

TMDs: Towards a Synergy between Lattice QCD and Global Analyses
June 21, 2023

Global extraction of Transverse Momentum Distributions

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on behalf of the MAP Collaboration



Extraction of unpolarized quark TMDs

What's new?

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📌 **GLOBAL ANALYSIS** of Drell-Yan and Semi-Inclusive DIS data sets \rightarrow **2031** data points

📌 Perturbative accuracy: **N^3LL^-**

📌 **Normalization** of SIDIS multiplicities beyond NLL

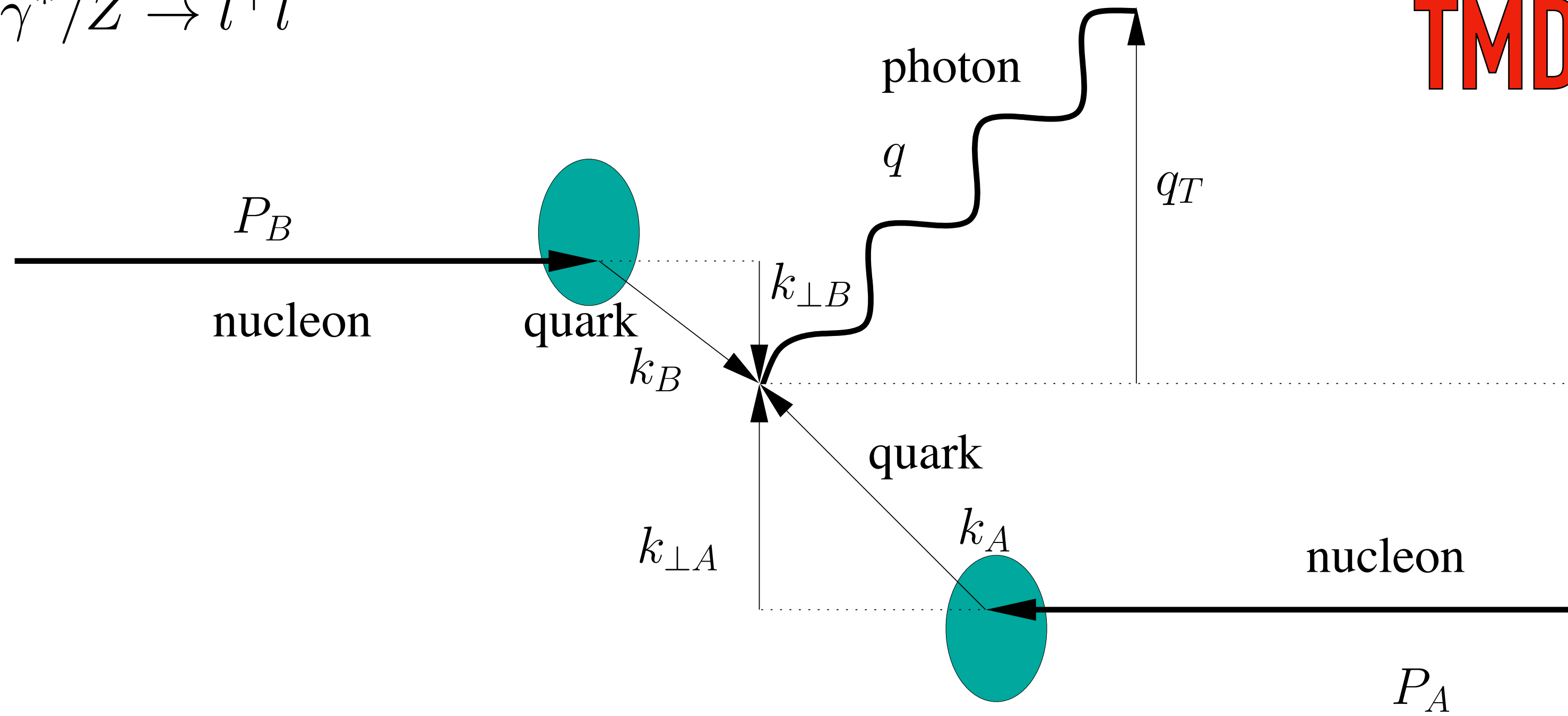
📌 Extremely good description: **$\chi^2/N_{\text{data}} = 1.06$**

Drell-Yan

$$N(P_A) + N(P_B) \rightarrow \gamma^*/Z \rightarrow l^+l^-$$

for $q_T \ll Q$

TMD factorization



$$F_{UU}^1(x_A, x_B, \mathbf{q}_T^2, Q^2)$$

$$= \sum_a \mathcal{H}_{UU}^{1a}(Q^2, \mu^2) \int d^2\mathbf{k}_{\perp A} d^2\mathbf{k}_{\perp B} f_1^a(x_A, \mathbf{k}_{\perp A}^2; \mu^2) f_1^{\bar{a}}(x_B, \mathbf{k}_{\perp B}^2; \mu^2) \delta^{(2)}(\mathbf{k}_{\perp A} - \mathbf{q}_T + \mathbf{k}_{\perp B})$$

$$+ Y_{UU}^1(Q^2, \mathbf{q}_T^2) + \mathcal{O}(M^2/Q^2)$$

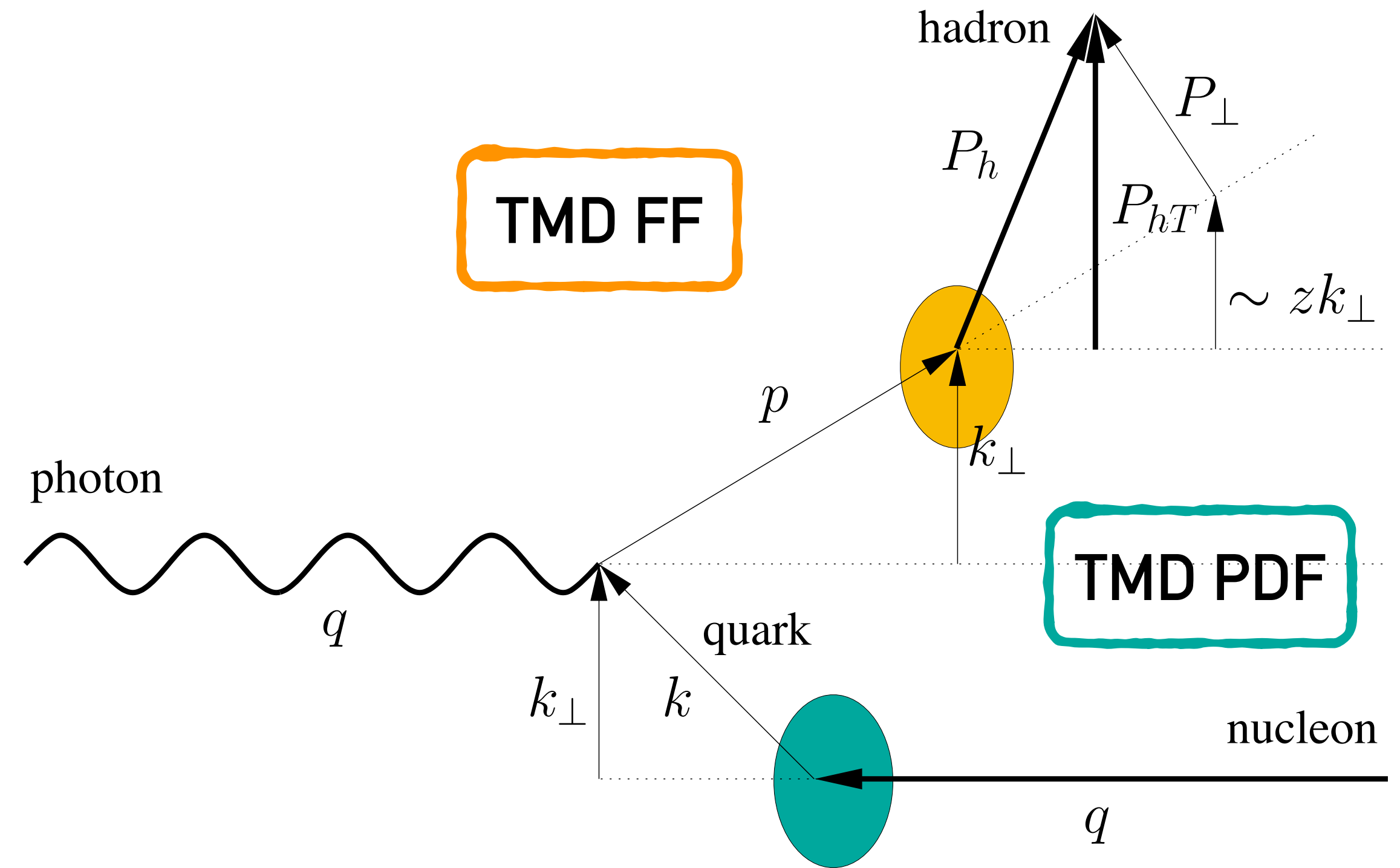
SIDIS

Semi-Inclusive Deep Inelastic Scattering

$$\ell(l) + N(p) \rightarrow \ell(l') + h(P_h) + X$$

TMD factorization

$$P_{hT}^2 \ll Q^2$$



$$F_{UU,T}(x \cdot z; \mu_F, \mathbf{P}_{hT}^2, Q^2) = x \sum_a H_{UU,T}^a(Q^2, \mu^2) \int d^2\mathbf{k}_\perp d^2\mathbf{P}_\perp f_1^a(x, \mathbf{k}_\perp^2; \mu^2) D_1^{a \rightarrow h}(z, \mathbf{P}_\perp^2; \mu^2) \delta^{(2)}(z\mathbf{k}_\perp - \mathbf{P}_{hT} + \mathbf{P}_\perp) + Y_{UU,T}(Q^2, \mathbf{P}_{hT}^2) + \mathcal{O}(M^2/Q^2)$$

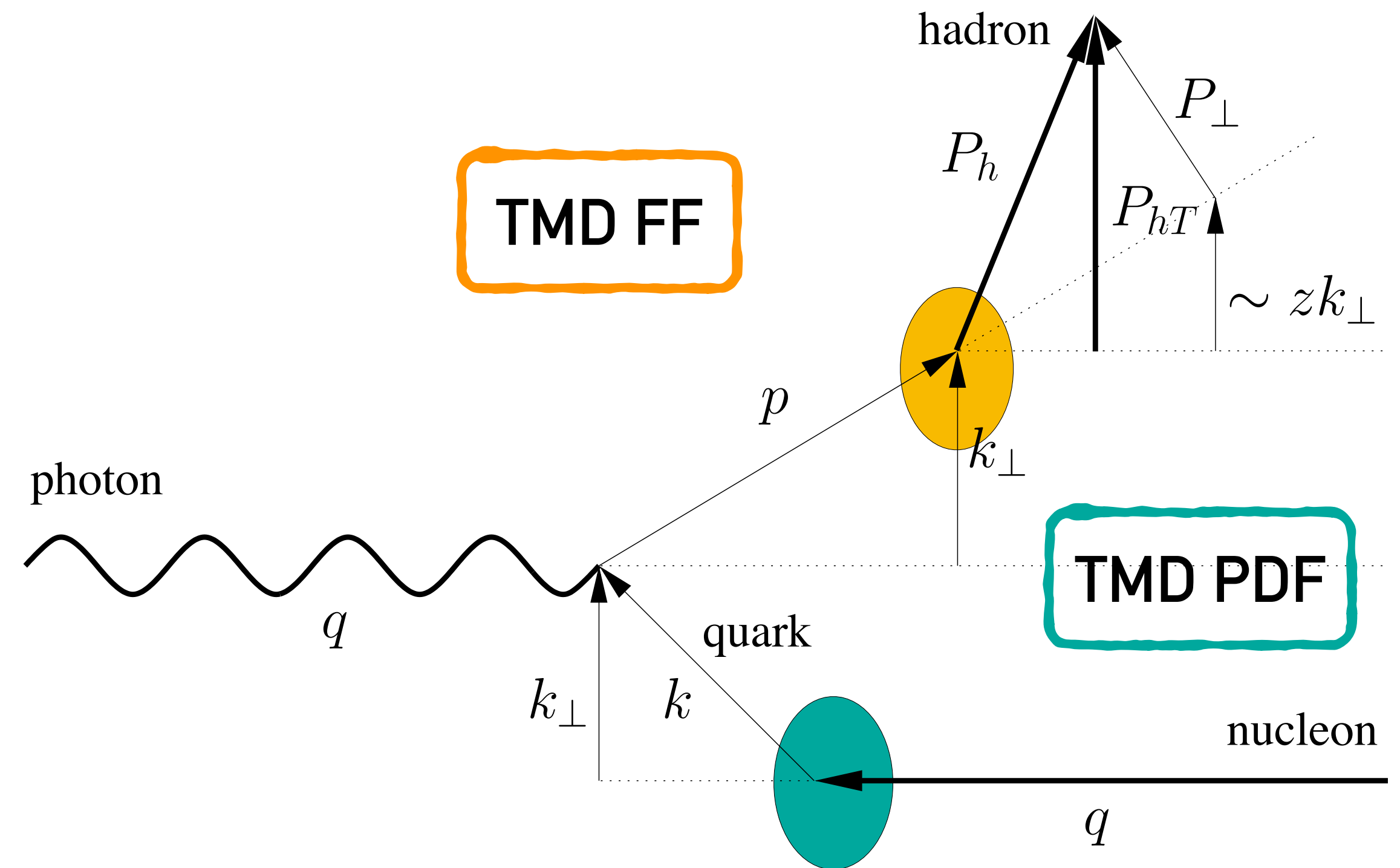
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Y term

W term

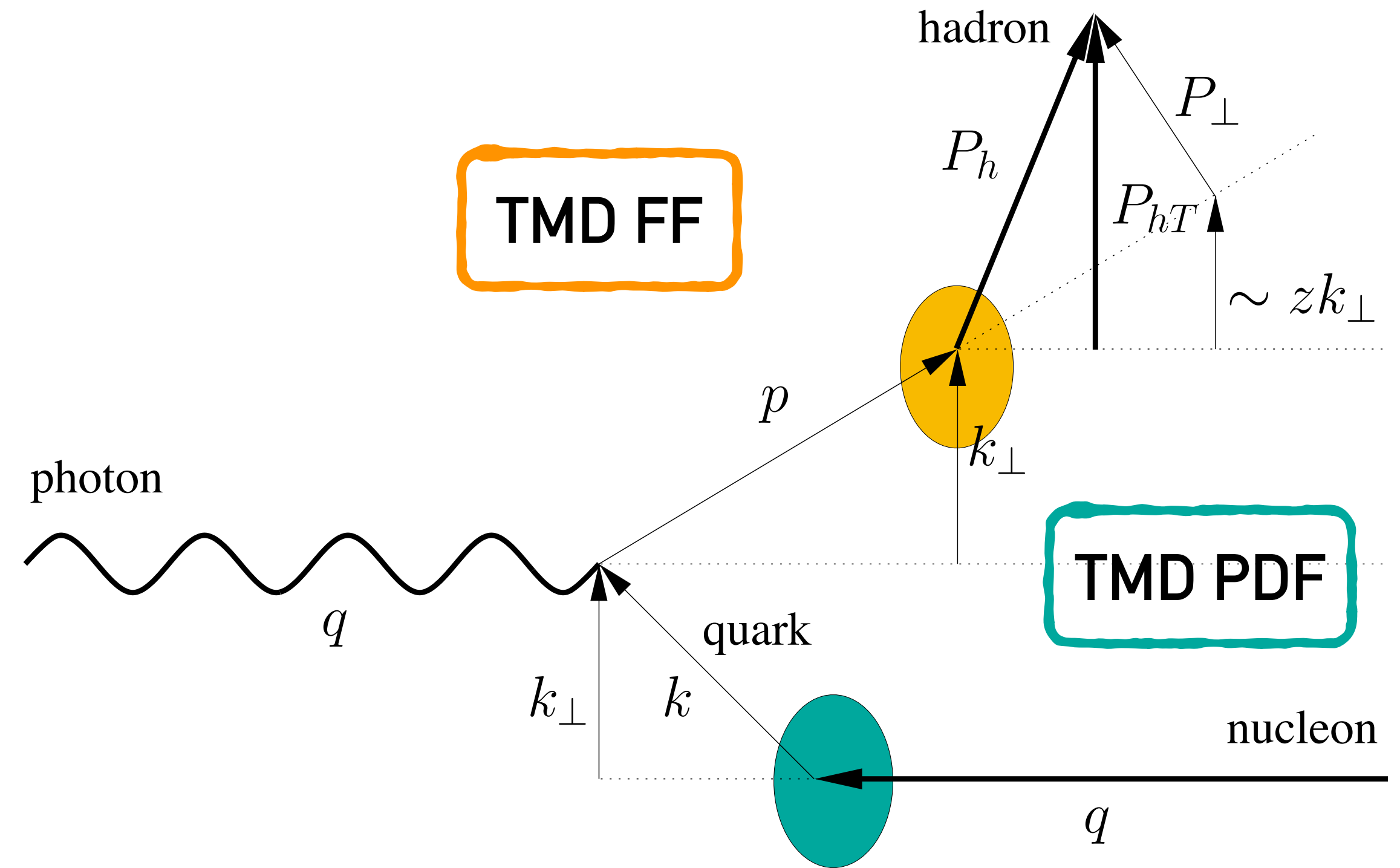
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$$+ Y_{UU,T}(Q^2, \mathbf{P}_{hT}^2) + \mathcal{O}(M^2/Q^2)$$

Y term

W term

- W term dominates in the region where $q_T \ll Q$
- Y term not included in the MAP analyses

unpolarized Transverse Momentum Dependent Parton Distribution Function

TMD PDFs

$$f_1^q(x, b; \mu, \zeta) = \sum_j (C_j \otimes f^j)(x, b_*; \mu_b) e^{R(b_*; \mu_b, \mu)} f_{\text{NP}}(x, b)$$

unpolarized Transverse Momentum Dependent Parton Distribution Function

TMD PDFs

collinear PDFs

$$f_1^q(x, b; \mu, \zeta) = \sum_j (C_j \otimes f^j)(x, b_*; \mu_b) e^{R(b_*; \mu_b, \mu)} f_{\text{NP}}(x, b)$$

unpolarized Transverse Momentum Dependent Parton Distribution Function

TMD PDFs

matching to the collinear region

collinear PDFs

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perturbative expansion
in $\alpha_s(\mu)$

perturbative evolution

unpolarized Transverse Momentum Dependent Parton Distribution Function

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perturbative expansion in $\alpha_s(\mu)$

perturbative evolution

resummation of large logarithms

$$L = \ln \frac{Q^2}{\mu_b^2}$$

log ordering

state of the art: N3LL

next-to-next-to-next leading log

unpolarized Transverse Momentum Dependent Parton Distribution Function

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perturbative expansion in $\alpha_s(\mu)$

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non perturbative transverse content

resummation of large logarithms

$$L = \ln \frac{Q^2}{\mu_b^2}$$

log ordering

state of the art: N3LL
next-to-next-to-next leading log

parametrized and fitted to data

Perturbative accuracy: N^3LL^-

Orders in powers of α_s

Accuracy	H and C	K and γ_F	γ_K	PDF and α_s evol.
LL	0	-	1	-
NLL	0	1	2	LO
NLL'	1	1	2	NLO
NNLL	1	2	3	NLO
NNLL'	2	2	3	NNLO
N^3LL^-	2	3	4	NLO (FF only)
N^3LL	2	3	4	NNLO
N^3LL'	3	3	4	N^3LO

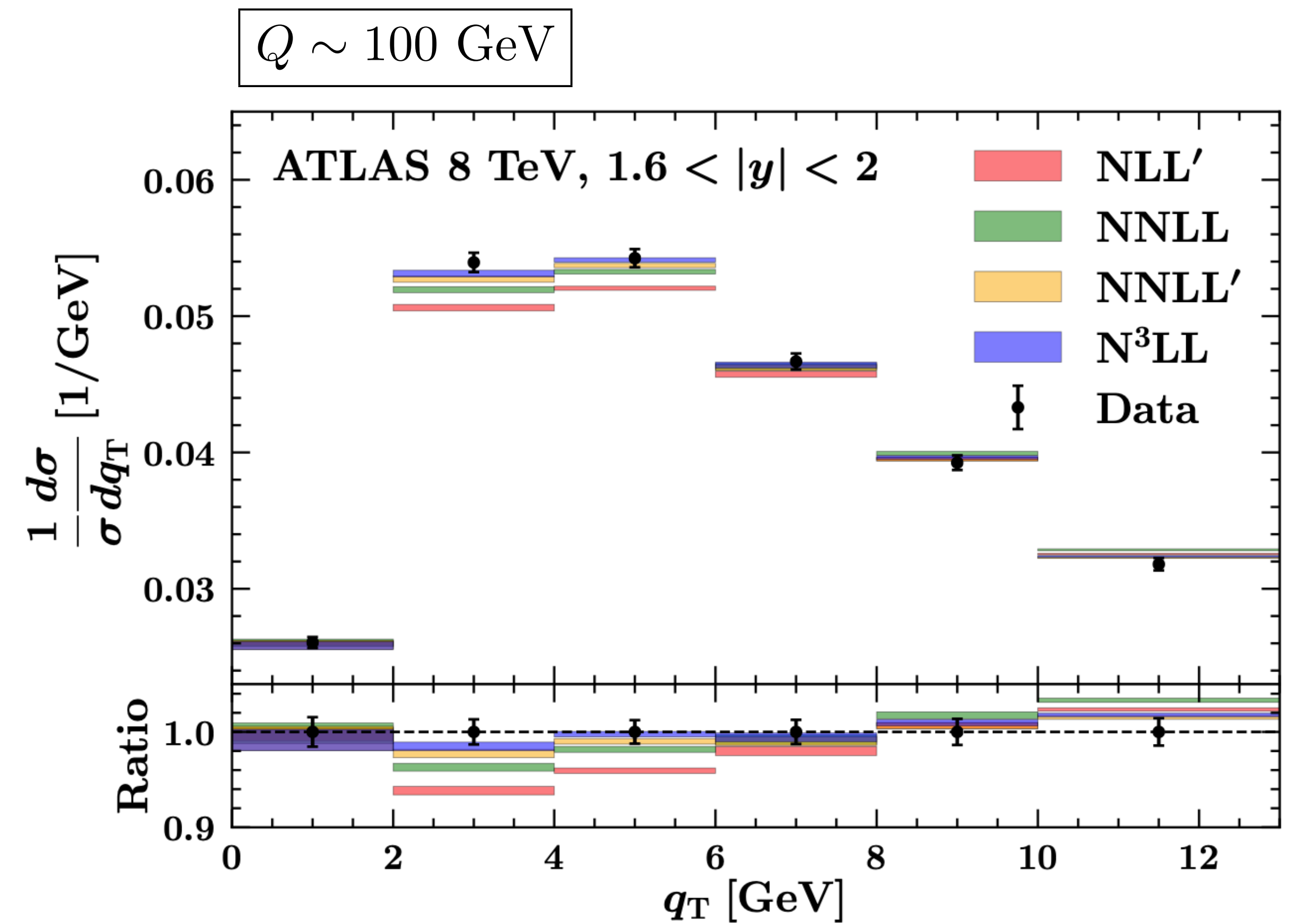
Perturbative accuracy: N^3LL^-

Orders in powers of α_s

Accuracy	Hard factor and matching coefficient	Ingredients in perturbative Sudakov form factor		PDF and α_s evol.
	H and C	K and γ_F	γ_K	
LL	0	-	1	-
NLL	0	1	2	LO
NLL'	1	1	2	NLO
NNLL	1	2	3	NLO
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Normalization of SIDIS multiplicities

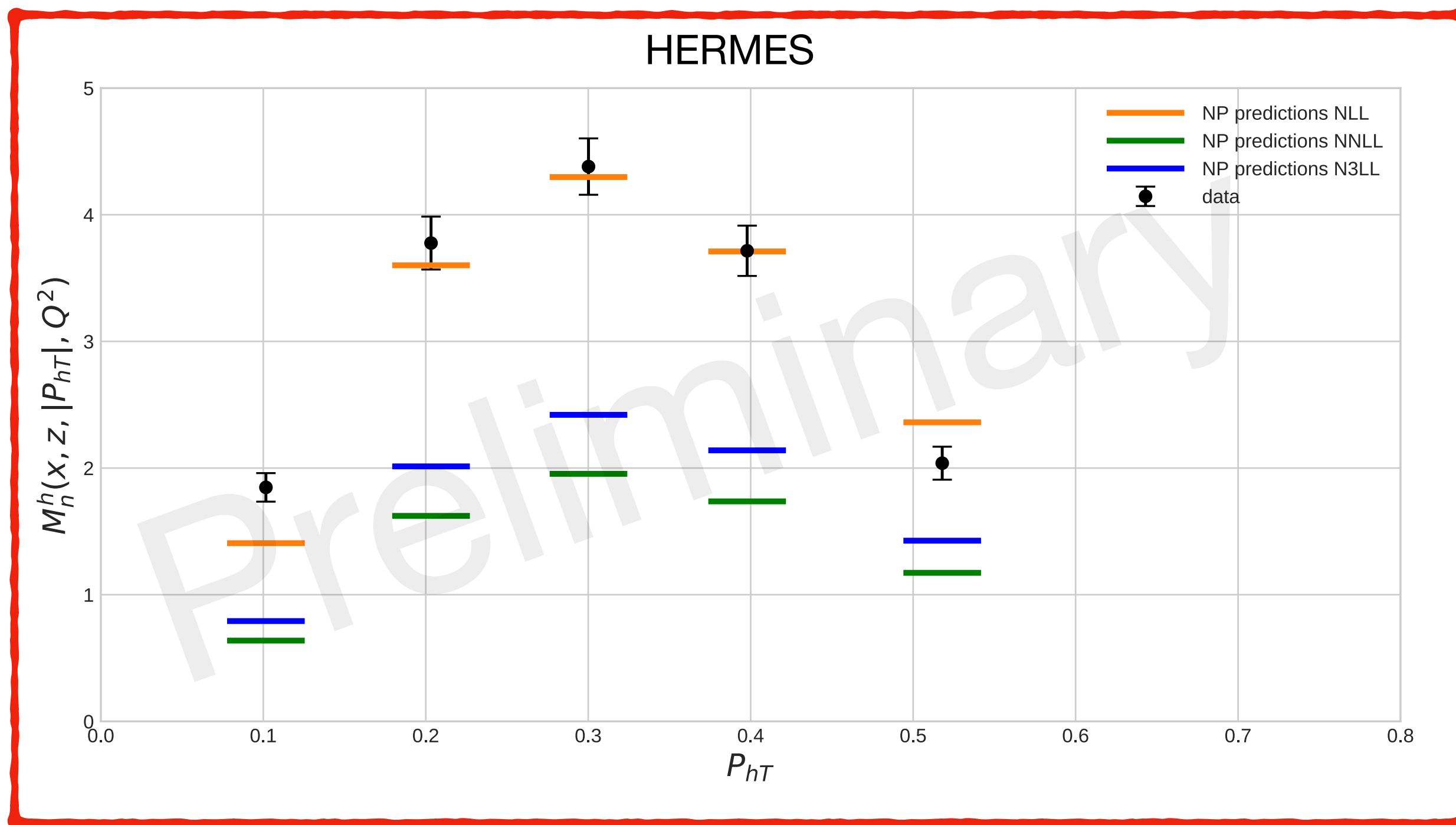
High-Energy Drell-Yan beyond NLL



Normalization of SIDIS multiplicities

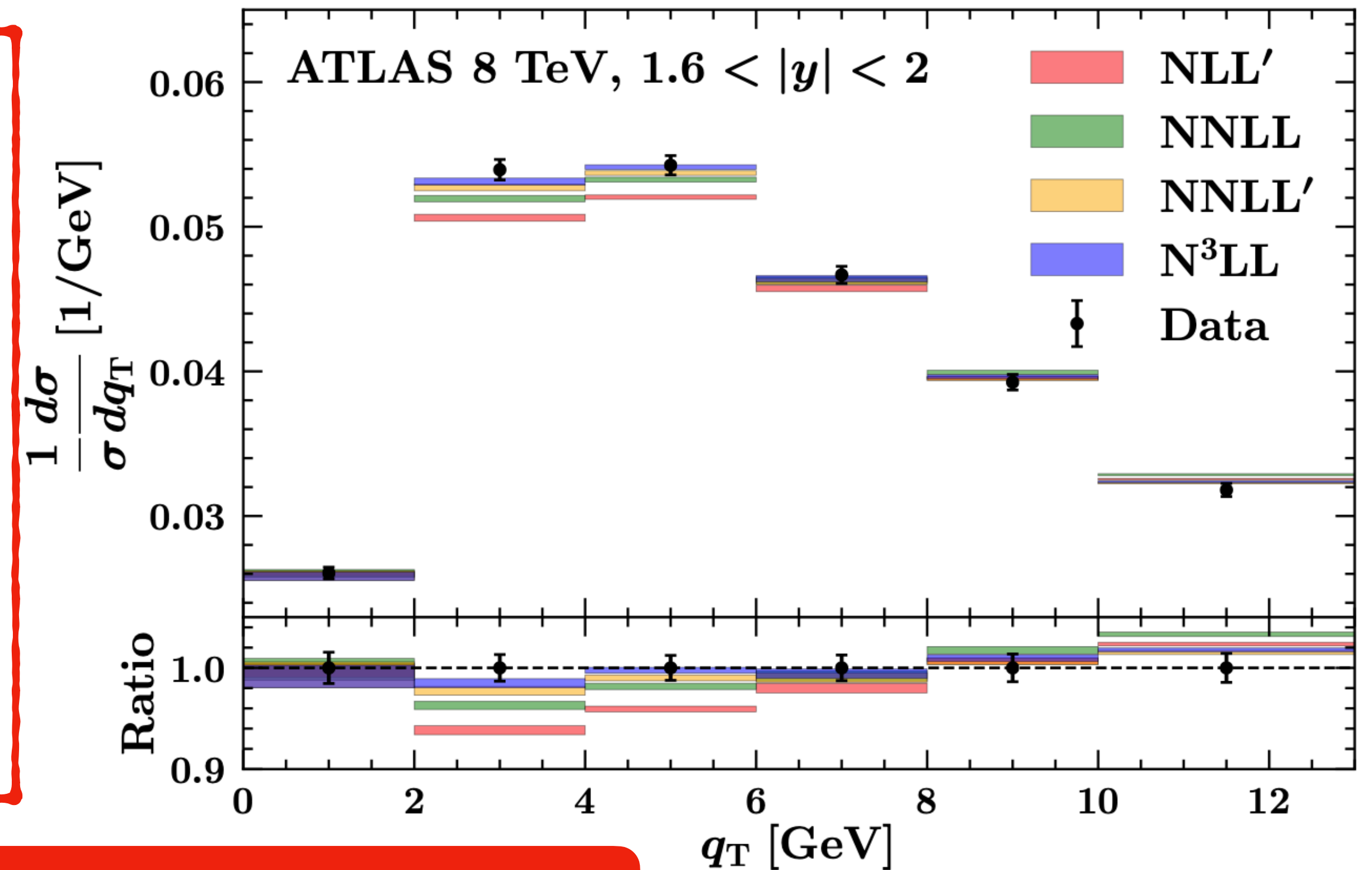
SIDIS multiplicities beyond NLL

$Q \sim 2 \text{ GeV}$



High-Energy Drell-Yan beyond NLL

$Q \sim 100 \text{ GeV}$



description considerably worsens at higher orders

Normalization of SIDIS multiplicities

Introduction of a normalization prefactor

SIDIS multiplicity

$$M(x, z, P_{hT}, Q) = \frac{d\sigma}{dx dQ dz dP_{hT}} \bigg/ \frac{d\sigma}{dx dQ}$$

Normalization of SIDIS multiplicities

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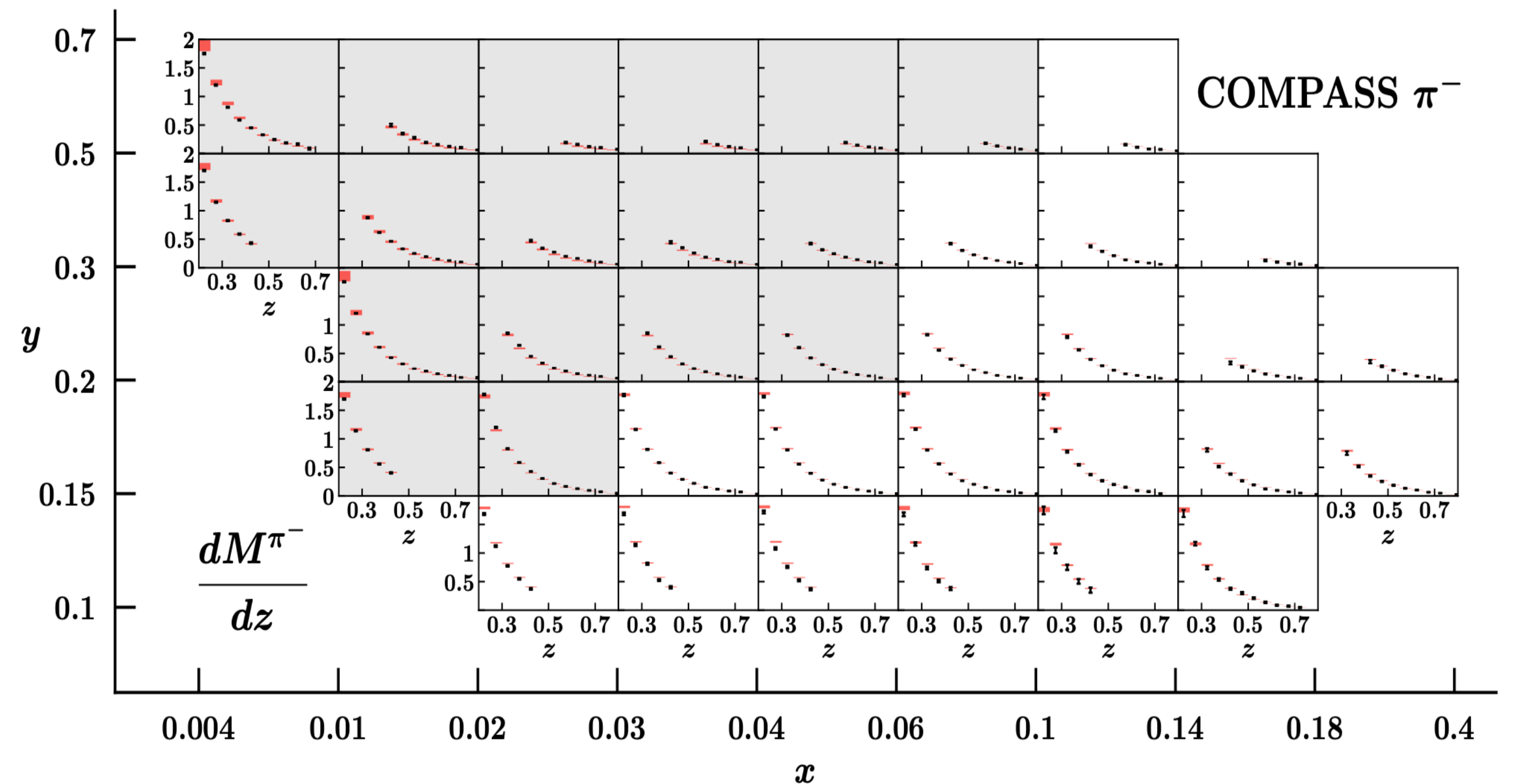
Collinear SIDIS cross section

$$\frac{d\sigma}{dx dQ dz}$$

No normalization problems

MAPFF1.0

MAP Collaboration collinear FFs fit



Normalization of SIDIS multiplicities

Introduction of a normalization prefactor

SIDIS multiplicity

$$M(x, z, P_{hT}, Q) = \frac{d\sigma}{dx dQ dz dP_{hT}} \bigg/ \frac{d\sigma}{dx dQ}$$

$$w(x, z, Q) = \frac{d\sigma}{dx dQ dz} \bigg/ \int dP_{hT} \frac{d\sigma}{dx dQ dz dP_{hT}}$$

computed a priori, before the fit

📌 **Depends on the collinear PDFs**

📌 **independent of the fitting parameters**

$$M(x, z, P_{hT}, Q) = w(x, z, Q) \frac{d\sigma}{dx dQ dz dP_{hT}} \bigg/ \frac{d\sigma}{dx dQ}$$

Global analysis of DY and SIDIS data sets

Cuts on kinematics

$$\langle Q \rangle > 1.3 \text{ GeV}$$

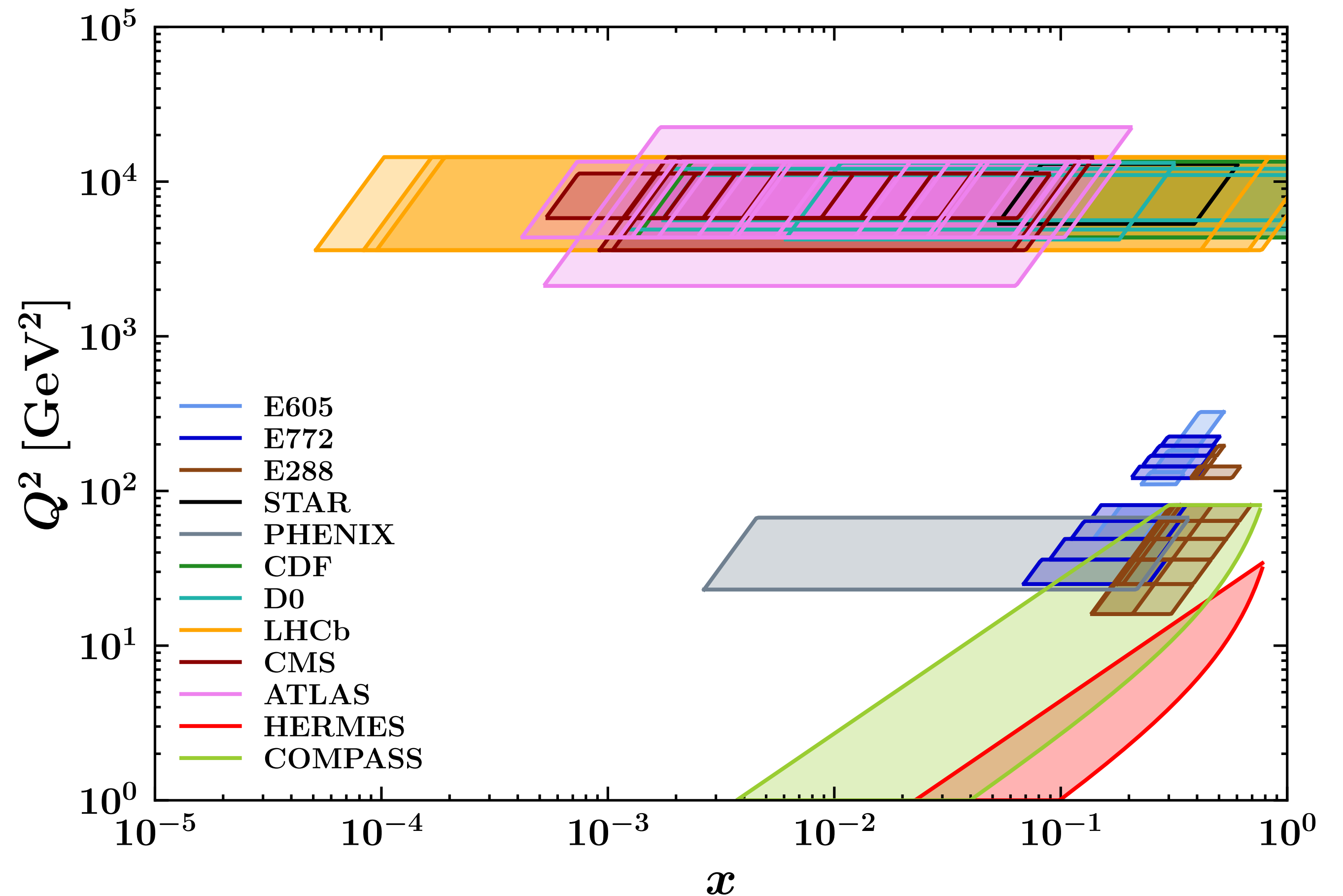
$$0.2 < \langle z \rangle < 0.7$$

DY

$$q_T|_{\max} = 0.2Q$$

SIDIS

$$P_{hT}|_{\max} = \min[\min[0.2Q, 0.5zQ] + 0.3 \text{ GeV}, zQ]$$



Total number of points = 2031

Non-perturbative part of TMDs

TMD PDF

$$f_{1\text{NP}}(x, b_T^2) \propto \text{F.T. of } \left(e^{-\frac{k_\perp^2}{g_{1A}}} + \lambda_B k_\perp^2 e^{-\frac{k_\perp^2}{g_{1B}}} + \lambda_C e^{-\frac{k_\perp^2}{g_{1C}}} \right)$$

$$g_1(x) = N_1 \frac{(1-x)^\alpha x^\sigma}{(1-\hat{x})^\alpha \hat{x}^\sigma}$$

TMD FF

$$D_{1\text{NP}}(x, b_T^2) \propto \text{F.T. of } \left(e^{-\frac{P_\perp^2}{g_{3A}}} + \lambda_{FB} k_\perp^2 e^{-\frac{P_\perp^2}{g_{3B}}} \right)$$

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NP evolution

$$g_K(b_T^2) = -g_2^2 \frac{b_T^2}{4}$$

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Gaussians

TMD FF

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Gaussians

weighted Gaussians

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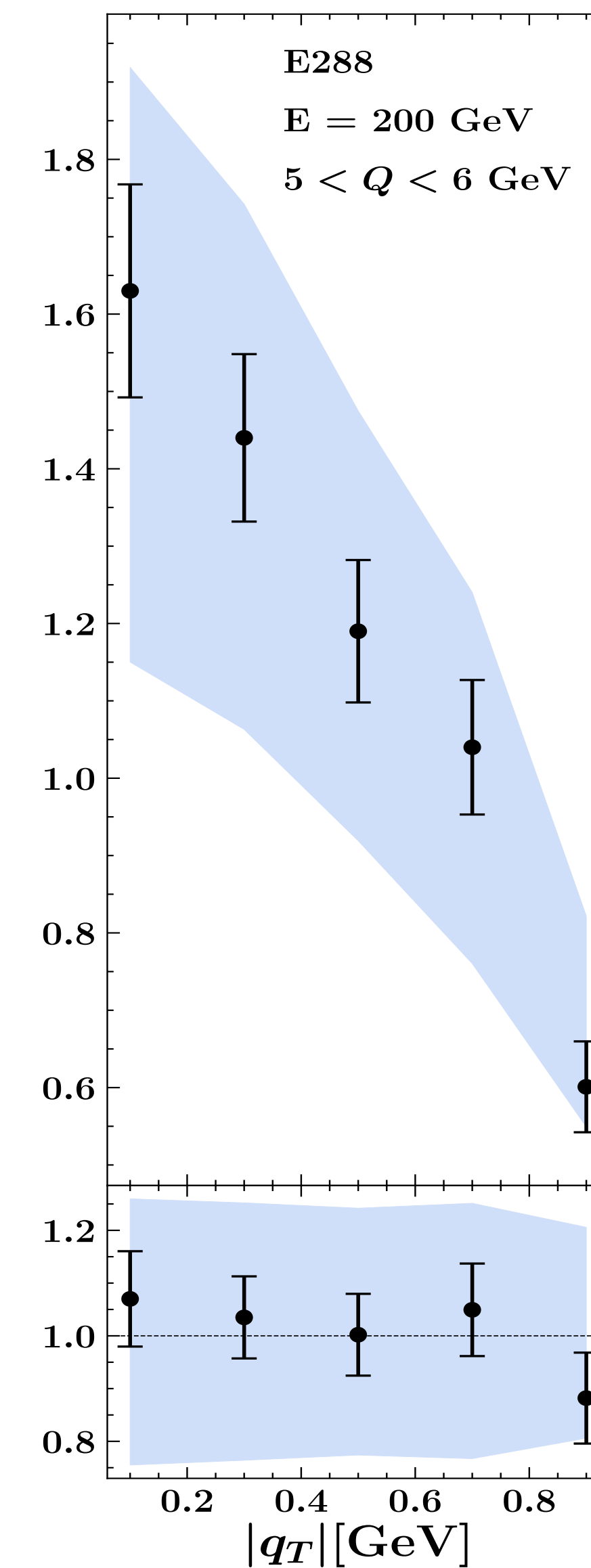
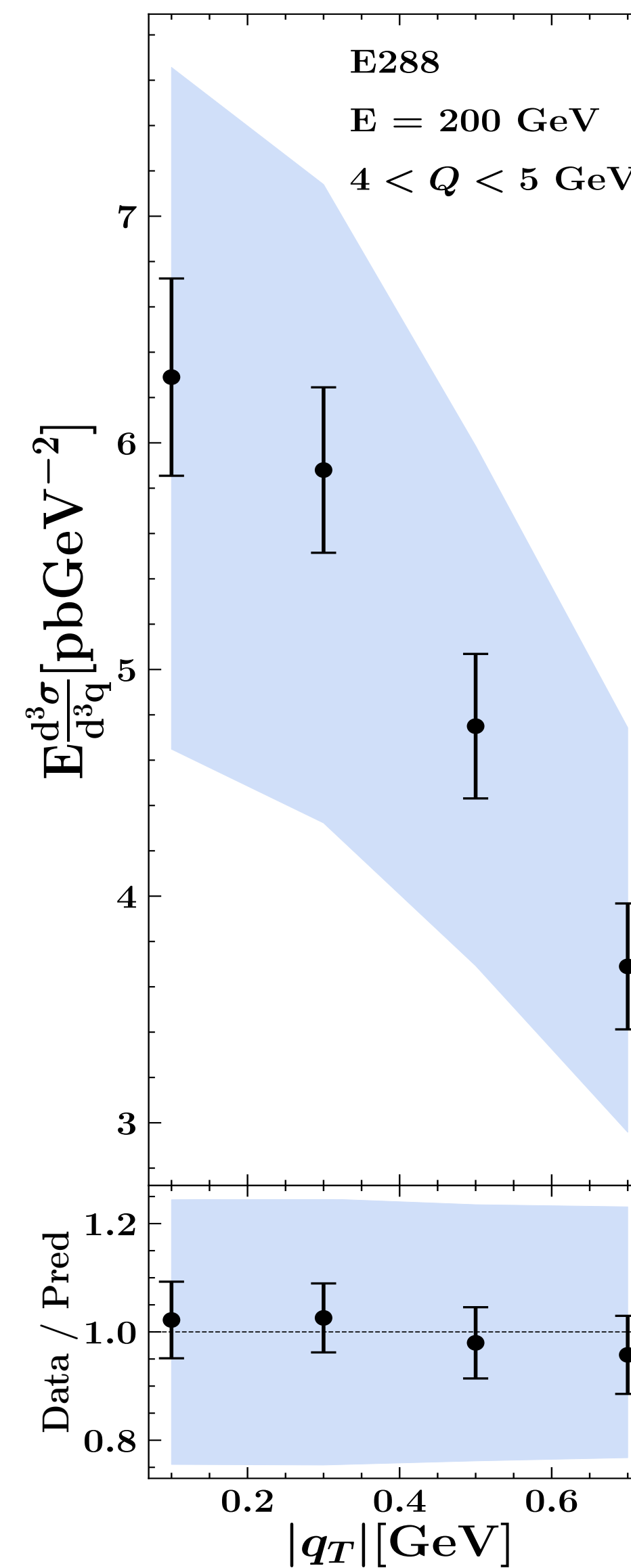
$$g_K(b_T^2) = -g_2^2 \frac{b_T^2}{4}$$

11 parameters for TMD PDF
 + 1 for NP evolution + 9 for TMD FF
 = 21 free parameters

Comparison with data

Drell-Yan

Data set	N_{dat}	χ_D^2/N_{dat}	$\chi_\lambda^2/N_{\text{dat}}$	χ_0^2/N_{dat}
DY collider total	251	1.86	0.2	2.06
DY fixed-target total	233	0.85	0.4	1.24
SIDIS total	1547	0.59	0.28	0.87
Total	2031	0.77	0.29	1.06



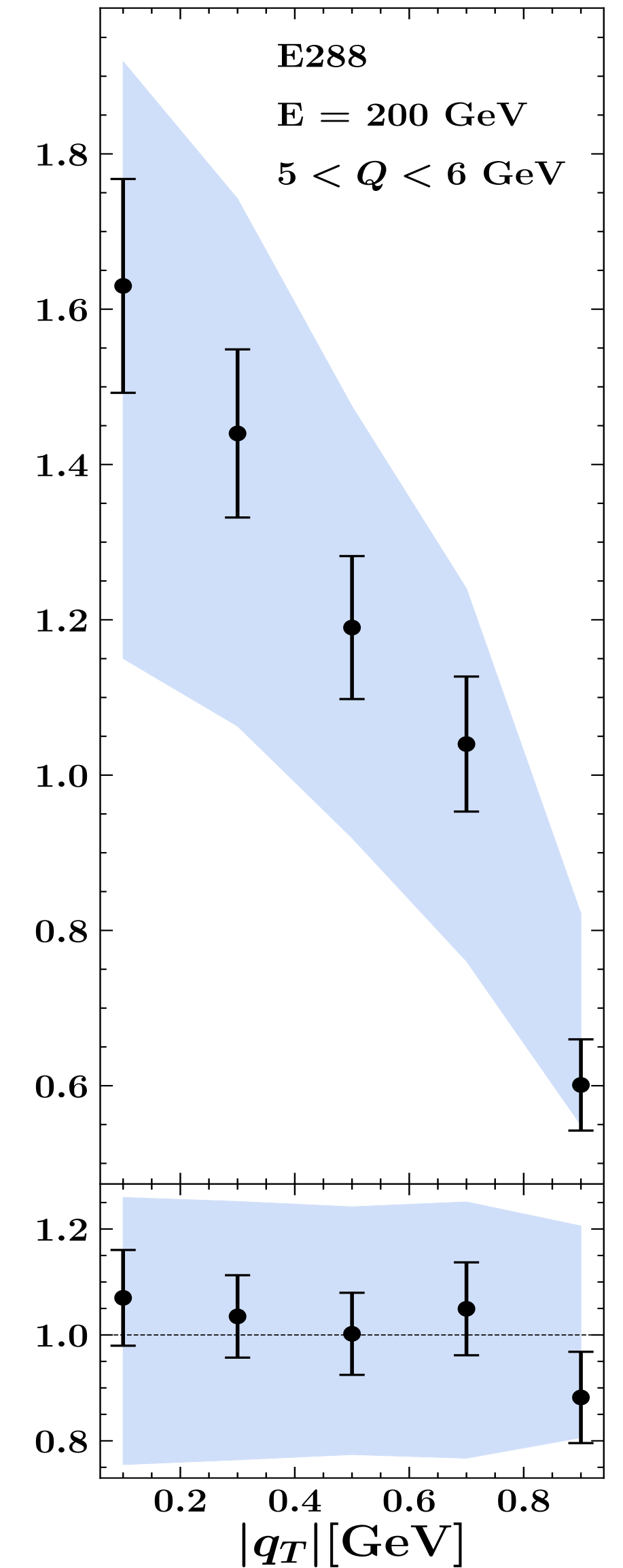
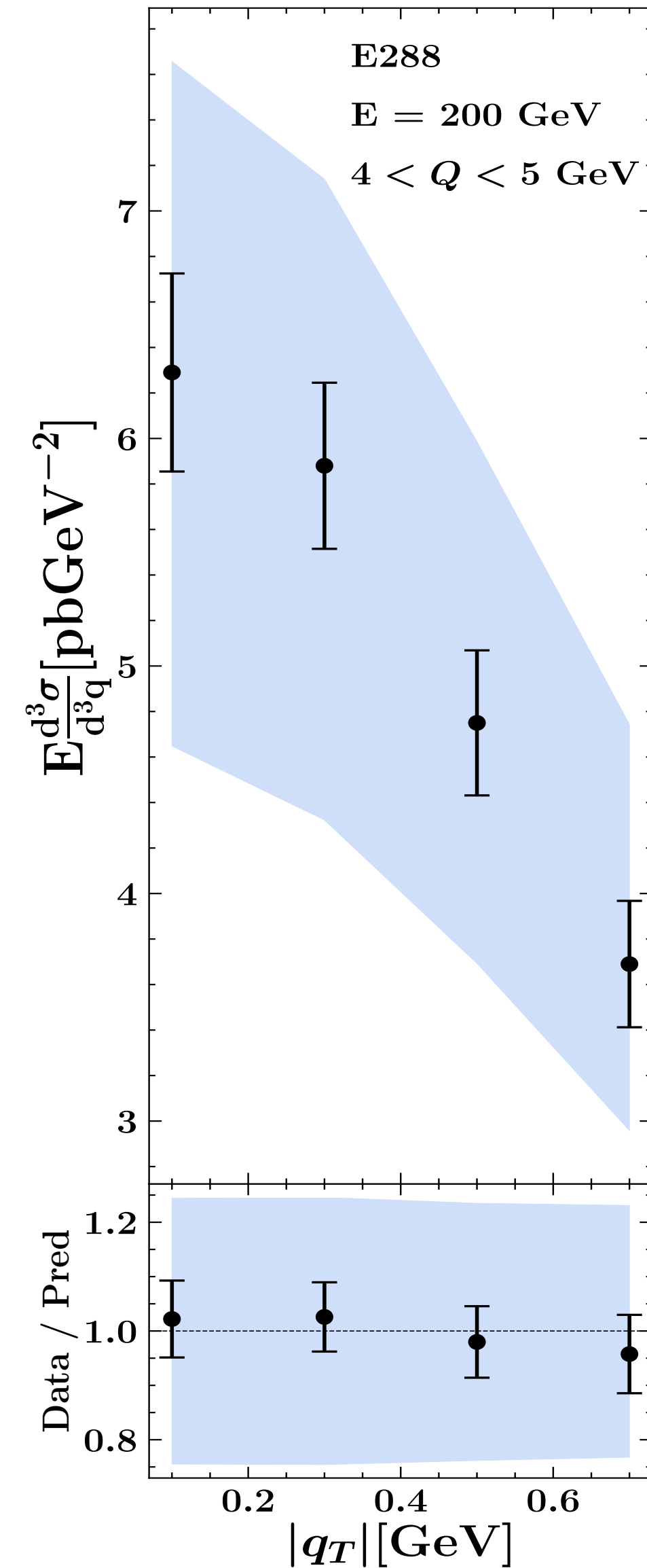
Comparison with data

Drell-Yan

N3LL⁻

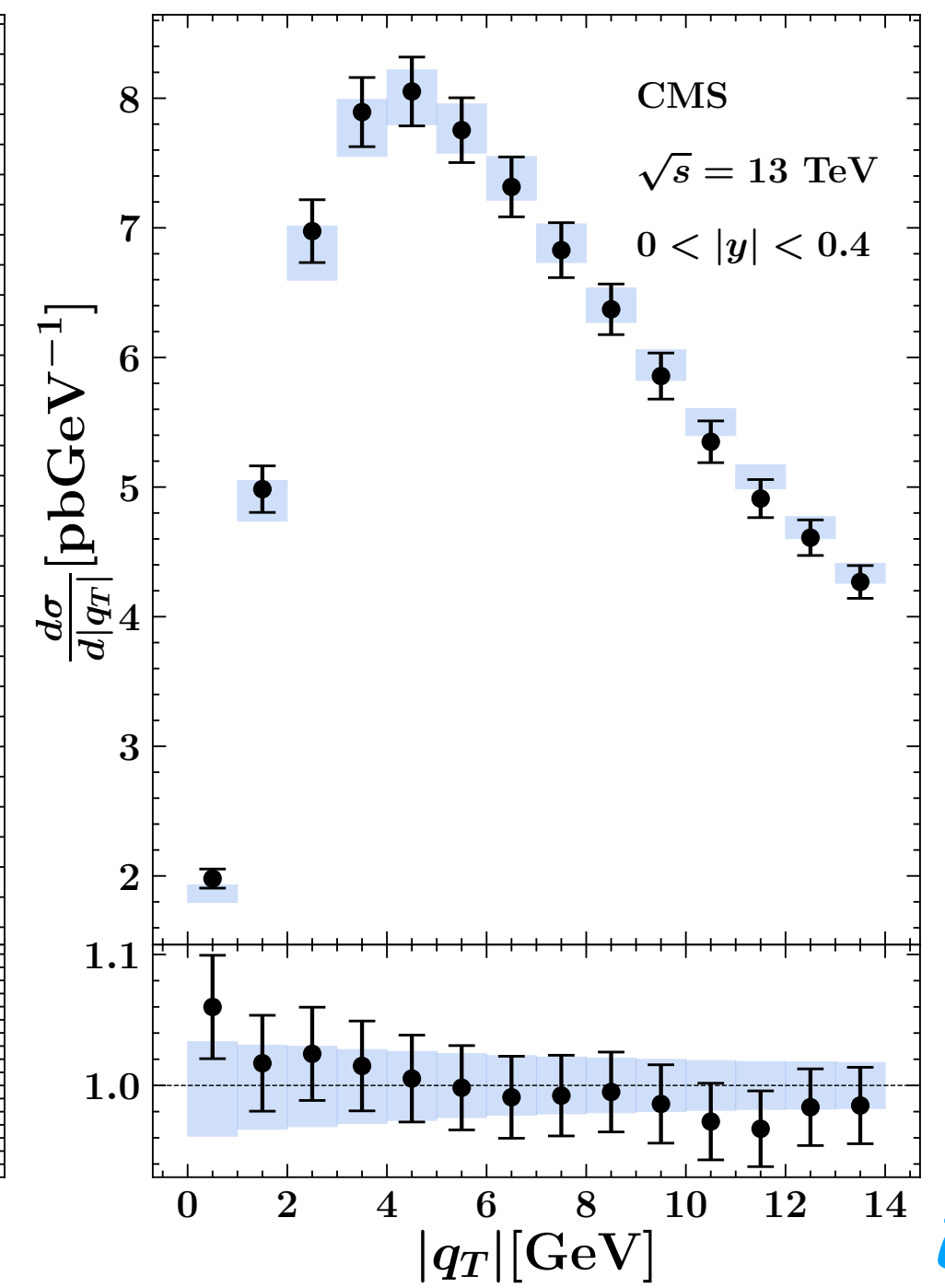
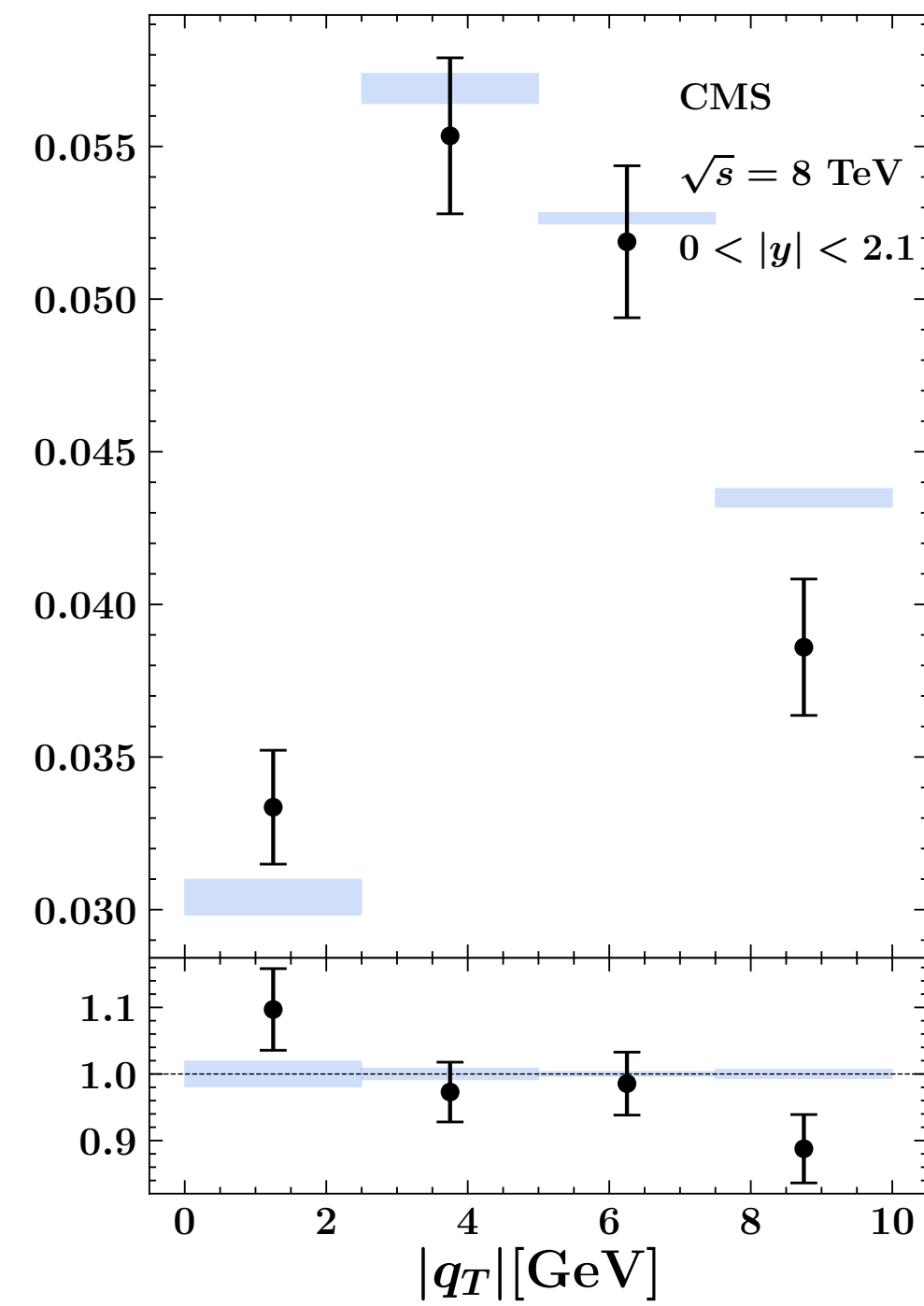
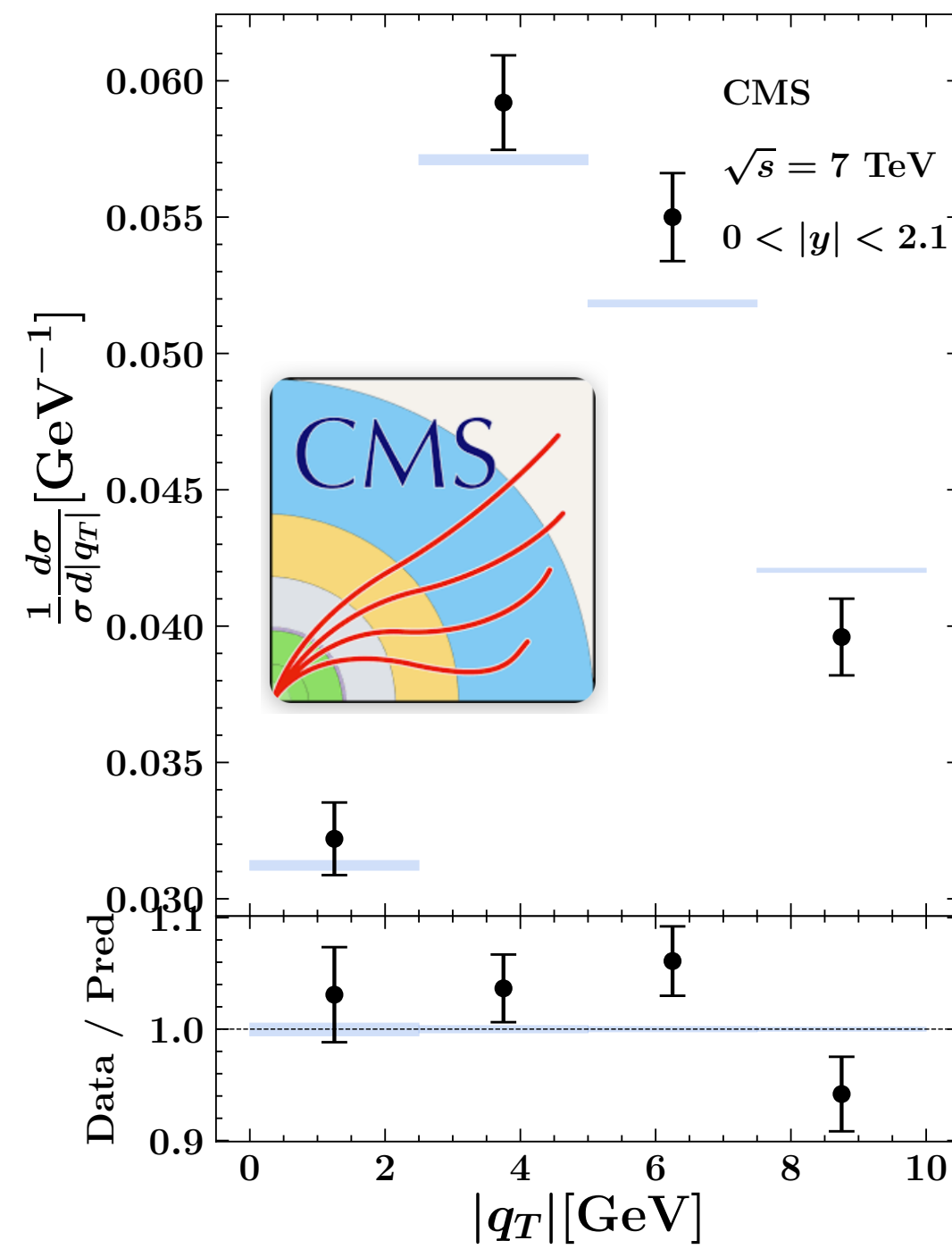
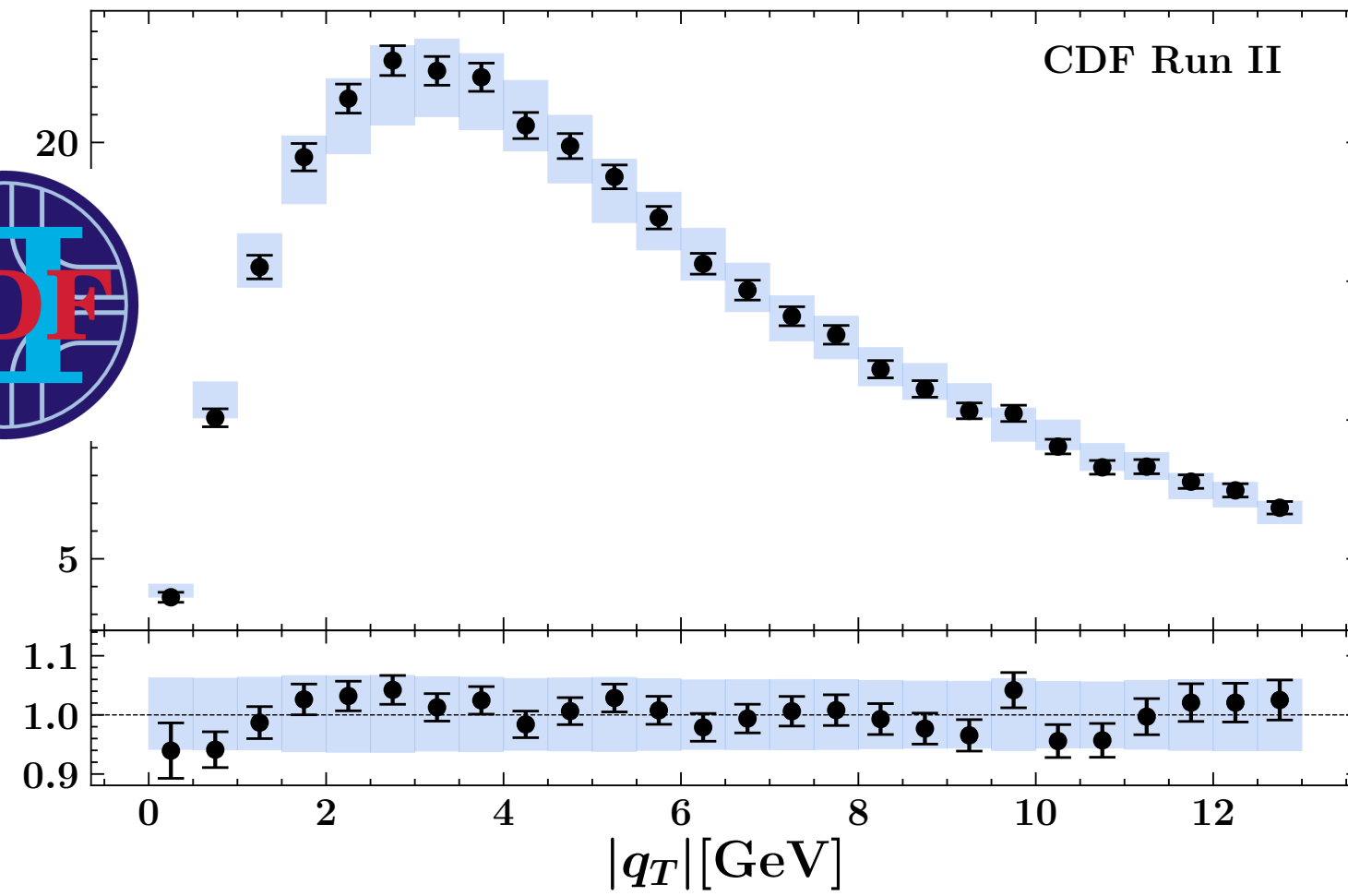
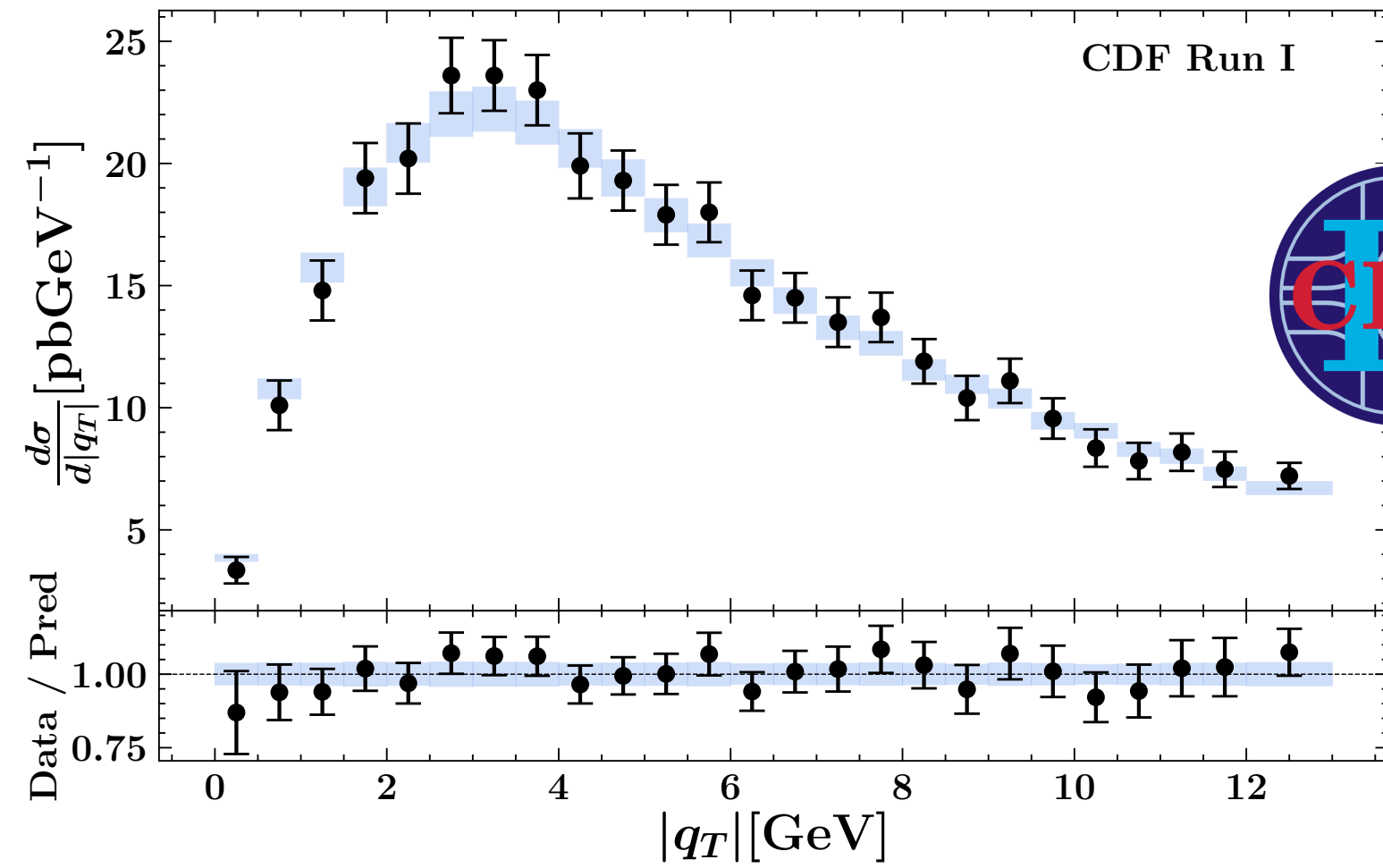
Good agreement
DY low energy

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Comparison with data **Drell-Yan**

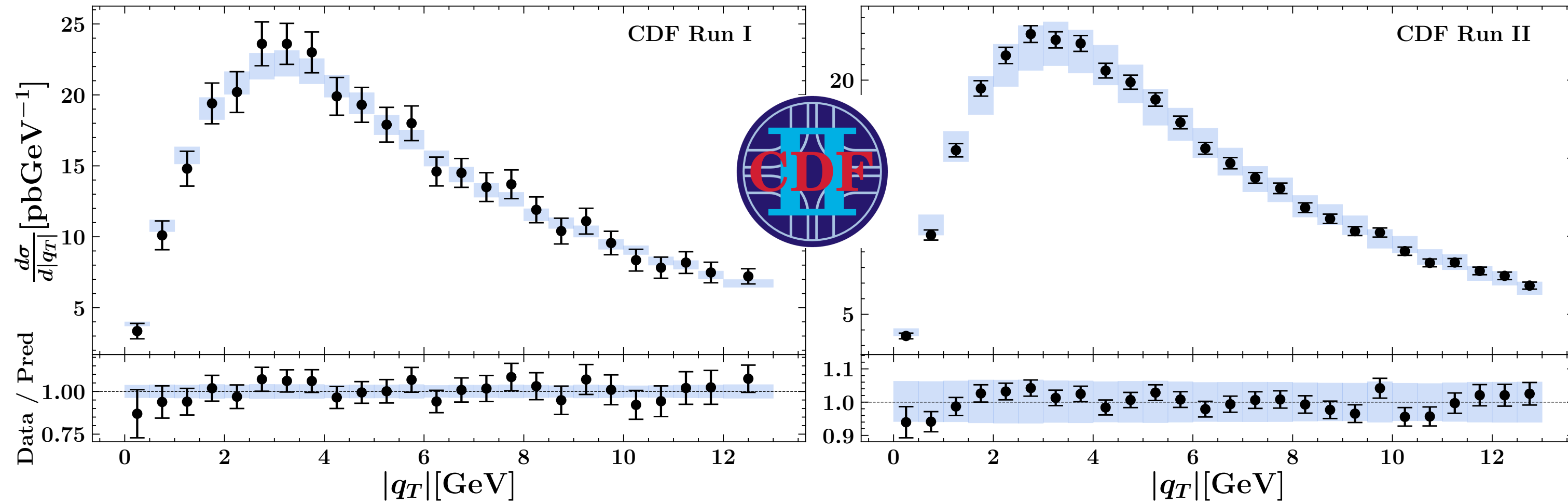
N3LL



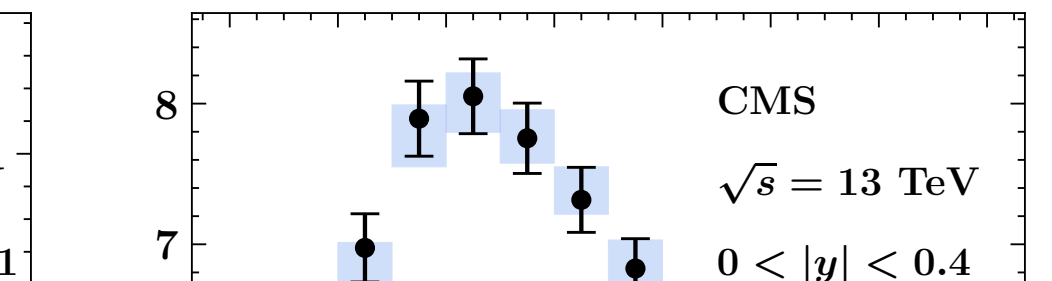
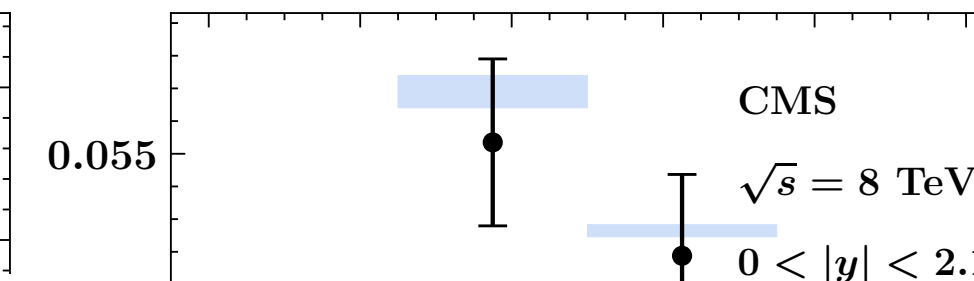
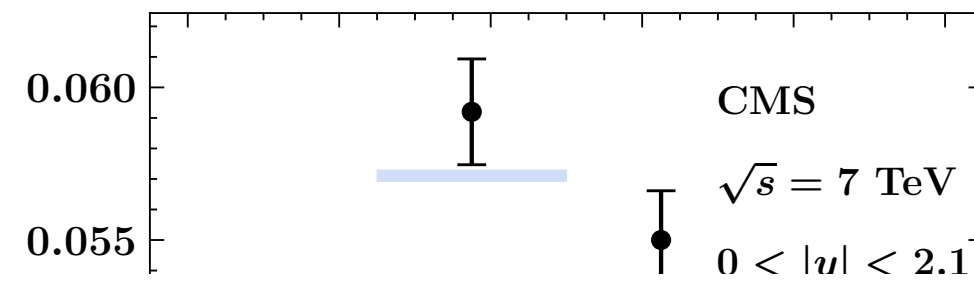
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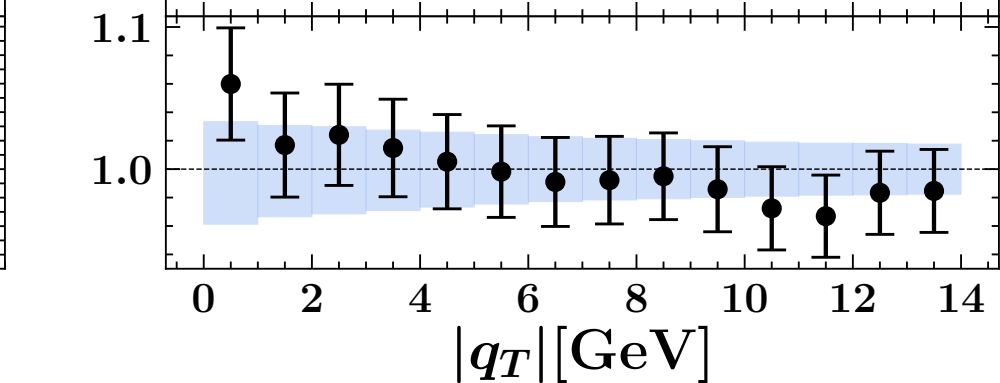
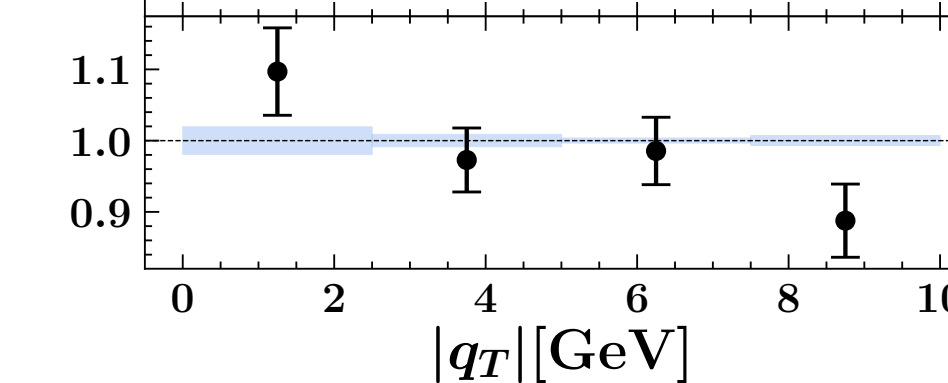
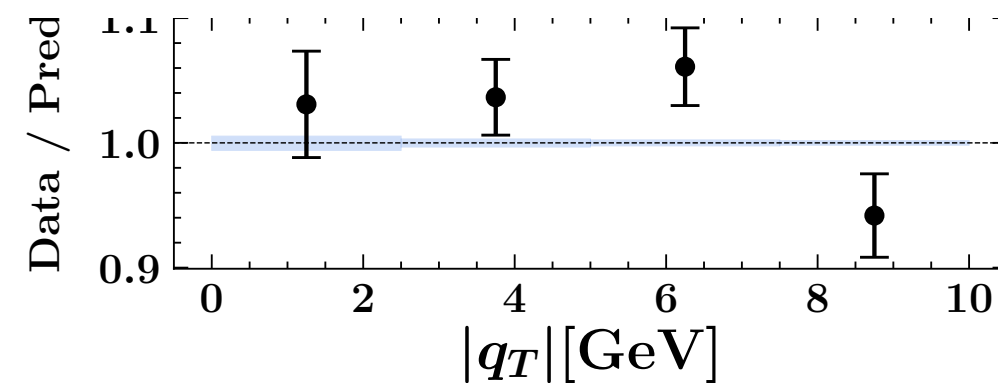
N3LL



Fairly good agreement
(DY high-energy)



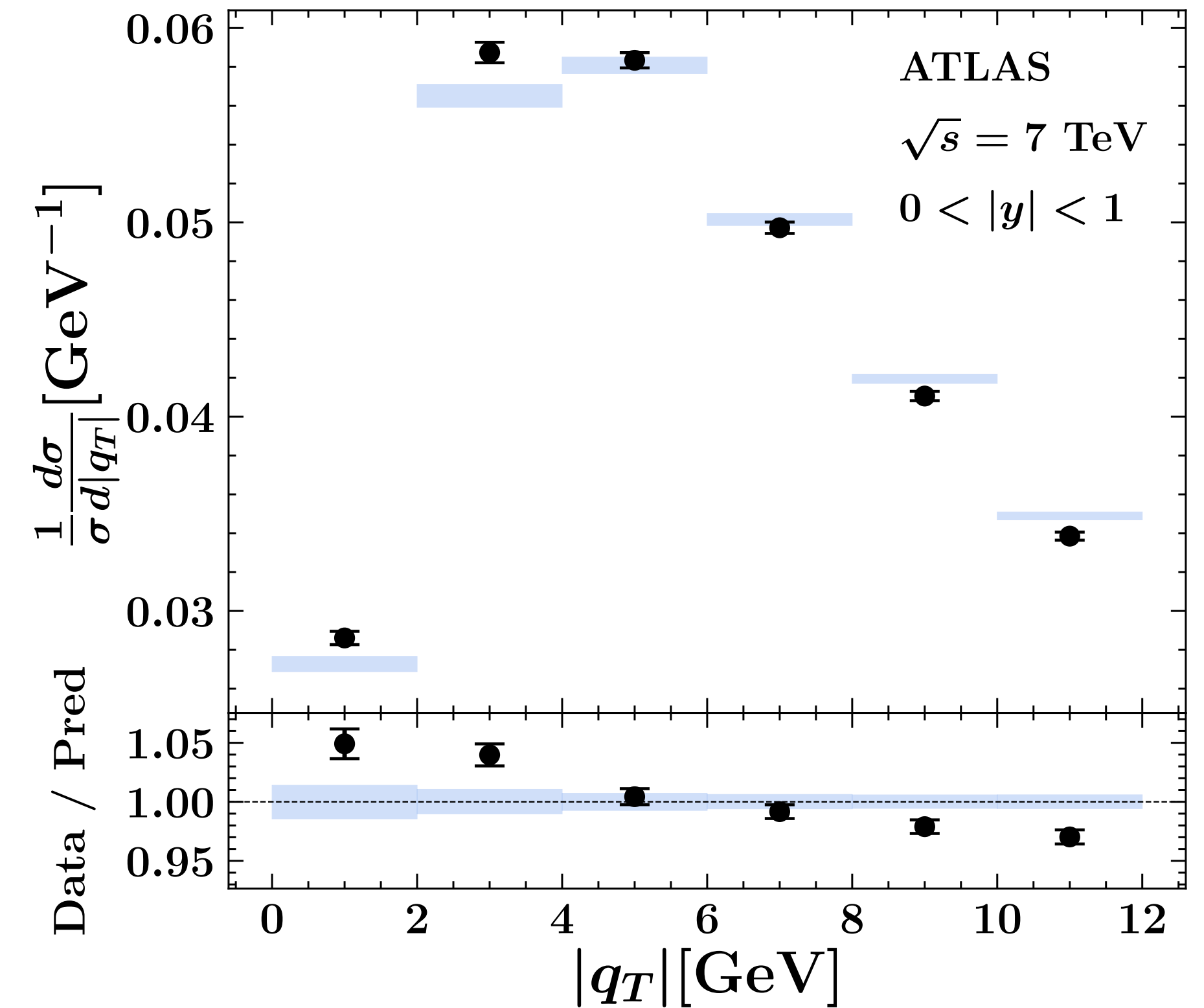
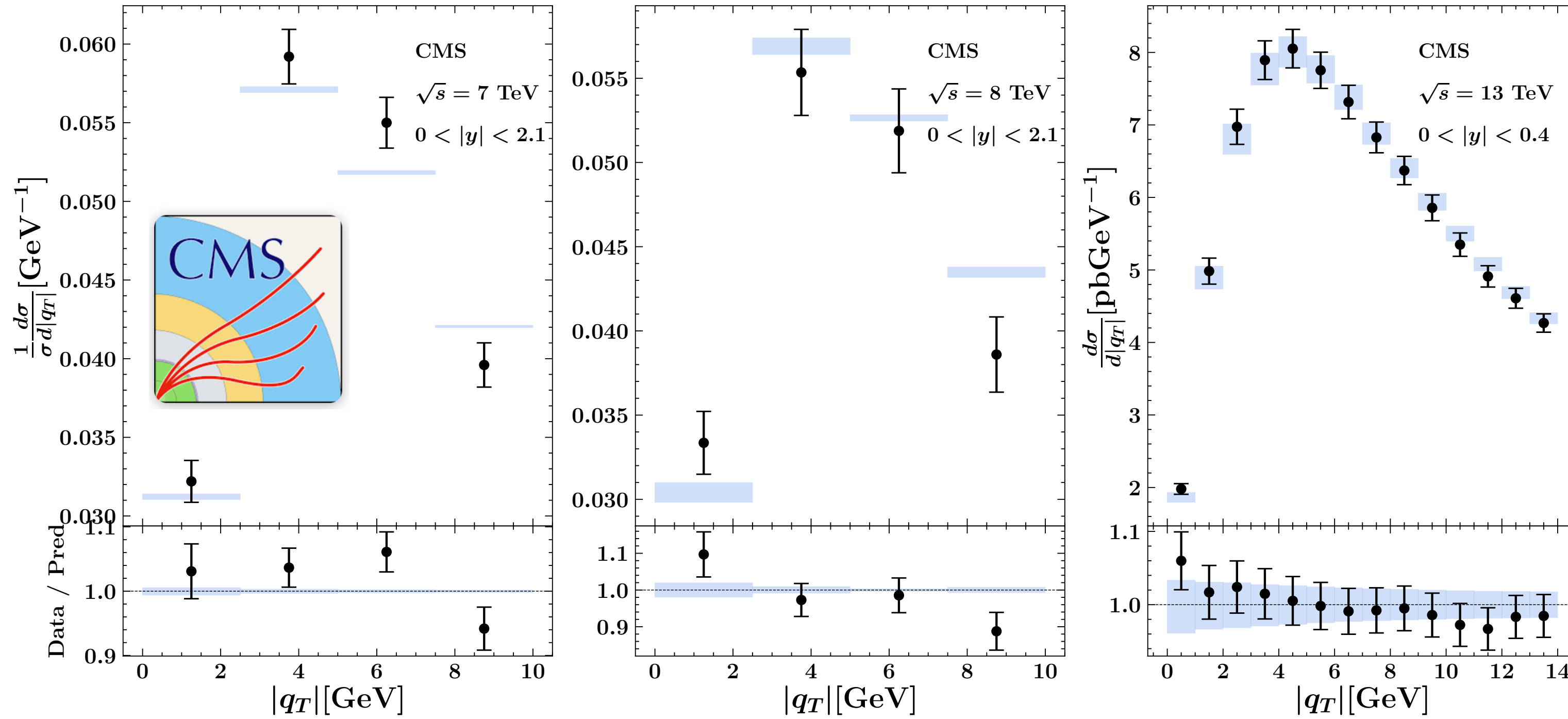
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N3LL⁻

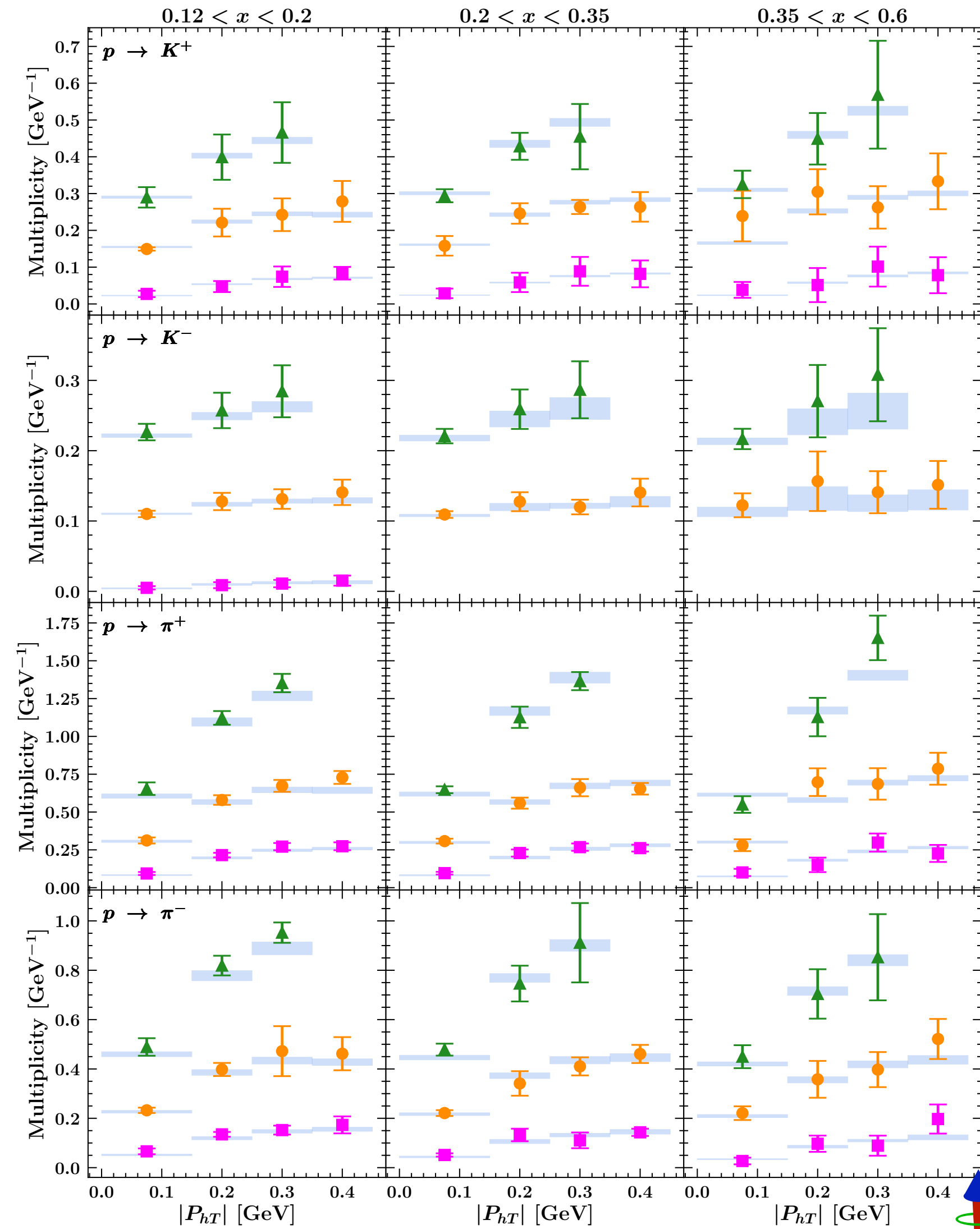


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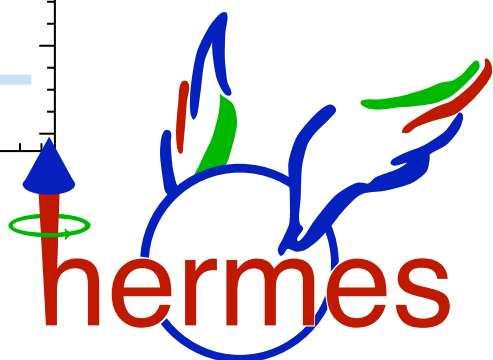
$\chi^2/N_{\text{data}} = 5.05$
(DY ATLAS)

Comparison with data

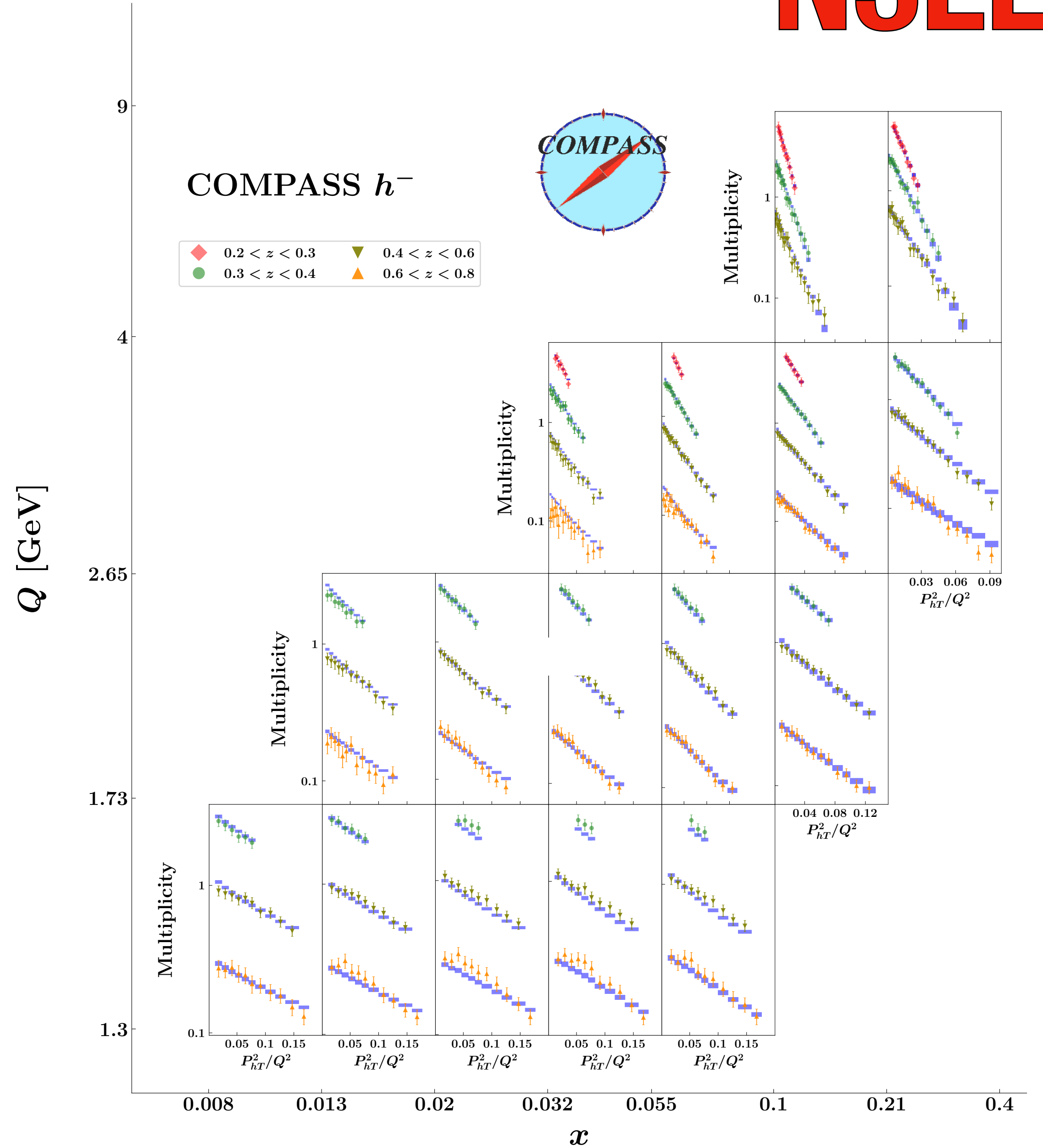
SIDIS



▲ $0.375 < z < 0.475$ (offset = 0.2)
 ● $0.475 < z < 0.6$ (offset = 0.1)
 ■ $0.6 < z < 0.8$ (offset = 0)



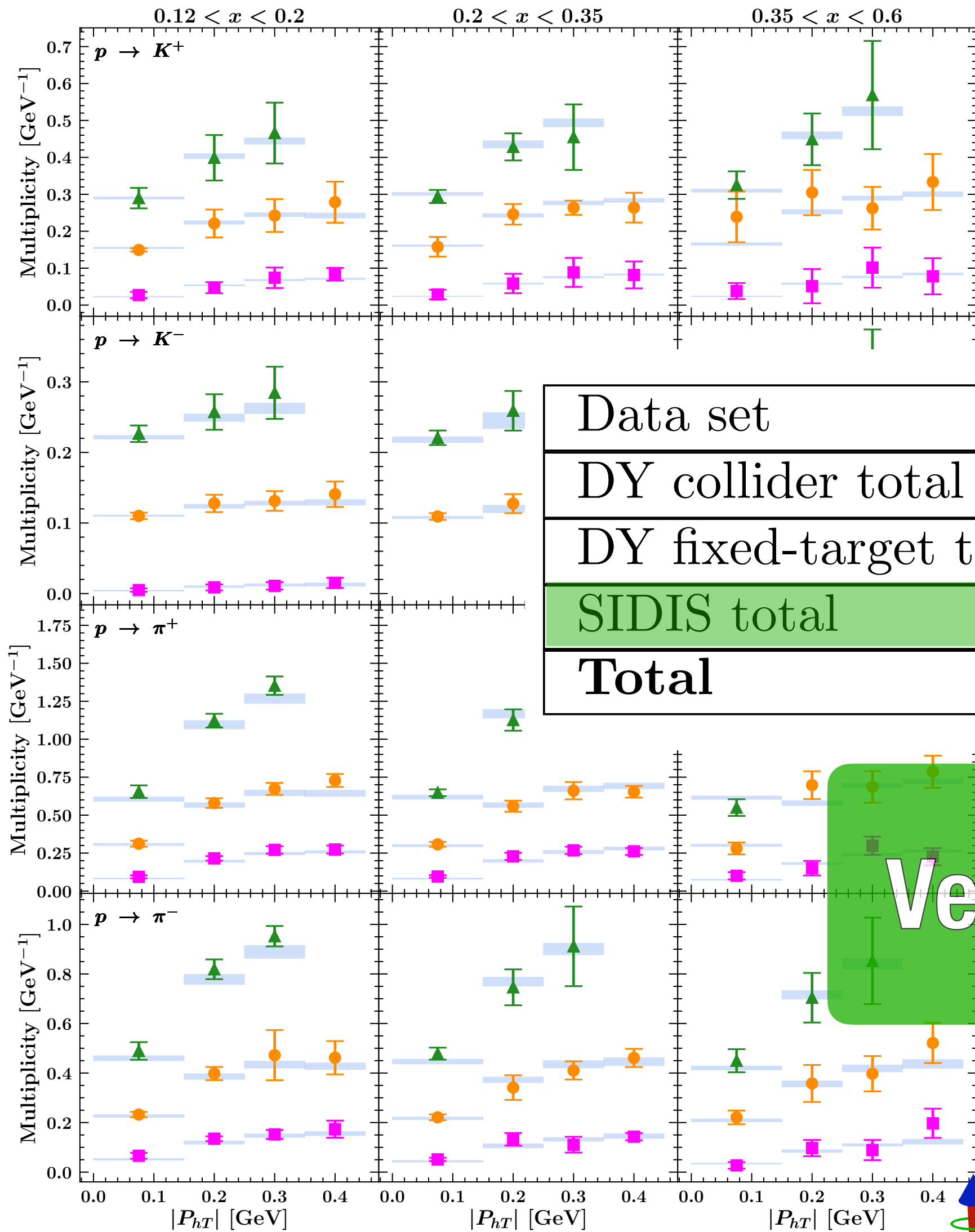
N3LL



Comparison with data

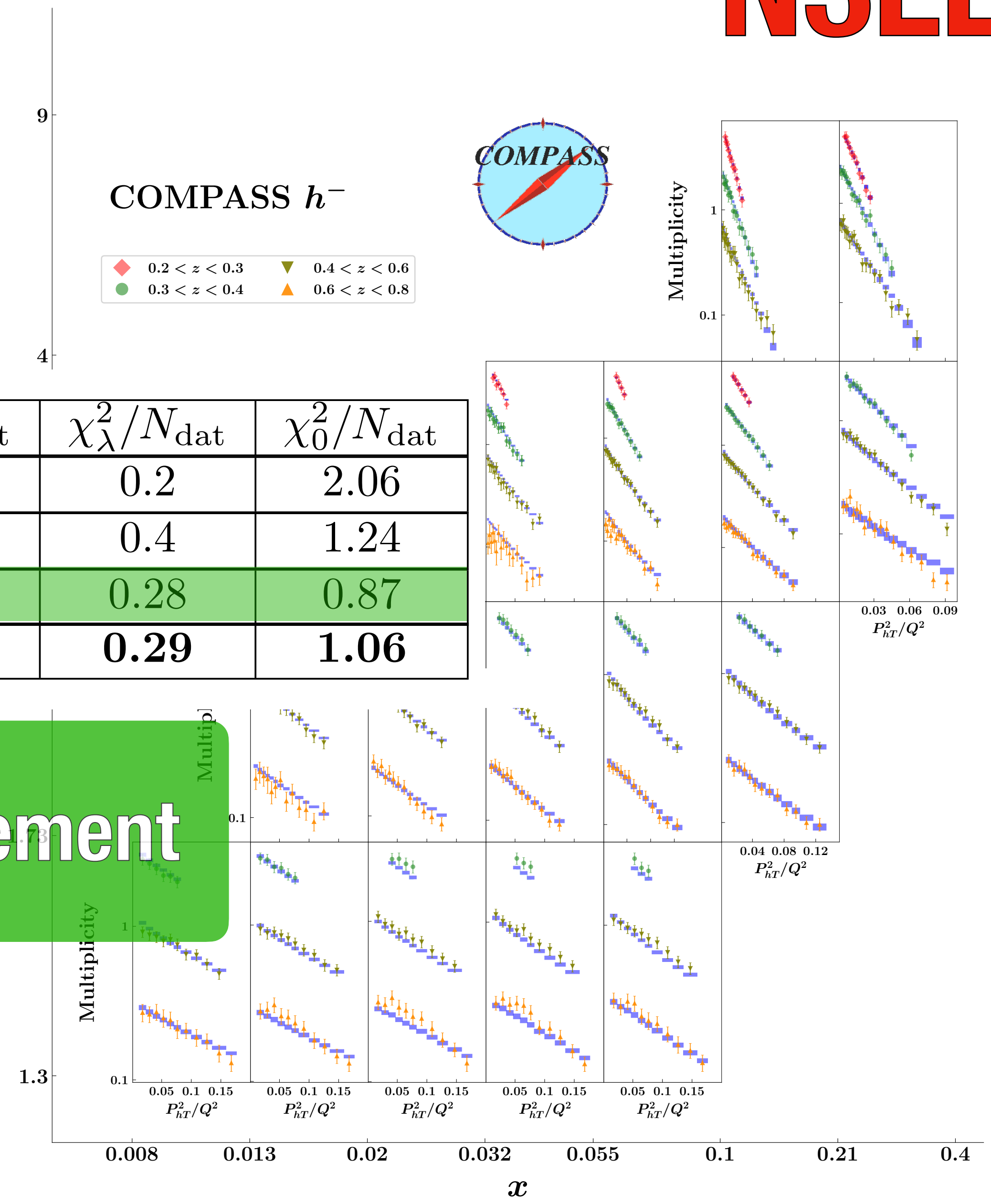
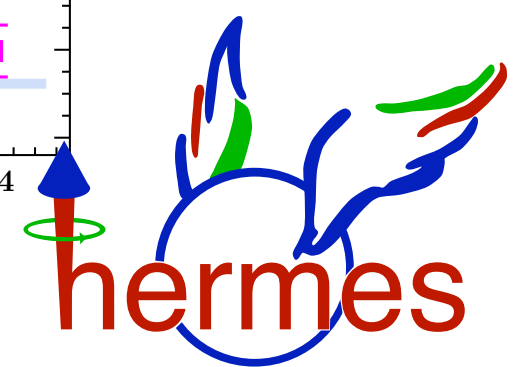
N3LL⁻

SIDIS



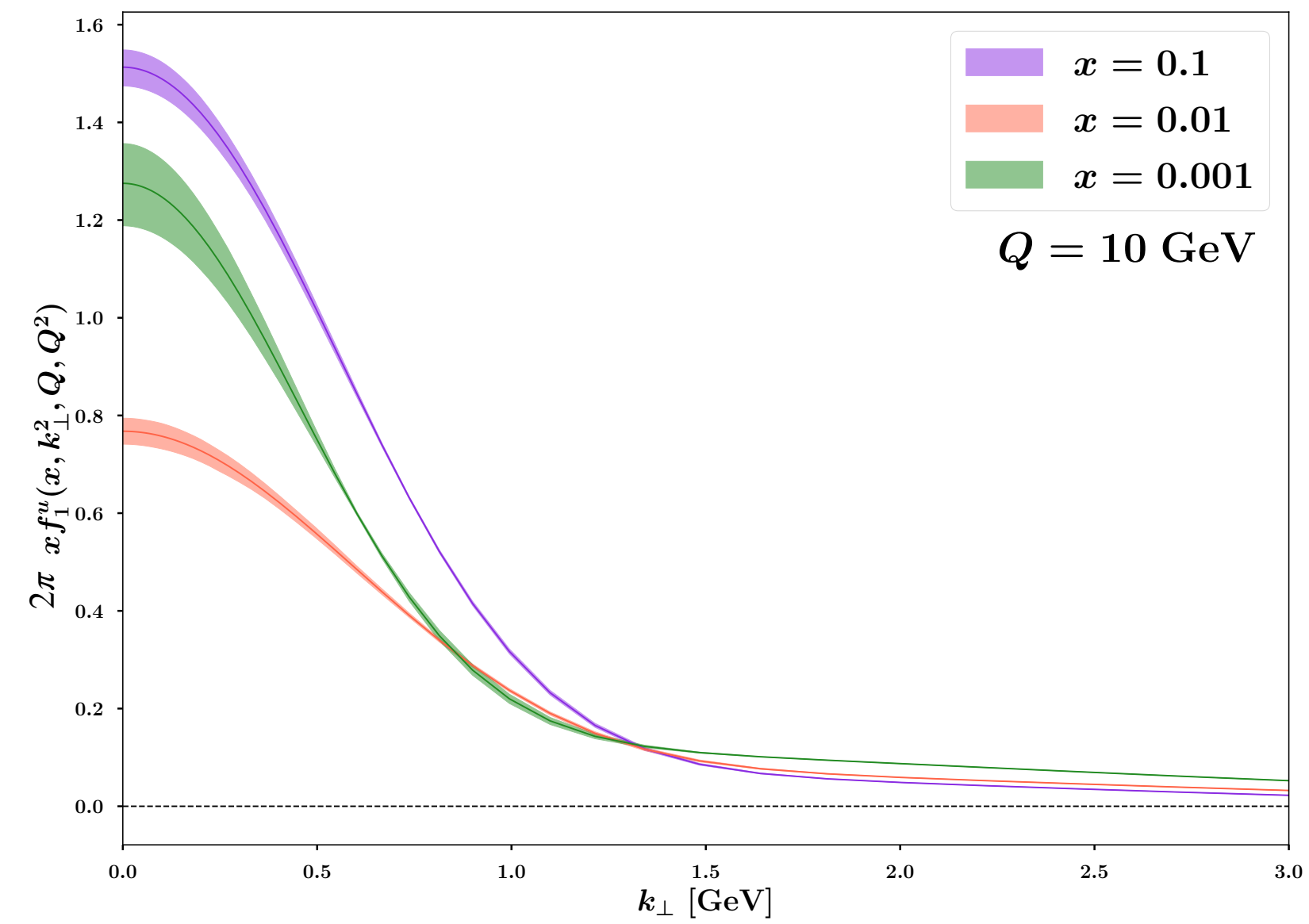
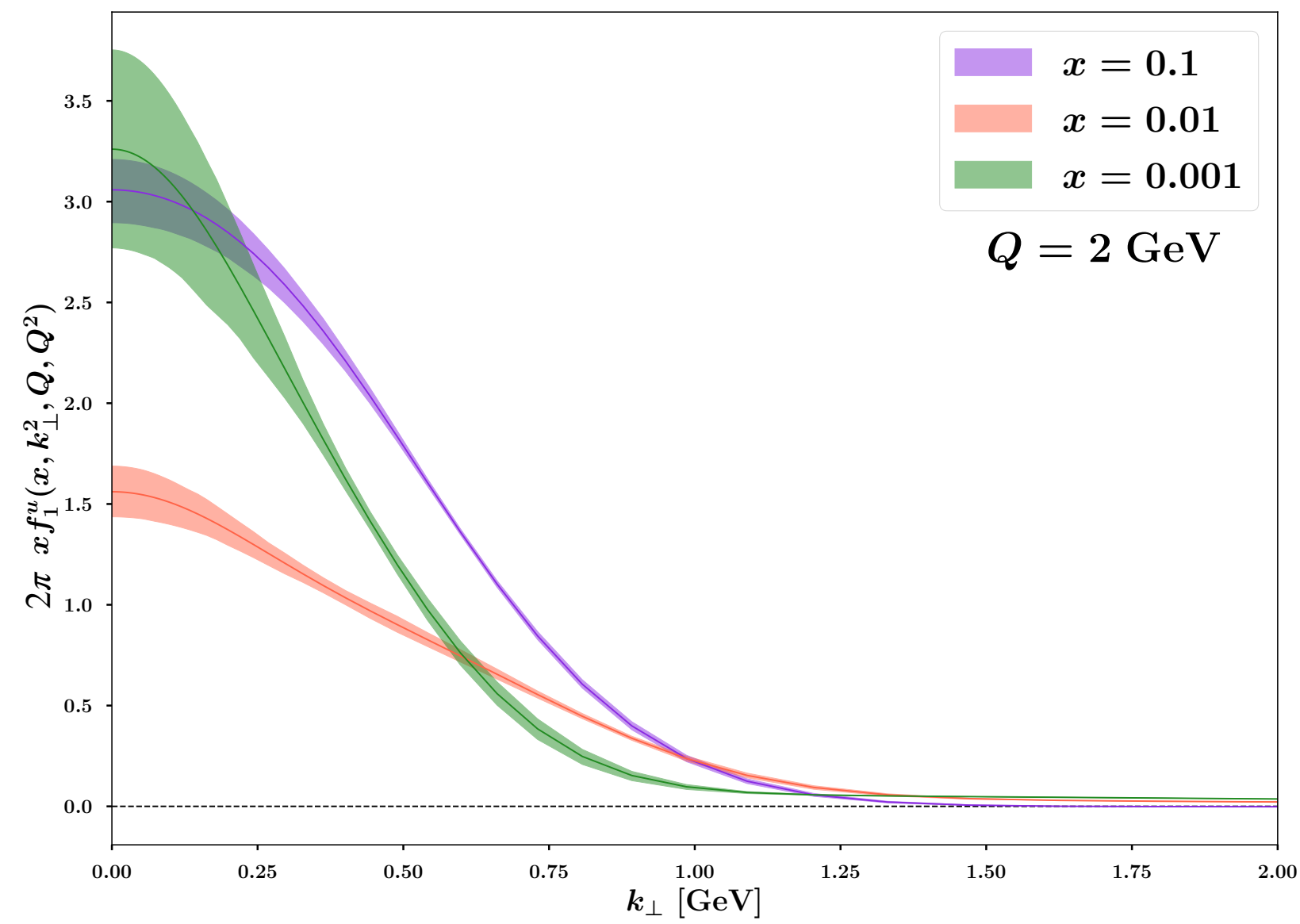
Data set	N_{dat}	χ_D^2/N_{dat}	$\chi_\lambda^2/N_{\text{dat}}$	χ_0^2/N_{dat}
DY collider total	251	1.86	0.2	2.06
DY fixed-target total	233	0.85	0.4	1.24
SIDIS total	1547	0.59	0.28	0.87
Total	2031	0.77	0.29	1.06

Very good agreement

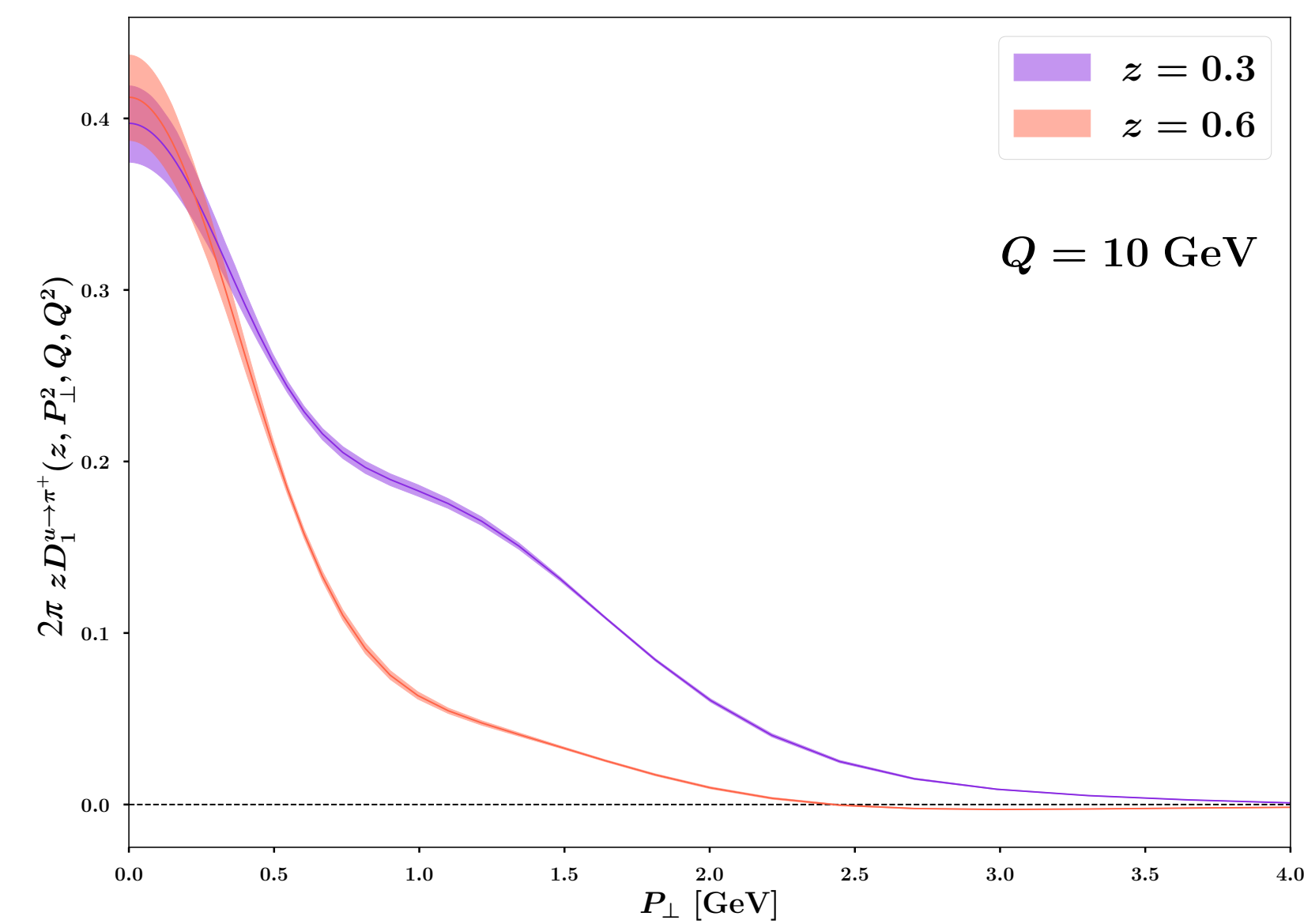
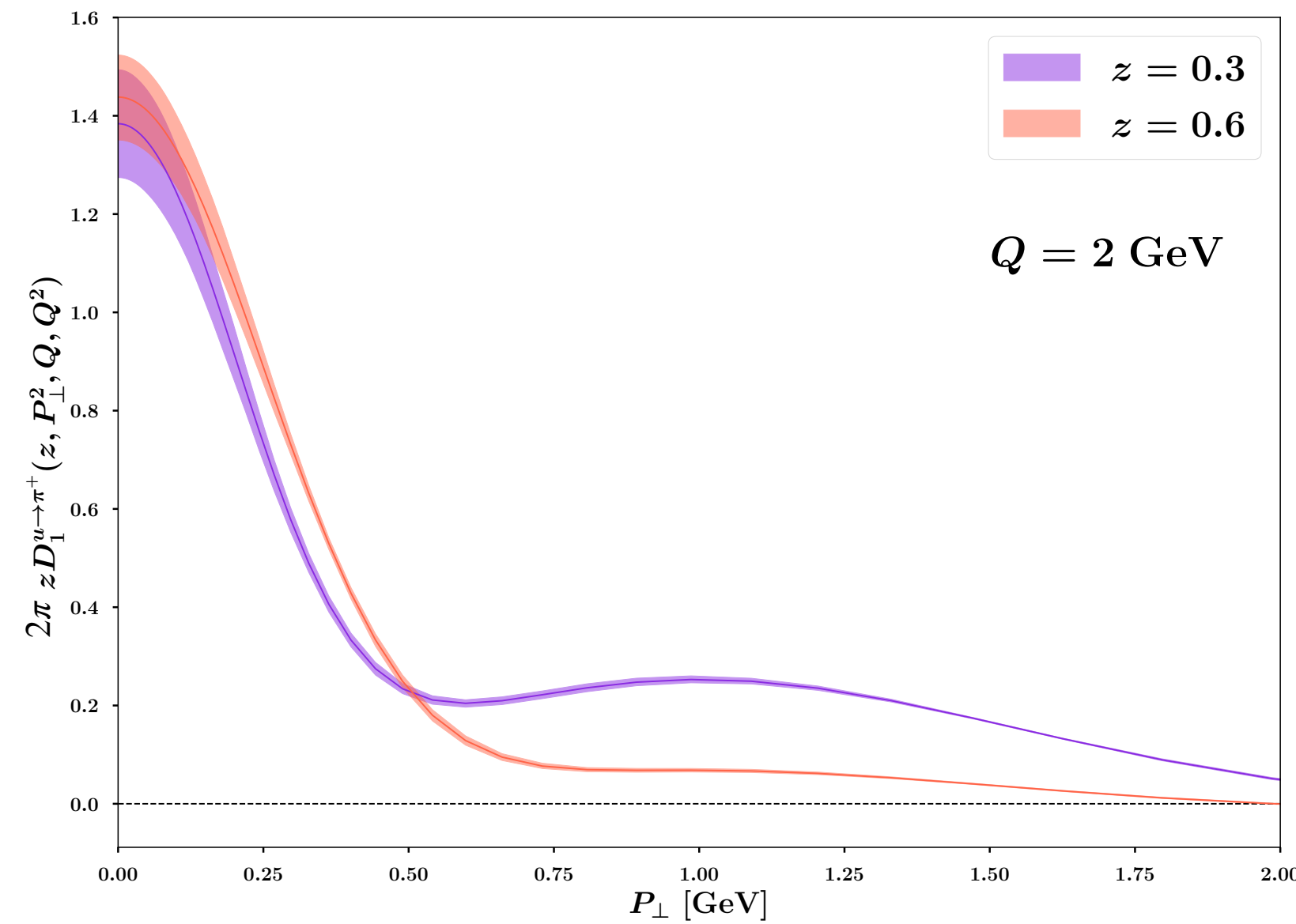


▲ $0.375 < z < 0.475$ (offset = 0.2)
 ● $0.475 < z < 0.6$ (offset = 0.1)
 ■ $0.6 < z < 0.8$ (offset = 0)

Fit results - TMDs



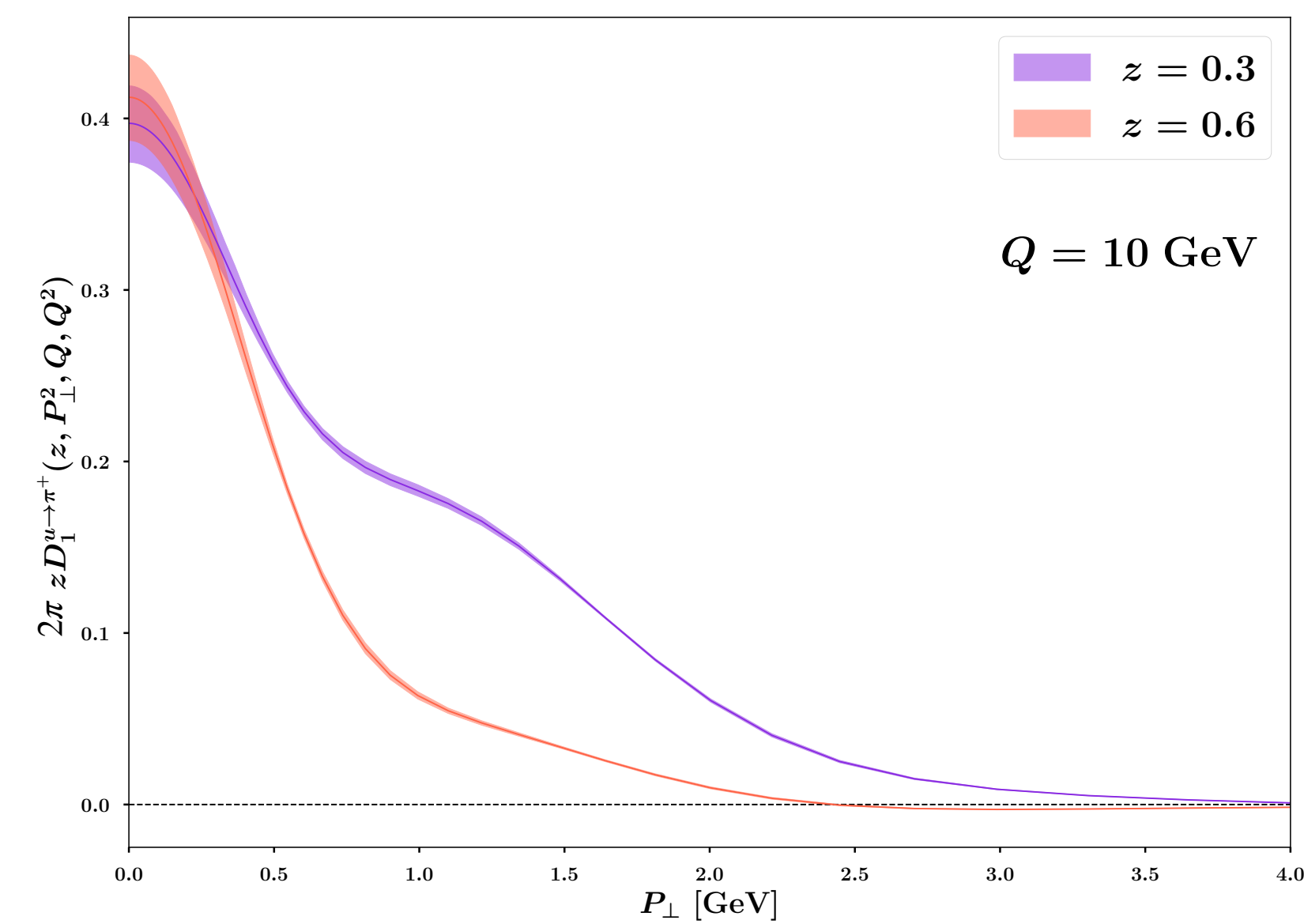
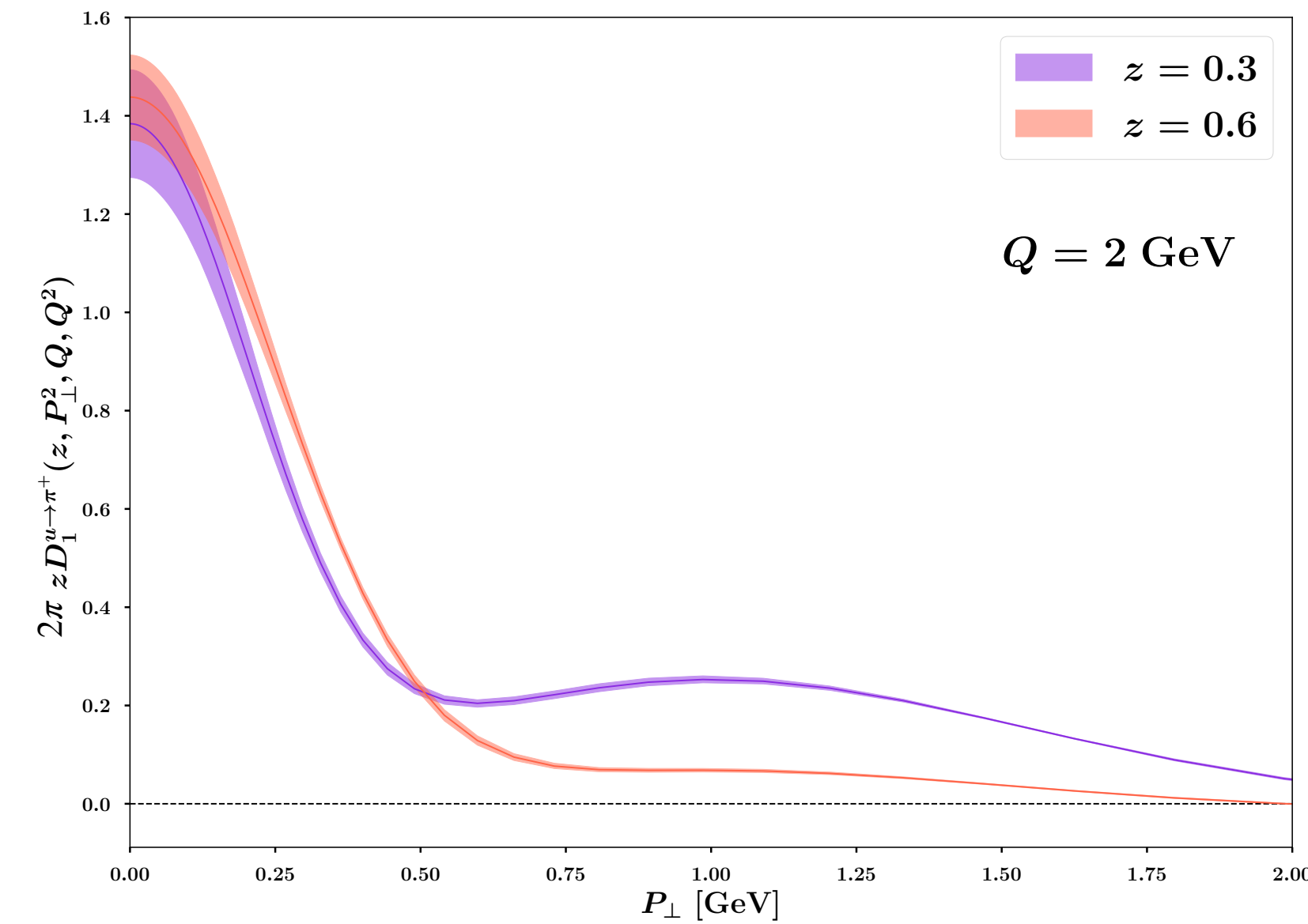
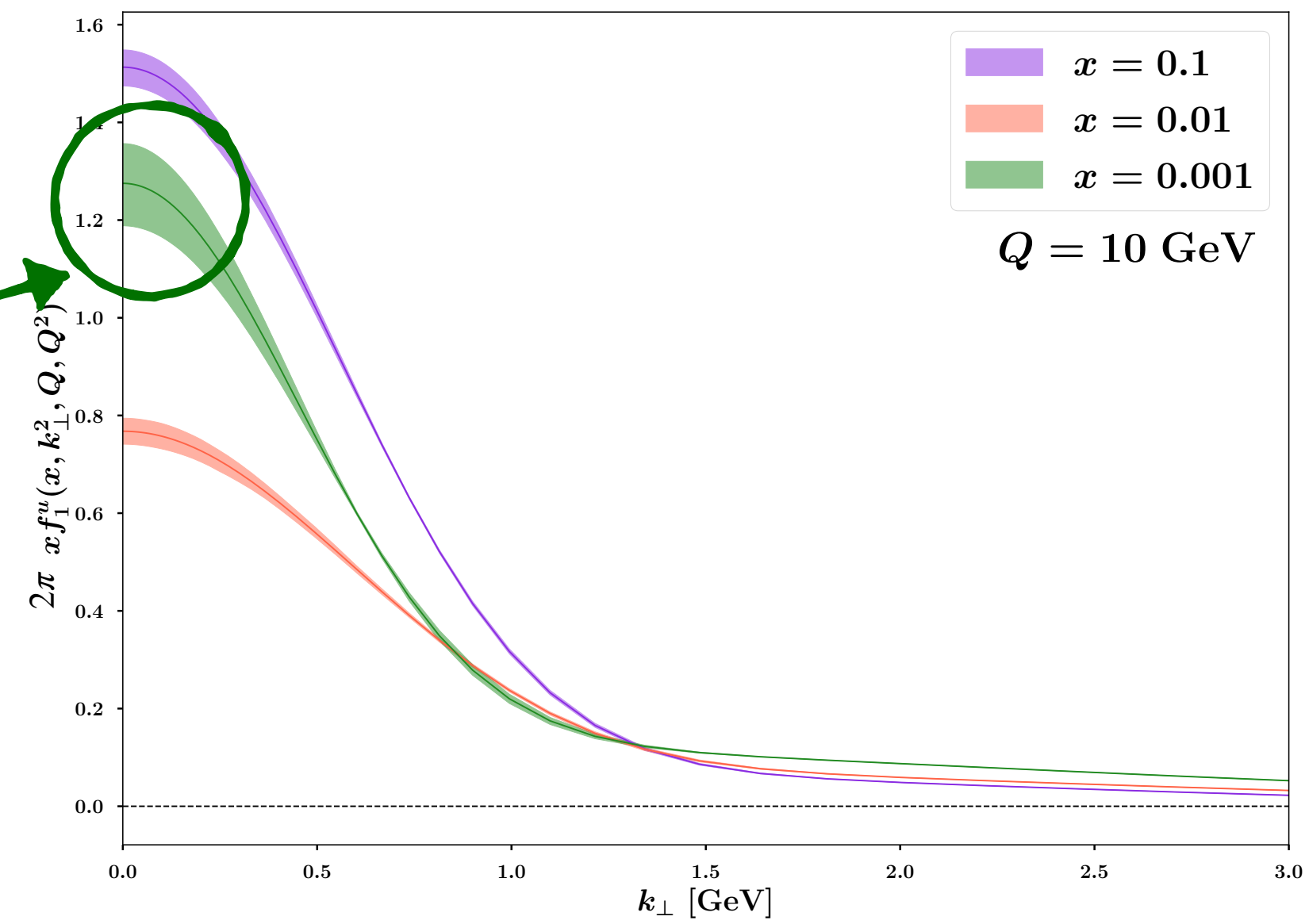
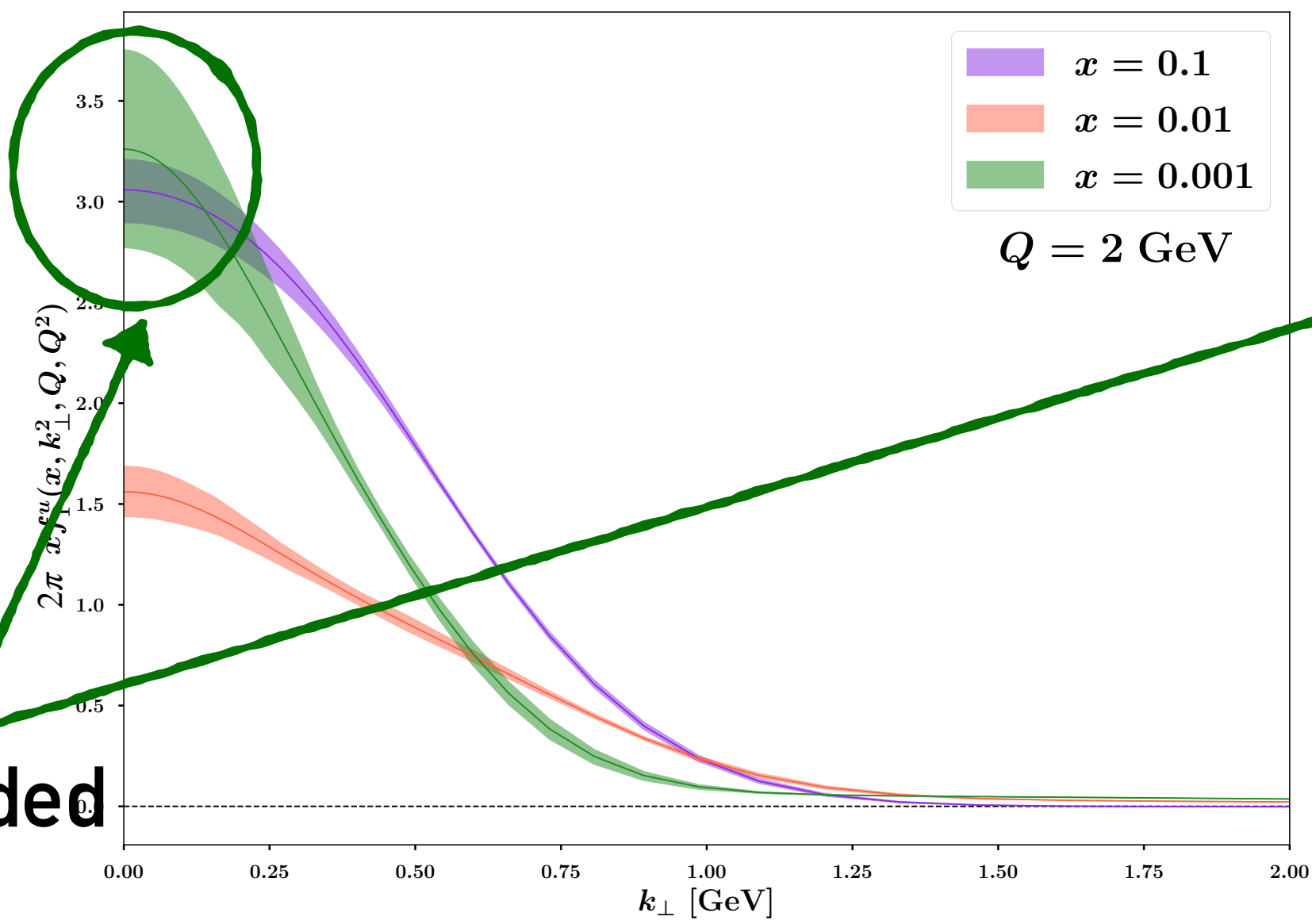
TMD PDFs



TMD FFs

Fit results - TMDs

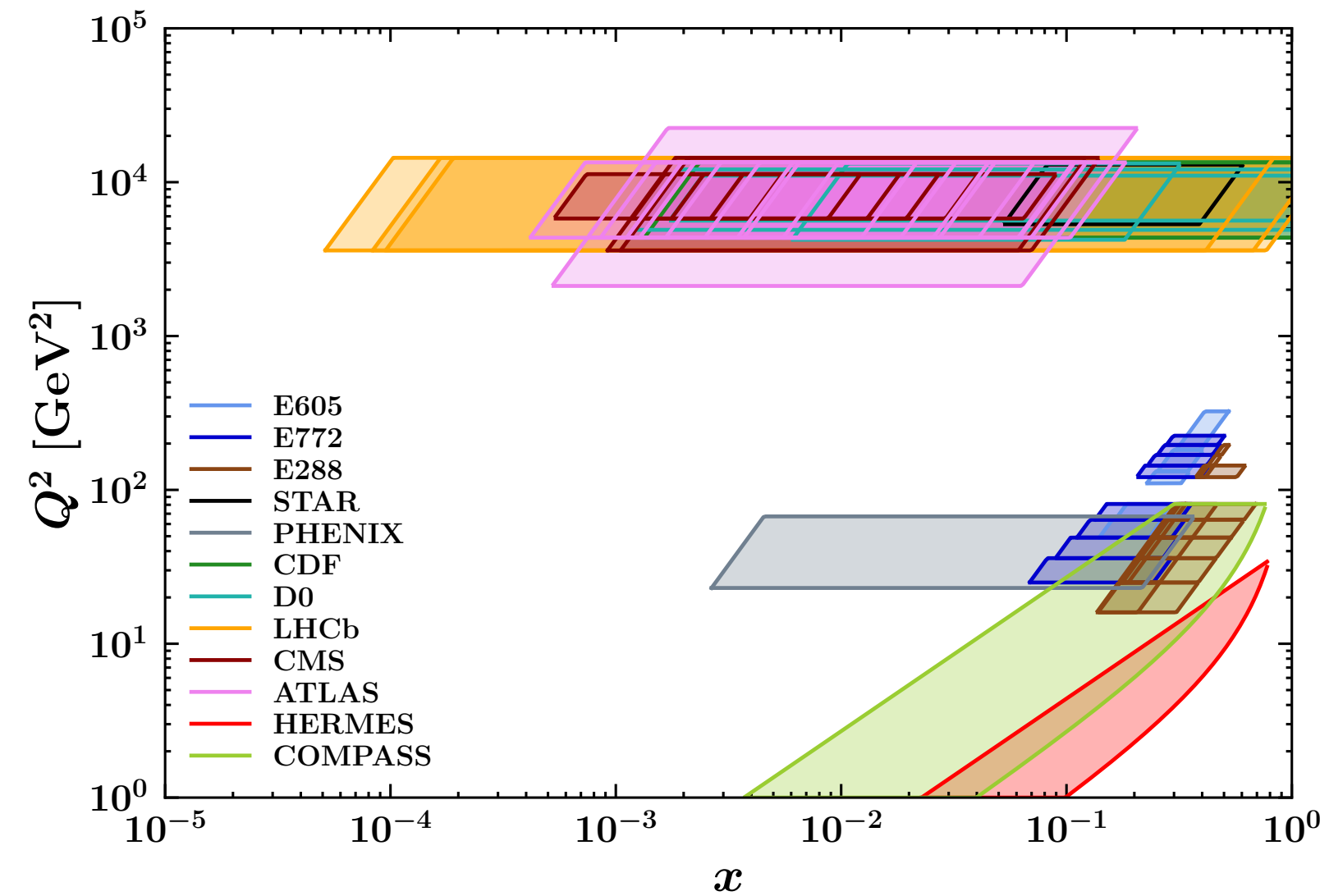
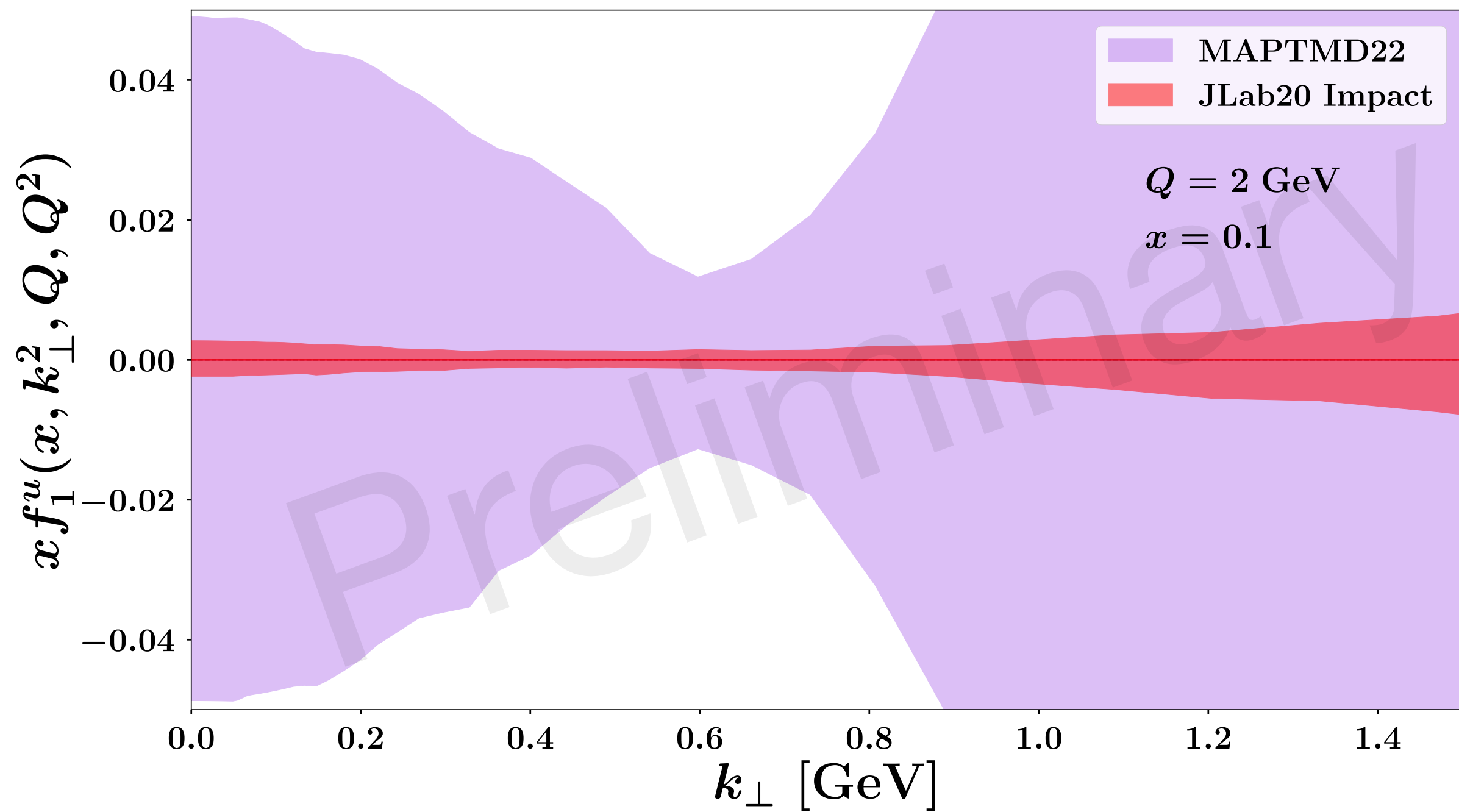
more data needed
at low x



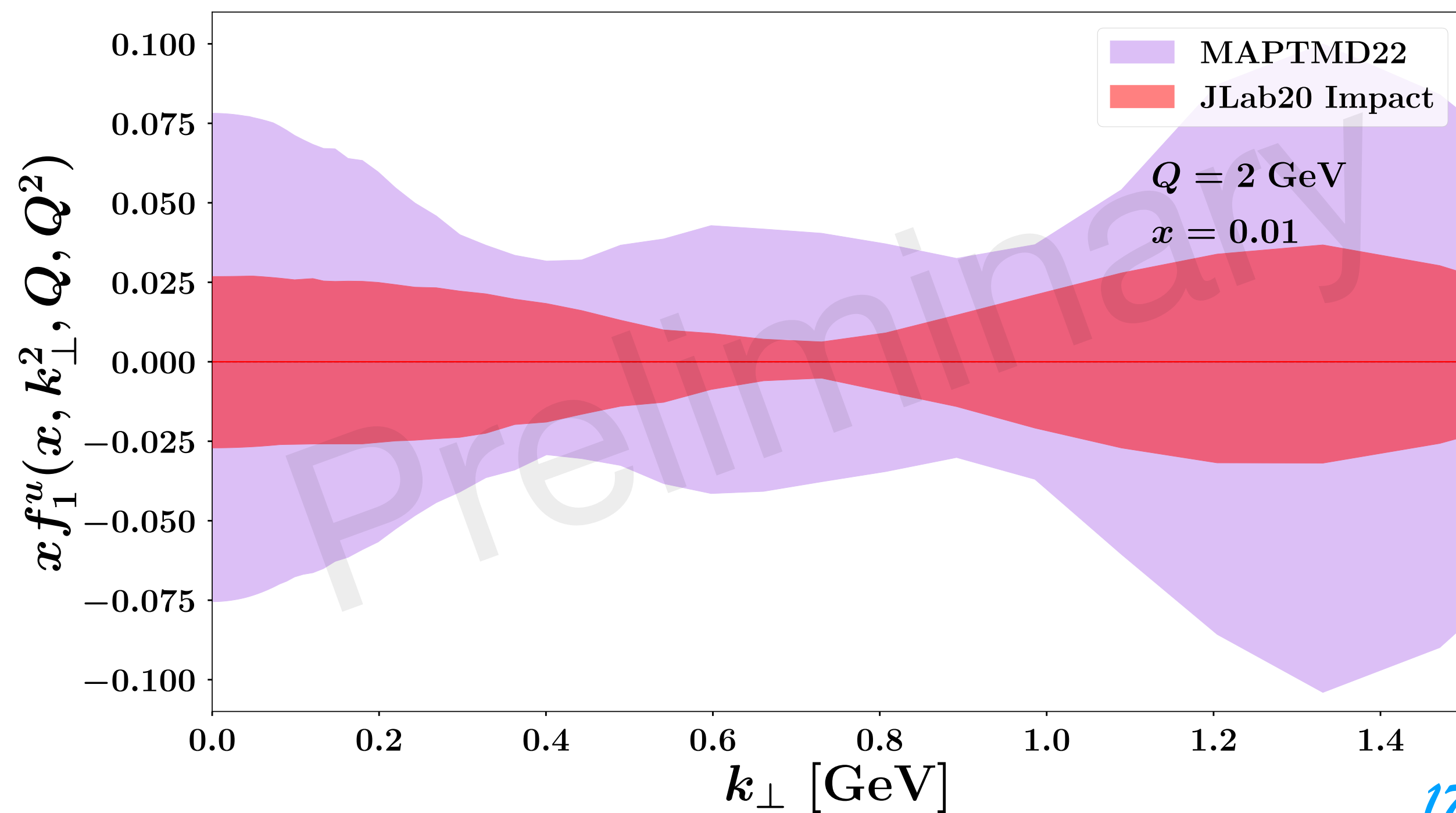
TMD PDFs

TMD FFs

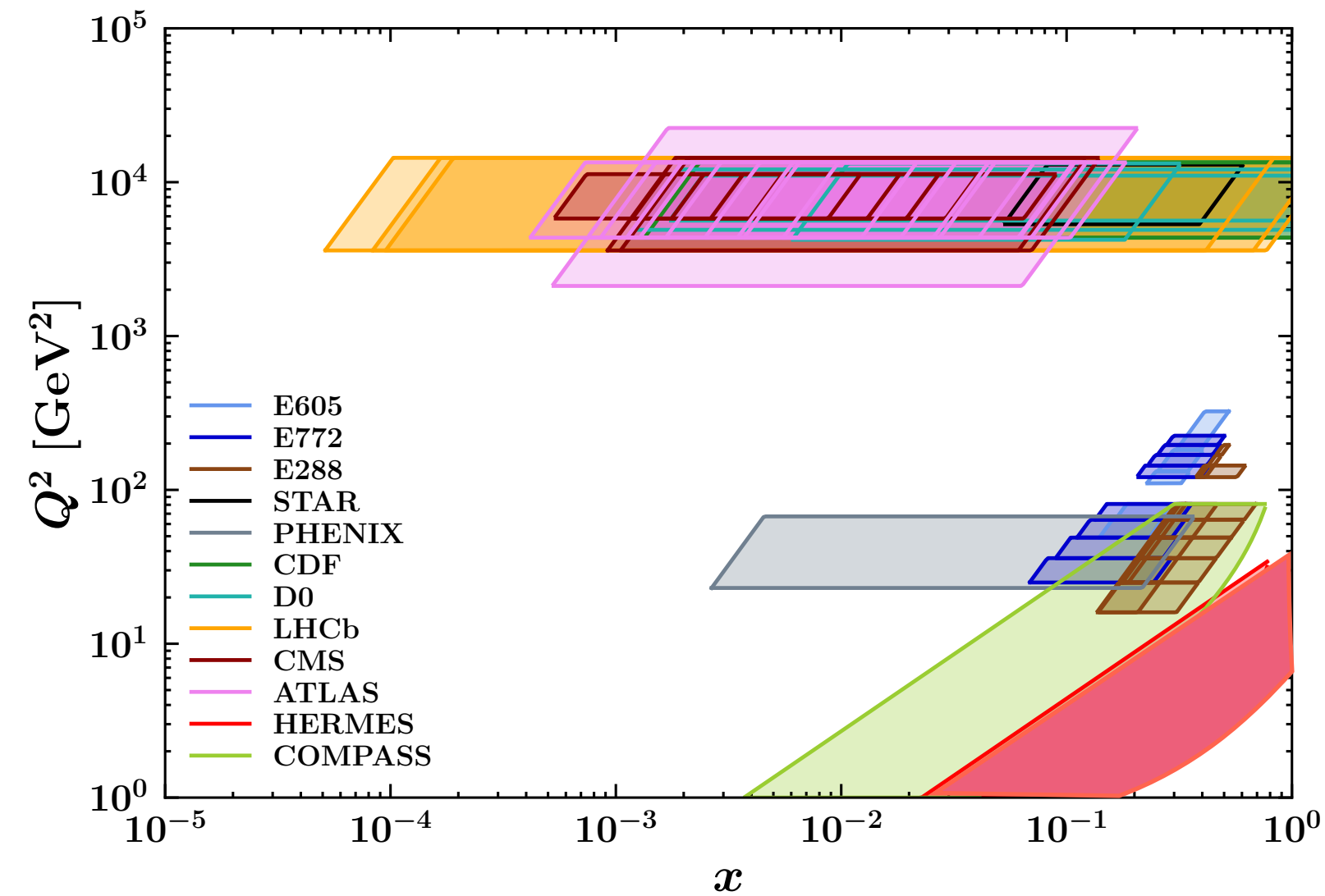
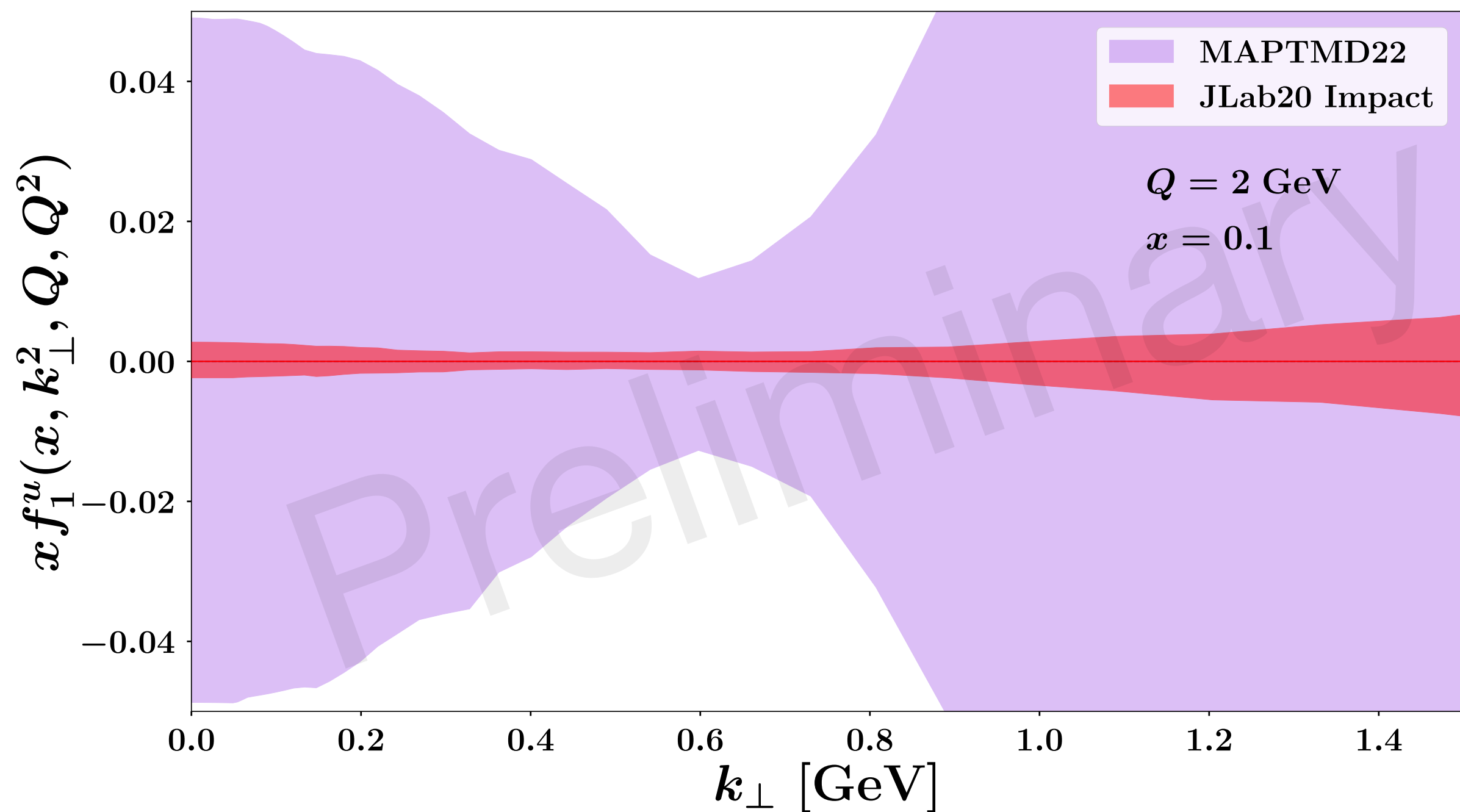
Impact studies - JLab 20+



better constrain at high x and low Q



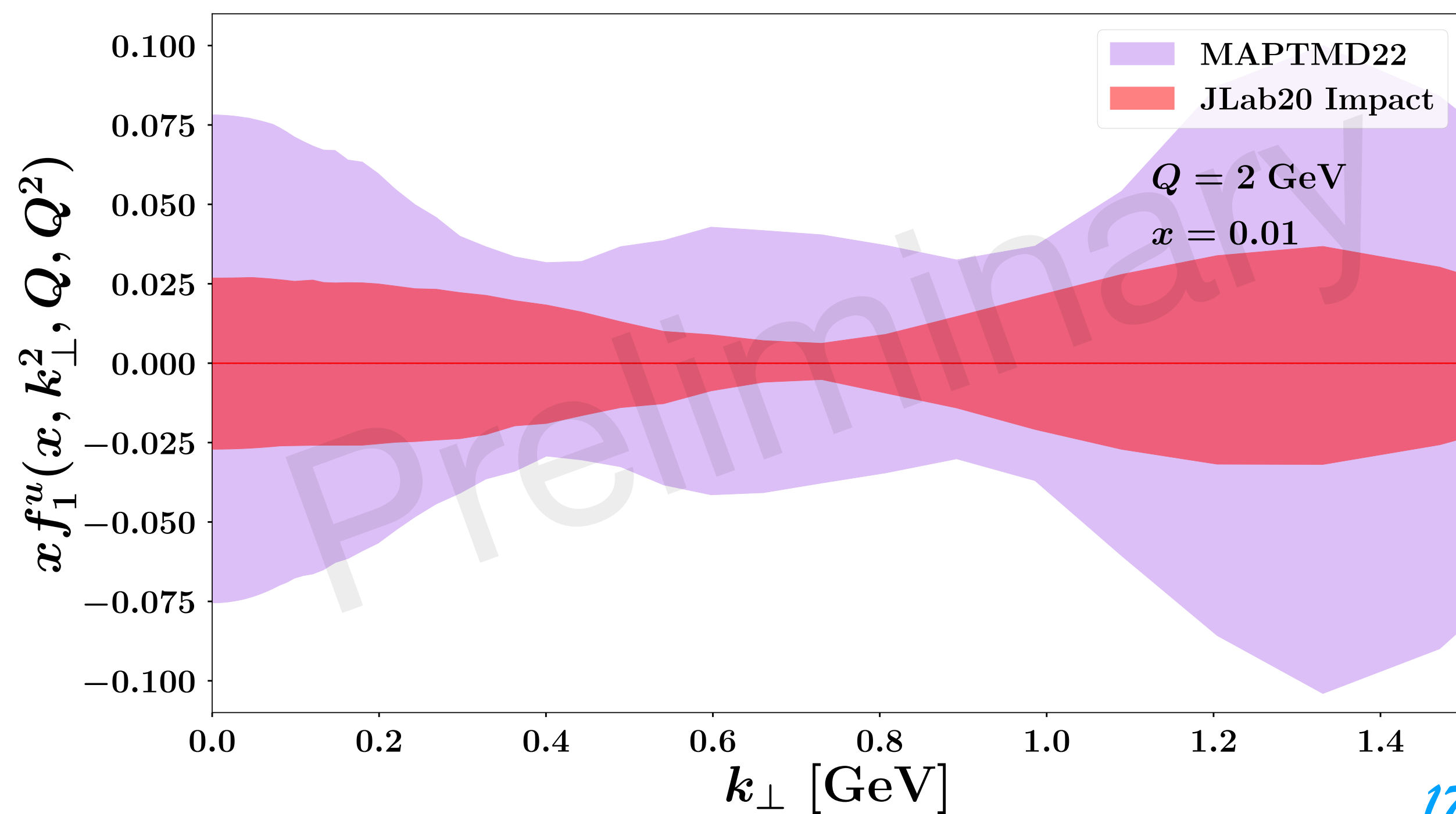
Impact studies - JLab 20+



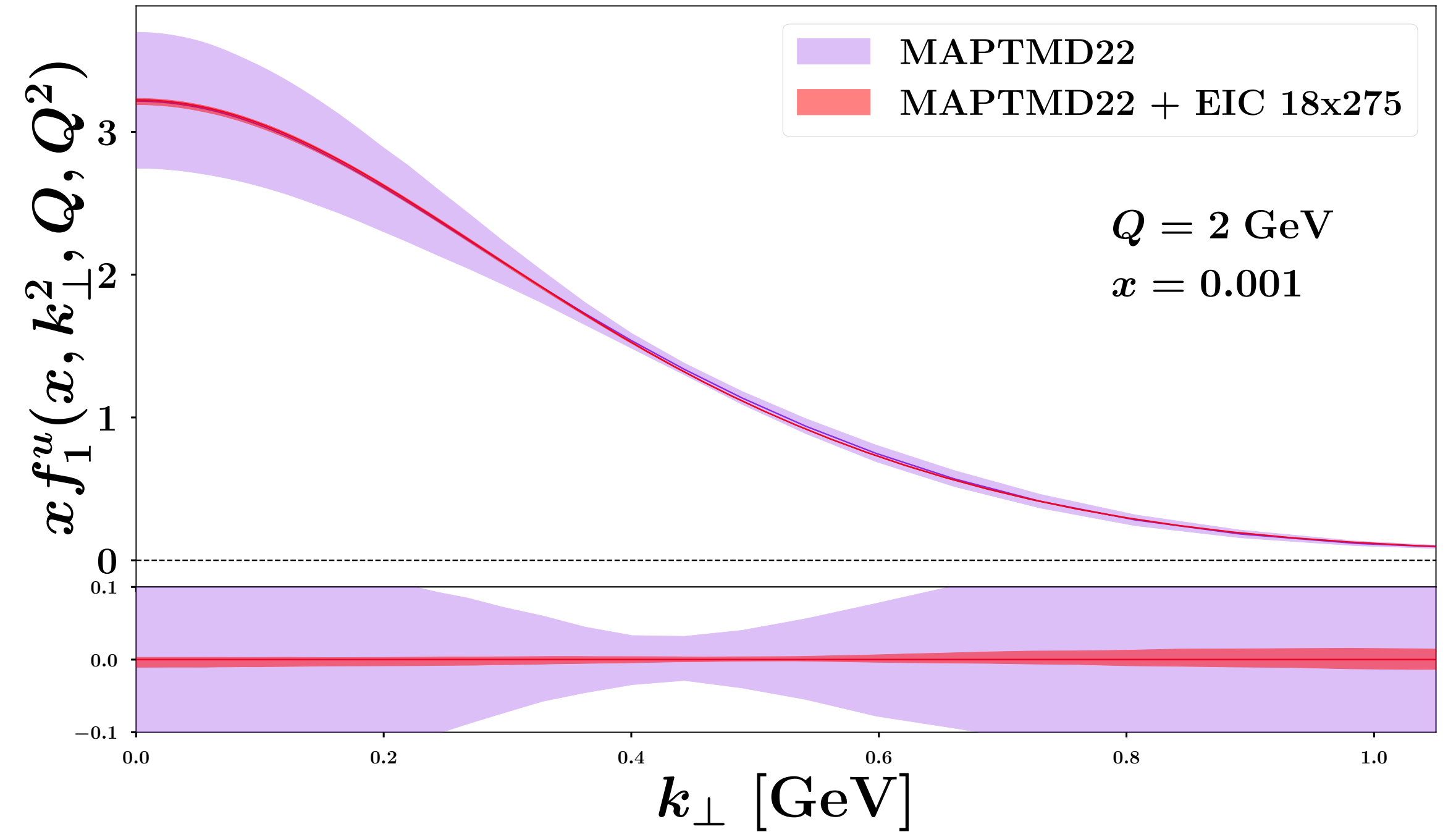
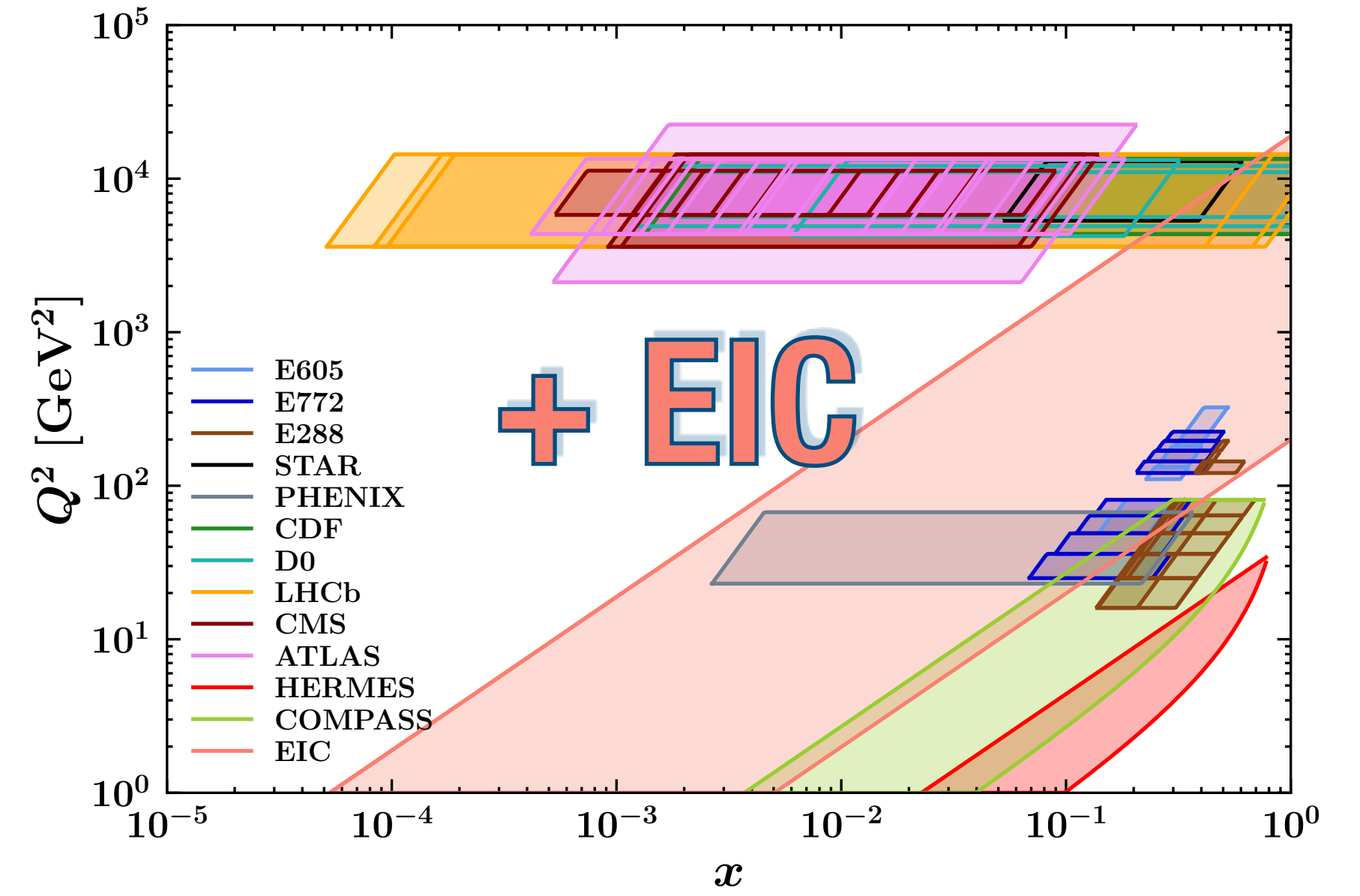
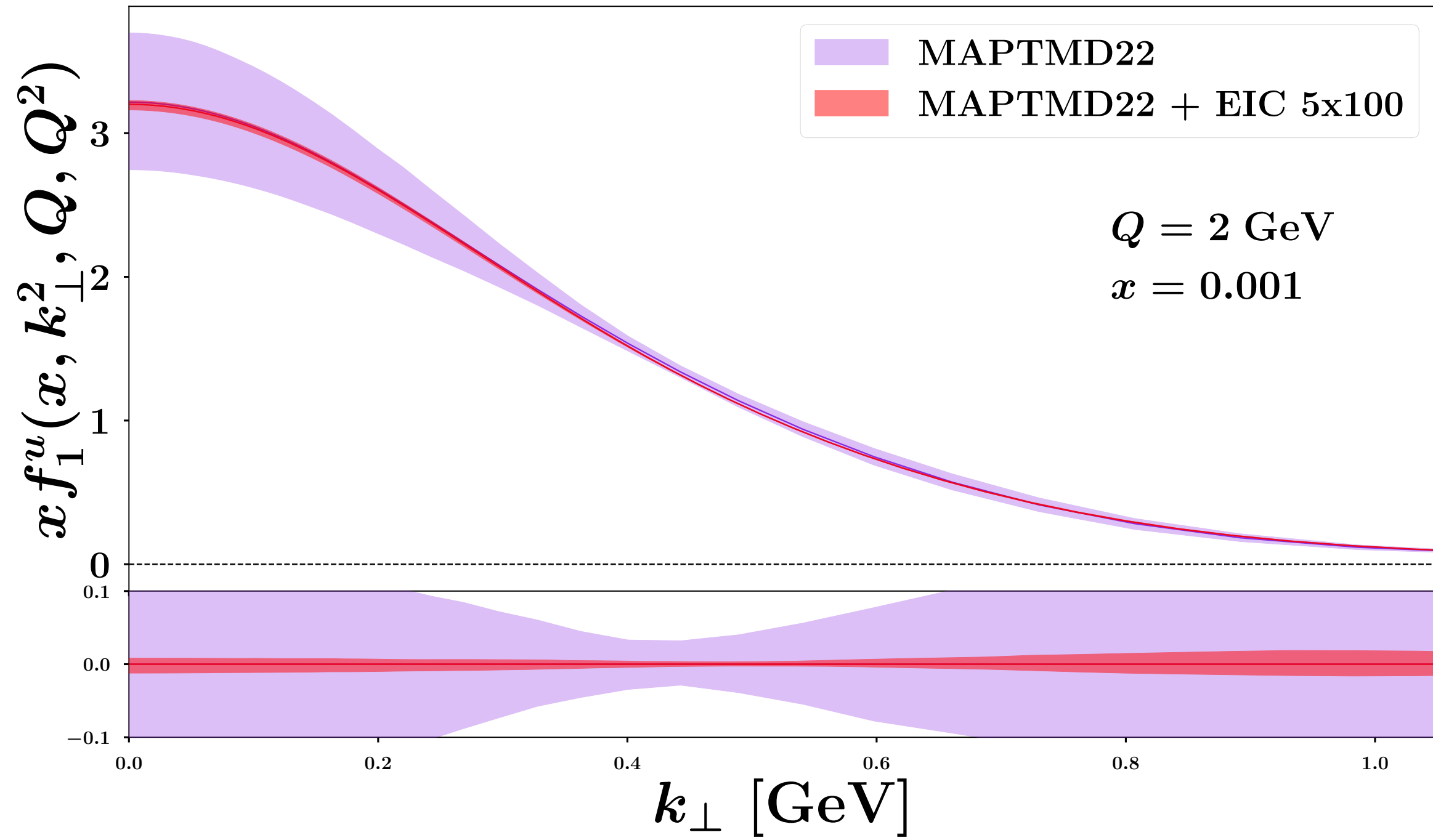
+ JLAB 20



better constrain at high x and low Q



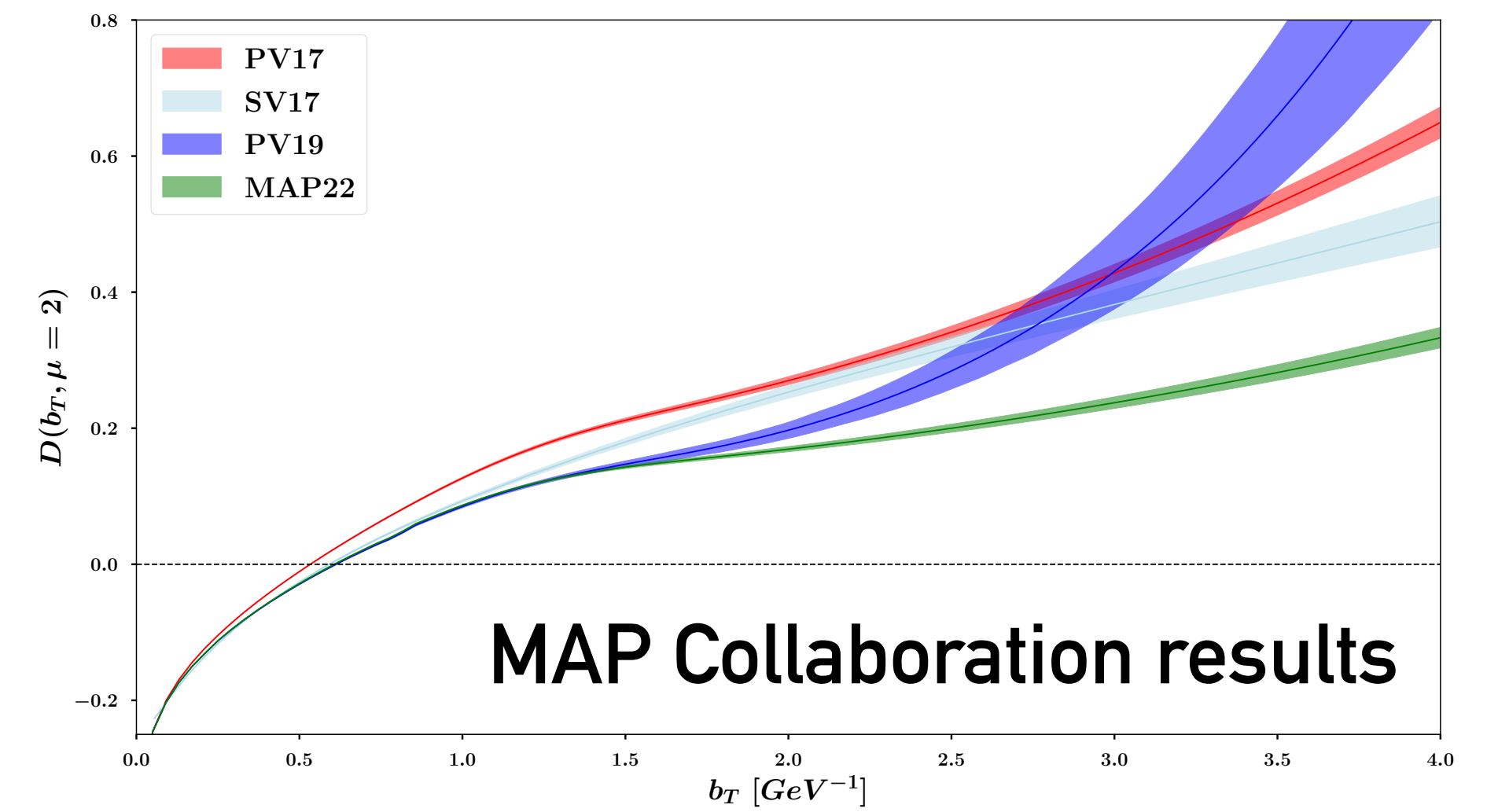
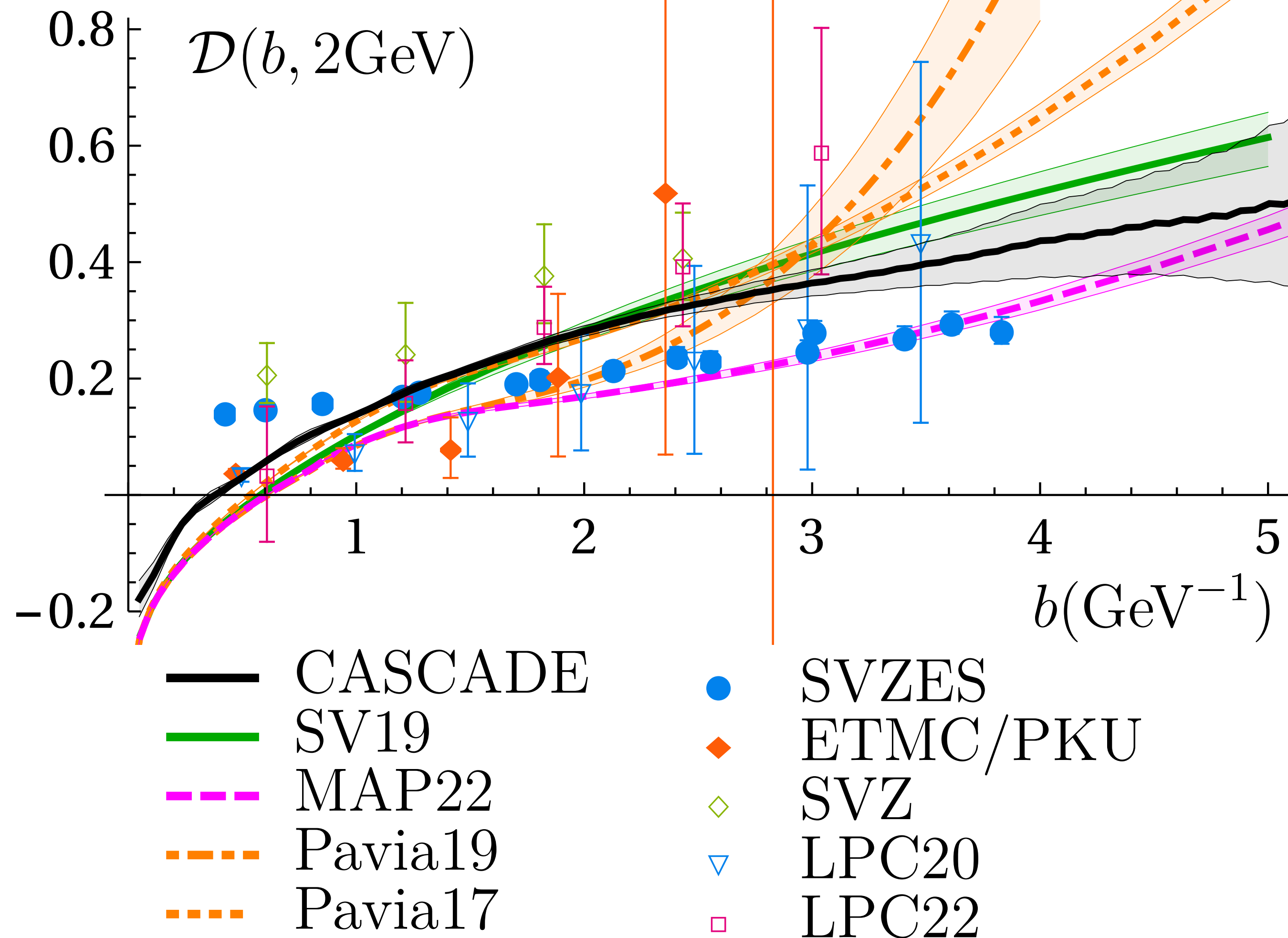
Impact studies - EIC



general reduction of the bands

Collins-Soper Kernel

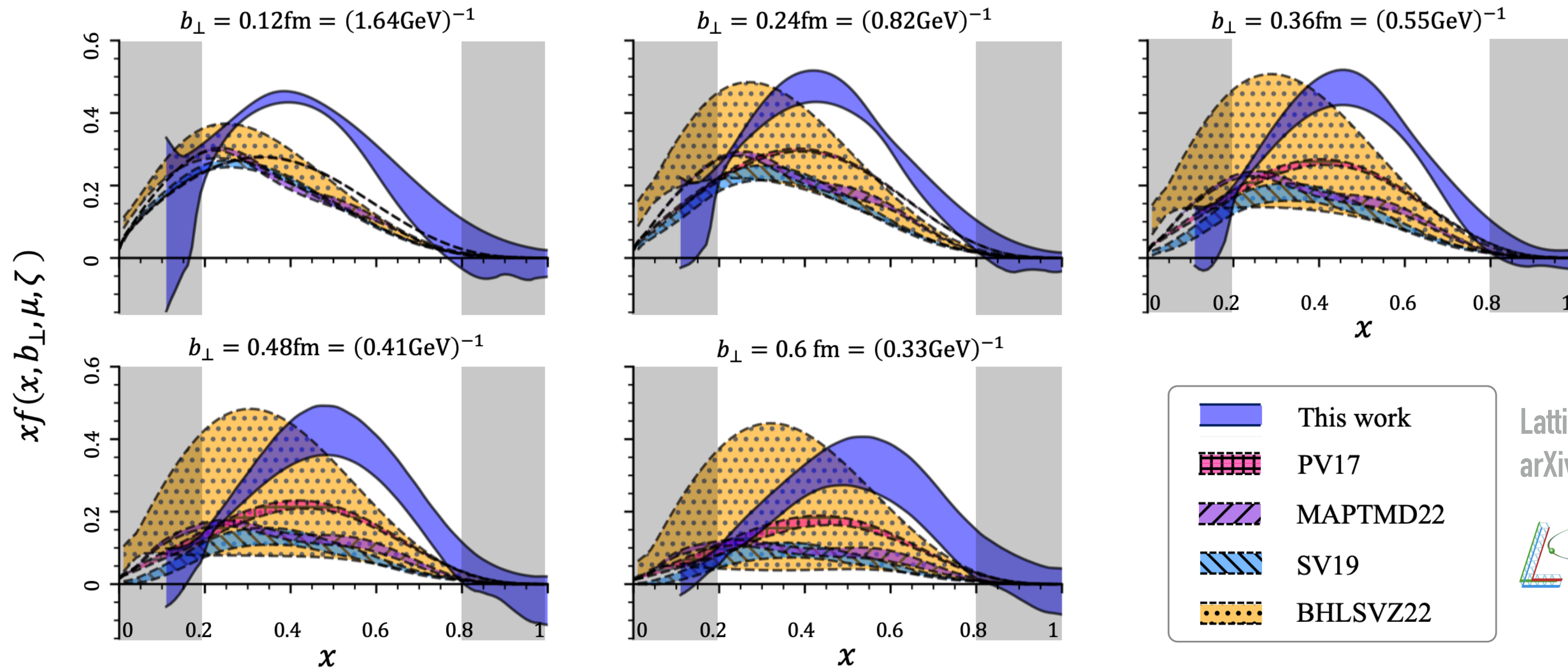
Comparison with lattice



TMD comparisons

in b_T space

📌 in the shaded grey regions LaMET predictions are not reliable



peak positions are not exactly the same

MAP22 global fit

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2031
data points

 **Global analysis** of Drell-Yan and Semi-Inclusive DIS data sets

 Perturbative accuracy: **N^3LL^-**

 **Normalization** of SIDIS multiplicities beyond NLL

 Number of parameters: **21**

 Extremely good description: **$\chi^2/N_{\text{data}} = 1.06$**

Backup

Source of W term suppression

Hard factor

$$\mathcal{H}_{ab}^{\text{SIDIS}}(Q, Q) = e_a^2 \delta_{ab} \left(1 + \frac{\alpha_S}{4\pi} C_F \left(-16 + \frac{\pi^2}{3} \right) \right)$$

Source of W term suppression

Hard factor

$$\mathcal{H}_{ab}^{\text{SIDIS}}(Q, Q) = e_a^2 \delta_{ab} \left(1 + \frac{\alpha_S}{4\pi} C_F \left(-16 + \frac{\pi^2}{3} \right) \right)$$

introducing $\mathcal{O}(\alpha_s)$ terms



reduces the structure function to about 60% of its original value.

Source of W term suppression

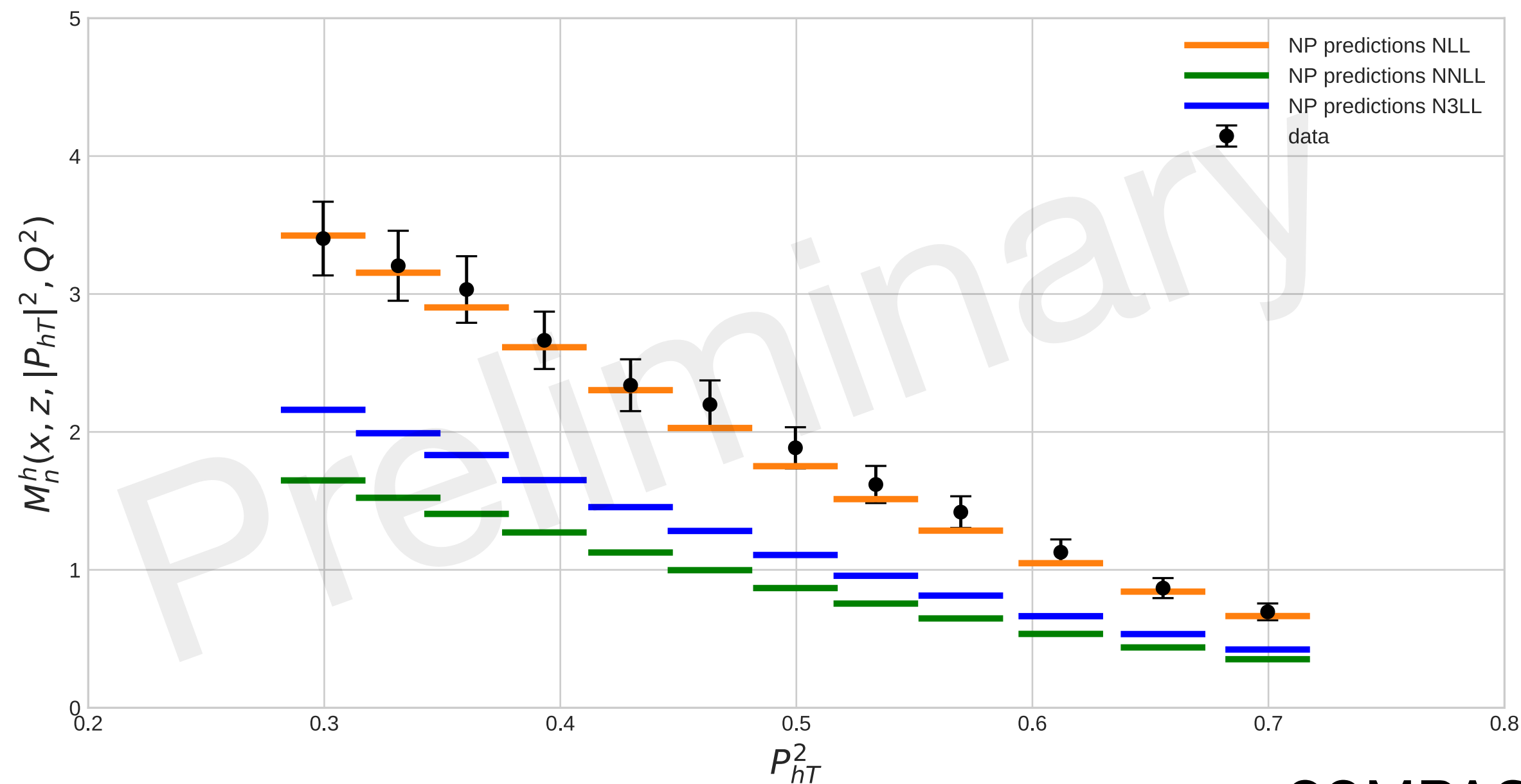
Hard factor

$$\mathcal{H}_{ab}^{\text{SIDIS}}(Q, Q) = e_a^2 \delta_{ab} \left(1 + \frac{\alpha_S}{4\pi} C_F \left(-16 + \frac{\pi^2}{3} \right) \right)$$

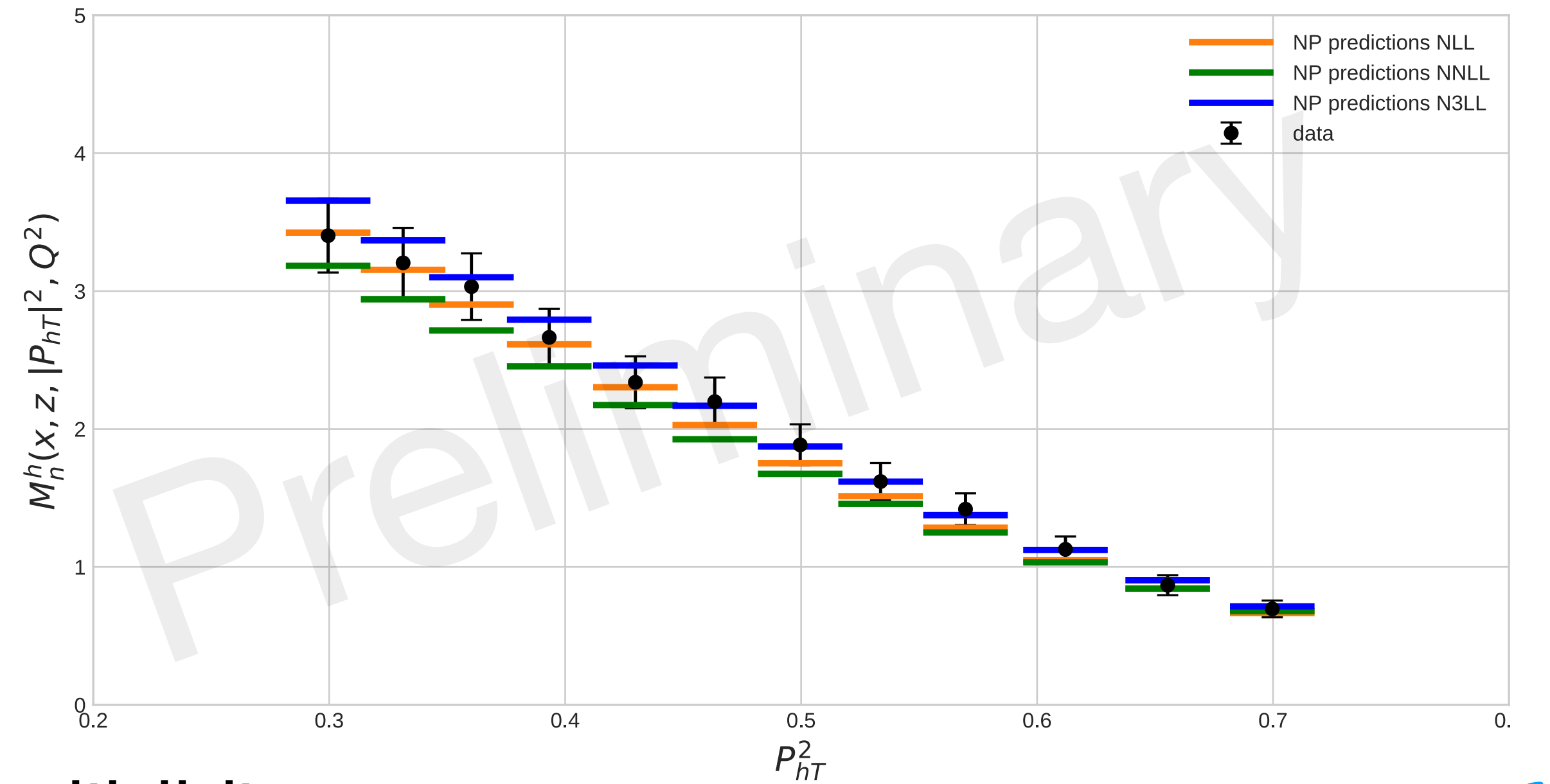
introducing $\mathcal{O}(\alpha_s)$ terms

reduces the structure function to about 60% of its original value.

Full Hard Factor

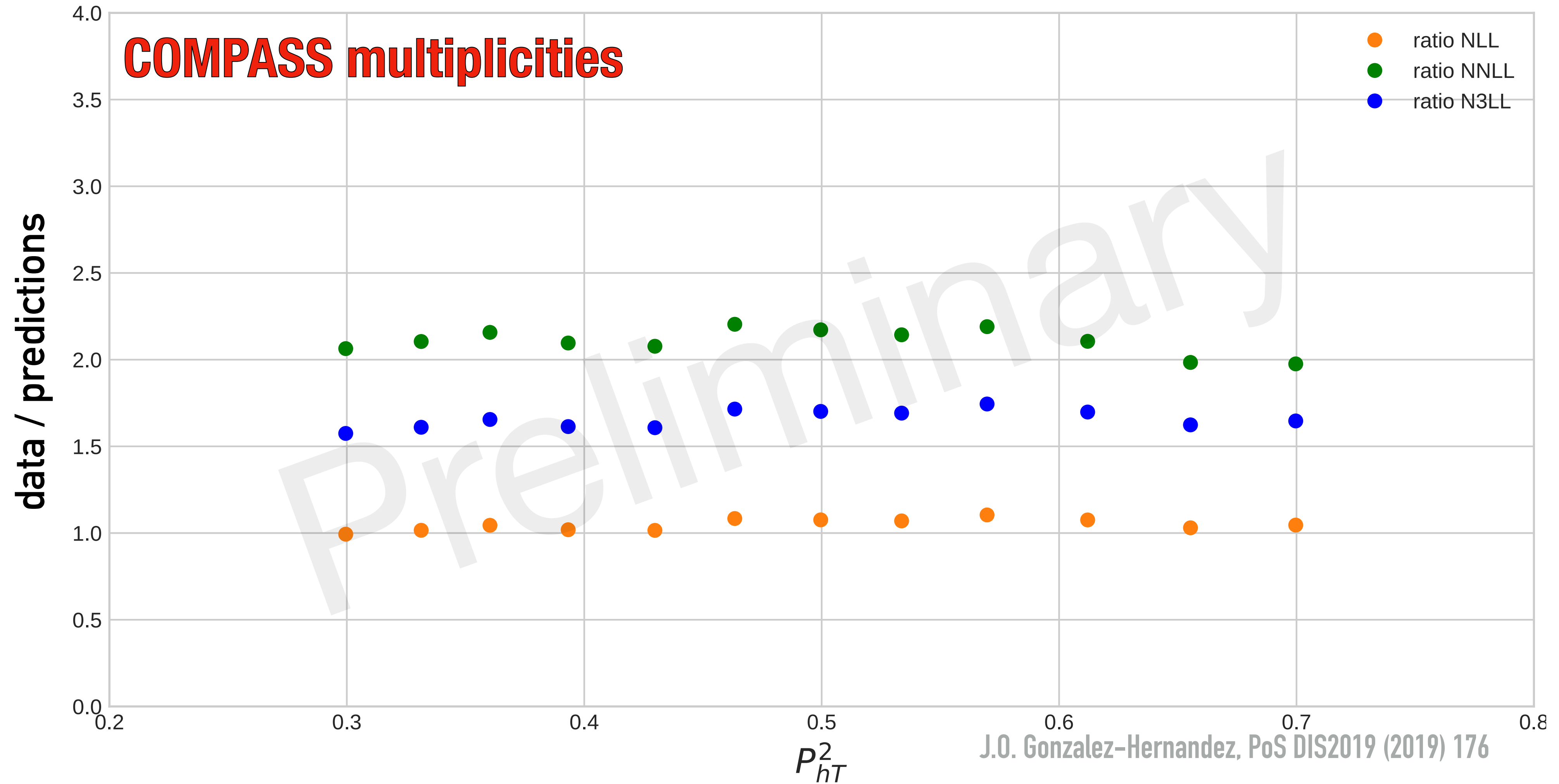


Hard Factor = 1



COMPASS multiplicity

Normalization of SIDIS multiplicities



The discrepancy amounts to an almost **constant factor**

Normalization of SIDIS multiplicities

Introduction of a normalization prefactor

(at low P_{hT})

$$\int W \Big|_{O(\alpha_S)} \quad \text{should} \quad \sim \quad \frac{d\sigma^h}{dx dQ^2 dz} \Big|_{O(\alpha_S)}$$

integral of the TMD formula

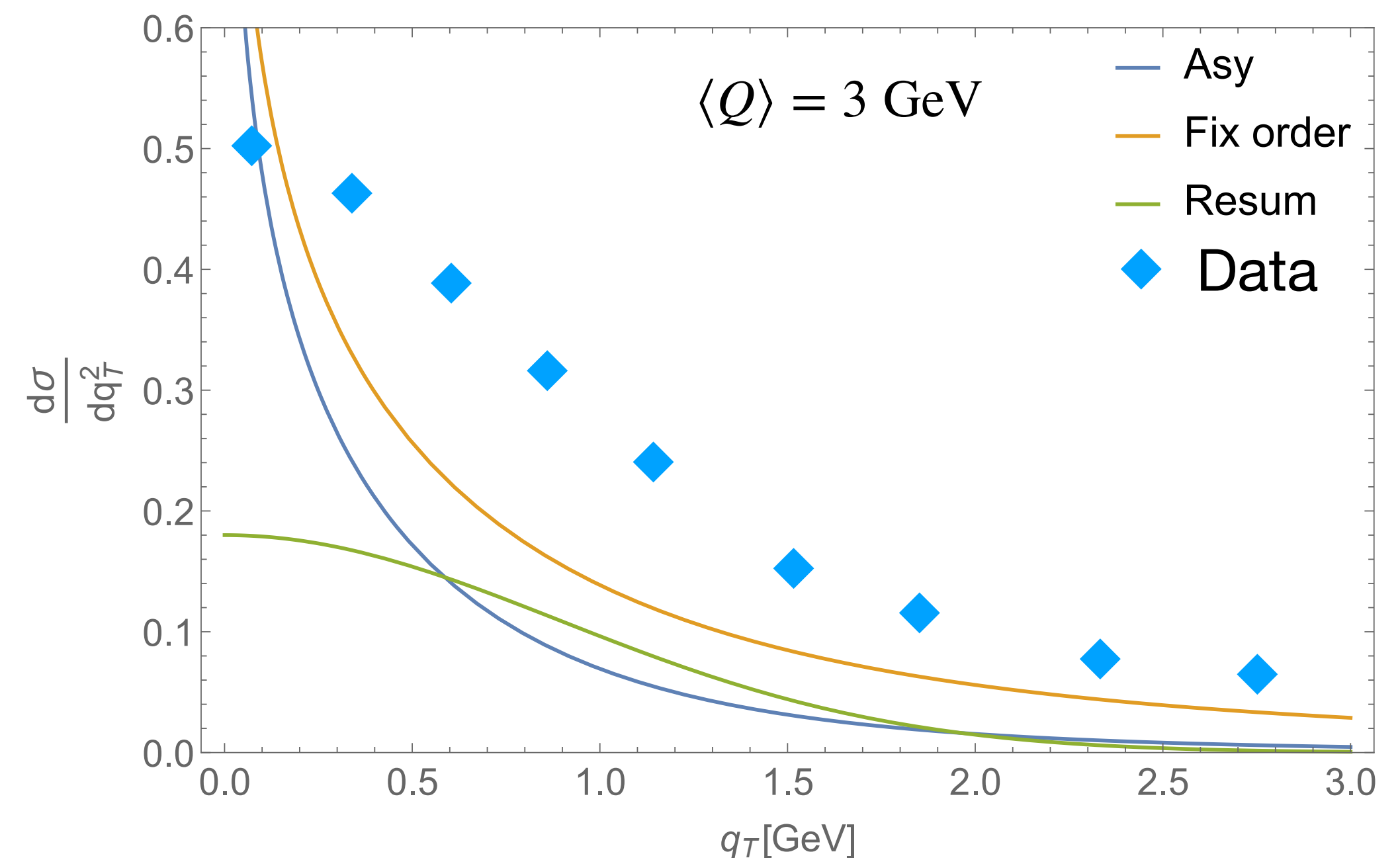
collinear cross section

BUT

this is not the case in the experimental regions under consideration

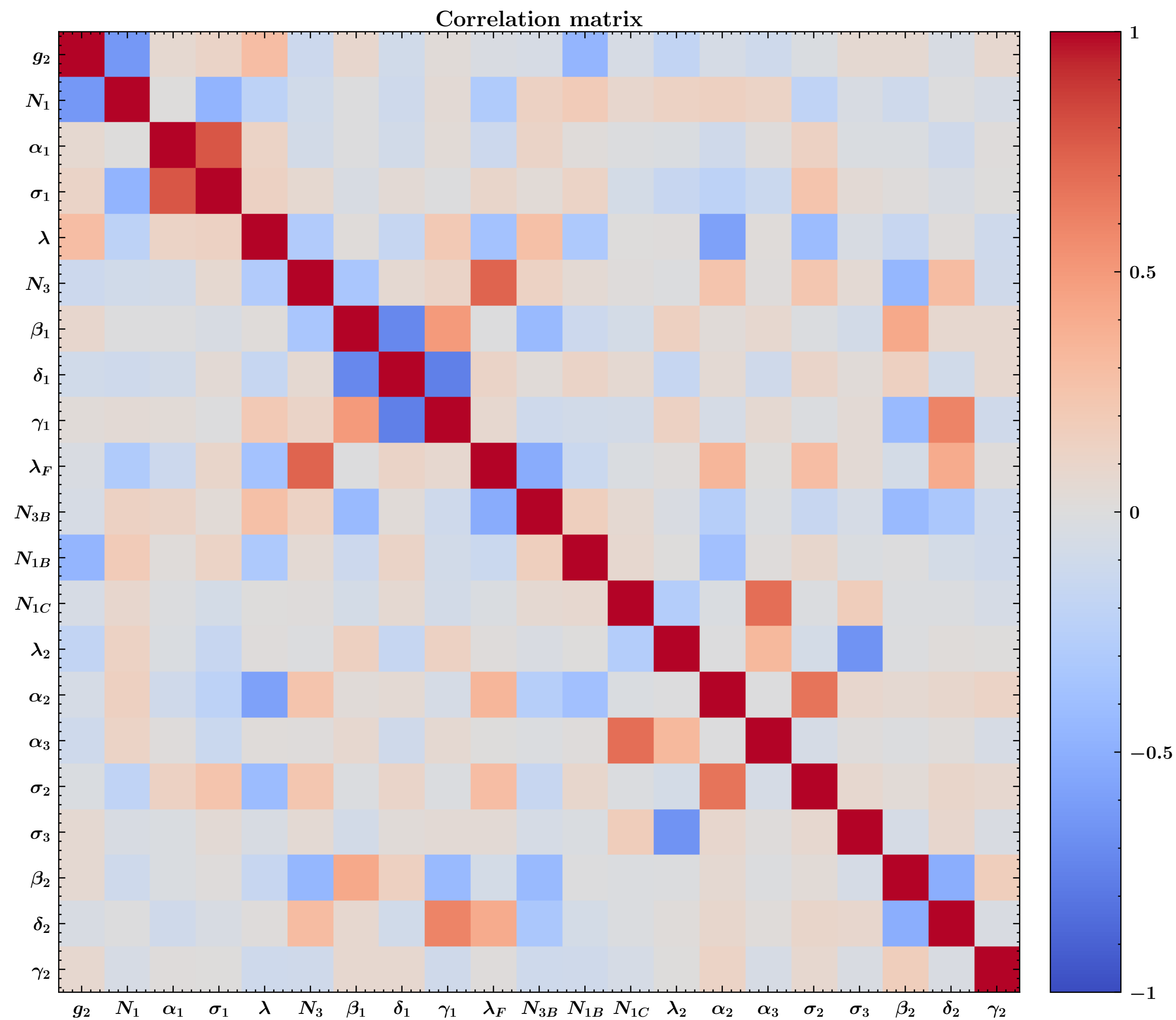
and we see

$$\int W \Big|_{O(\alpha_S)} \ll \frac{d\sigma^h}{dx dQ^2 dz} \Big|_{O(\alpha_S)}$$



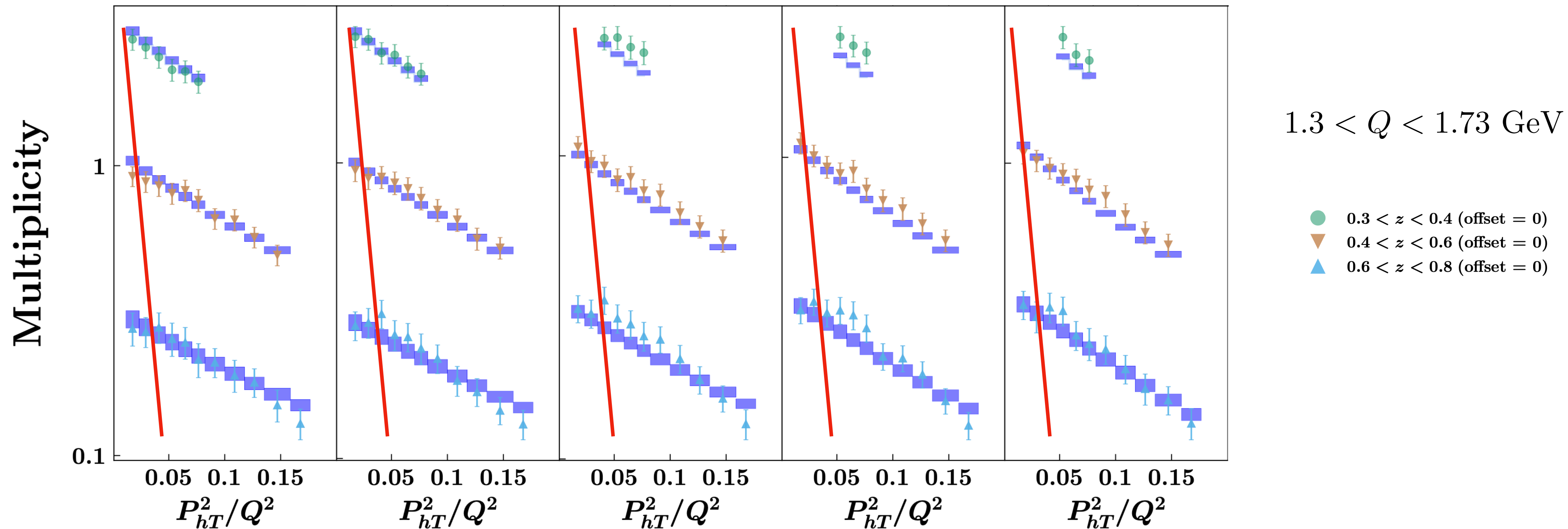
Fit results: correlation matrix

250 Montecarlo replicas



SIDIS cut for data selection

COMPASS multiplicities



$$P_{hT}|_{\max} = \min[\min[0.2Q, 0.5zQ] + 0.3 \text{ GeV}, zQ] \quad \text{vs} \quad \left. \frac{P_{hT}}{zQ} \right|_{\max} = 0.25$$

Cut q_T/Q for SIDIS dataset

