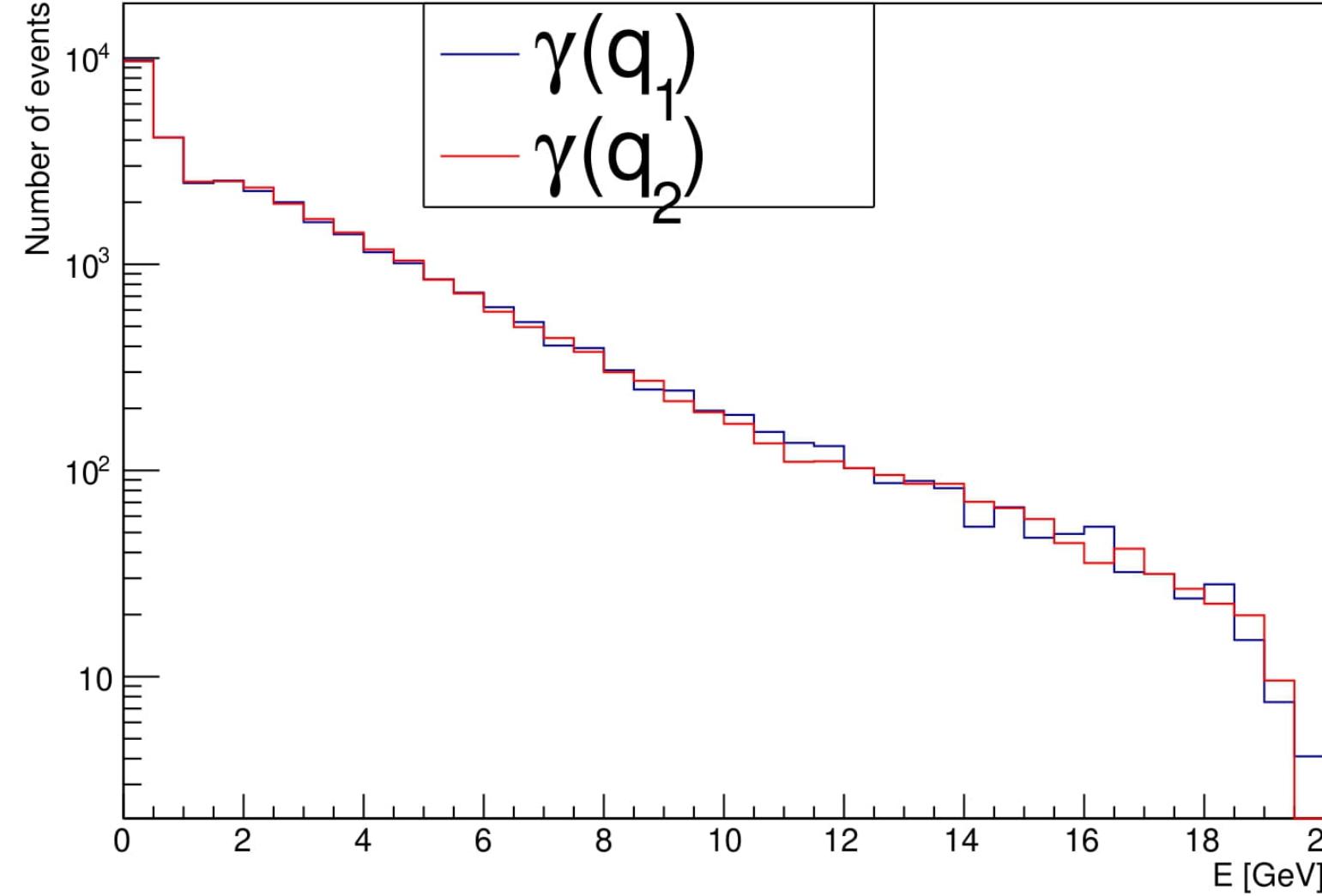
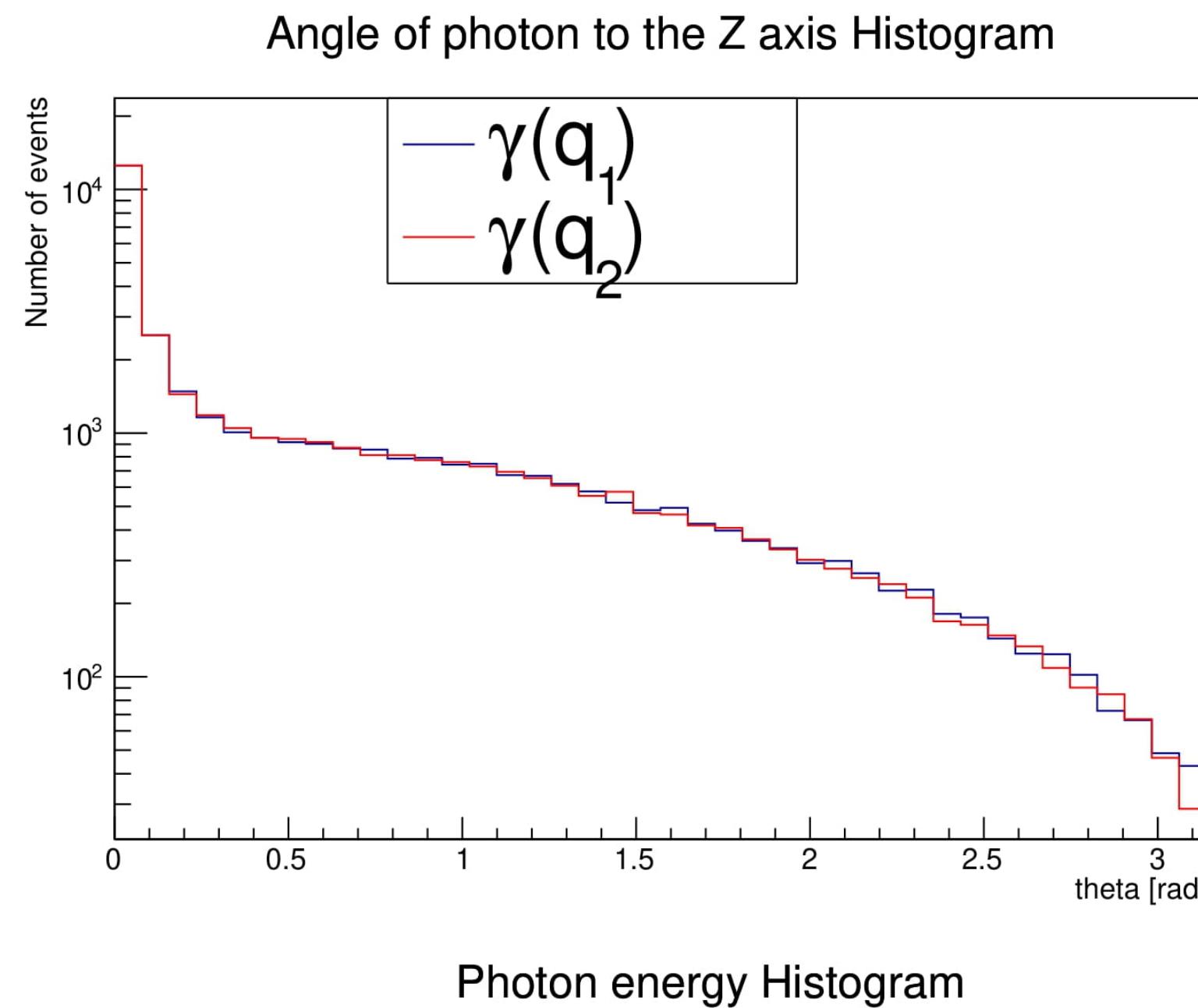
**Preliminary results:**  
BH cross-section in DVCS and TCS limits



- Process probes C-odd GPDs
- No contribution of D-term
- No non-perturbative ingredients other than GPDs
- Both LO and NLO description available
- Gluons do not contribute also at NLO
- Description already available in PARTONS (not released yet), soon will be available in EpIC



	GK	MMS
LO	—	—
NLO	.....	— - -



- The process implemented in EpIC MC generator with equivalent-photon approximation

$$\frac{d^6\sigma}{dQ^2 dy dt du' dM_{\gamma\gamma}^2 d\phi} = \Gamma(y, Q^2) \times \frac{d^4\sigma_{2\gamma}}{dt du' dM_{\gamma\gamma}^2 d\phi}$$

- Condition used in generation of events

$$E = 20 \text{ GeV}$$

$$0 < -t < 1 \text{ GeV}^2$$

$$0 < -u < 6 \text{ GeV}^2$$

$$1 \text{ GeV}^2 < M_{\gamma\gamma}^2 < 5 \text{ GeV}^2$$

$$0 < \phi < 2\pi$$

$$0 < y < 1$$

$$0 < Q^2 < 0.01 \text{ GeV}^2$$

- Event counts are scaled to  $10 \text{ fb}^{-1}$

## **4. Tools**

- PARTONS - open-source framework to study GPDs  
→ <http://partons.cea.fr>
- Come with number of available physics developments implemented
- Written in C++, also available via virtual machines (VirtualBox) and containers (Docker)
- Addition of new developments as easy as possible
- Developed to support effort of GPD community,  
can be used by both theorists and experimentalists
- v3 version of PARTONS is now available!





- Novel MC generator called EpIC released  
→ <https://pawelsznajder.github.io/epic>
- EpIC is based on PARTONS
- EpIC is characterised by:
  - flexible architecture that utilises a modular programming paradigm
  - a variety of modelling options, including radiative corrections
  - multichannel capability (now: DVCS, TCS, DV $\pi^0$ P, diphoton; coming soon: DDVCS, J/ $\psi$ )
- This is the new tool to be used in the precision era commenced by the new generation of experiments

- Substantial progress in:
  - understanding of fundamental problems, like deconvolution of CFFs, and analysis methods  
→ important for extraction of GPDs
  - modelling of GPD, fulfilling all theory-driven constraints (including positivity)  
→ subject not touched enough in the current literature  
→ developed in mind for easy inclusion of latticeQCD data
  - addressing the long-standing problem of model dependency of GPDs  
→ nontrivial and timely analysis
  - description of exclusive processes  
→ new sources of GPD information
  - delivering open-source tools for the community  
→ to support both experimentalists and theoreticians

This progress is important for the precision era of GPD extraction allowed by the new generation of experiments