



# TMDs with flavor dependence: the MAPTMD24 extraction

arXiv:2405.138833

Matteo Cerutti - MAP Collaboration

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EIC/ePIC Collaboration Meeting



# Transverse-Momentum Distributions (TMDs)

3-*dimensional map* of the internal structure of the nucleon

Non-collinear framework

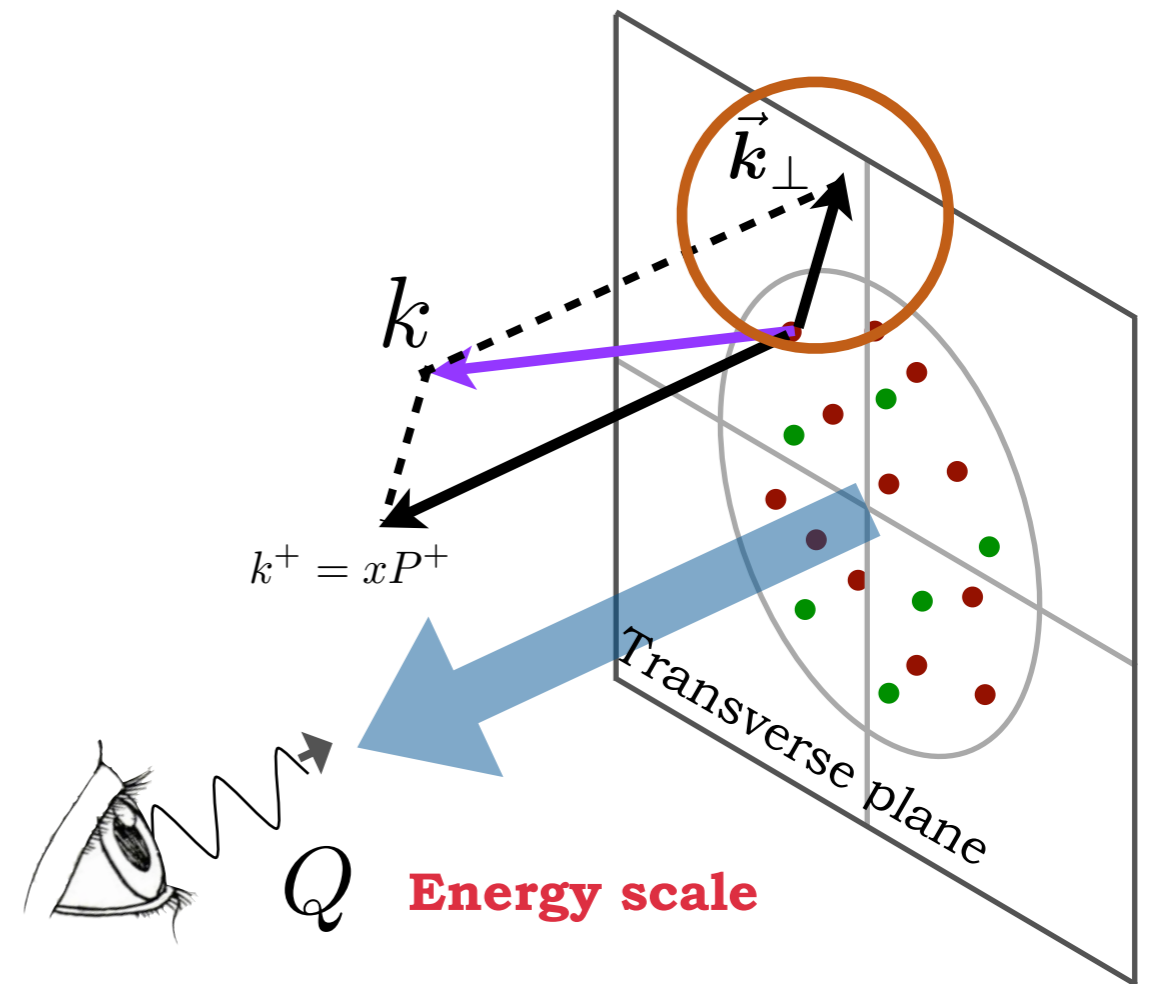
Quark Polarization

Nucleon Pol.

	U	L	T
U	$f_1$		$h_1^\perp$
L		$g_1$	$h_{1L}^\perp$
T	$f_{1T}^\perp$	$g_{1T}$	$h_1 h_{1T}^\perp$

Time-reversal odd

Time-reversal even



TMD PDFs

$$F(x, \mathbf{k}_\perp^2, \mu, \zeta)$$

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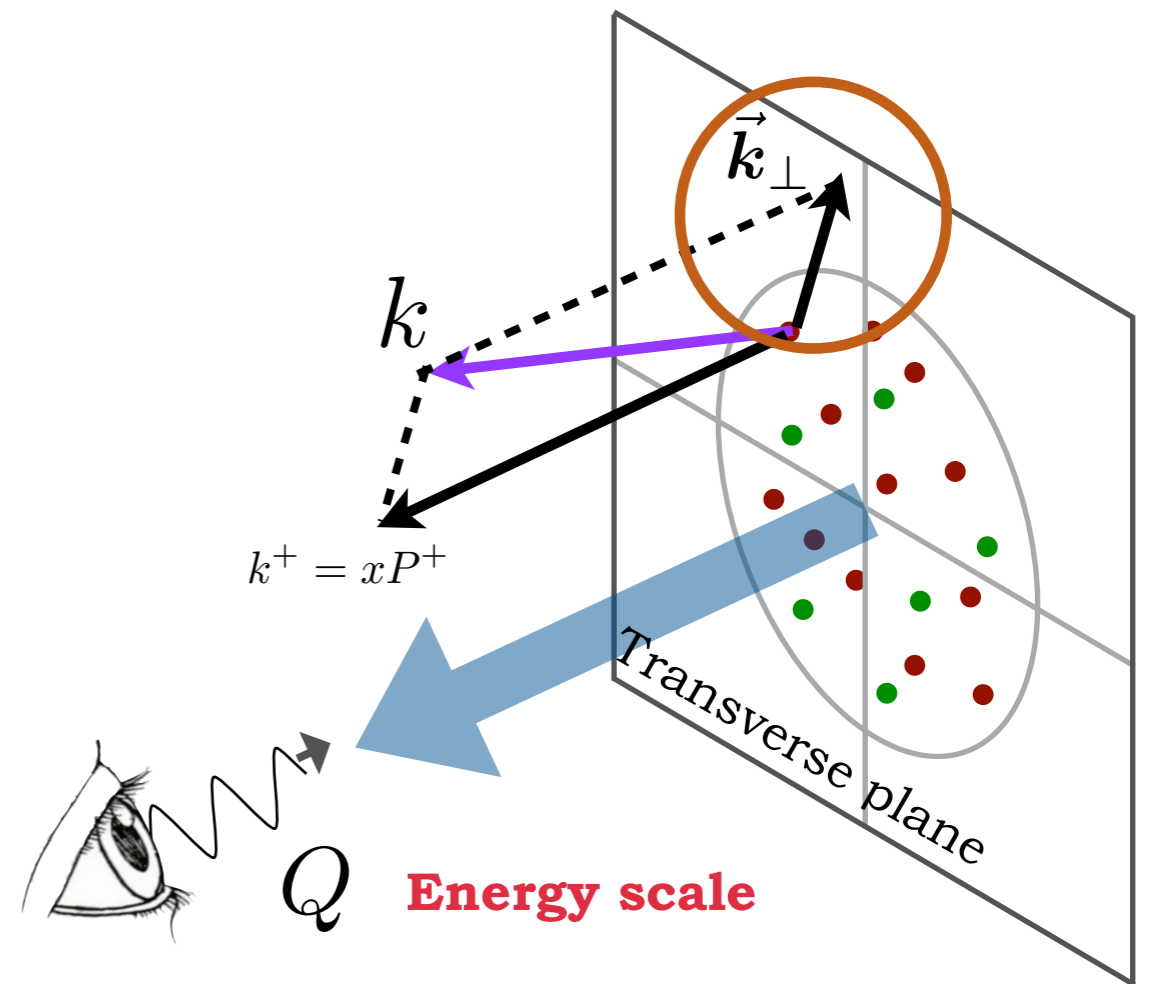
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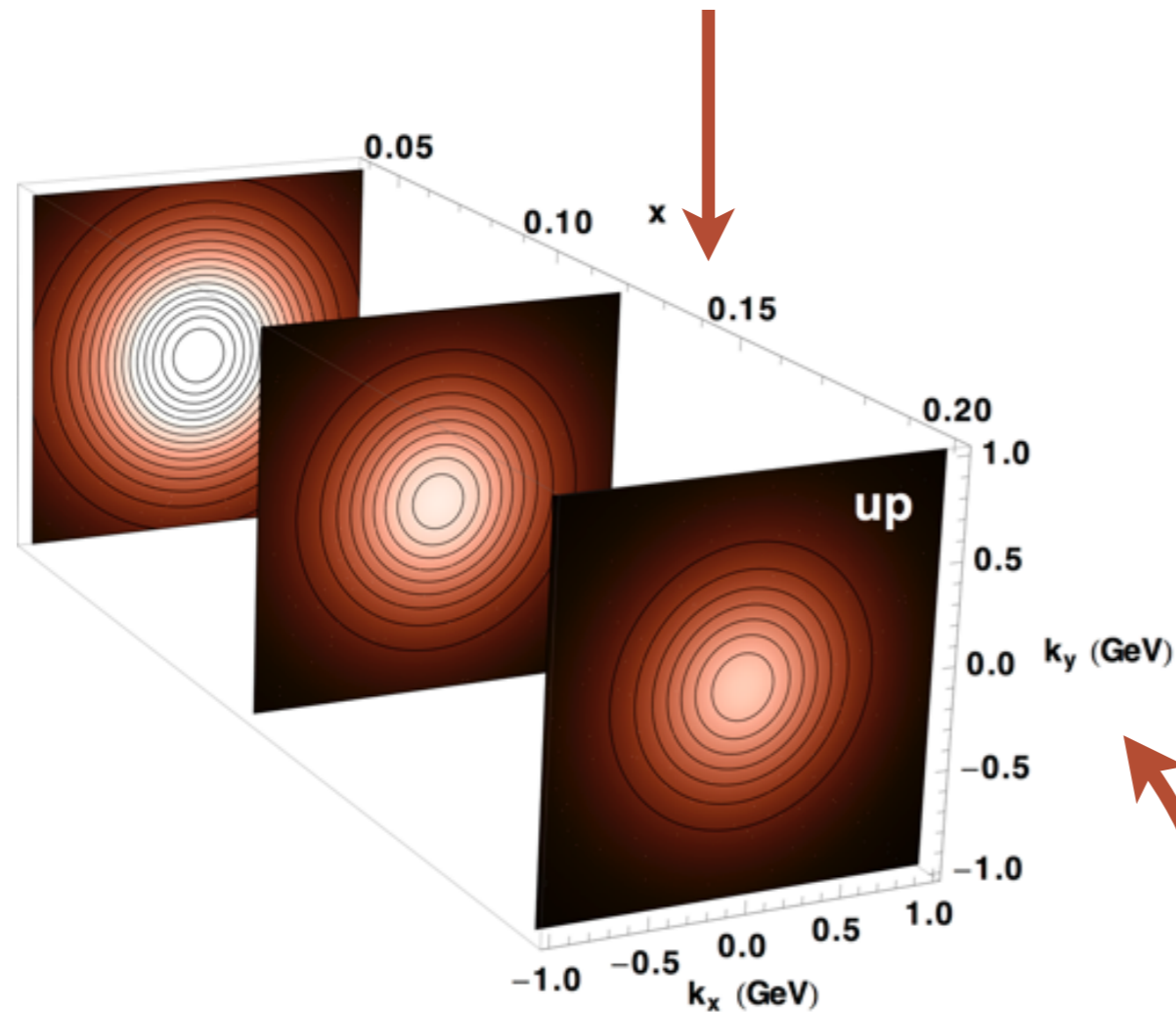


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# Transverse-Momentum Distributions (TMDs)

Fraction of longitudinal momentum



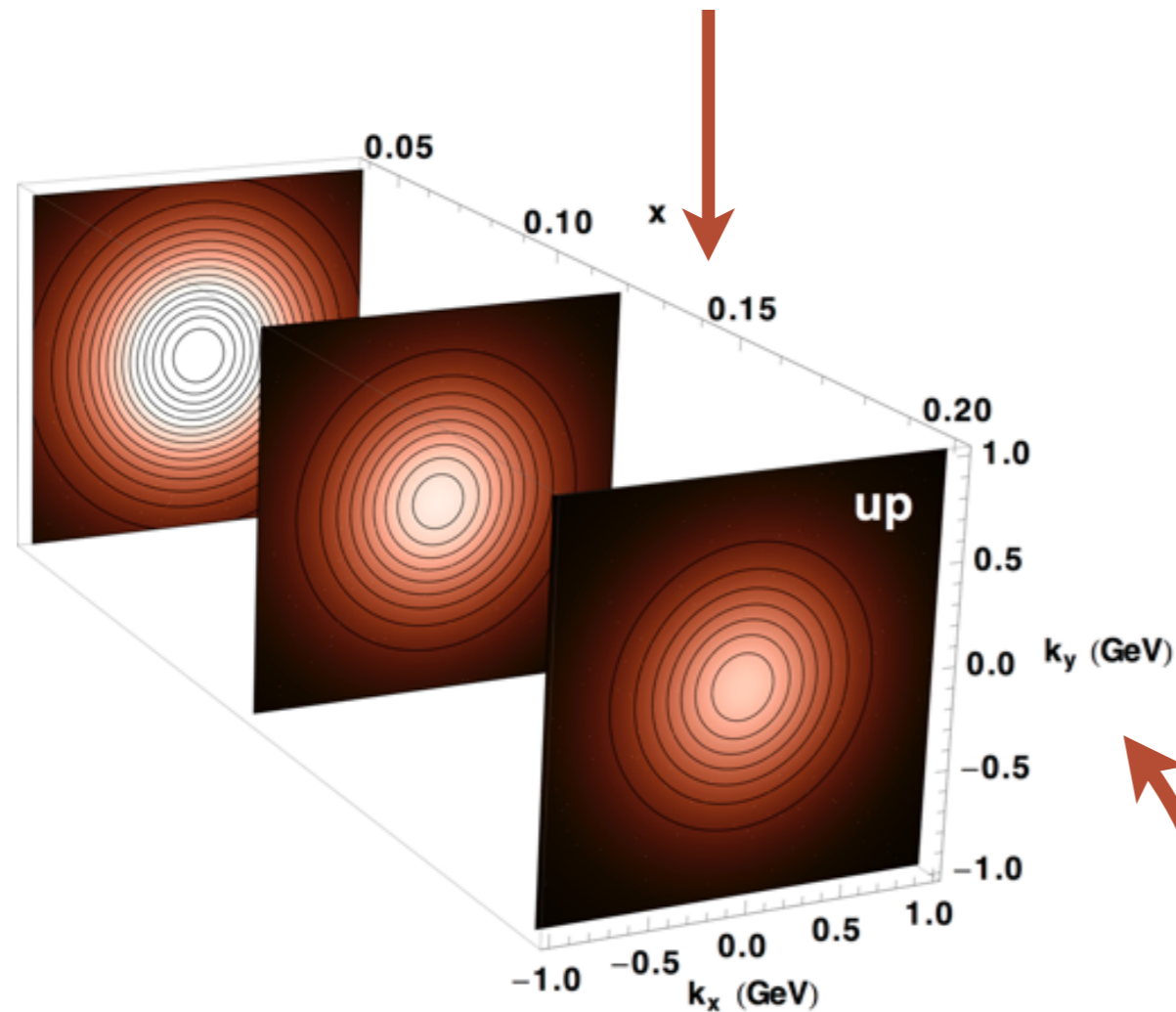
**TMDs** map the distribution of partons inside the nucleon in 3D in momentum space.

They can be extracted through **global fits**  
There are attempts to calculate them in lattice QCD

Transverse momentum

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**Are TMDs universal?**

**Do they depend on  $x$ ?**

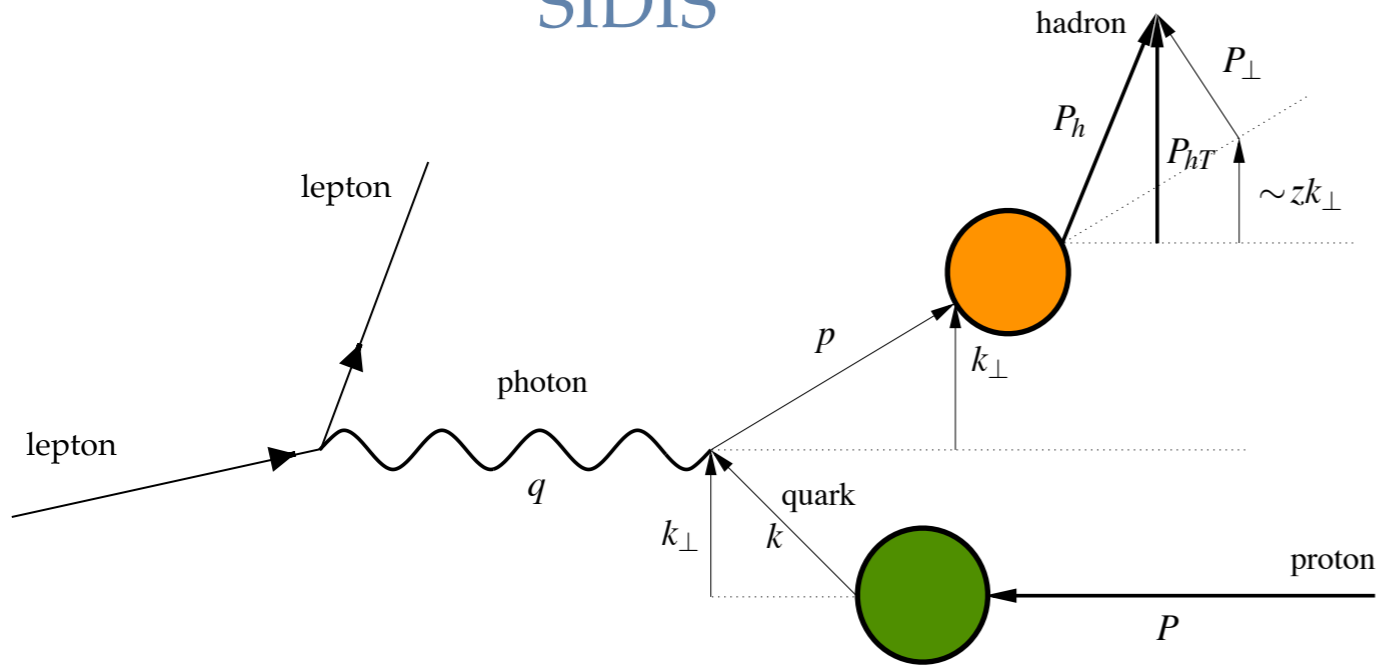
**Do they depend on the quark flavor?**

Transverse momentum

# TMD factorization: Universality

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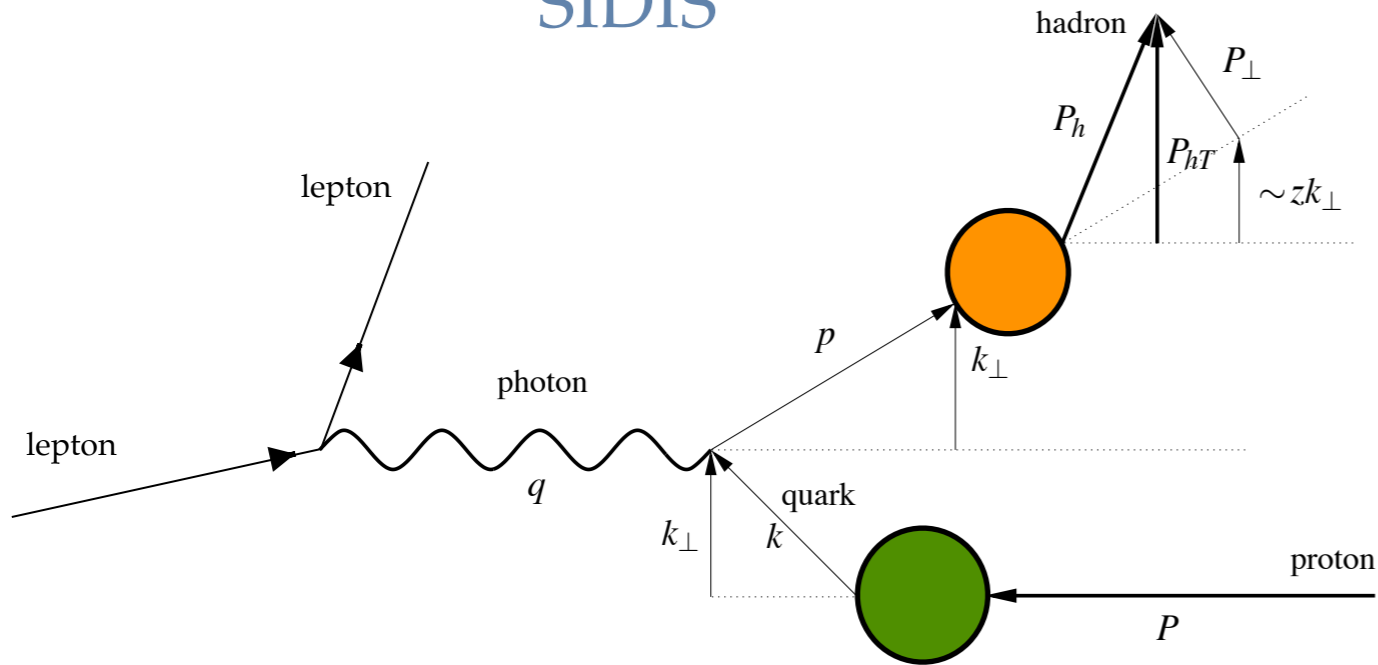
SIDIS



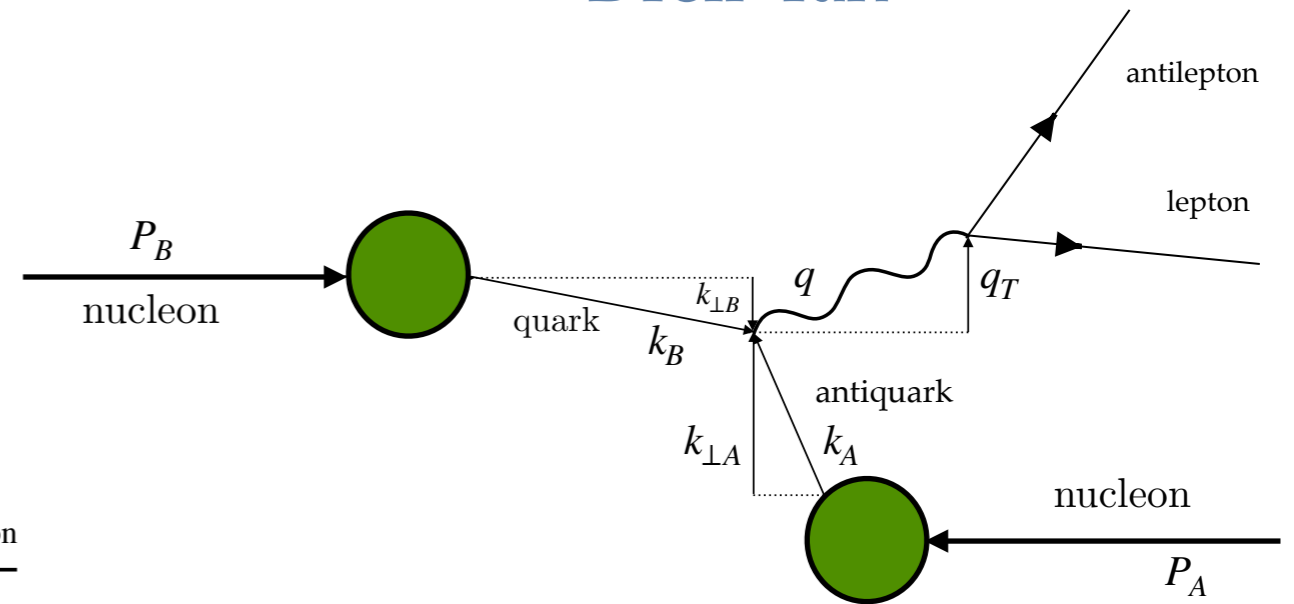
$$F_{UU,T}(x, z, |\mathbf{q}_T|, Q) \sim \int_0^{+\infty} d|\mathbf{b}_T| |\mathbf{b}_T| J_0(|\mathbf{b}_T| |\mathbf{q}_T|) \hat{f}_1^a(x, b_T^2; \mu, \zeta_A) \hat{D}_1^{a \rightarrow h}(z, b_T^2; \mu, \zeta_B)$$

# TMD factorization: Universality

SIDIS



Drell-Yan

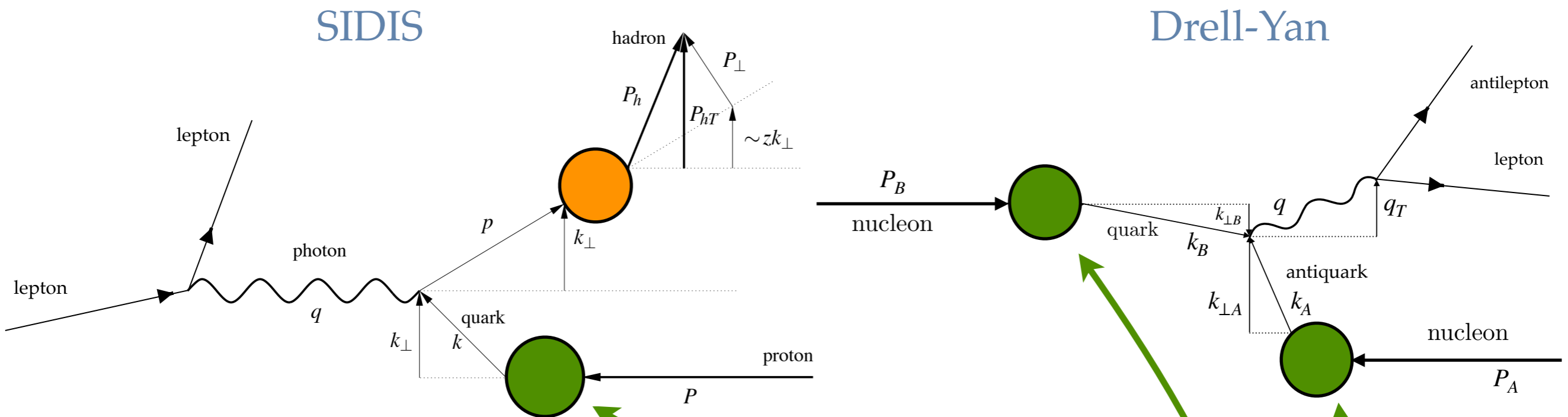


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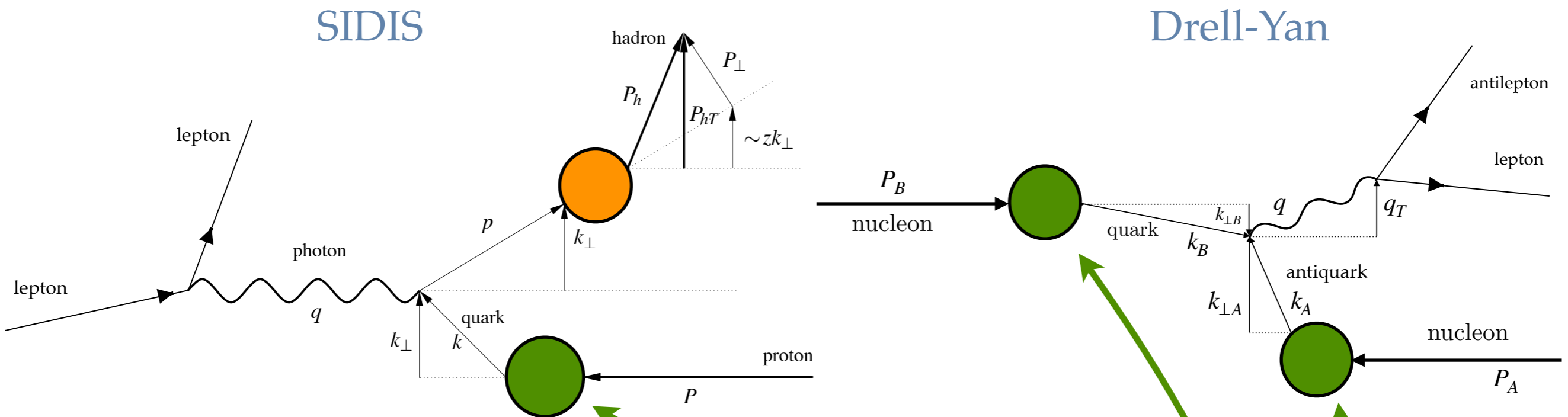


*Same functions*

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**GLOBAL FITs**

# Structure of a TMD

TMD in Fourier space

$$\hat{F}(x, b_T^2; \mu, \zeta) = \int \frac{d^2 \mathbf{k}_\perp}{(2\pi)^2} e^{i\mathbf{b}_T \cdot \mathbf{k}_\perp} F(x, k_\perp^2; \mu, \zeta)$$

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$$\hat{f}_1^q(x, b_T^2; \mu, \zeta) = \sum_j C_{q/j}(x, b_*; \mu_{b_*}, \mu_{b_*}^2) \otimes f_1^j(x, \mu_{b_*})$$

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Perturbative TMD at the initial scale

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Perturbative TMD at the initial scale

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Evolution to final scale (of the process)

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Non-perturbative part of the TMD

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Non-perturbative part of the TMD

Parameterization

# MAP TMD fitting framework

<https://github.com/MapCollaboration/NangaParbat>



☰ README.md



Nanga Parbat is a fitting framework aimed at the determination of the non-perturbative component of TMD distributions.

## Download

You can obtain NangaParbat directly from the github repository:

<https://github.com/MapCollaboration/NangaParbat>

For the last development branch you can clone the master code:

```
git clone git@github.com:MapCollaboration/NangaParbat.git
```

# Available Global Fits

	Accuracy	SIDIS	DY	N of points	$\chi^2/N_{\text{data}}$
<b>Pavia 2017</b> Bacchetta, Delcarro, et al., JHEP 06 (2017)	NLL	✓	✓	8059	1.55
<b>SV 2019</b> Scimemi, Vladimirov, JHEP 06 (2020)	$N^3LL^-$	✓	✓	1039	1.06
<b>MAPTMD22</b> Bacchetta, Bertone, et al., JHEP 10 (2022)	$N^3LL^-$	✓	✓	<b>2031</b>	<b>1.06</b>



# MAPTMD22 global fit: features

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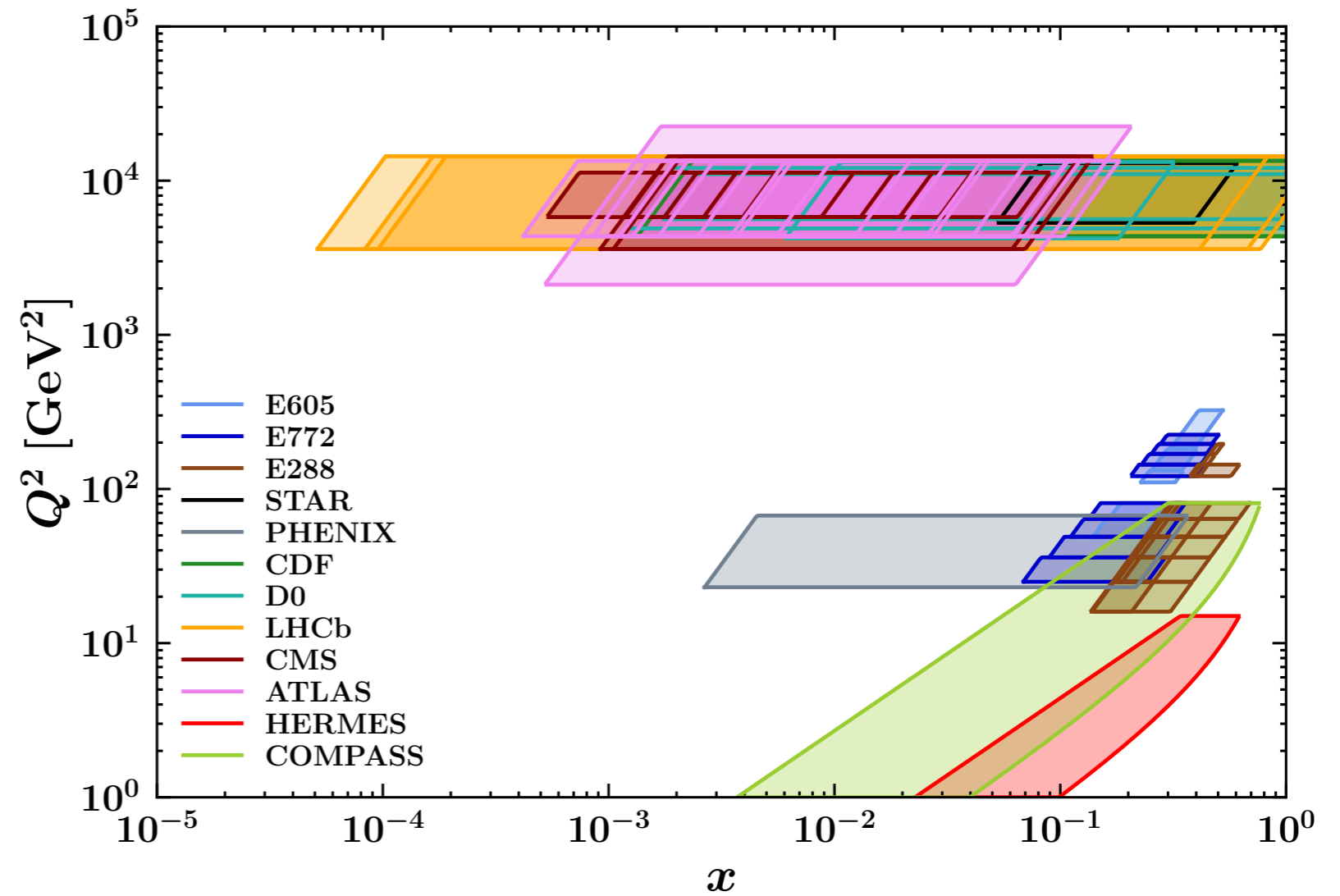
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- Agreement with data:  $\chi^2/N_{data} = 1.06$

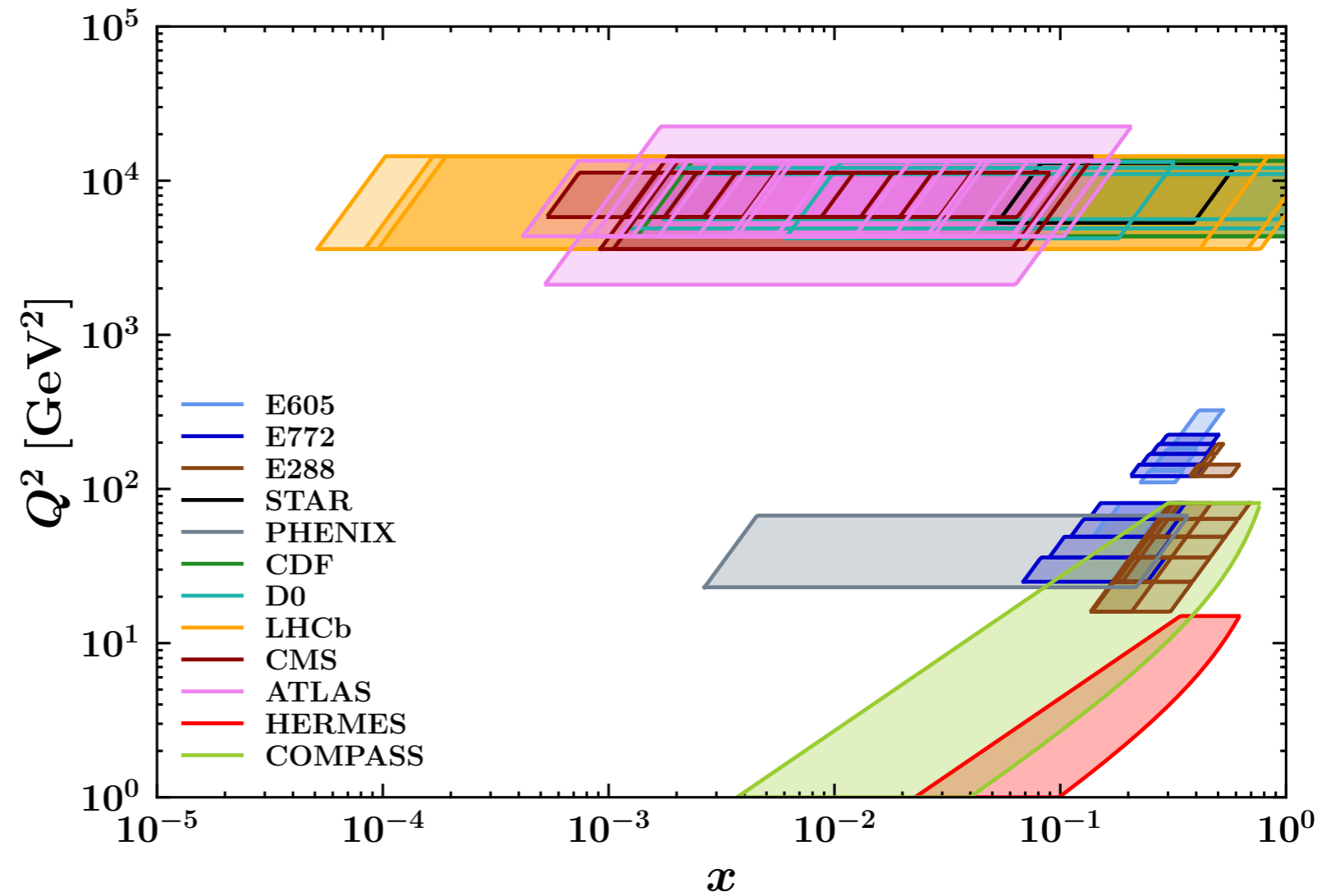
# MAP22: included data sets



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Drell-Yan data

484

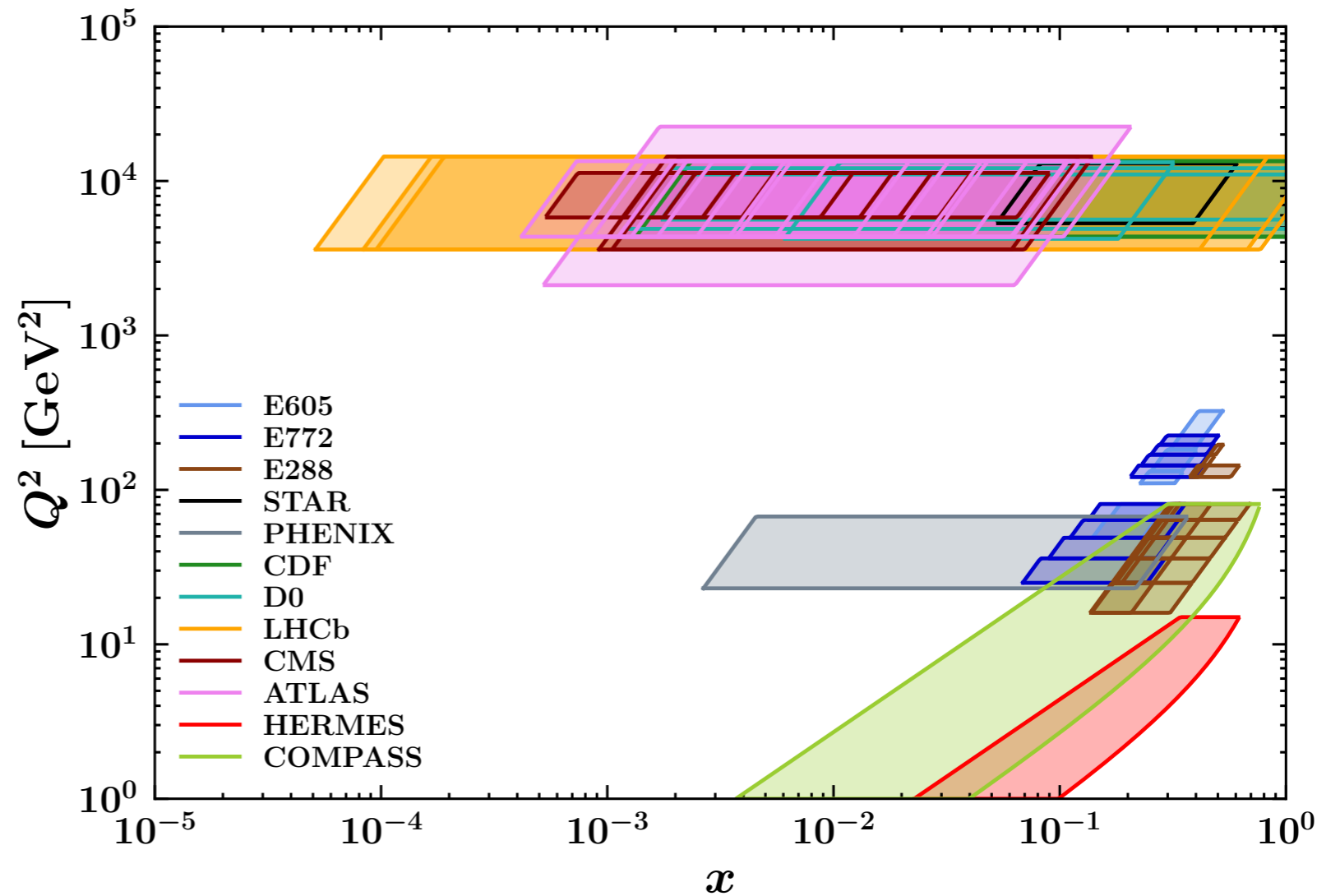


# MAP22: included data sets

**Drell-Yan data**      **484**

Fixed-target:  
E288, E605, E772

Collider mode:  
RHIC, Tevatron, LHC



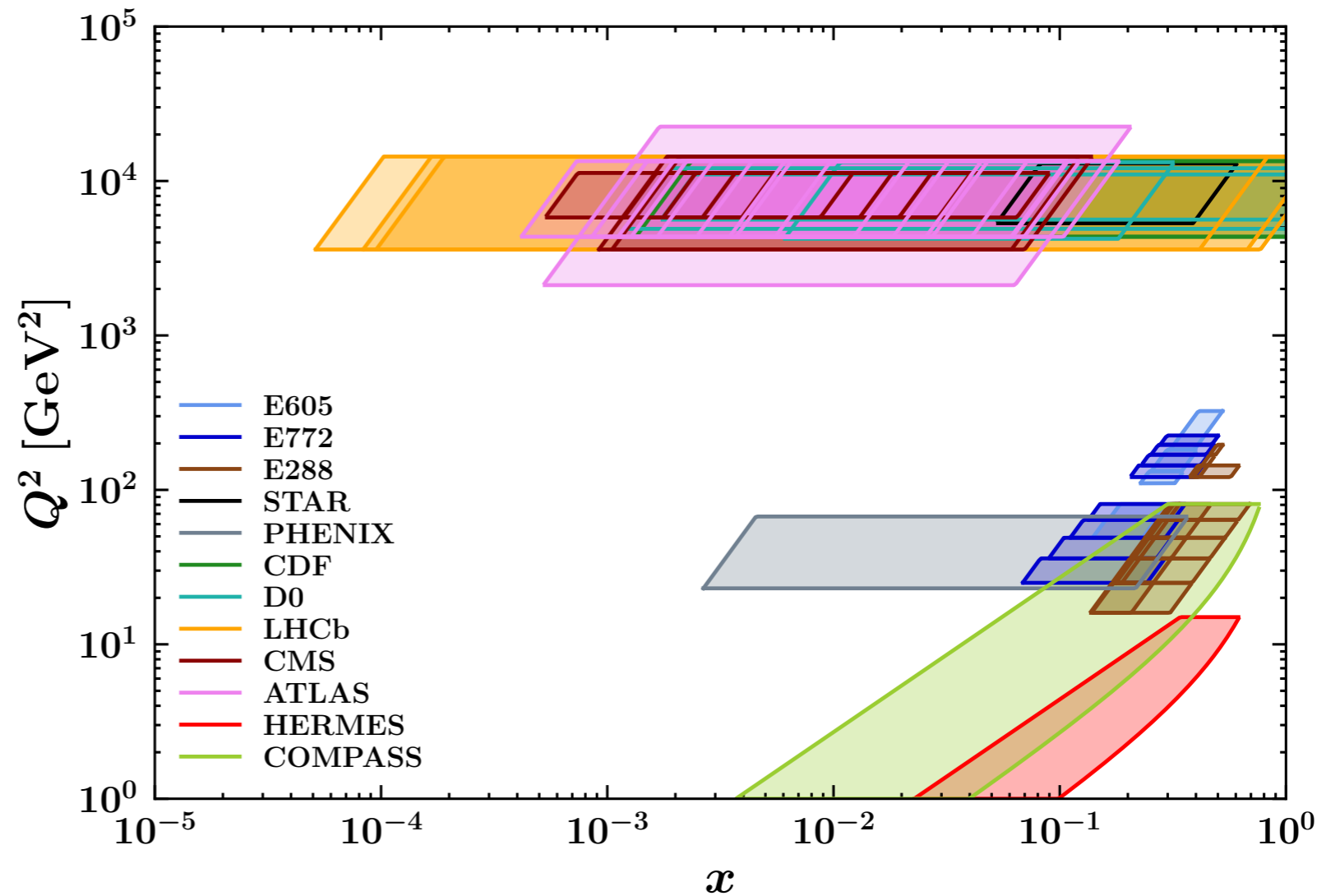
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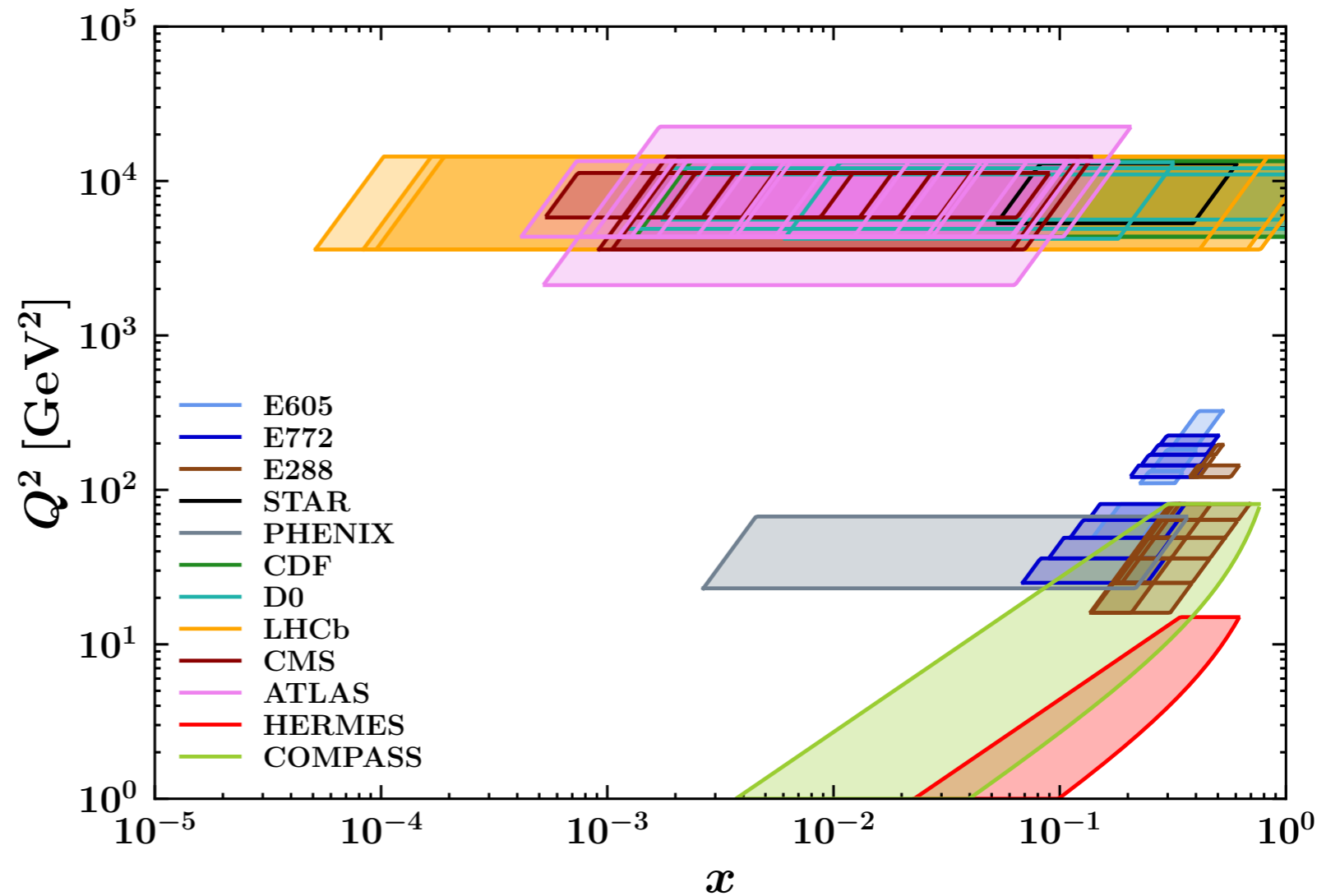
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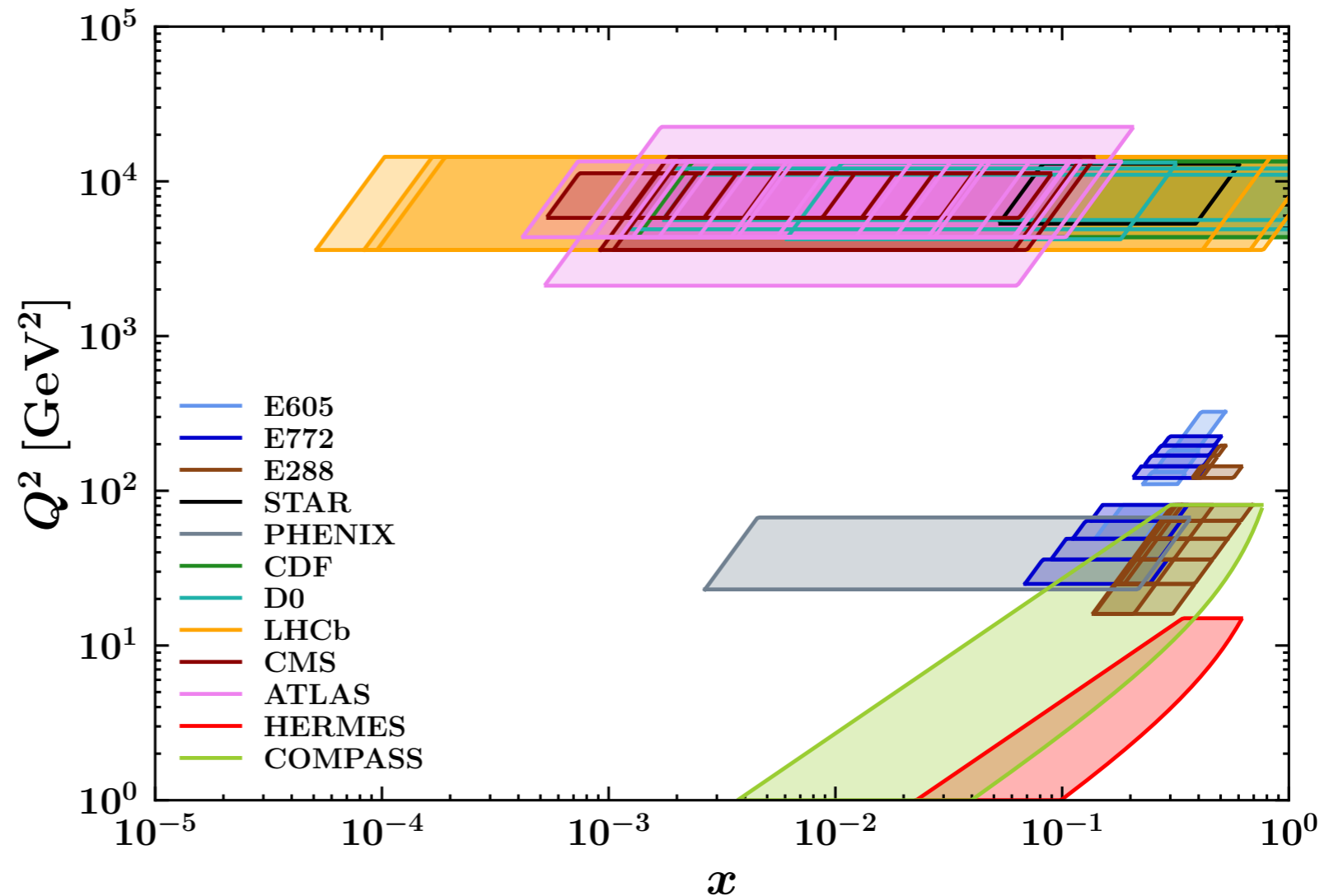
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# MAP22: NP parametrization

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

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

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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

# MAPTMD22 global fit $\Rightarrow$ MAPTMD24

- Global analysis of Drell-Yan and SIDIS data sets: **2031** data points
- Perturbative accuracy:  $N^3LL^-$
- *Normalization prefactor* for SIDIS observables
- Number of fitted parameters: **21**
- Agreement with data:  $\chi^2/N_{data} = 1.06$

# MAPTMD22 global fit $\Rightarrow$ MAPTMD24

- Global analysis of Drell-Yan and SIDIS data sets: **2031** data points  
**Same global data set**
- Perturbative accuracy:  $N^3LL^-$
- *Normalization prefactor* for SIDIS observables
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 $N^3LL$
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- Perturbative accuracy:  $N^3LL^-$       MMHT      **NNPDF**  
 $N^3LL$                       DSS                   $\Rightarrow$       **MAPFF**
- *Normalization prefactor* for SIDIS observables
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 $N^3LL$  DSS  $\Rightarrow$  **MAPFF**
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**Same approach**
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- Global analysis of Drell-Yan and SIDIS data sets: **2031** data points  
**Same global data set**
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 $N^3LL$  DSS  $\Rightarrow$  **MAPFF**
- *Normalization prefactor* for SIDIS observables  
**Same approach**
- Number of fitted parameters: **21**  
**Same parameterization**
- Agreement with data:  $\chi^2/N_{data} = 1.06$

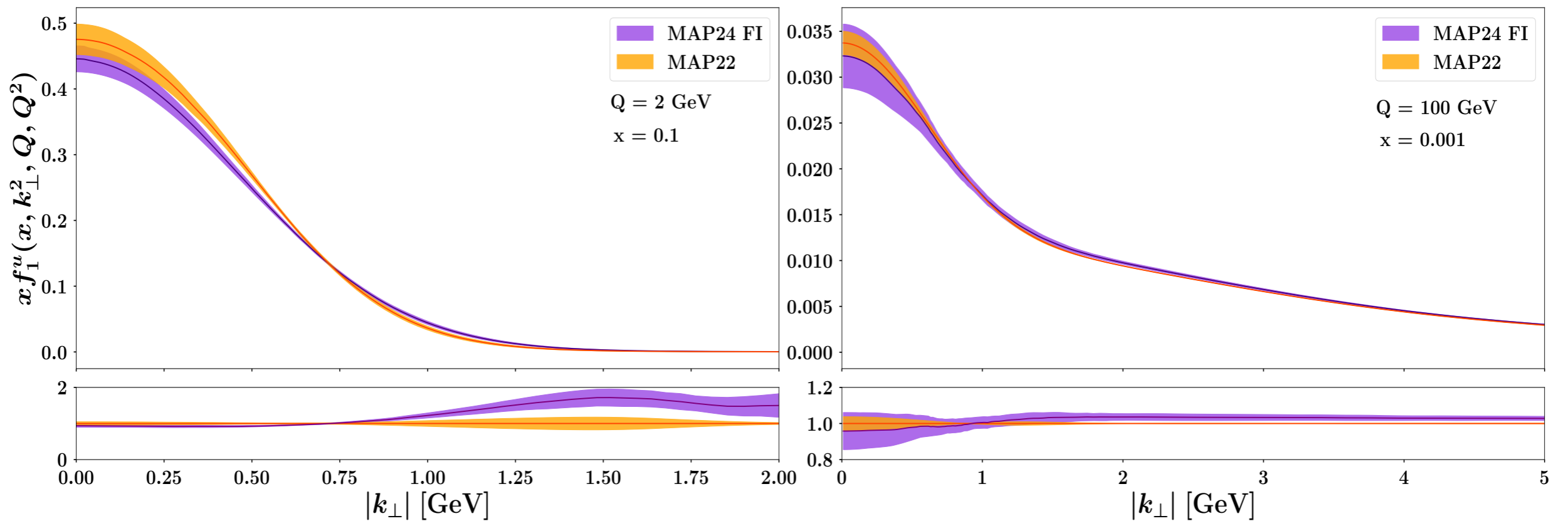


# MAPTMD24

Configuration	$\chi^2/N_{data}$		
	DY	SIDIS	Total
<b>MMHT</b> +DSS (MAP22)	1.66	0.87	<b>1.06</b>
<b>NNPDF</b> +MAPFF (MAP24 FI)	1.58	1.34	<b>1.40</b>

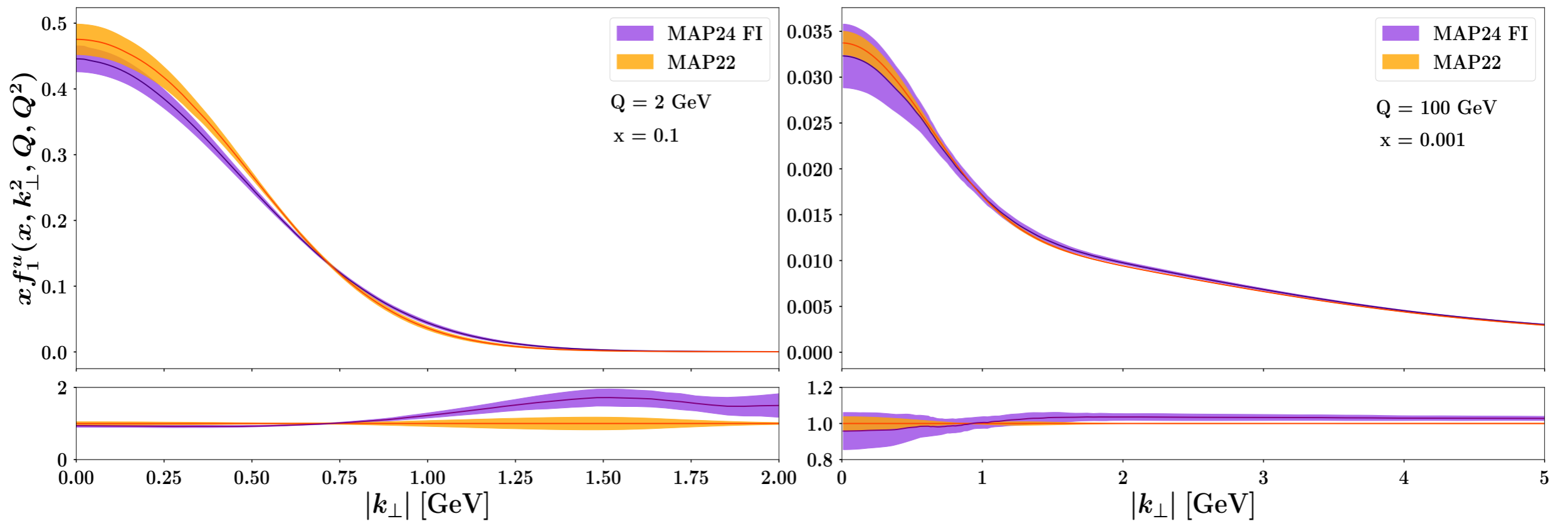
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**TMD PDFs are compatible with MAP22**

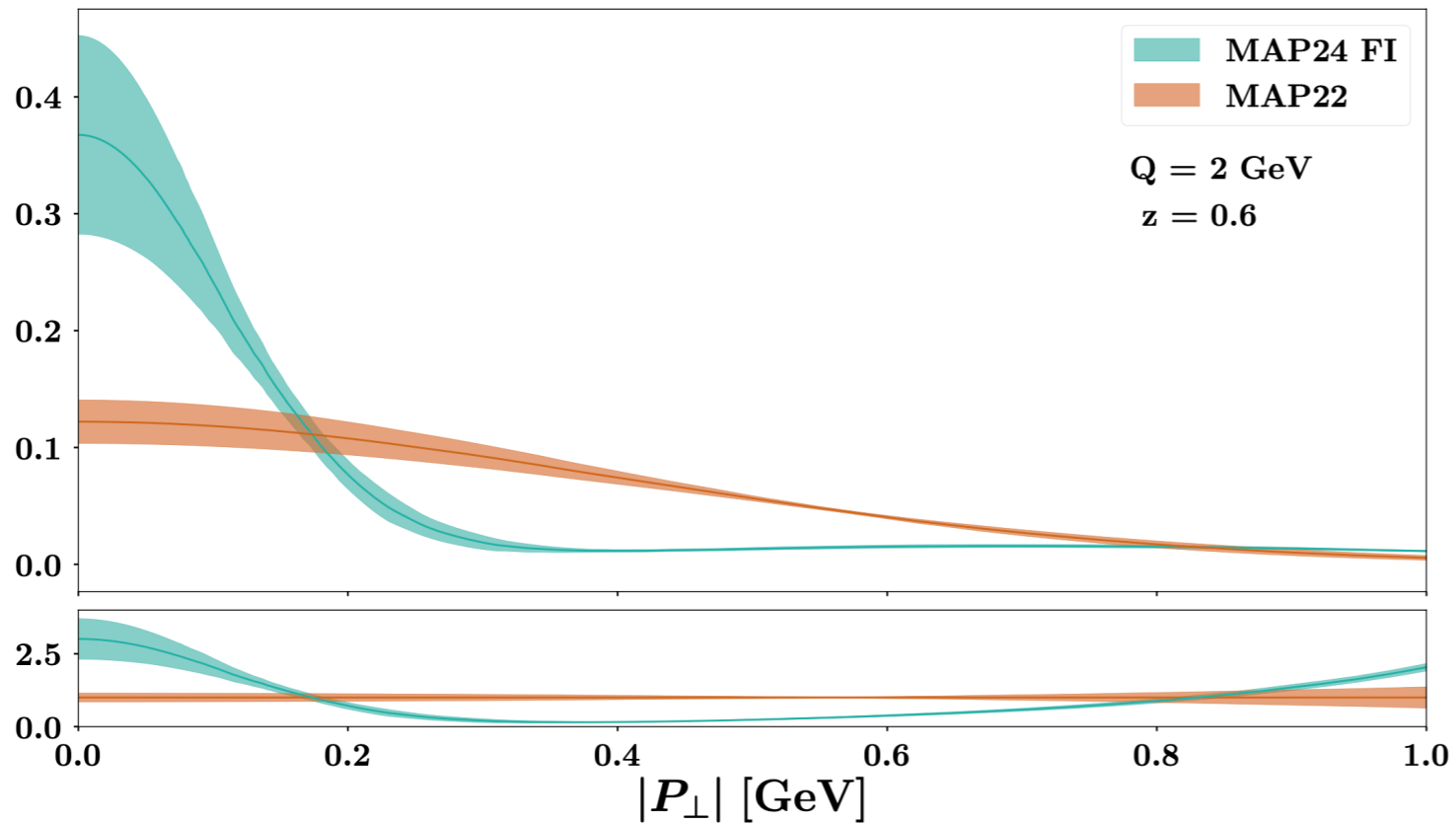


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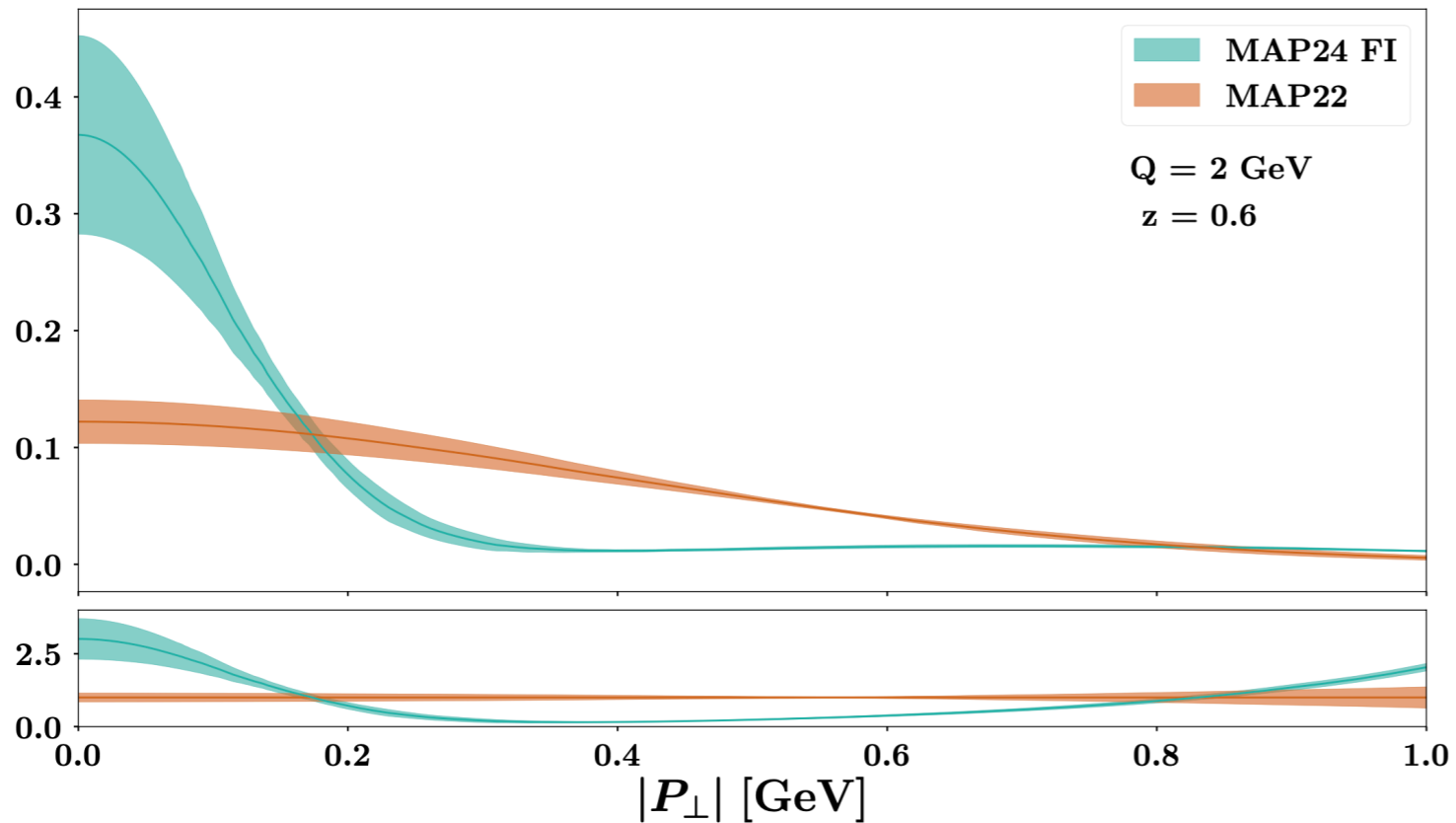
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## MAPFF1.0nnlo

- approx NNLO
- NN approach
- New behaviors
- Smaller uncertainties

# MAPTMD24

## NNPDF + MAPFF (MAP24 FI)

Data set	$N_{\text{dat}}$	$\chi_0^2/N_{\text{dat}}$
DY collider total	251	2.14
Dy fixed target total	233	0.68
HERMES total	344	2.72
COMPASS total	1203	0.99
SIDIS total	1547	1.38
Total	2031	1.40

## MMHT + MAPFF

Data set	$N_{\text{dat}}$	$\chi_0^2/N_{\text{dat}}$
DY collider total	251	2.01
Dy fixed target total	233	1.11
HERMES total	344	2.51
COMPASS total	1203	0.99
SIDIS total	1547	1.33
Total	2031	1.39

Data set	$N_{\text{dat}}$	$\chi_0^2/N_{\text{dat}}$
DY collider total	251	2.43
Dy fixed target total	233	0.75
HERMES total	344	0.95
COMPASS total	1203	0.88
SIDIS total	1547	0.90
Total	2031	1.07

Data set	$N_{\text{dat}}$	$\chi_0^2/N_{\text{dat}}$
DY collider total	251	2.06
Dy fixed target total	233	1.24
HERMES total	344	0.71
COMPASS total	1203	0.92
SIDIS total	1547	0.87
Total	2031	1.06

## NNPDF + DSS

## MMHT + DSS (MAP22)

# MAPTMD24

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←→  
**COMPATIBILITY**

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# MAPTMD24

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## NNPDF + DSS

## MMHT + DSS (MAP22)

# MAPTMD24

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## NNPDF + DSS

## MMHT + DSS (MAP22)

TU QUOQUE,  
BRUTE  
HERMES



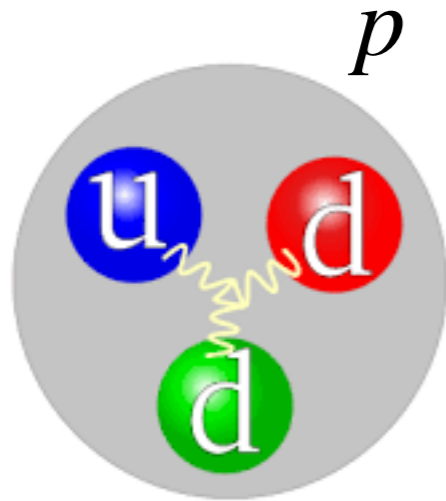


# MAPTMD24: new approach

Solution: we need **flavor dependence** to obtain a good agreement between theory and experiments

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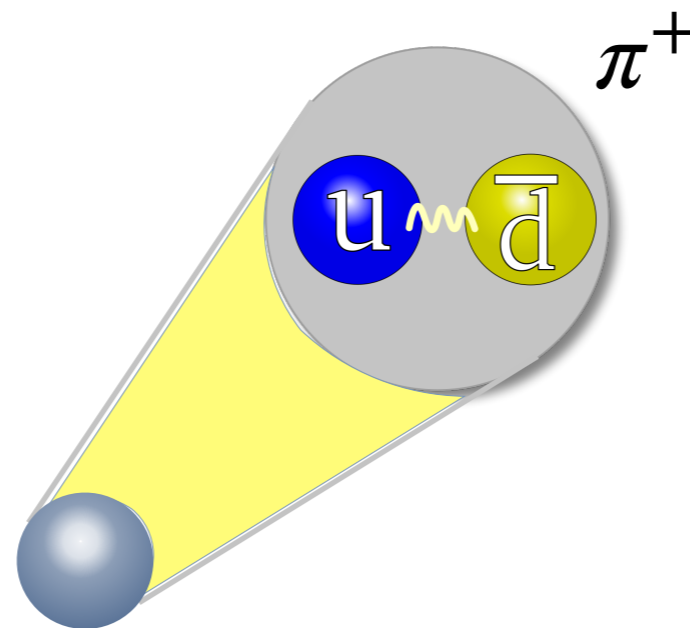
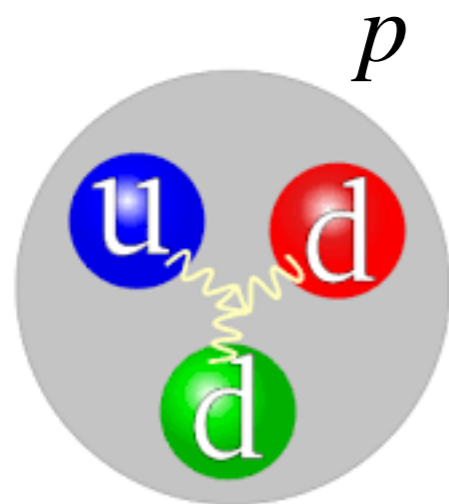
$u, d$

$\bar{u}, \bar{d}$

$s$  (*sea*)

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$u, d$

$\bar{u}, \bar{d}$

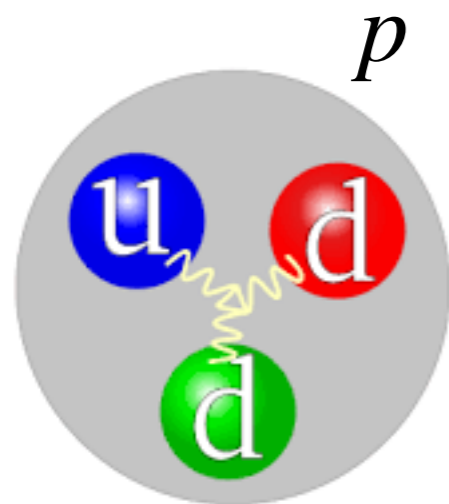
$s$  (*sea*)

$u \rightarrow \pi^+, \dots$

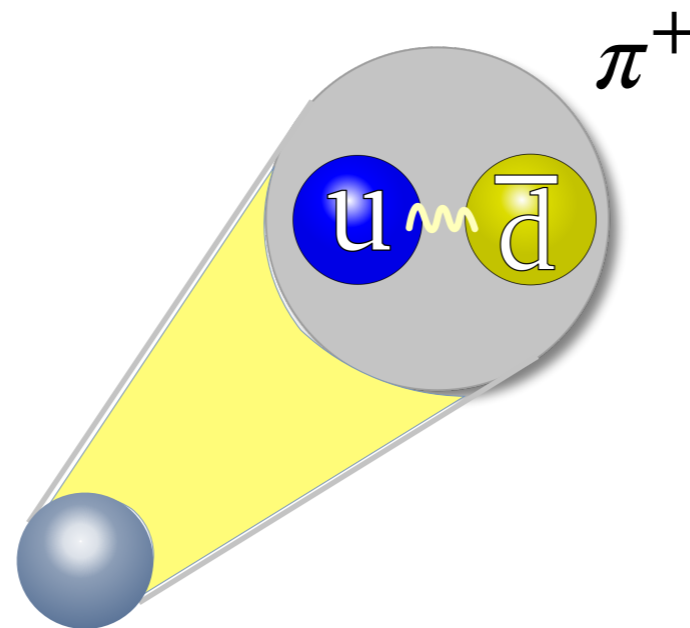
$d \rightarrow \pi^+, \dots$

# MAPTMD24: new approach

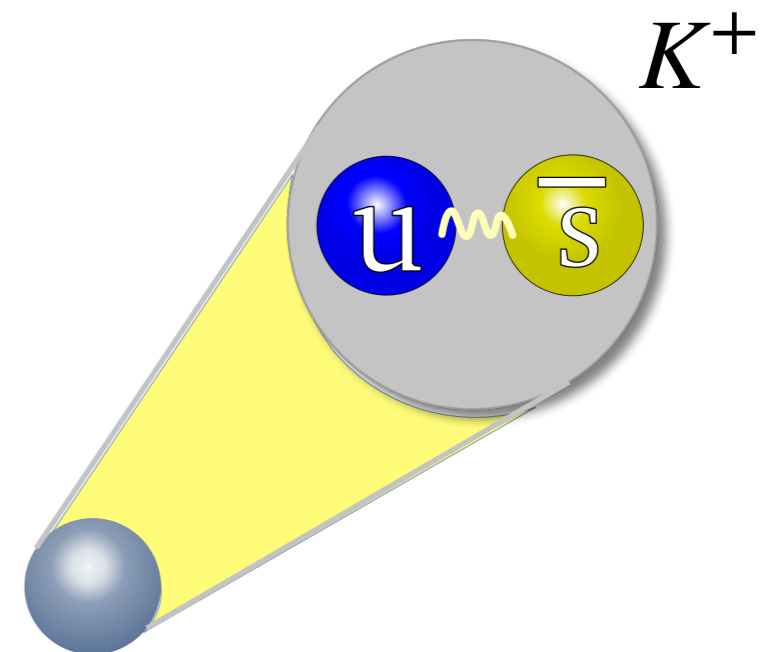
Solution: we need **flavor dependence** to obtain a good agreement between theory and experiments



$u, d$   
 $\bar{u}, \bar{d}$   
 $s$  (*sea*)



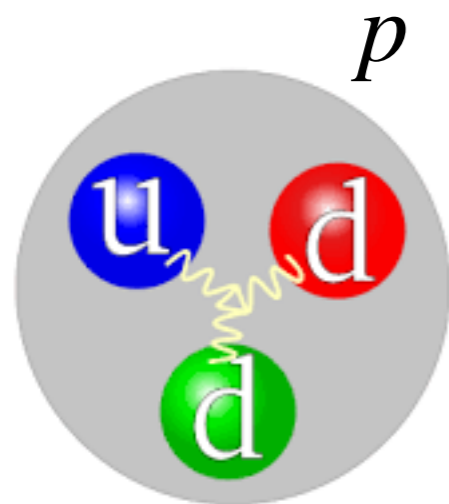
$u \rightarrow \pi^+, \dots$   
 $d \rightarrow \pi^+, \dots$



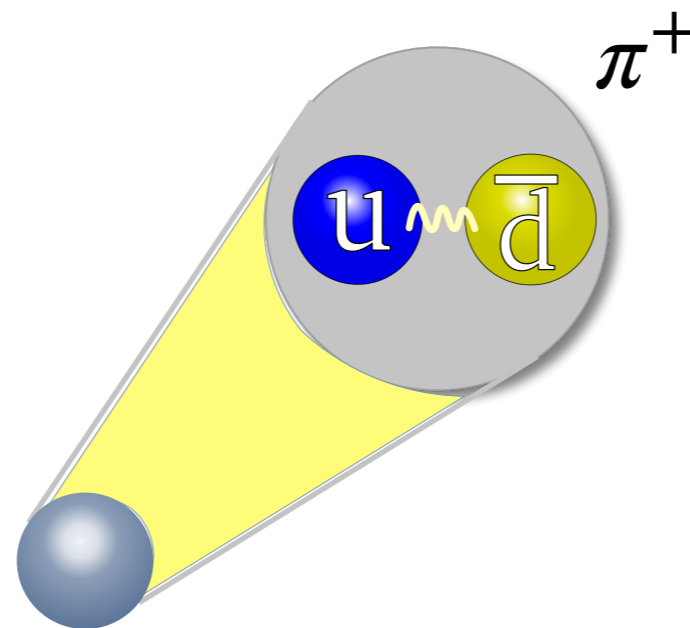
$u \rightarrow K^+, \dots$   
 $\bar{s} \rightarrow K^+, \dots$   
 $d \rightarrow K^+, \dots$

# MAPTMD24: new approach

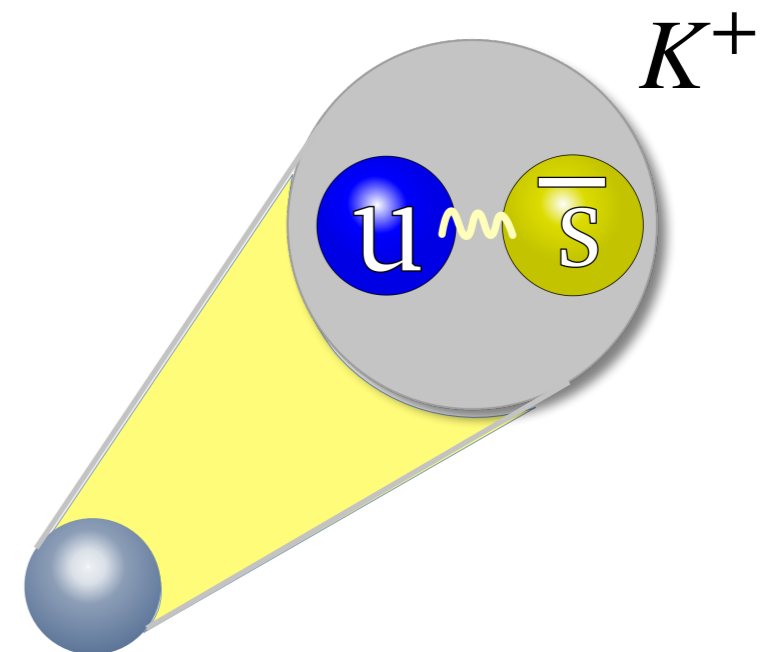
Solution: we need **flavor dependence** to obtain a good agreement between theory and experiments



$u, d$   
 $\bar{u}, \bar{d}$   
 $s$  (*sea*)



$u \rightarrow \pi^+, \dots$   
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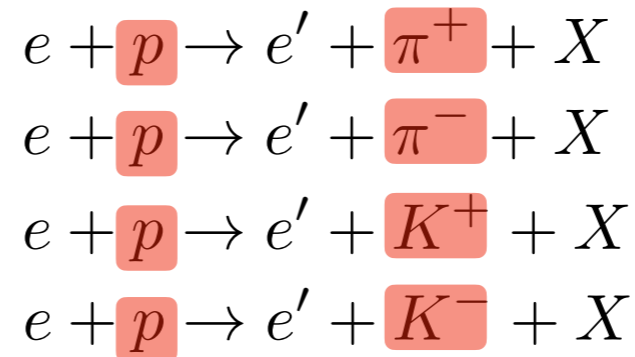


$u \rightarrow K^+, \dots$   
 $\bar{s} \rightarrow K^+, \dots$   
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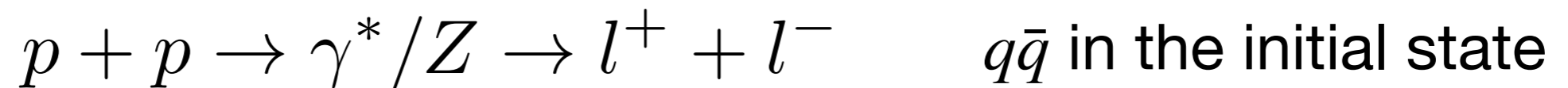
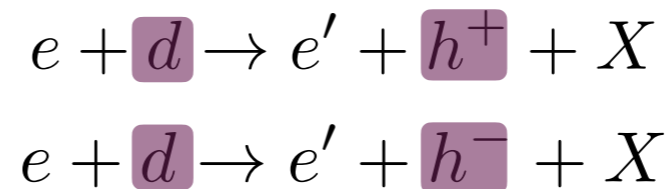
**charge conjugation**

# MAPTMD24: new approach

## high sensitivity to flavor dependence



+ deuteron target



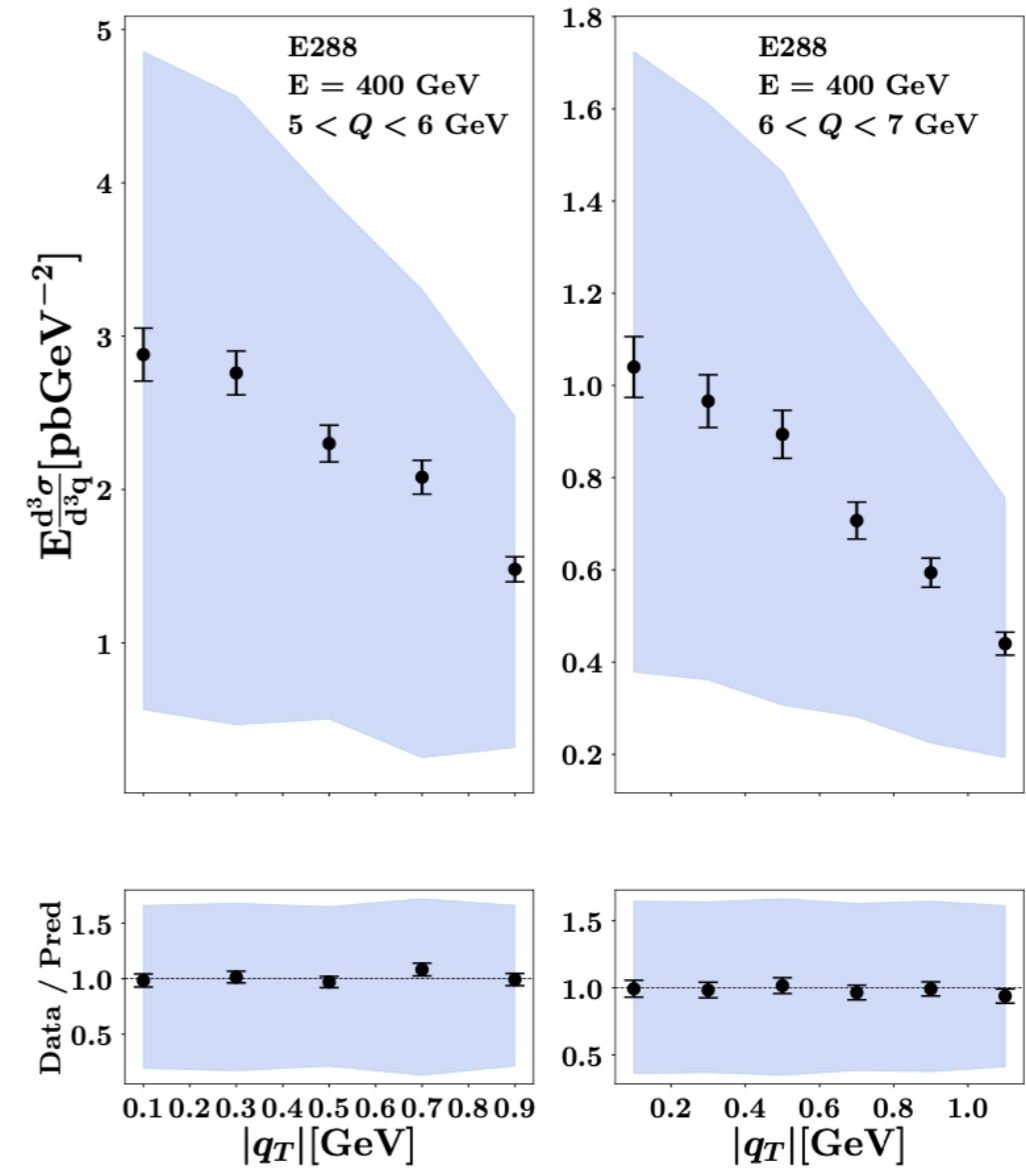
## low sensitivity to flavor dependence

# MAPTMD24: results

Data set	N <sup>3</sup> LL			
	$N_{\text{dat}}$	$\chi_D^2$	$\chi_\lambda^2$	$\chi_0^2$
DY collider total	251	1.37	0.28	1.65
DY fixed-target total	233	0.63	0.31	0.94
<i>HERMES total</i>	344	0.81	0.24	1.05
<i>COMPASS total</i>	1203	0.67	0.27	0.94
SIDIS total	1547	0.70	0.26	0.96
<b>Total</b>	<b>2031</b>	<b>0.81</b>	<b>0.27</b>	<b>1.08</b>

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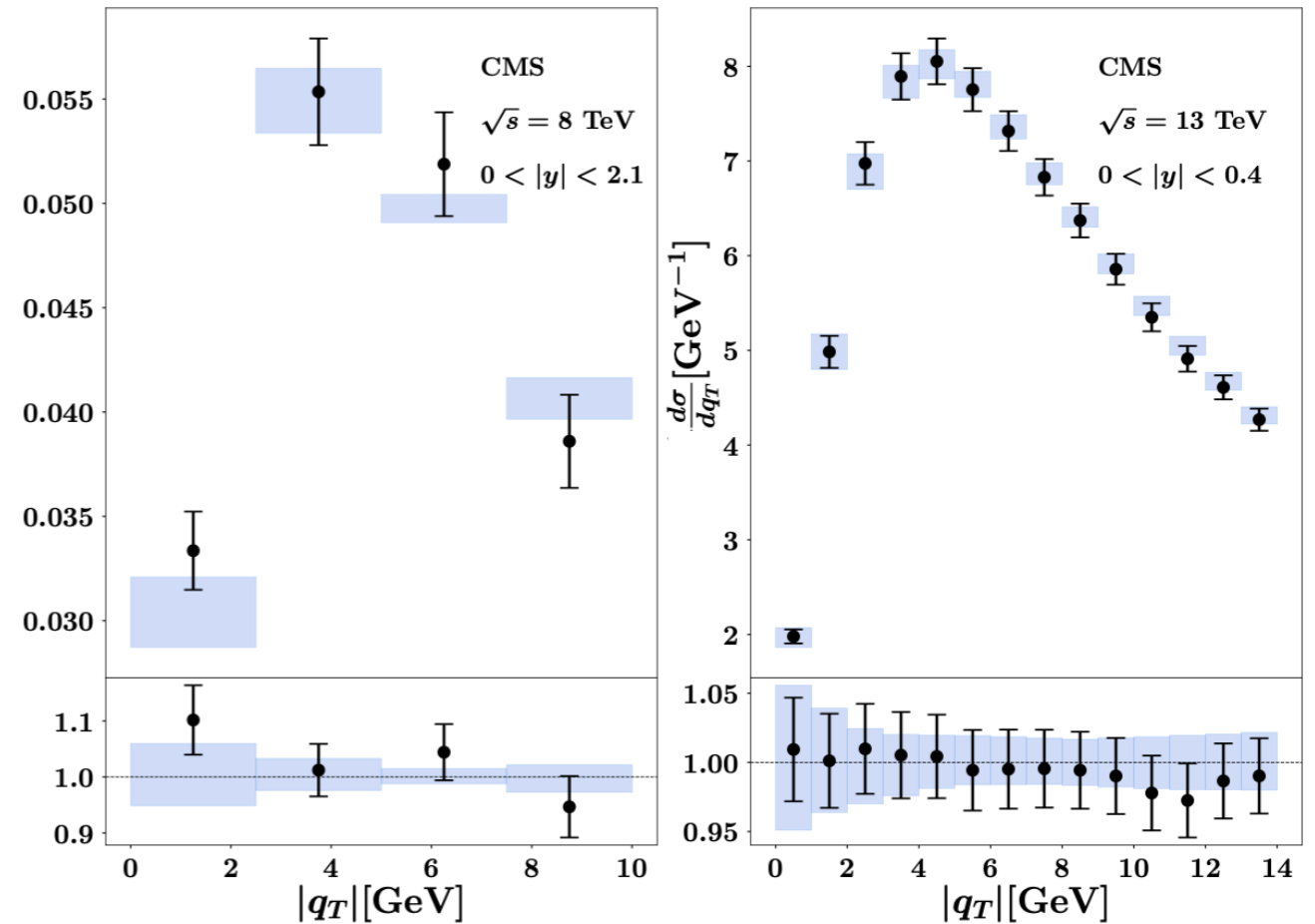
Data set	$N^3LL$			
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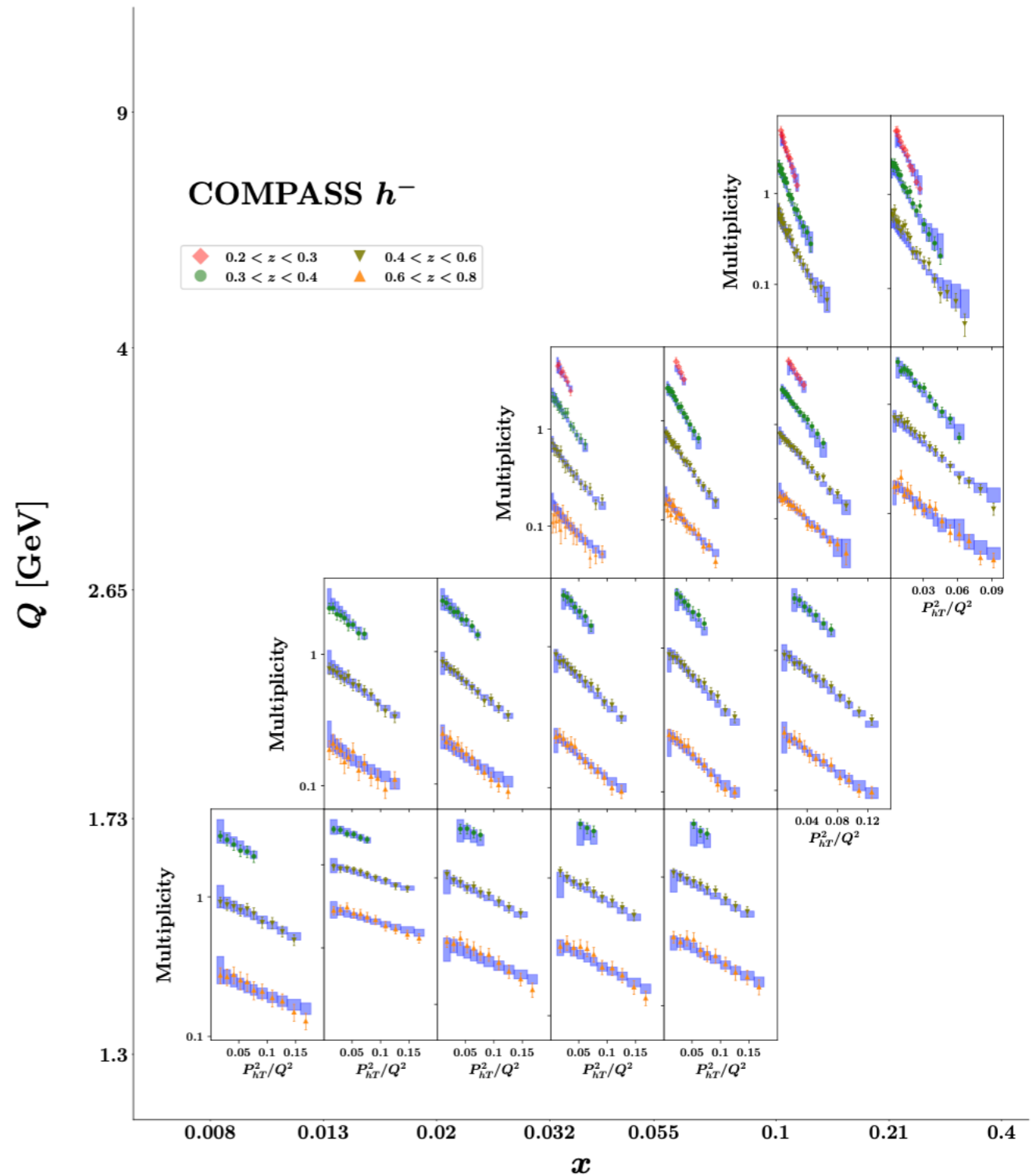
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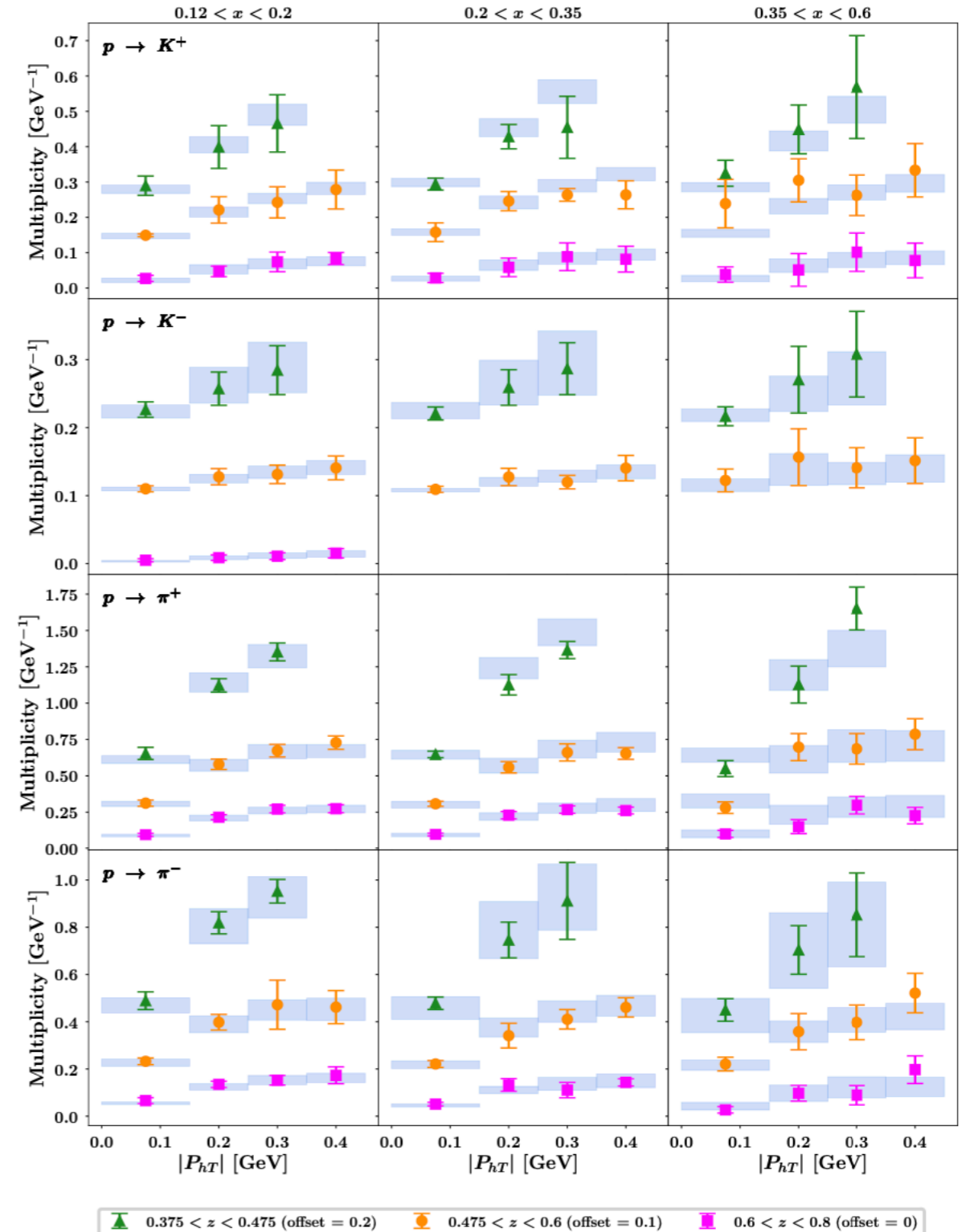
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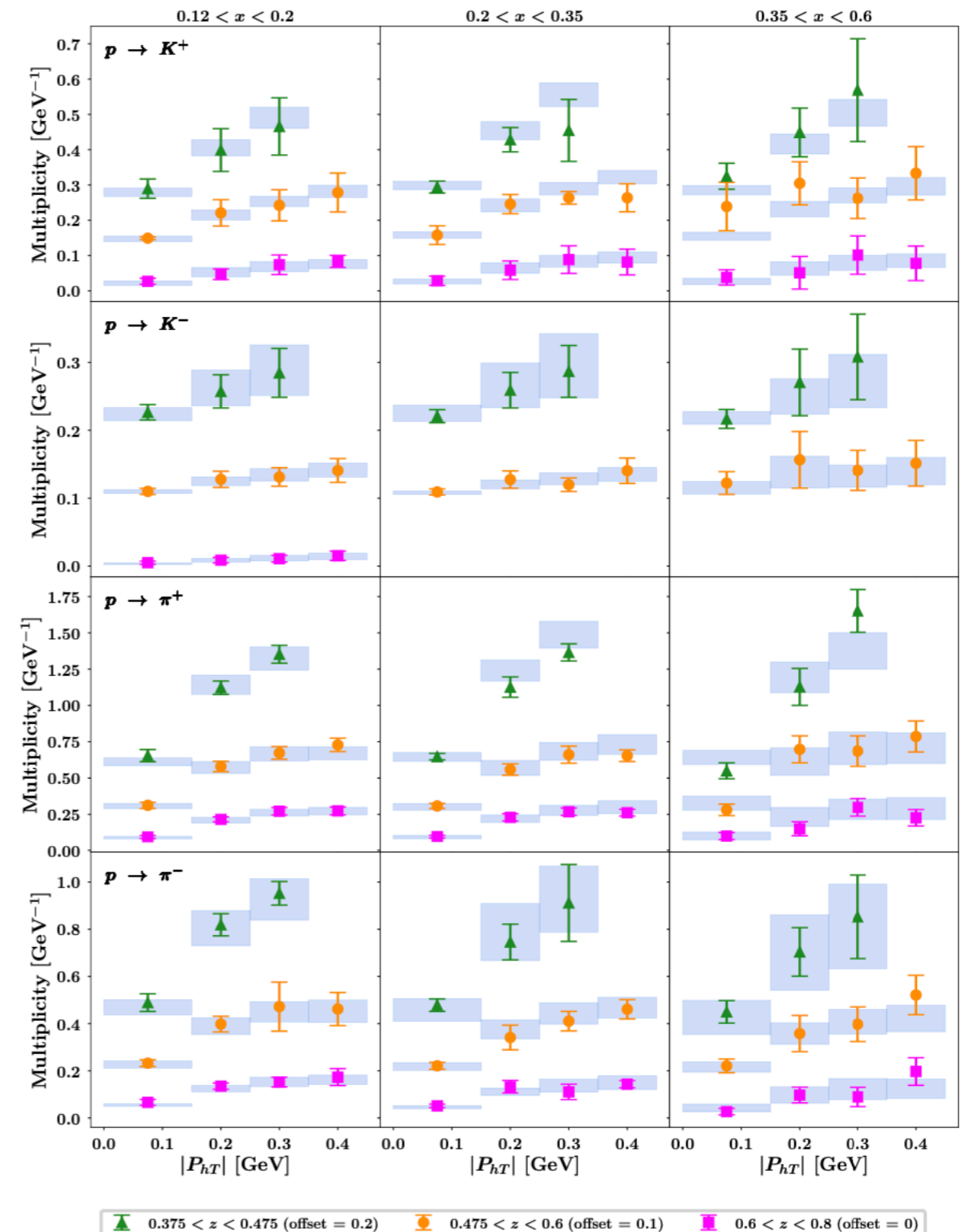
Data set	$N^3LL$			
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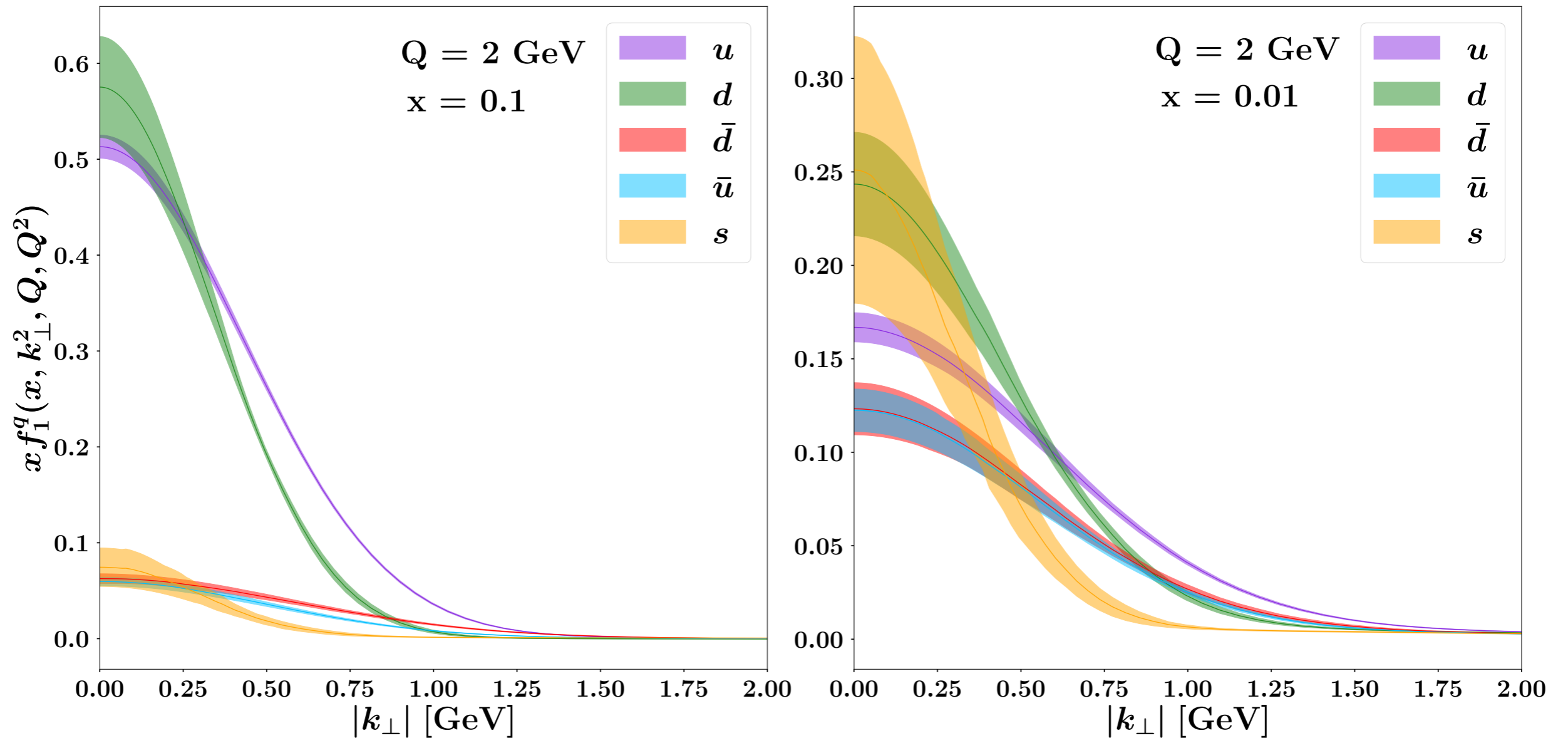
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The agreement between theory and HERMES data has increased a lot!



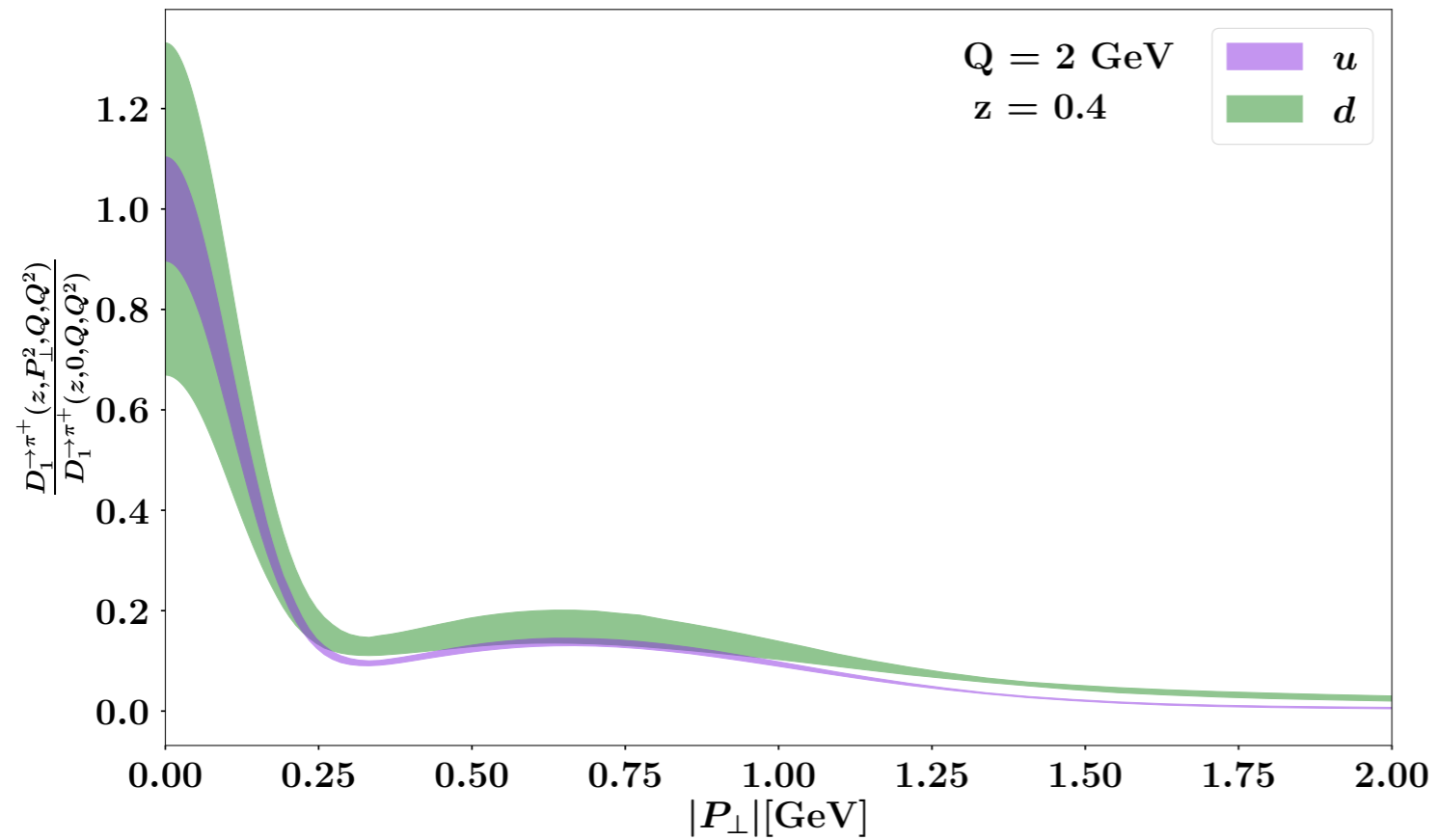
## Flavor-dependent TMD PDFs



Evidence of different behaviors for different flavors

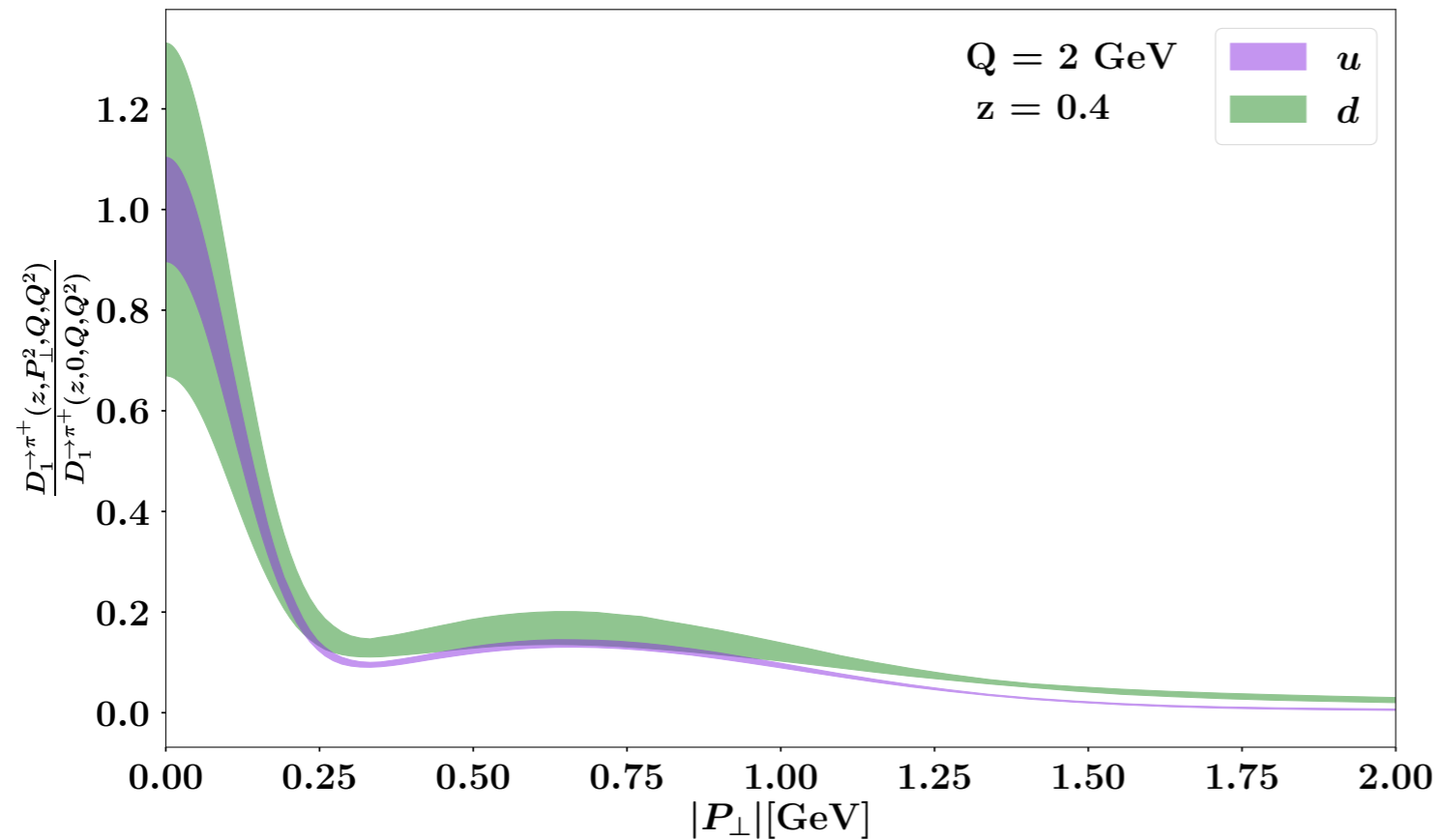
# MAPTMD24: results

## Flavor-dependent TMD FFs



# MAPTMD24: results

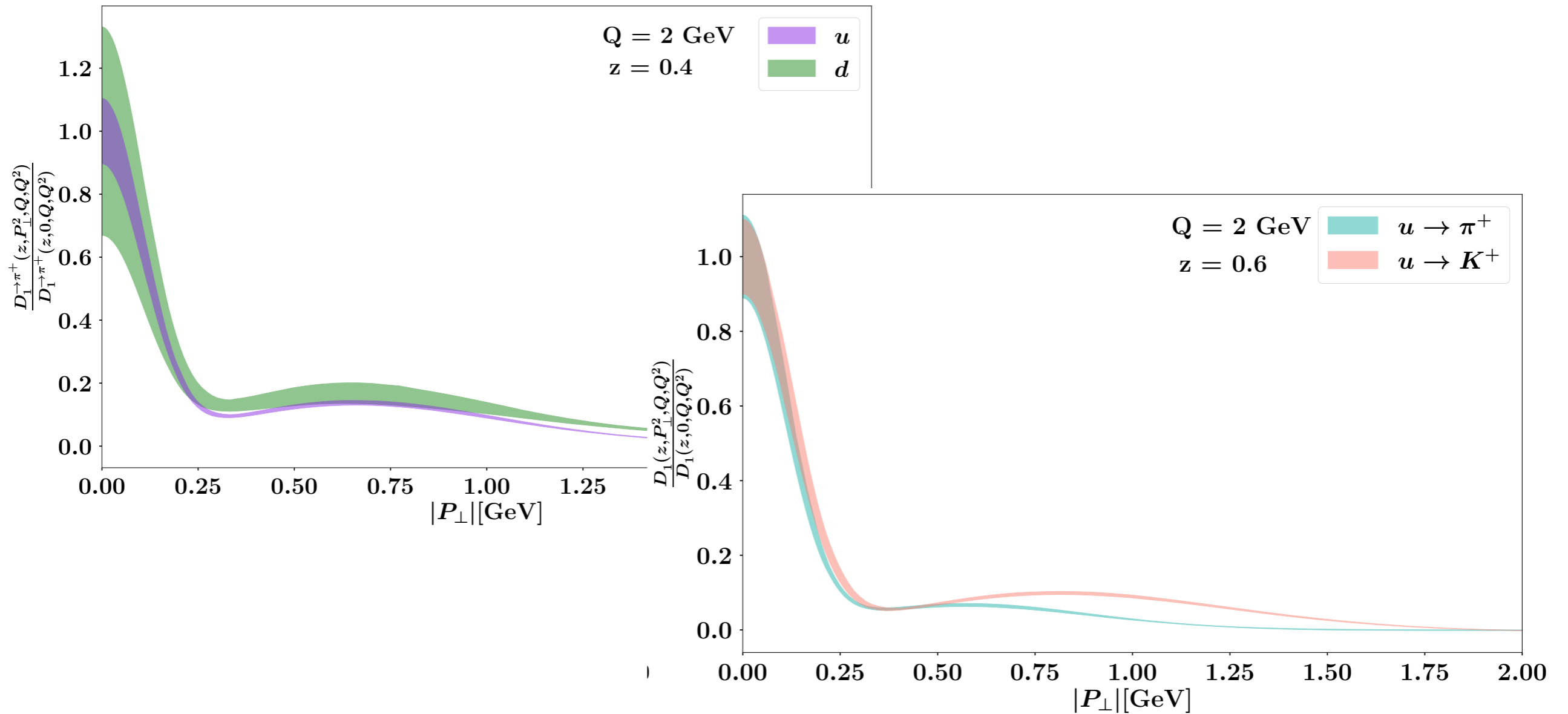
## Flavor-dependent TMD FFs



Small evidence of different behaviors for different flavors

# MAPTMD24: results

## Flavor-dependent TMD FFs

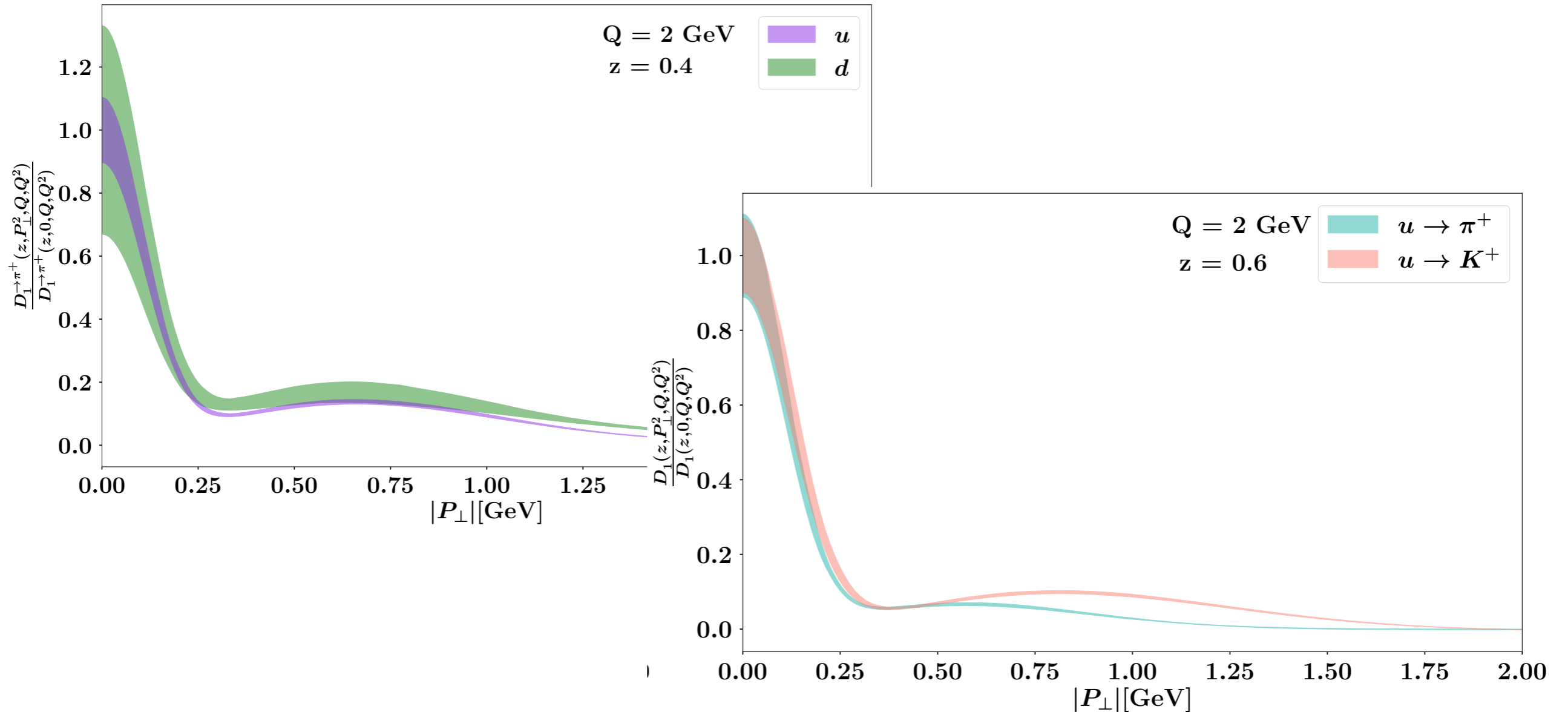


Small evidence of different behaviors for different flavors



# MAPTMD24: results

## Flavor-dependent TMD FFs

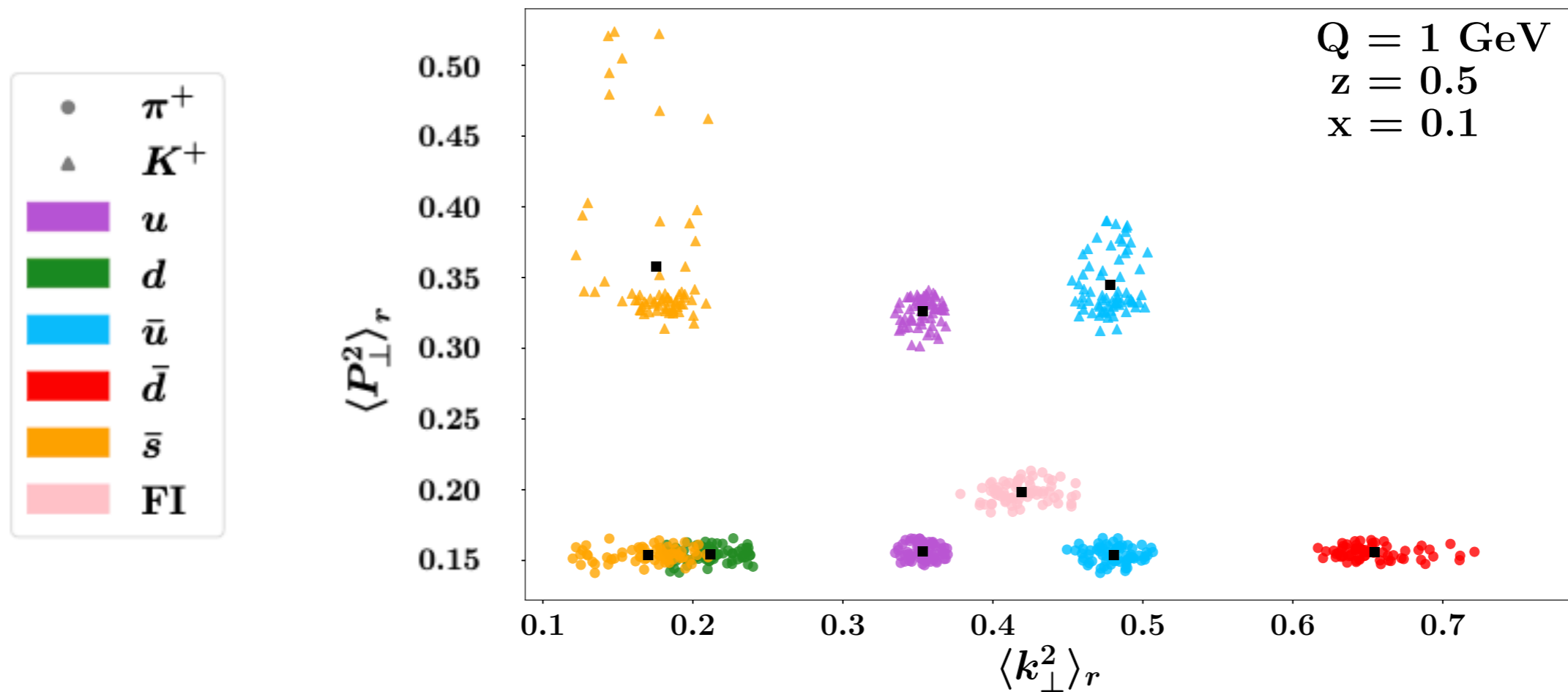


Small evidence of different behaviors for different flavors

Some evidence of different behaviors for different measured hadrons

# MAPTMD24: results

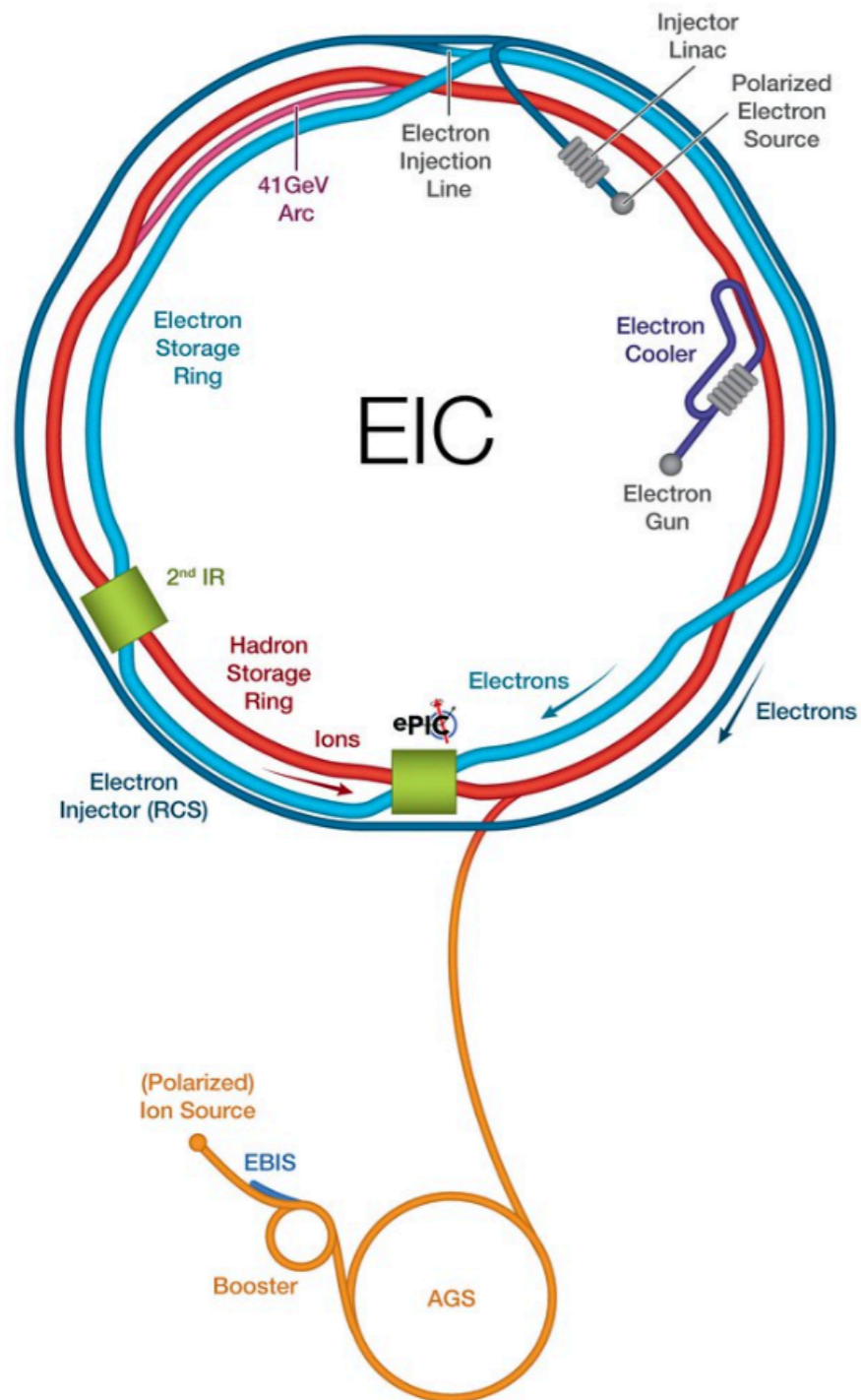
## TMD's “effective width”



Evidence of different behaviors for different flavors

Evidence of different behaviors for different measured hadrons

# Future perspectives



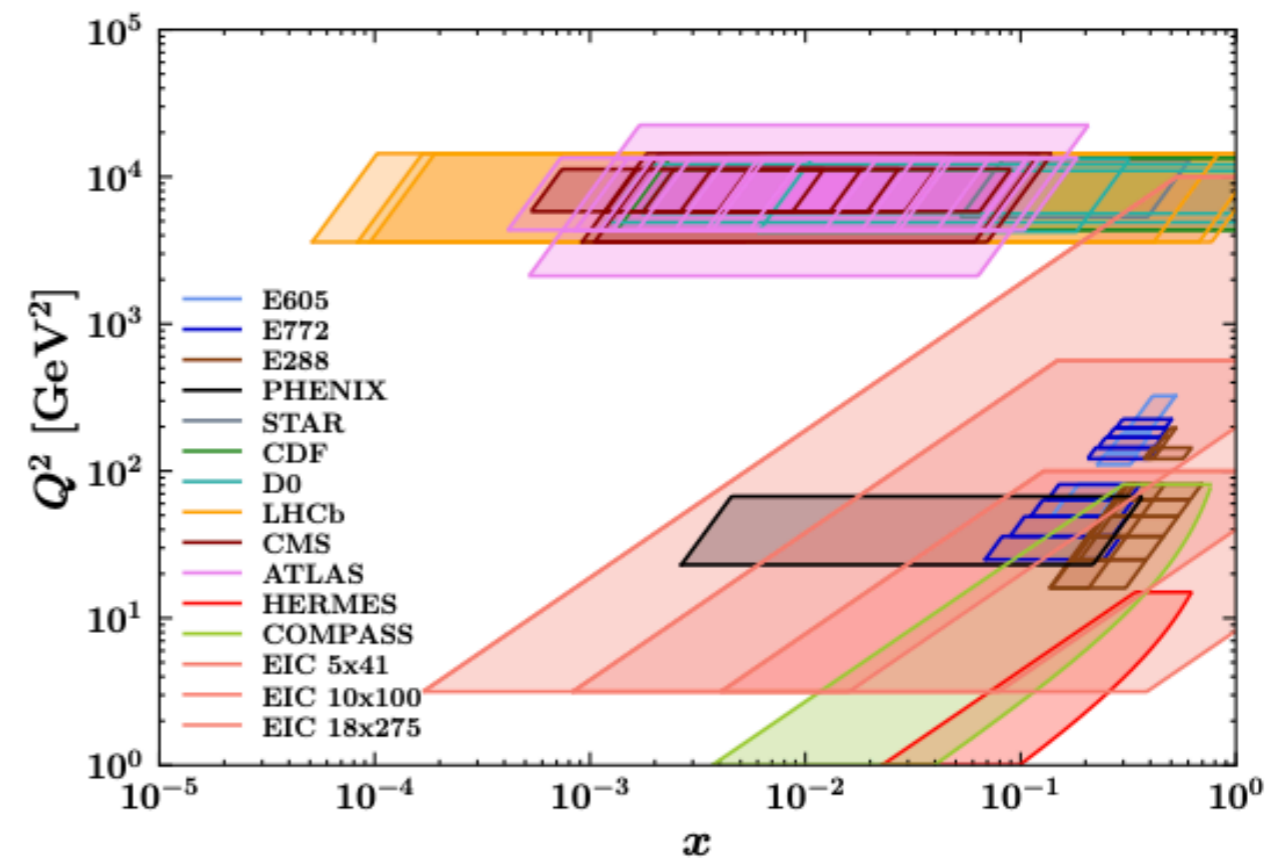
**What can we learn  
from new experiments?**

open channel with EIC  
SIDIS Working group

# Future perspectives

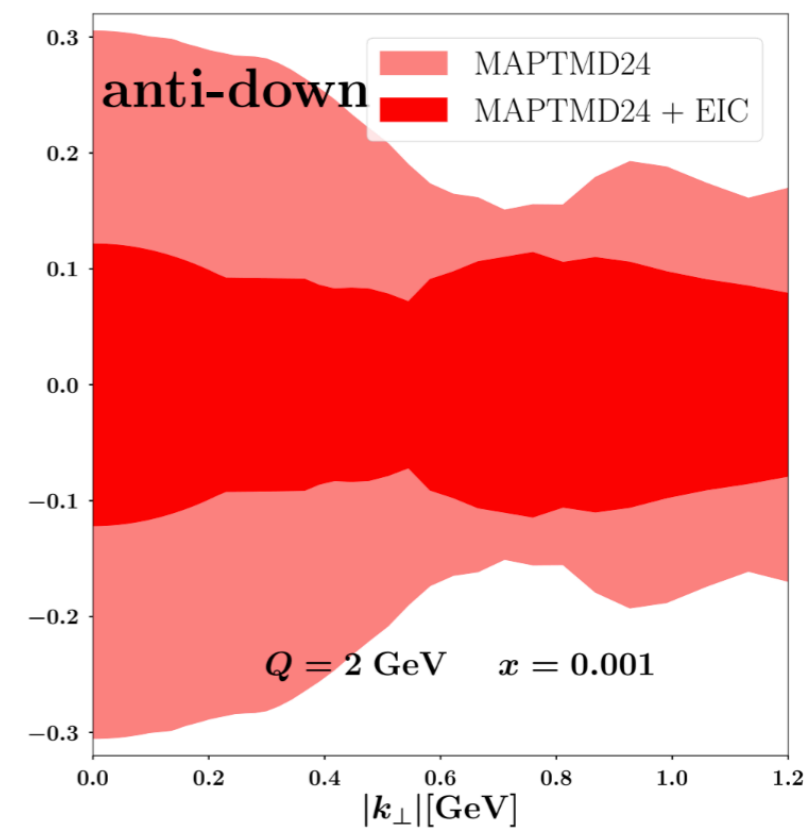
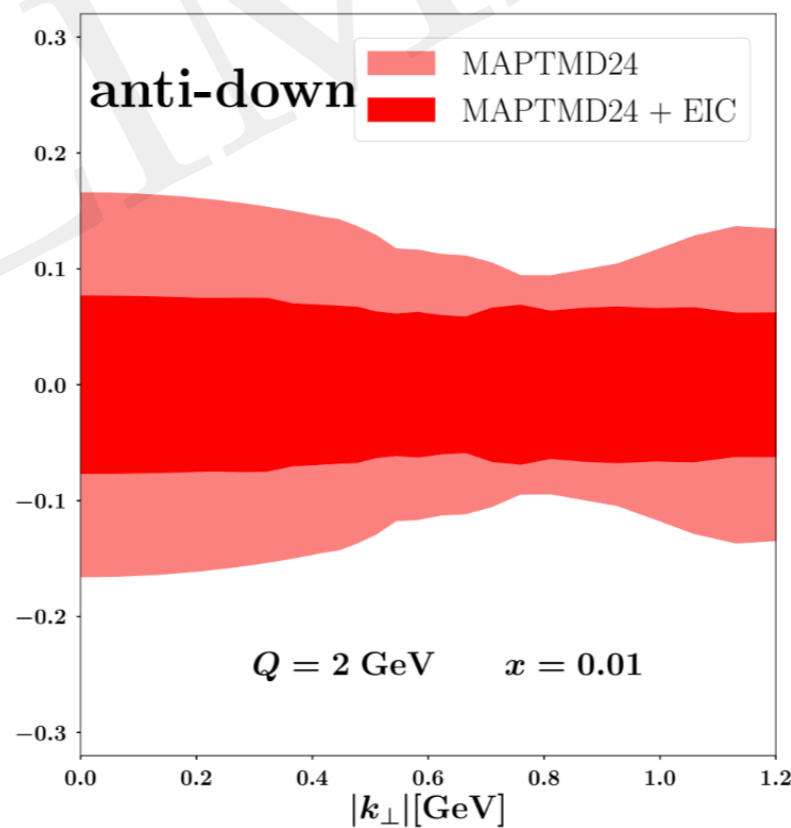
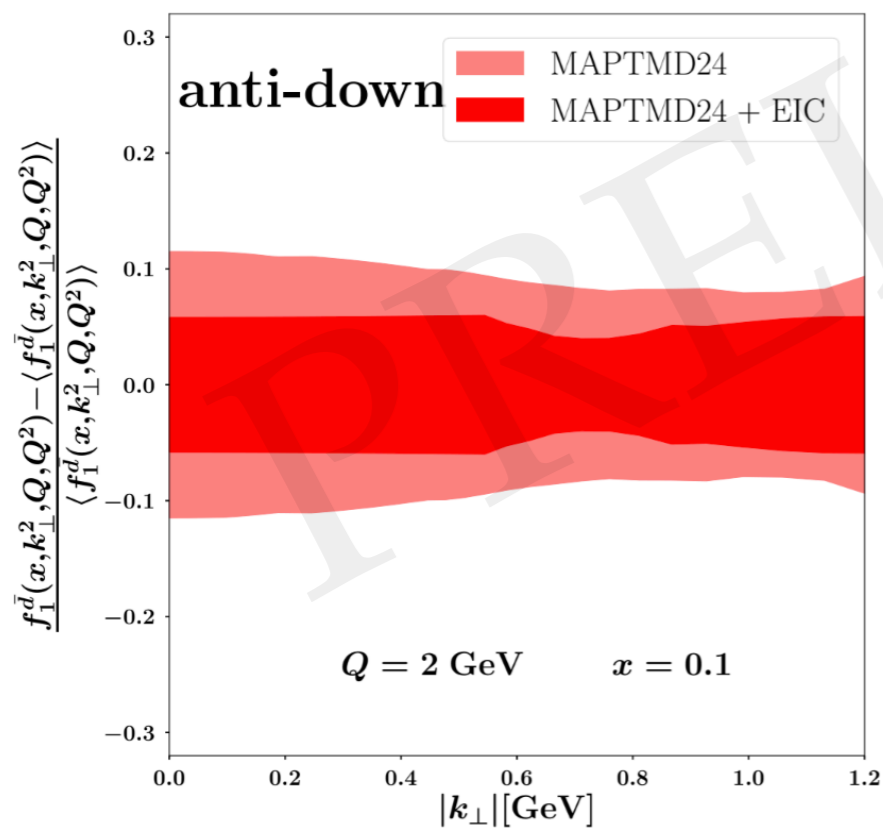
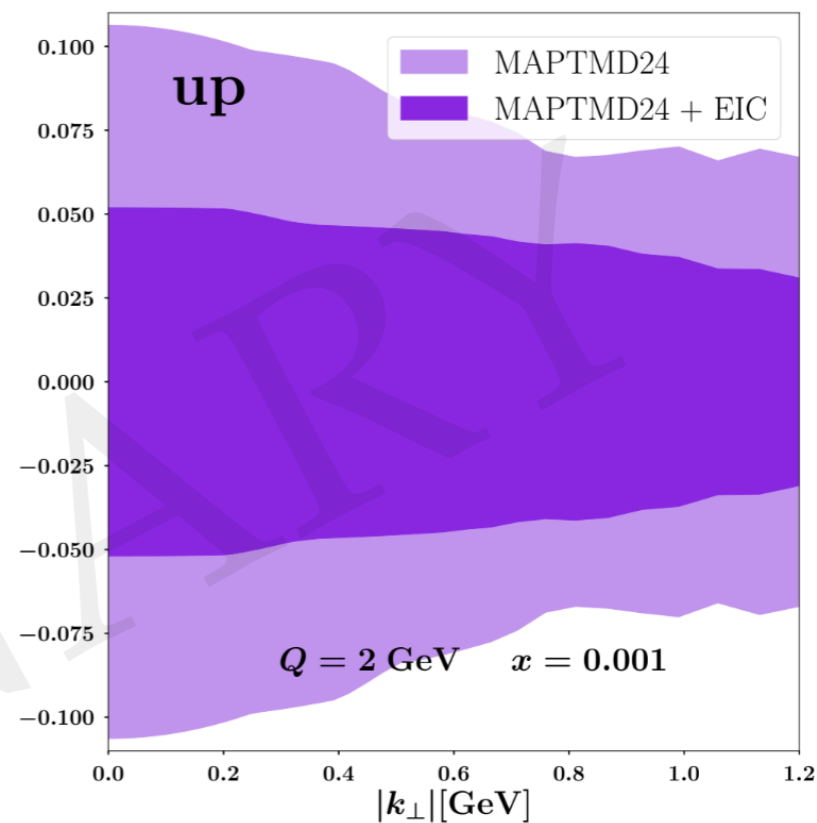
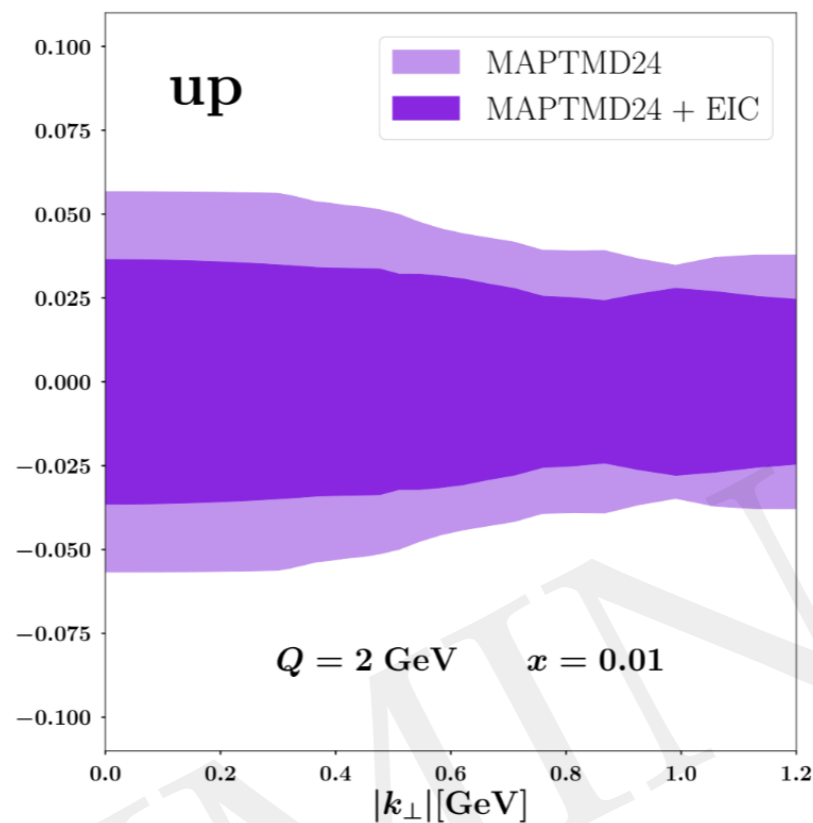
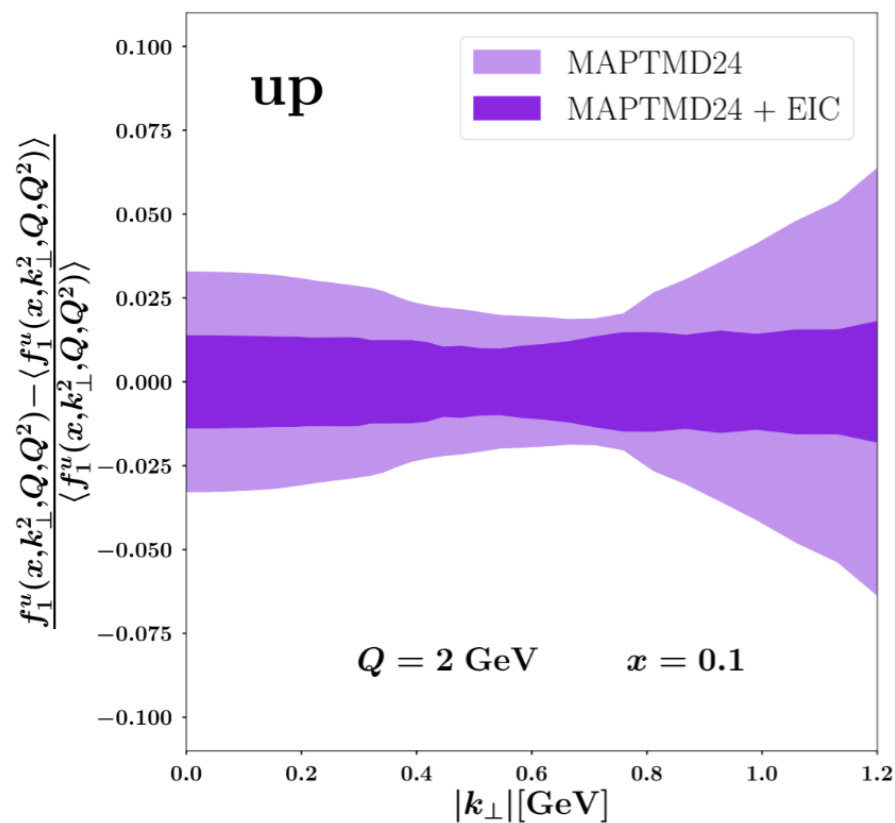
## Impact of the EIC on TMDs extractions

	Data points
MAPTMD24	2031
$5 \times 41 \pi^+$	1273
$10 \times 100 \pi^+$	1611
$18 \times 275 \pi^+$	1648
	<b>6563</b>



The number of data points has been **TRIPLED**

# Future perspectives



# Conclusions

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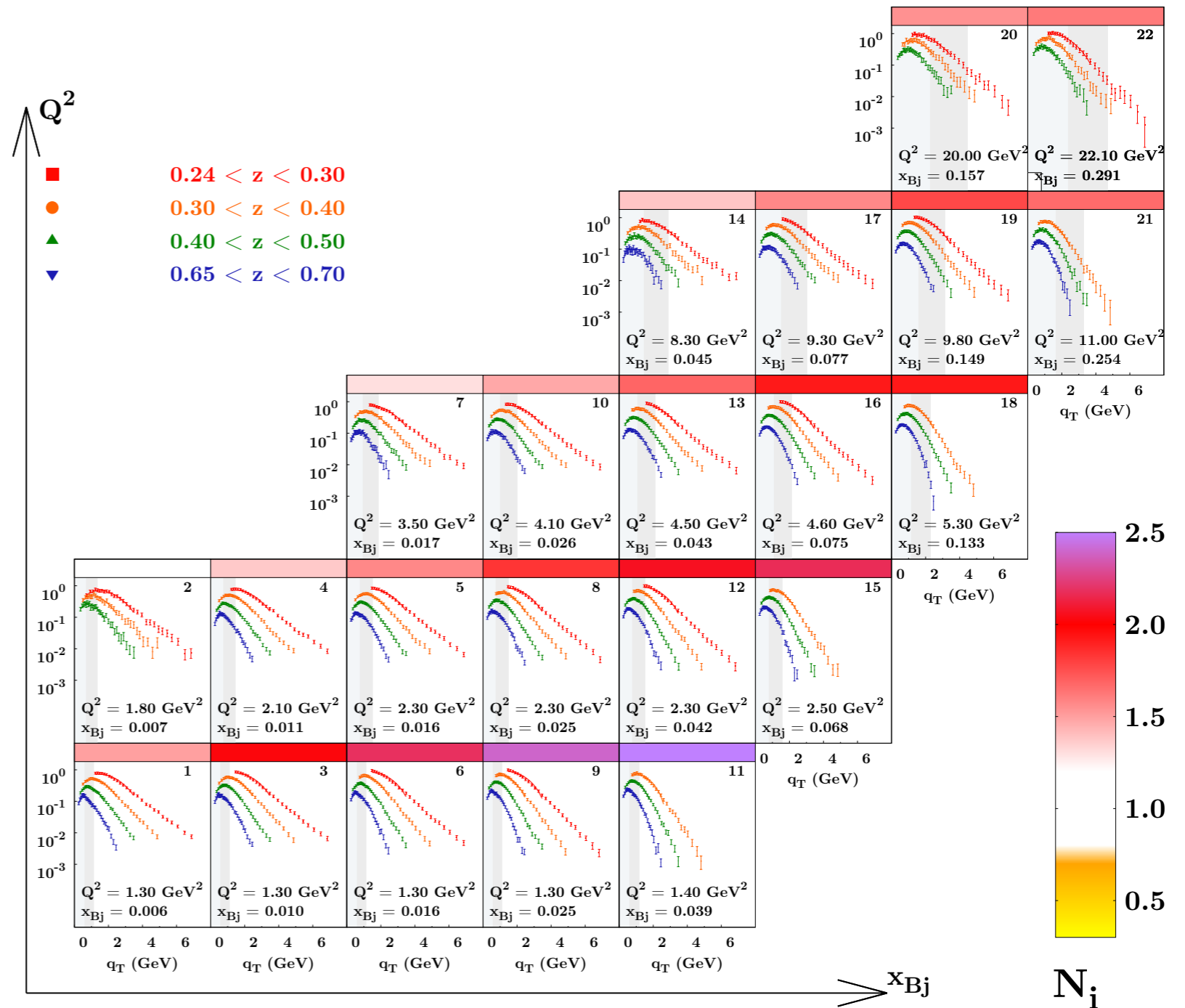
# Conclusions

- The extractions of **unpolarized quark TMDs** through global fits of experimental data have reached NNNLL accuracy
- **MAPTMD24** is the **first** simultaneous **extraction** of flavor-dependent unpolarized TMD PDFs and FF **through a global fit**
- We observe **non-trivial differences** in the transverse momentum distribution of partons inside hadrons
- New data from the **EIC** will be very important to **reduce the uncertainties** of extracted TMDs **by almost 50%**

**Backup**

# Normalization of SIDIS calculation

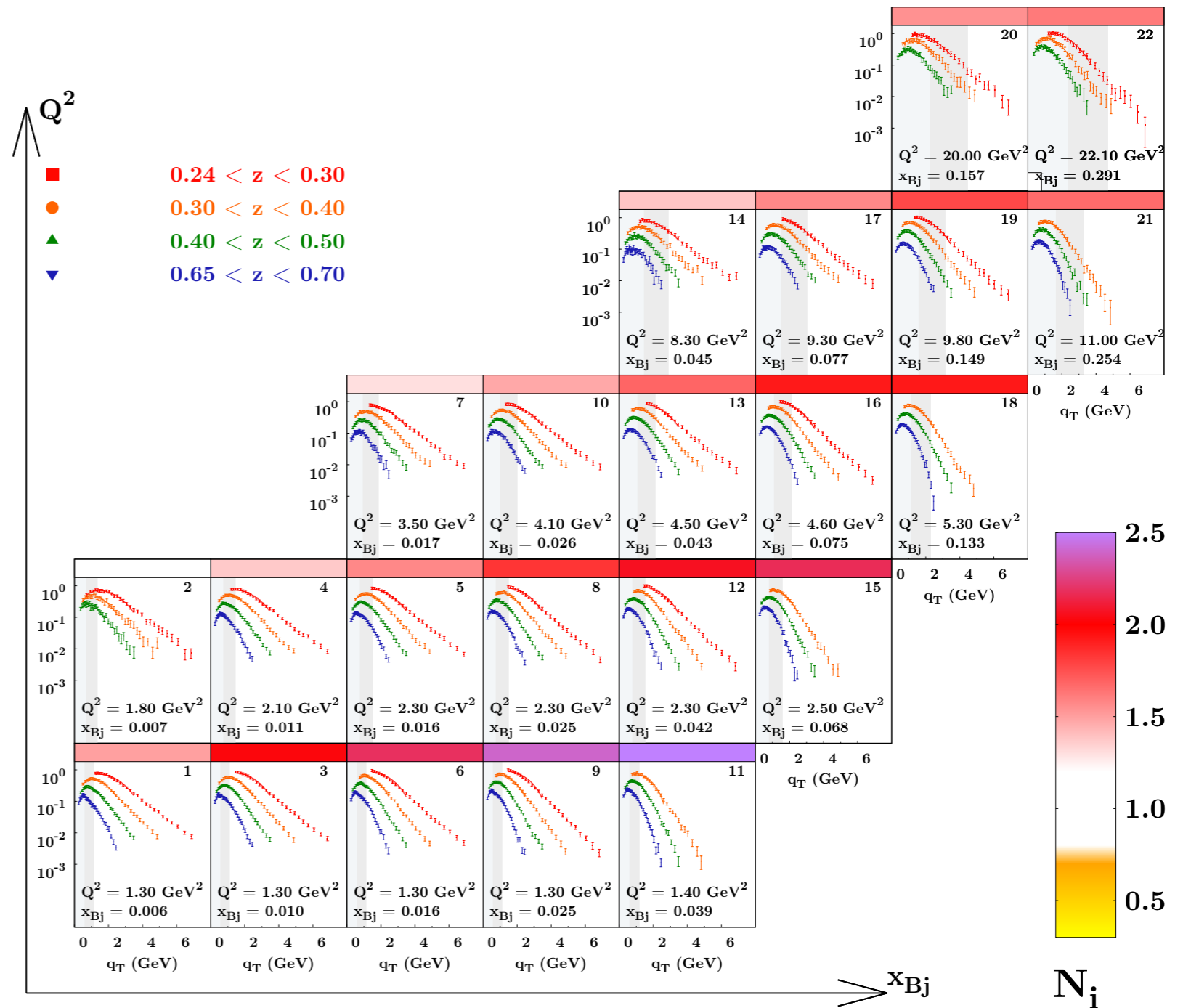
Normalization issue confirmed also in other analyses from different collaborations



# Normalization of SIDIS calculation

Normalization issue confirmed also in other analyses from different collaborations

Sun, Isaacson, Yuan, Yuan, IJNP A (2014)  
 Gonzalez-Hernandez, PoS DIS2019 (2019)  
 Vladimirov, JHEP 12 (2023)



# Normalization of SIDIS calculation

## MAP22 work solution

Good agreement for almost all bins

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SIDIS multiplicity

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

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Collinear SIDIS cross section

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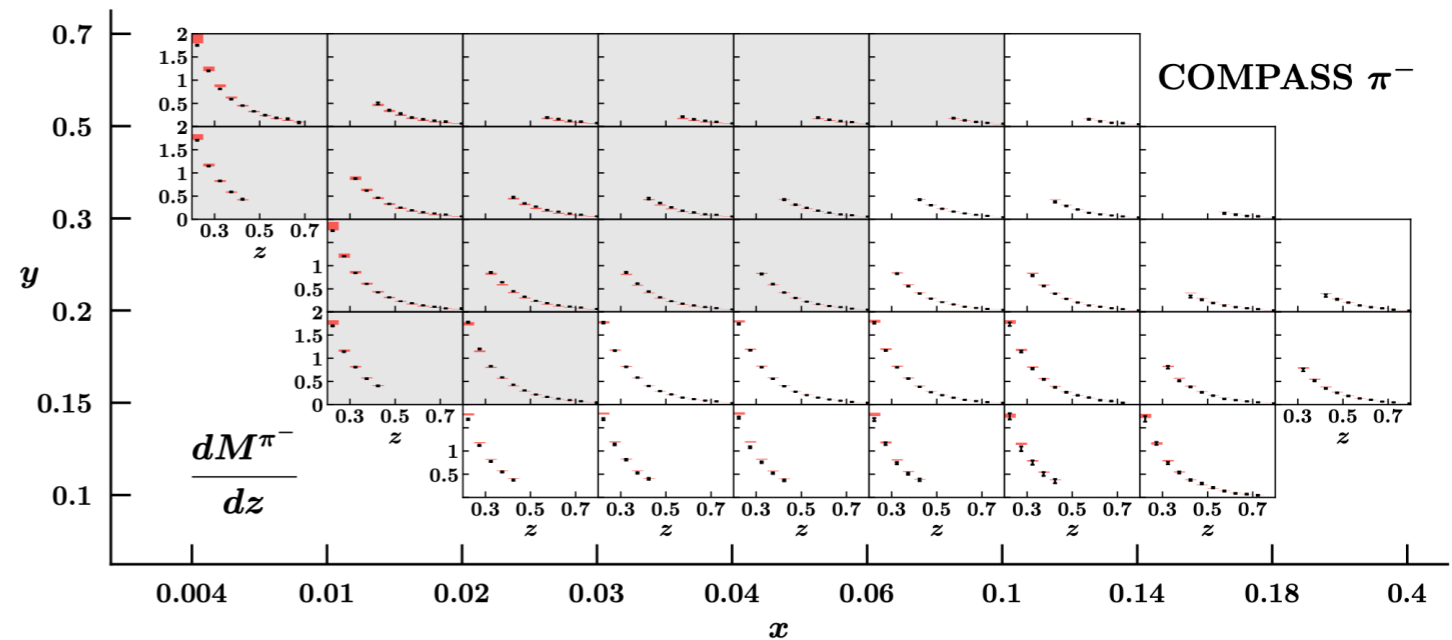
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Good agreement theory/data



Khalek, Bertone, Nocera, et al., PRD 104 (2021)

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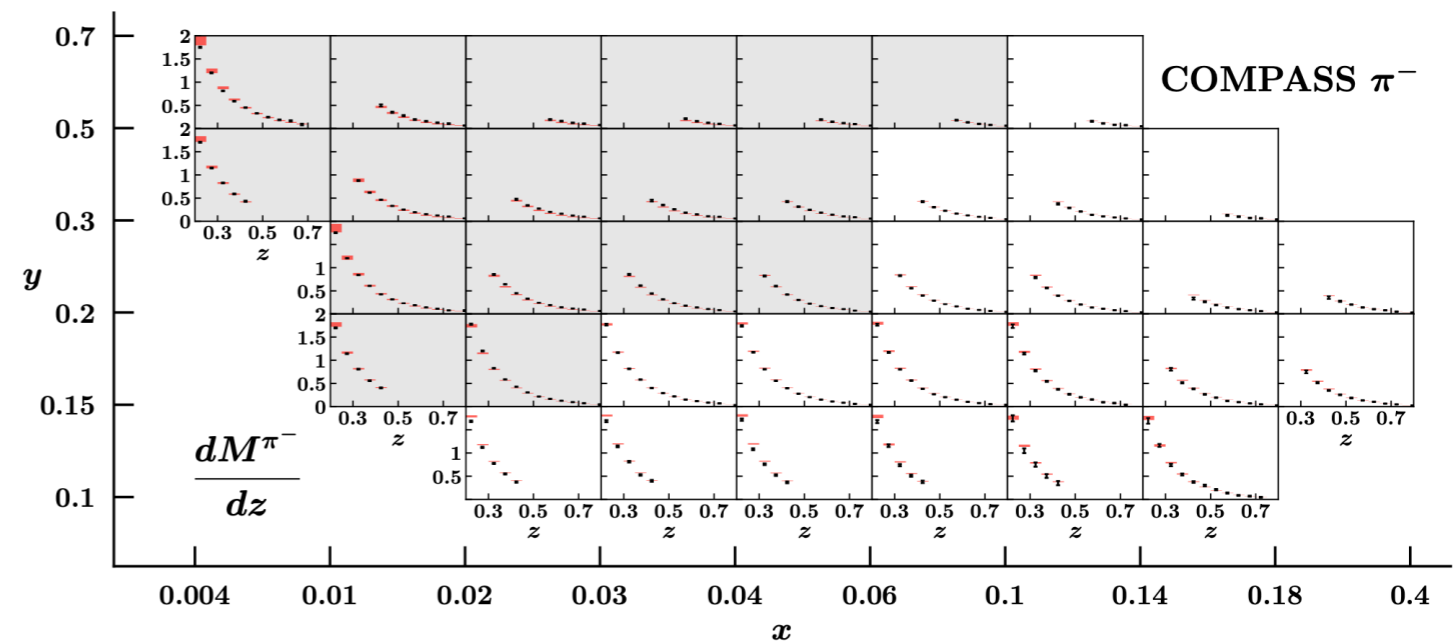
Collinear SIDIS cross section

Normalization of prediction such that

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

Piacenza, PhD thesis (2020)

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Khalek, Bertone, Nocera, et al., PRD 104 (2021)

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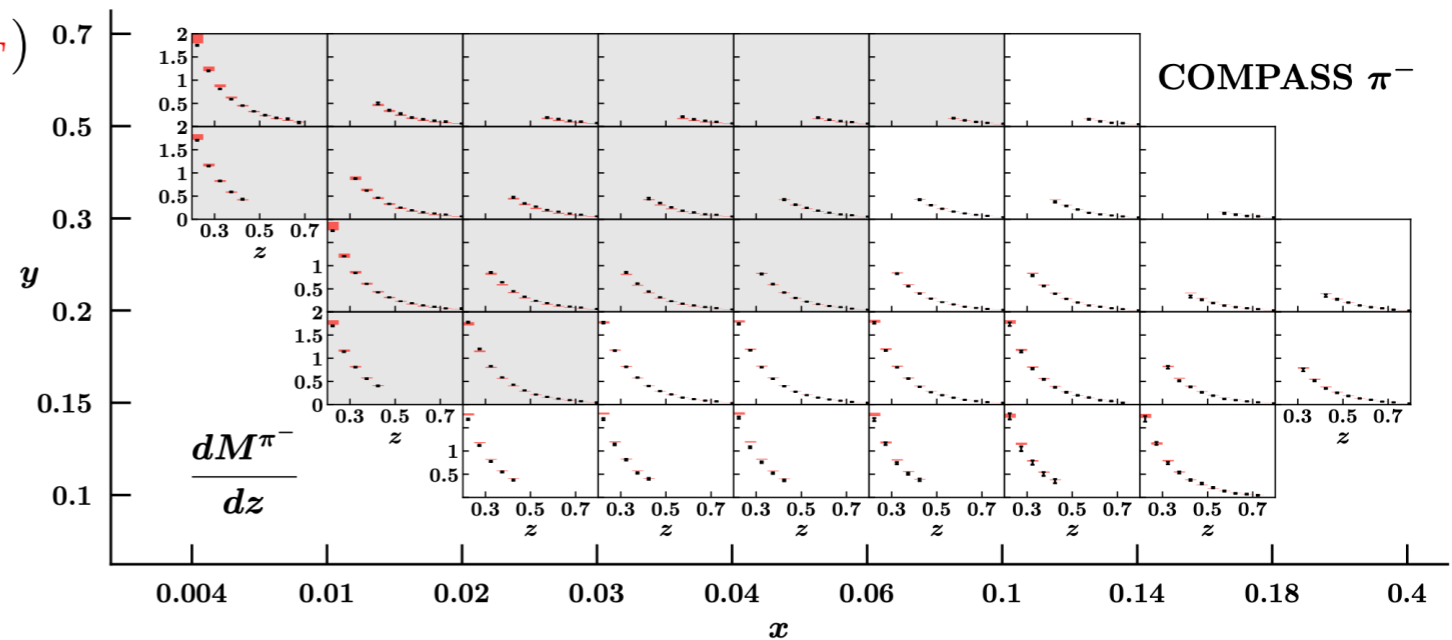
Collinear SIDIS cross section

Normalization of prediction such that

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

Piacenza, PhD thesis (2020)

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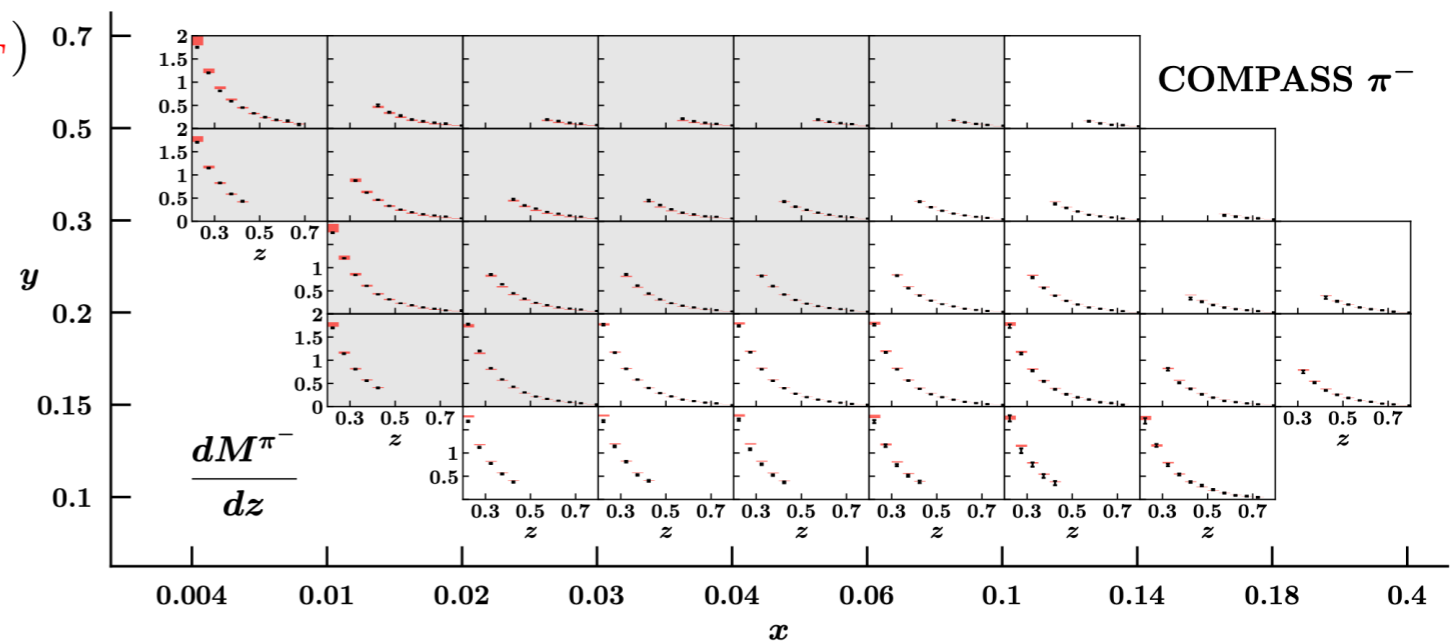
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Khalek, Bertone, Nocera, et al., PRD 104 (2021)

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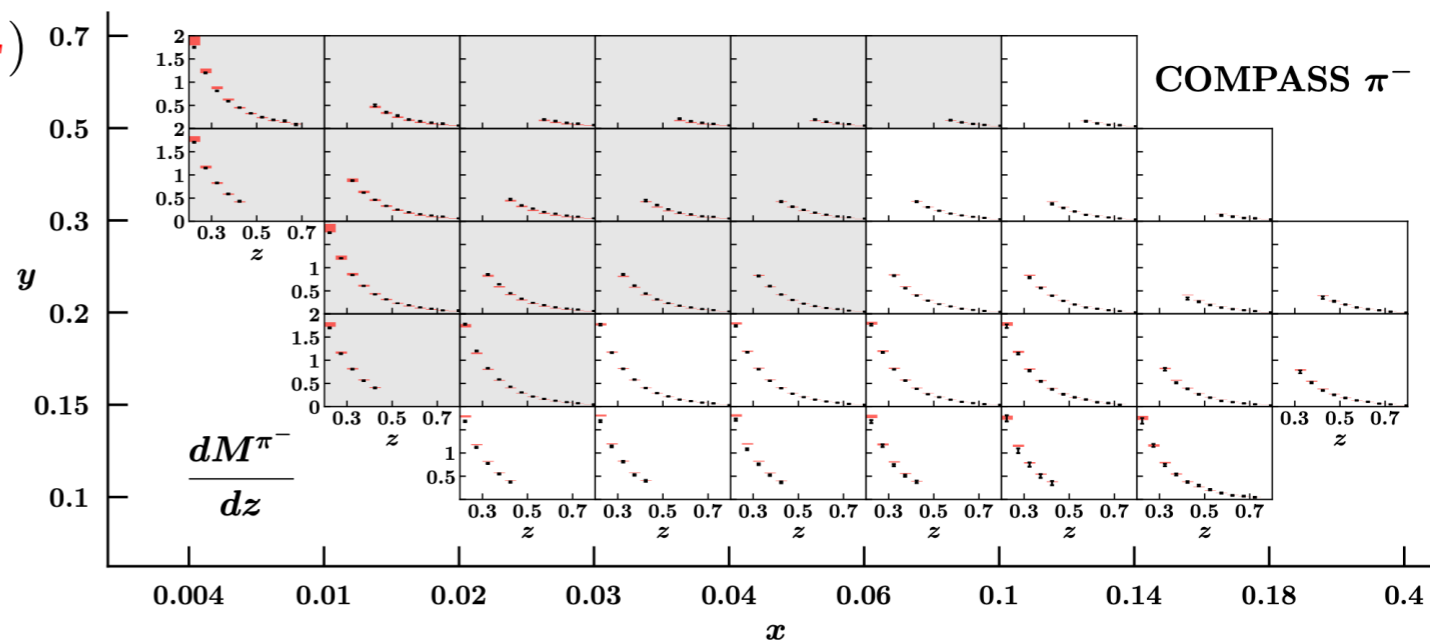
Piacenza, PhD thesis (2020)

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

**Calculable before the fit**

**Good agreement for almost all bins**

**Good agreement theory/data**



Khalek, Bertone, Nocera, et al., PRD 104 (2021)

# MAPTMD22 — Error analysis

Error propagation



**100 Monte Carlo replicas of data**

100 Monte Carlo replicas of PDFs

100 Monte Carlo replicas of FFs

