



# New insights on flavor dependence in TMD extractions from global fits

arXiv:2405.138833

Matteo Cerutti - MAP Collaboration



# Transverse-Momentum Distributions (TMDs)

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

Non-collinear framework

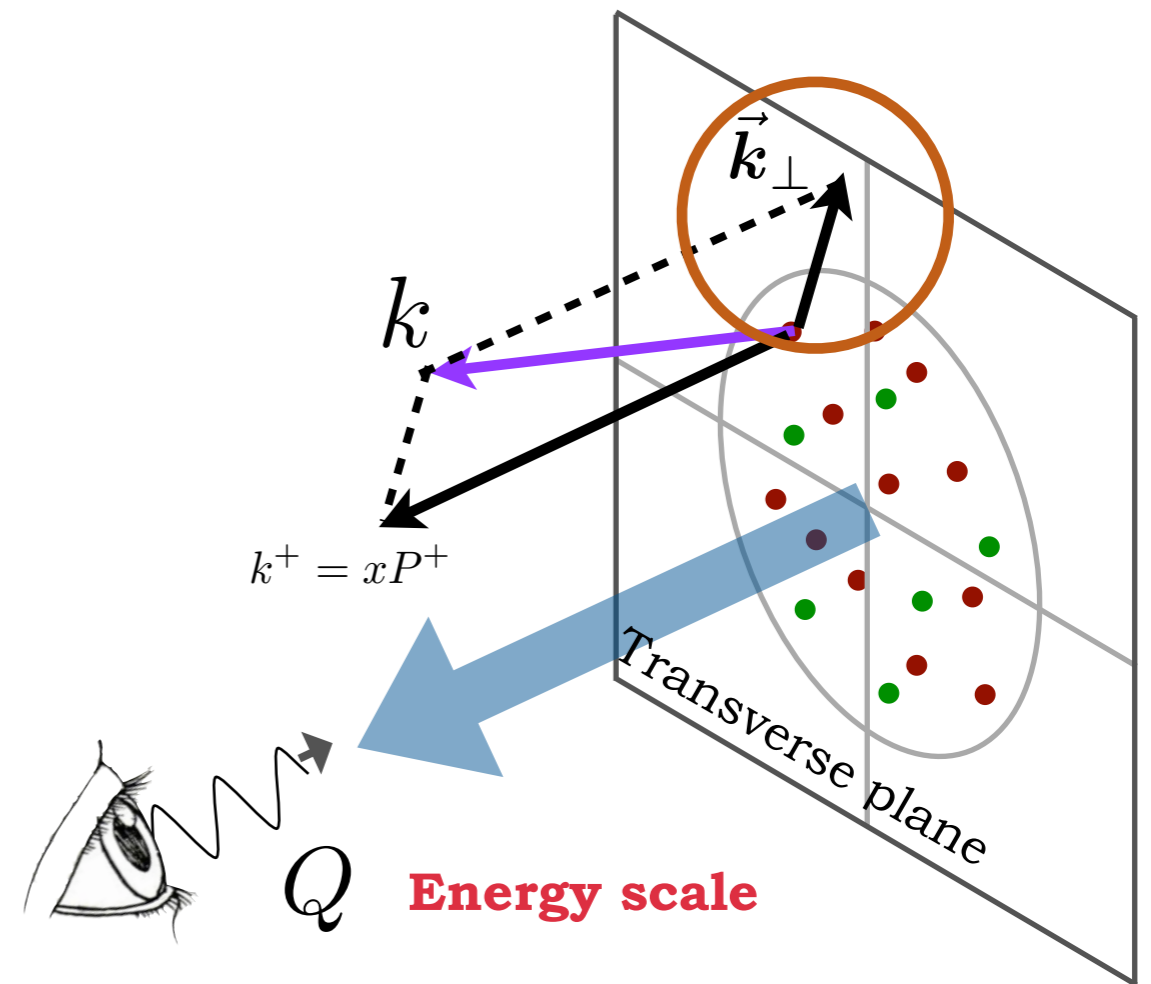
Quark Polarization

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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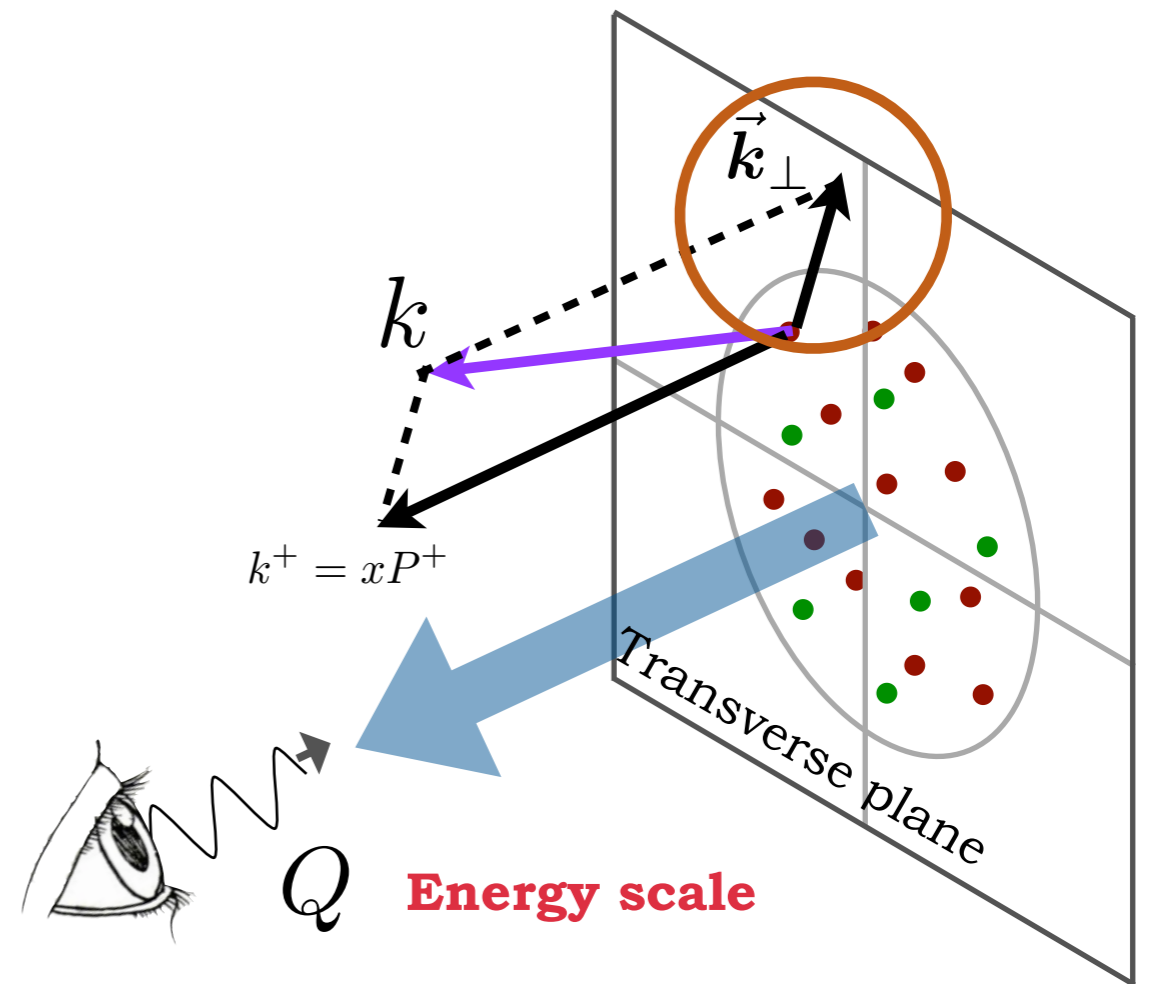
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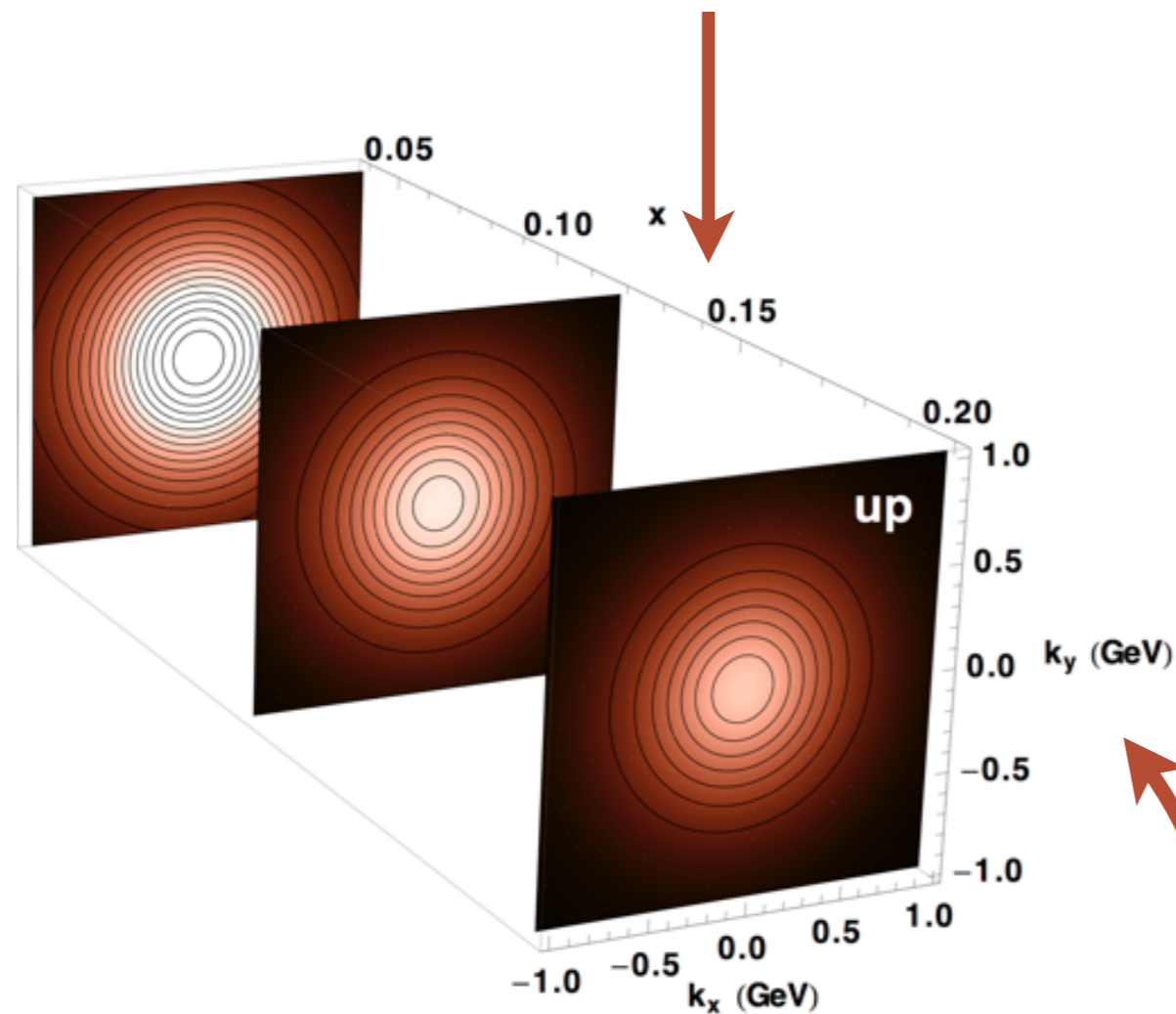


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Fraction of longitudinal momentum



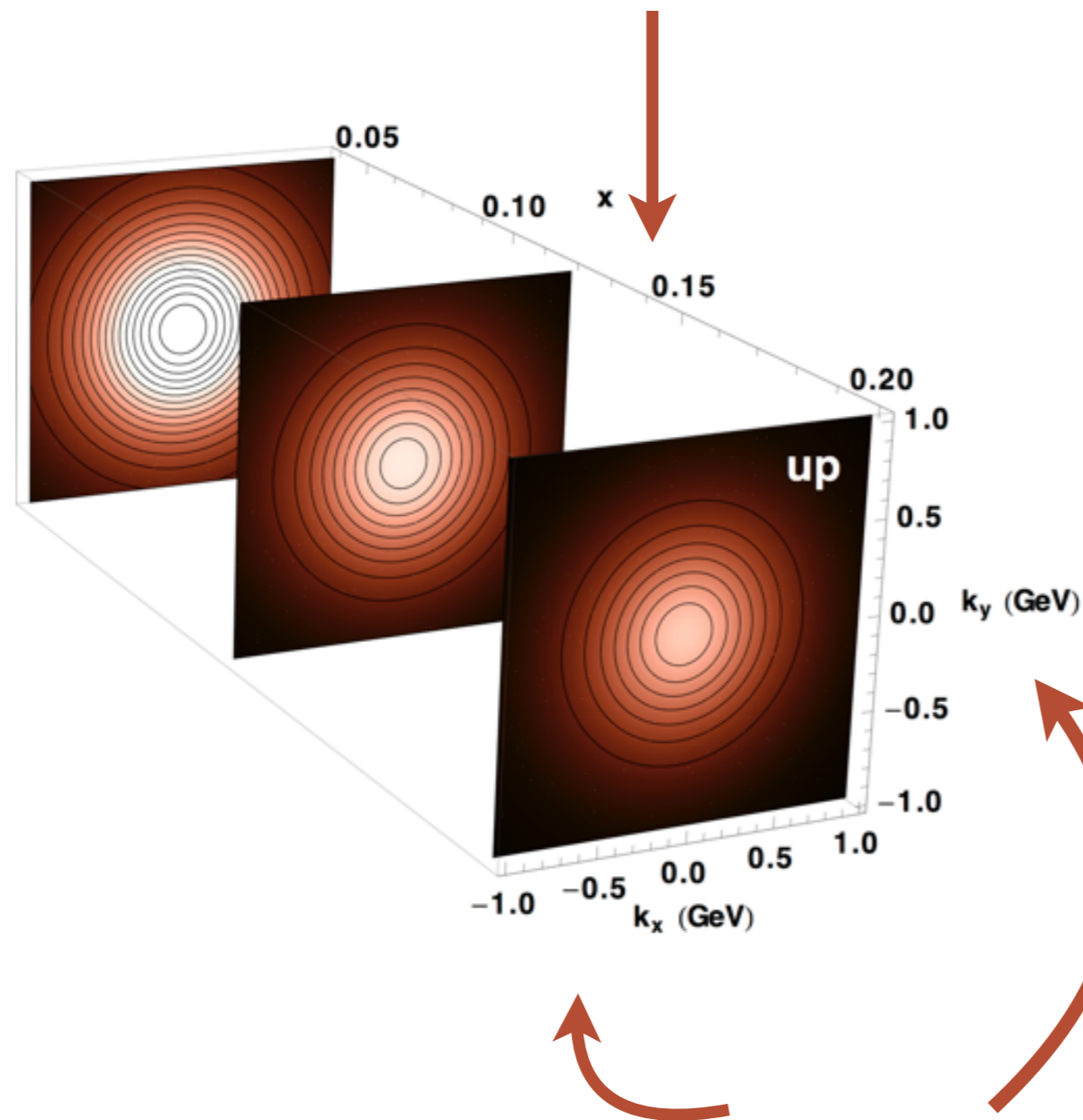
**TMDs** map the distribution of partons inside the nucleon in 3D in momentum space.

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  $x$ ?**

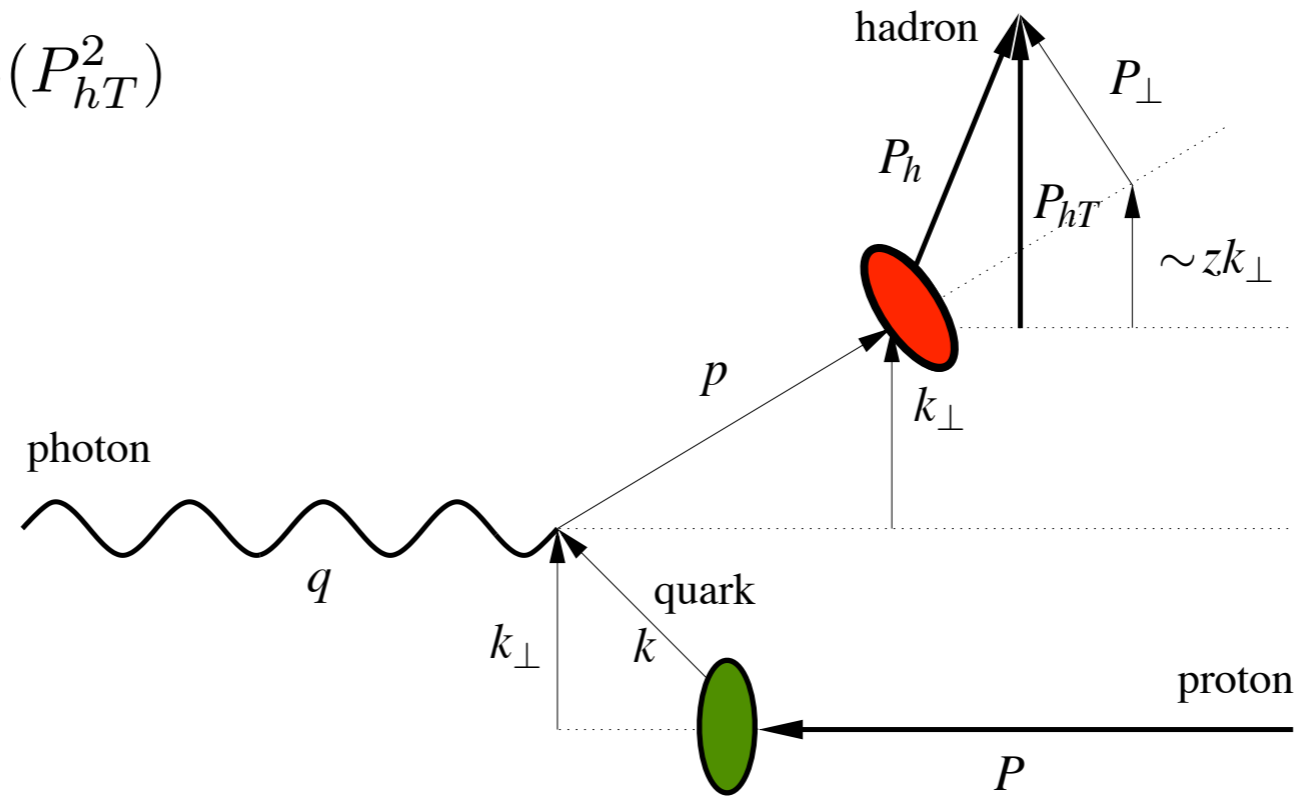
**Do they depend on the quark flavor?**

Transverse momentum

# TMD factorization: SIDIS

## Semi-Inclusive Deep-Inelastic Scattering

If  $Q^2 \gg M^2$  and  $Q^2 \gg q_T^2 (P_{hT}^2)$



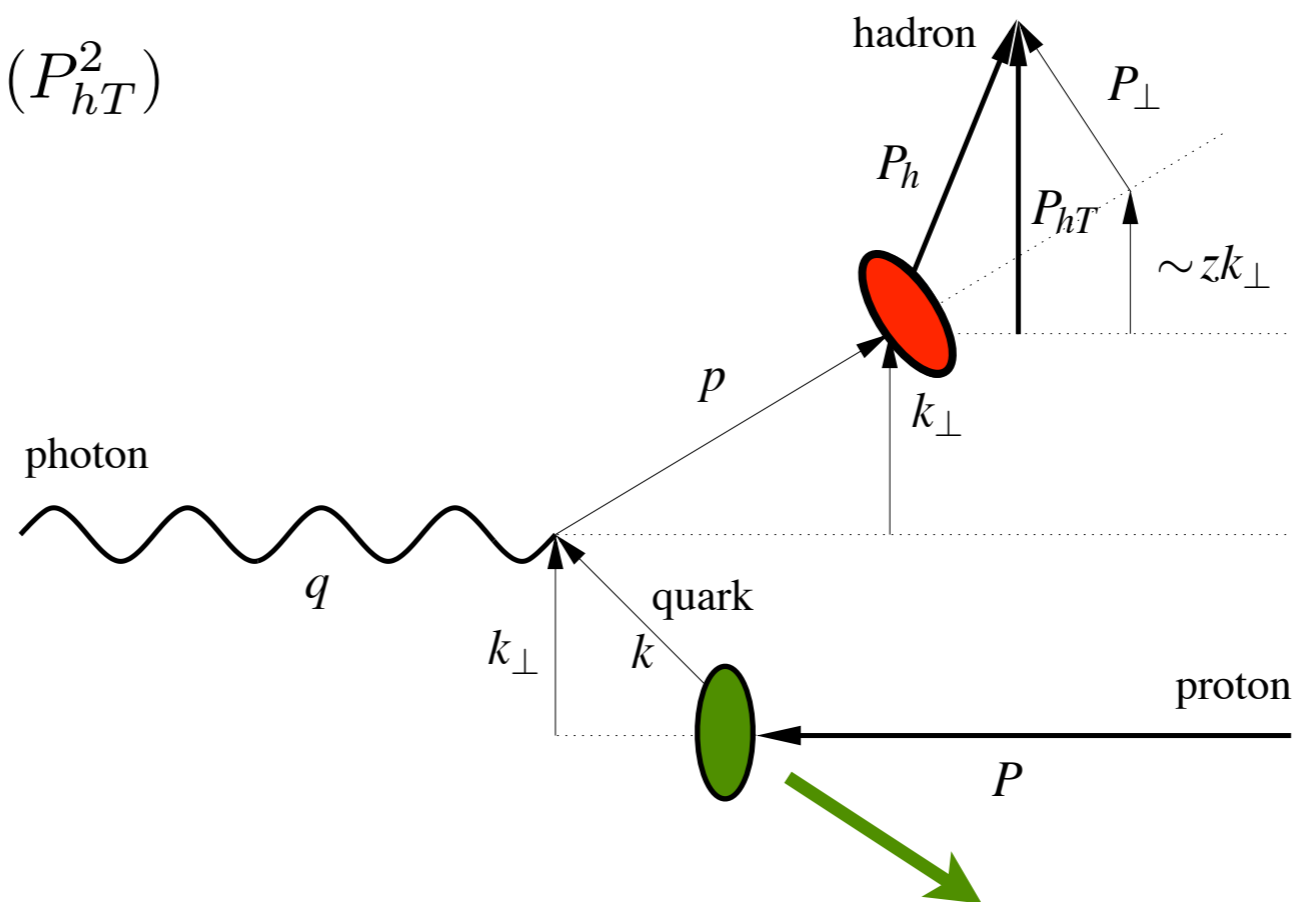
$$F_{UU,T}(x, z, |\mathbf{q}_T|, Q) = \frac{x}{2\pi} \mathcal{H}^{SIDIS}(Q, \mu) \sum_{a=q, \bar{q}} e_a^2 \int_0^{+\infty} d|\mathbf{b}_T| |\mathbf{b}_T| J_0(|\mathbf{b}_T| |\mathbf{q}_T|) \hat{f}_1^a(x, b_T^2; \mu, \zeta_A) \hat{D}_1^{a \rightarrow h}(z, b_T^2; \mu, \zeta_B) + Y_{UU,T}(Q^2, \mathbf{P}_{hT}^2) + \mathcal{O}(M^2/Q^2)$$

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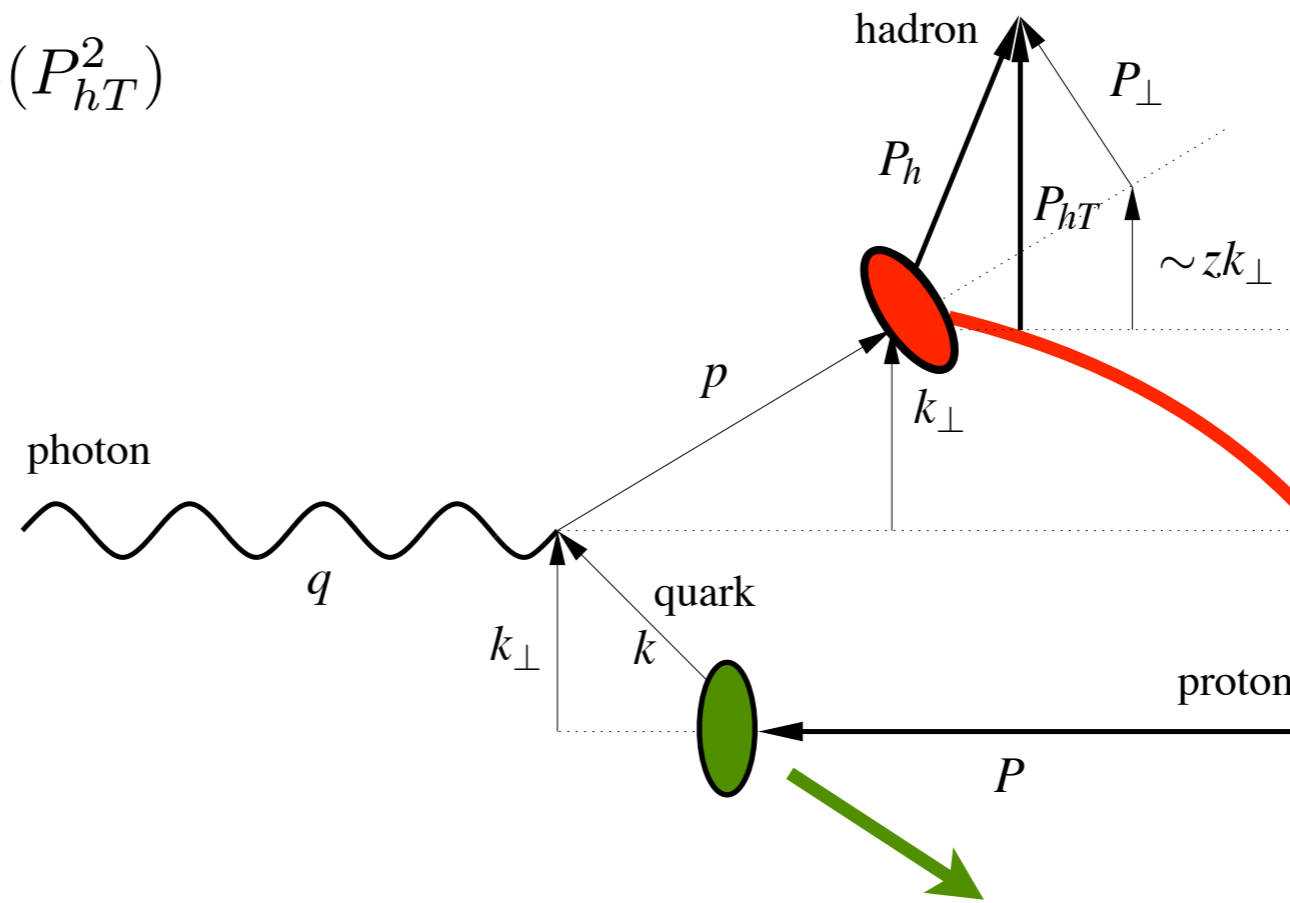
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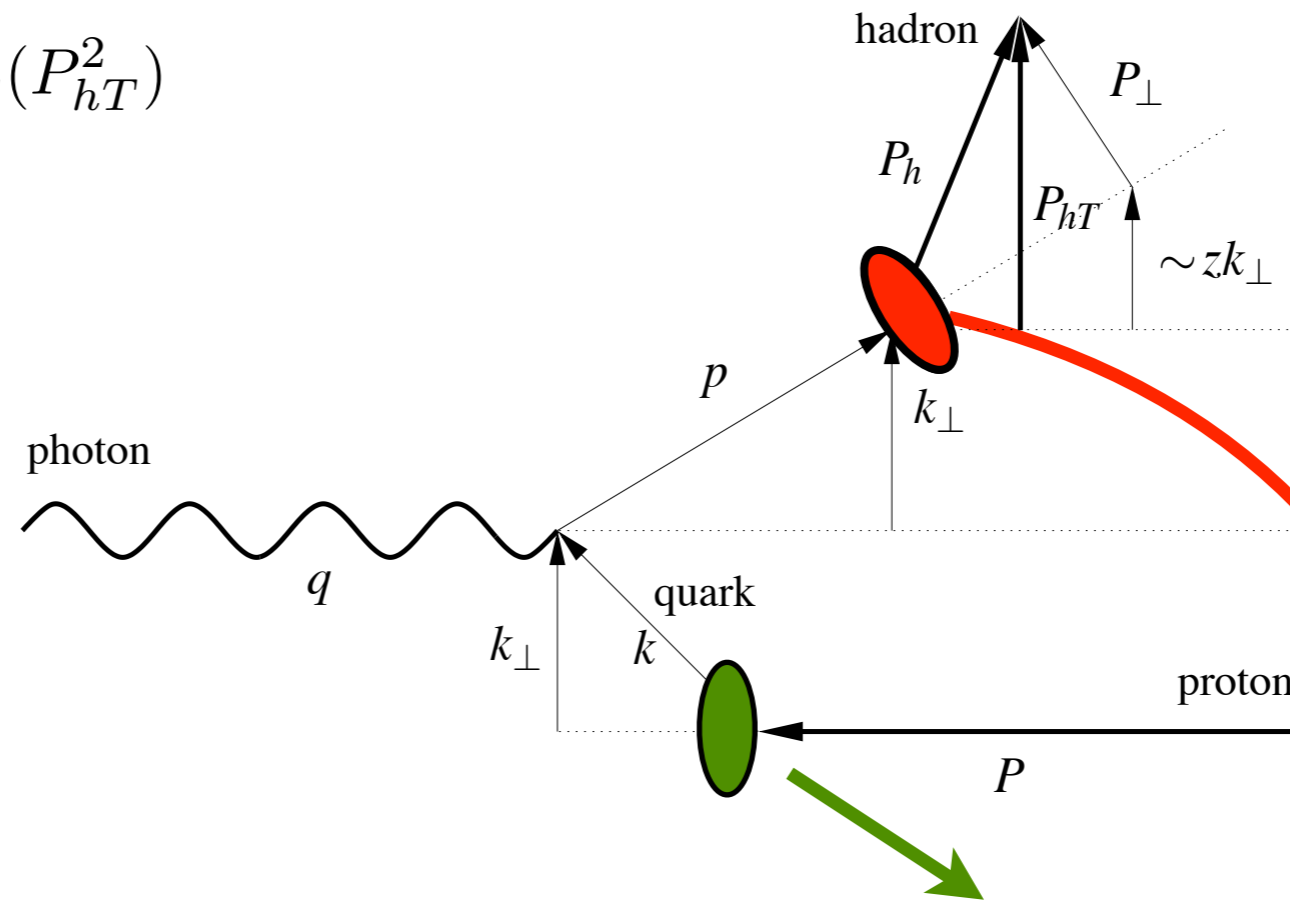
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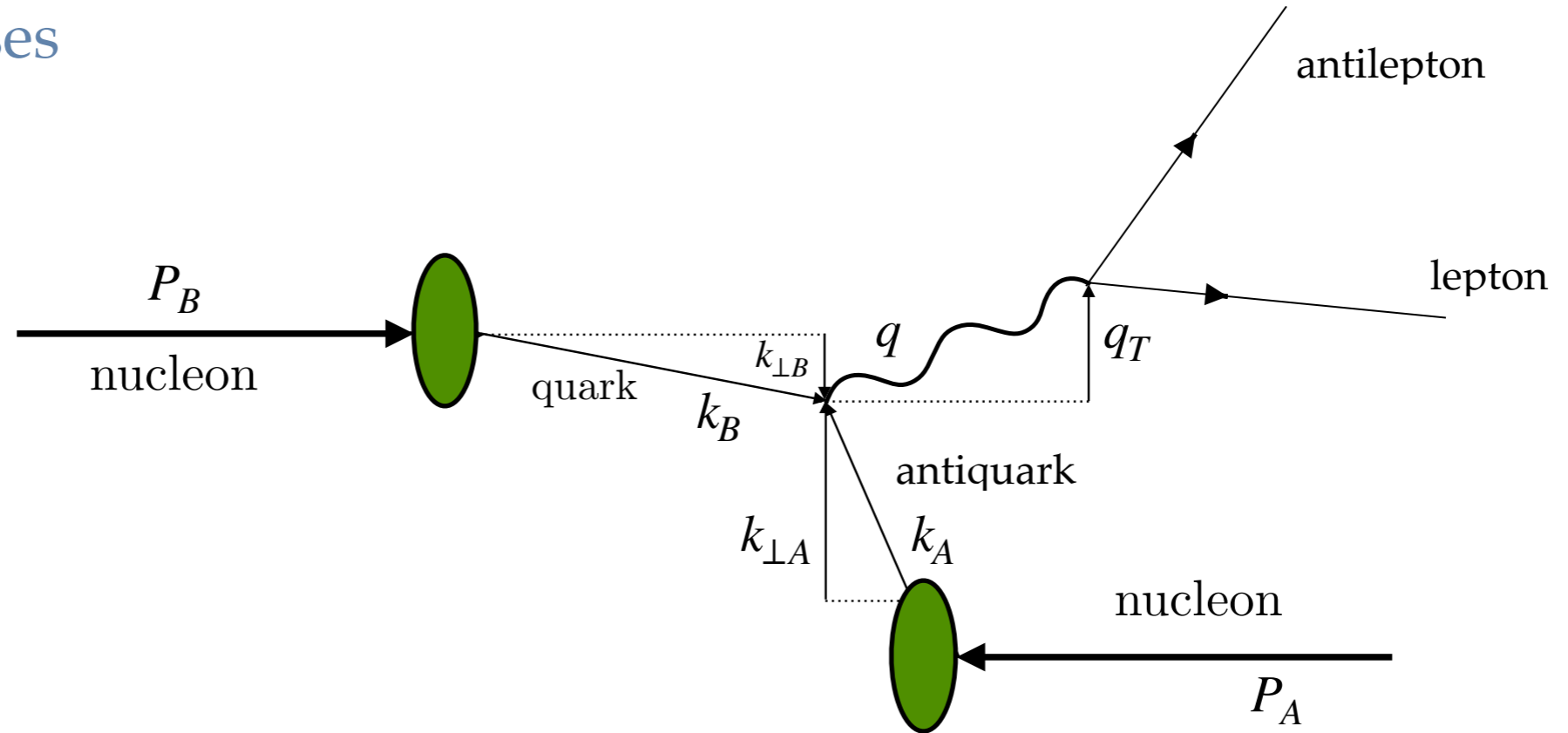


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- The **W term** dominates in the region where  $\mathbf{q}_T \ll Q$
- The Y term has been excluded in the analysis

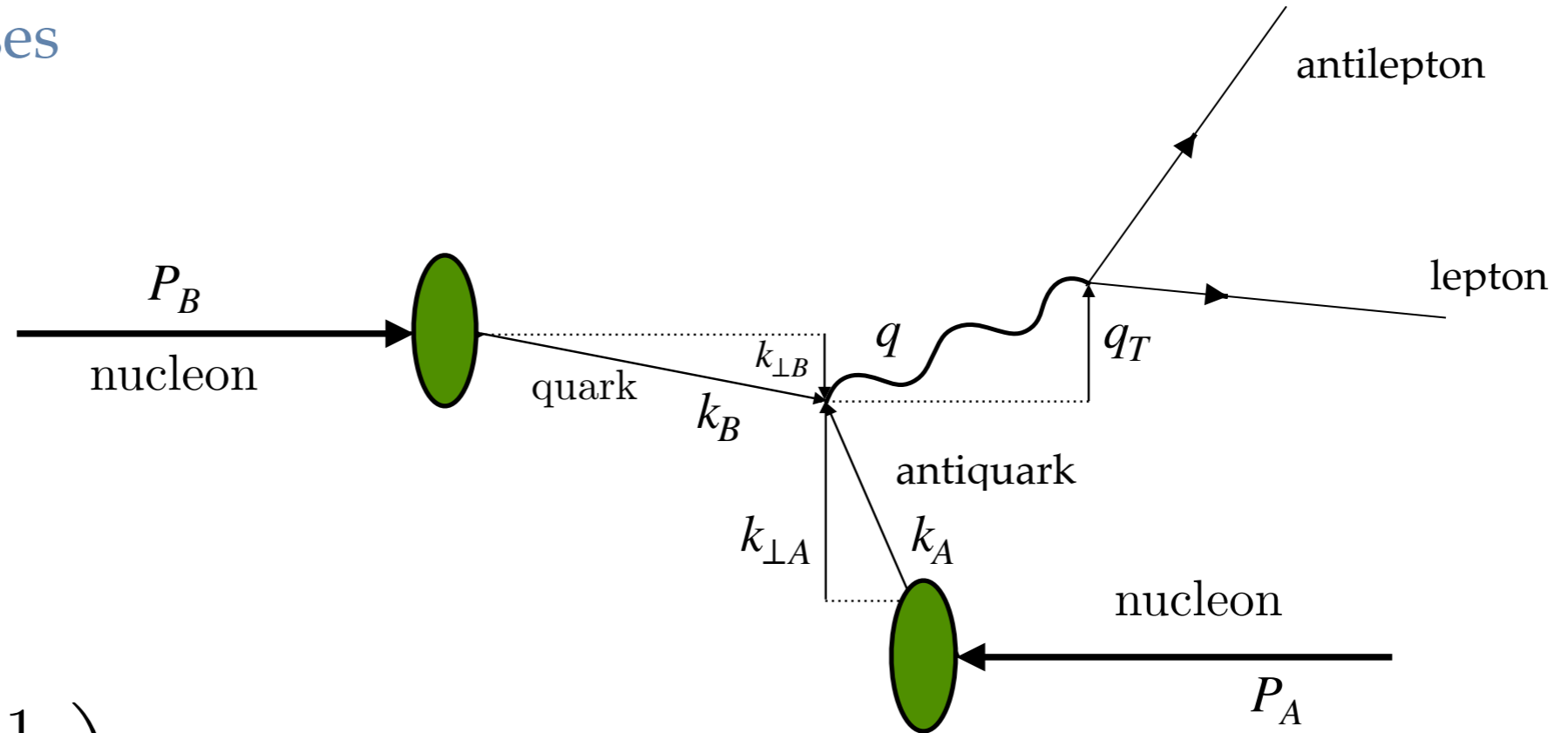
# TMD factorization: Drell-Yan

## Drell-Yan processes



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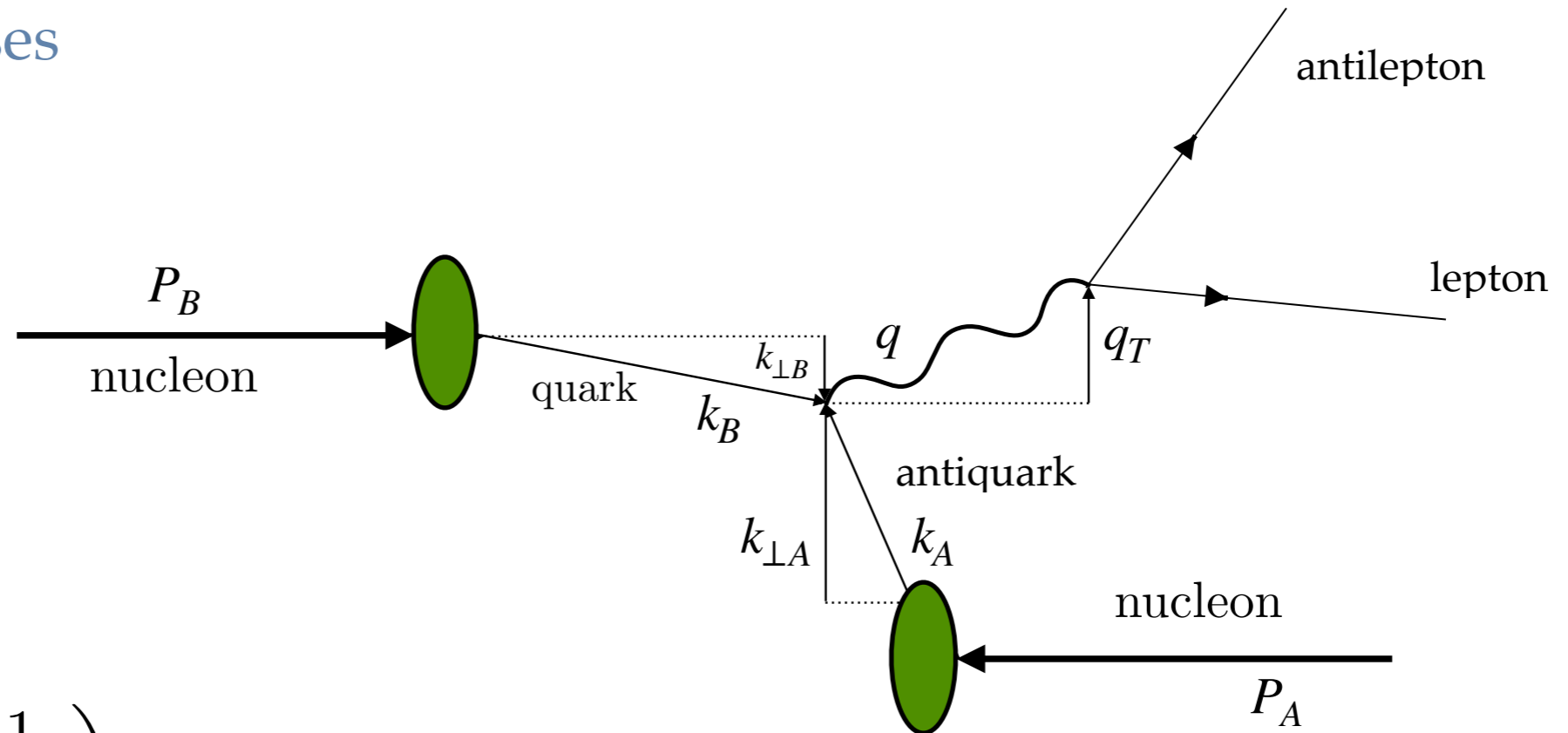
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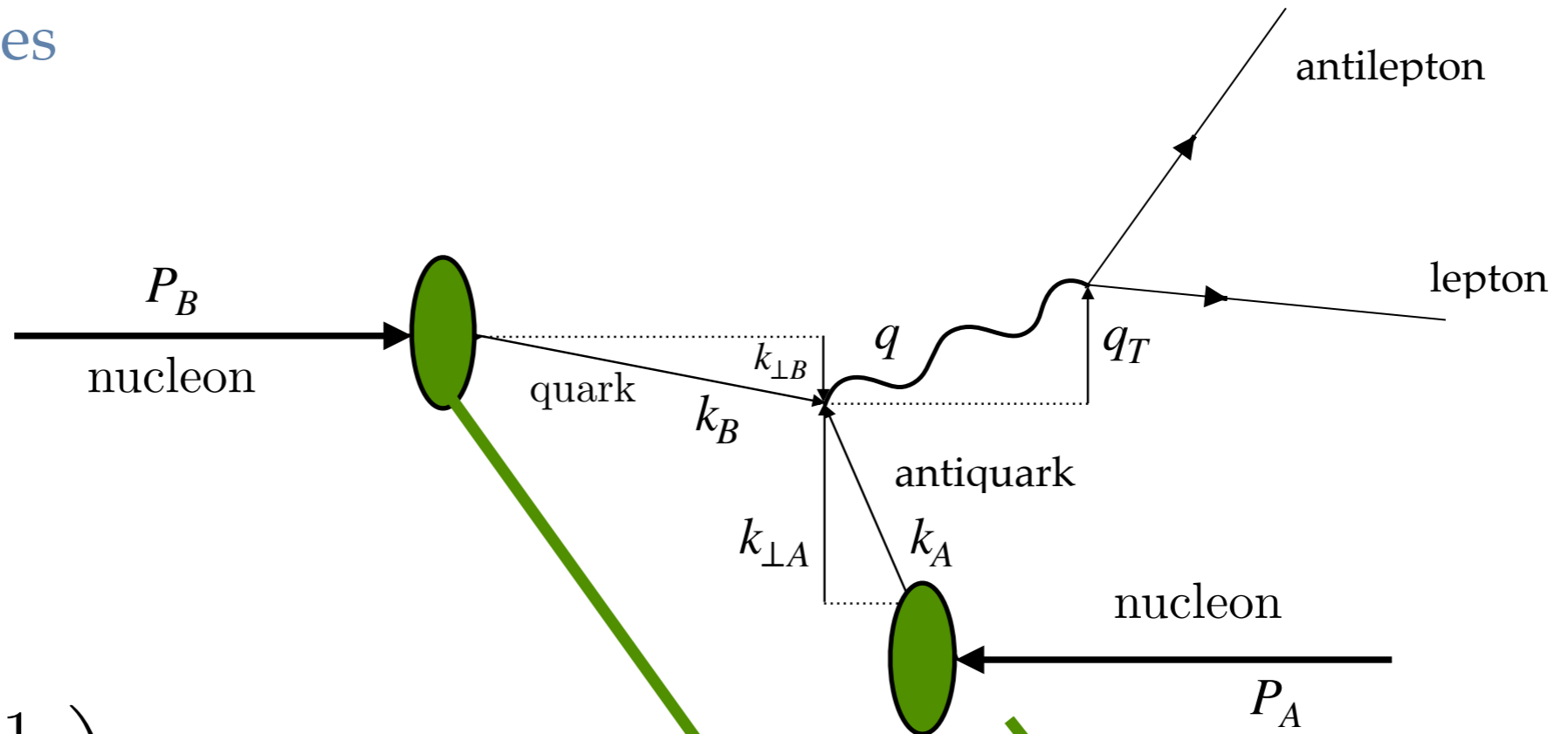
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# Structure of a TMD

See R. Kishore's talk

TMD in Fourier space

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Perturbative TMD at the initial scale



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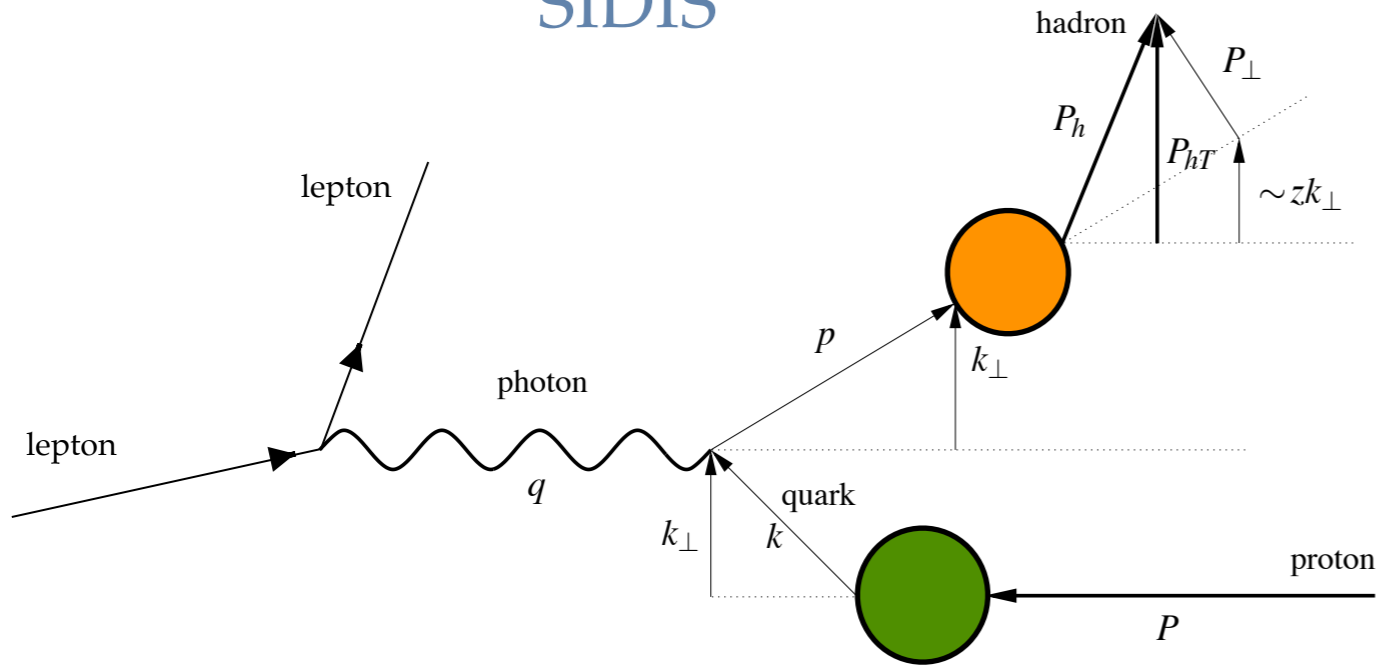
Parameterization

# TMD factorization: Universality

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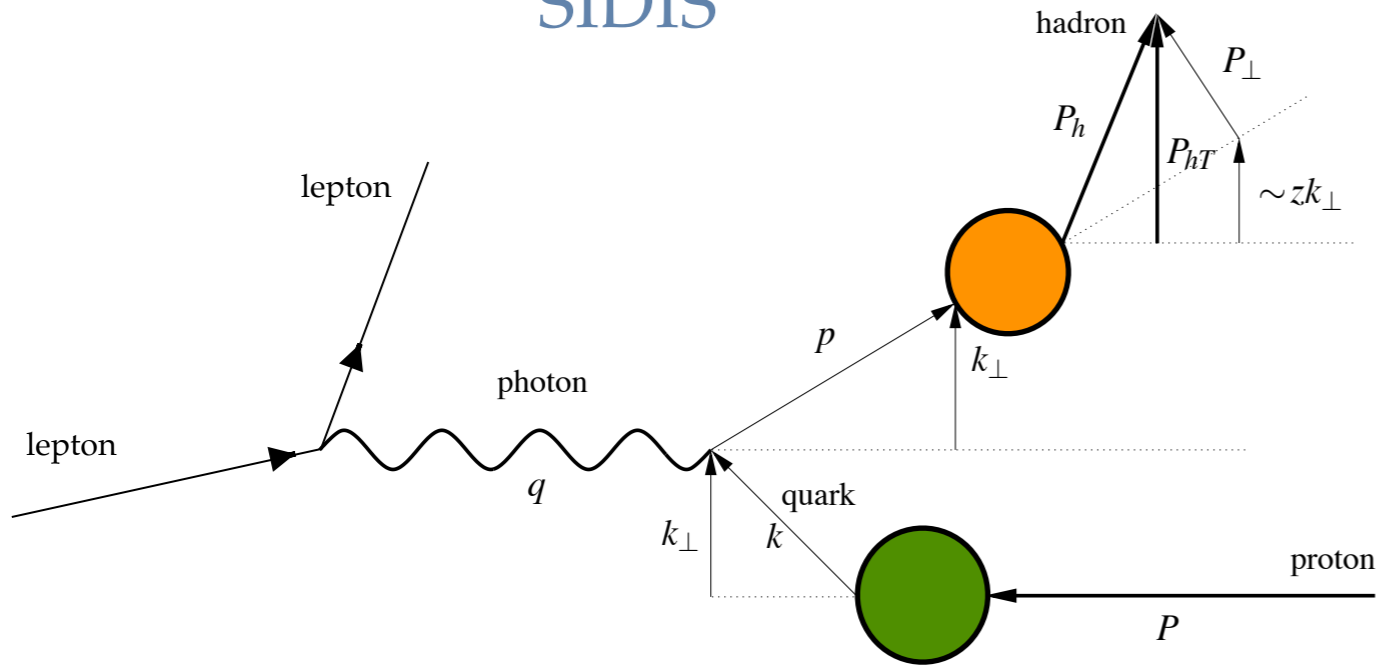
SIDIS



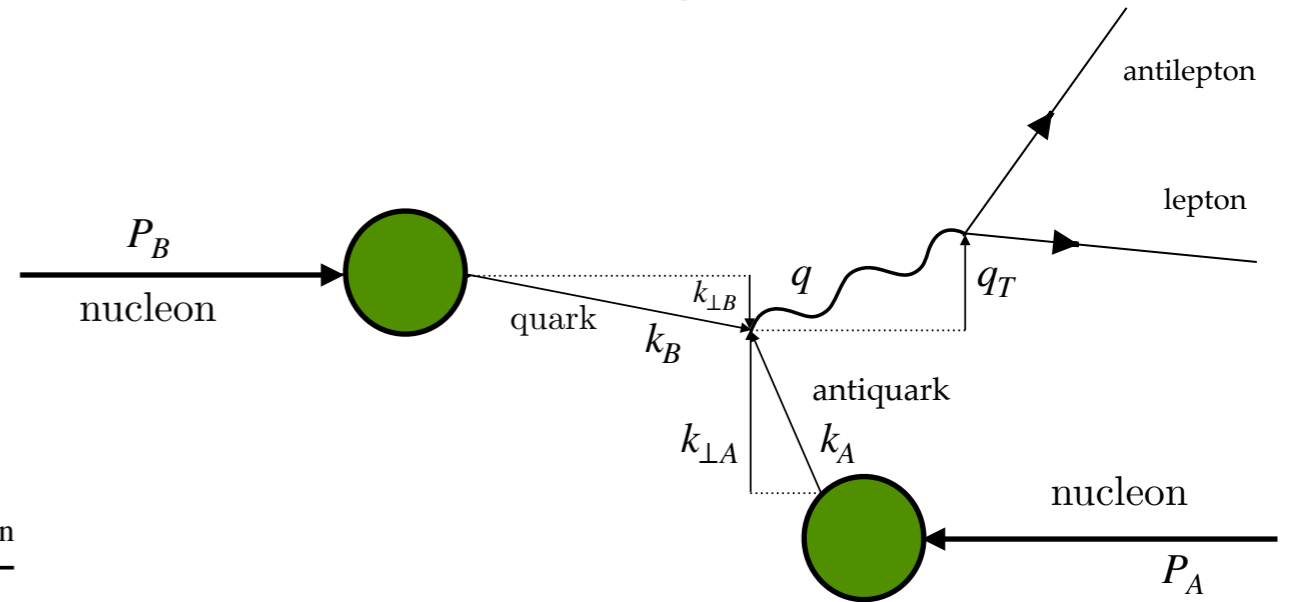
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SIDIS



Drell-Yan

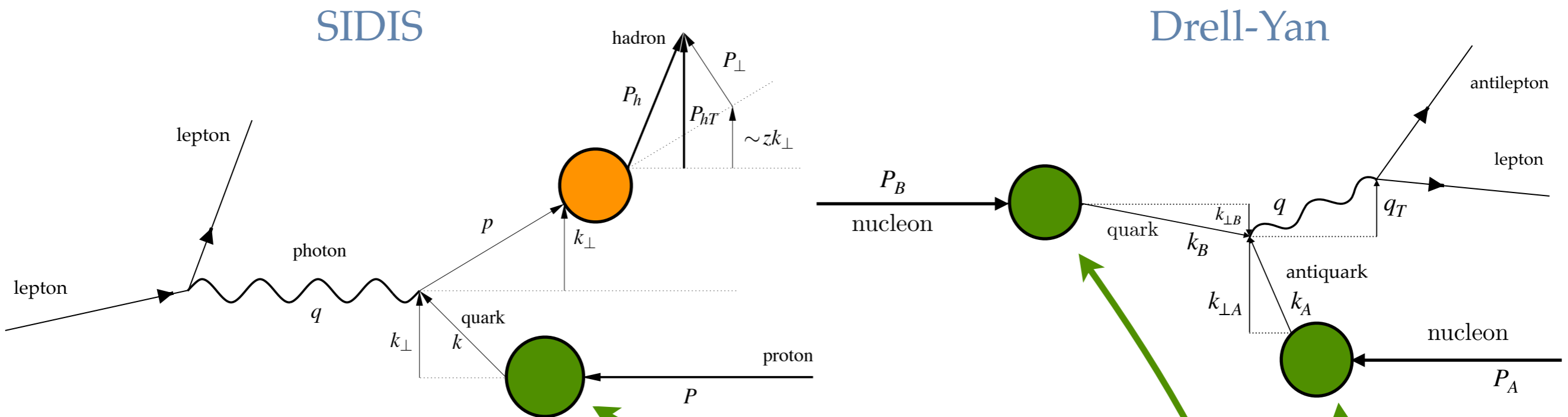


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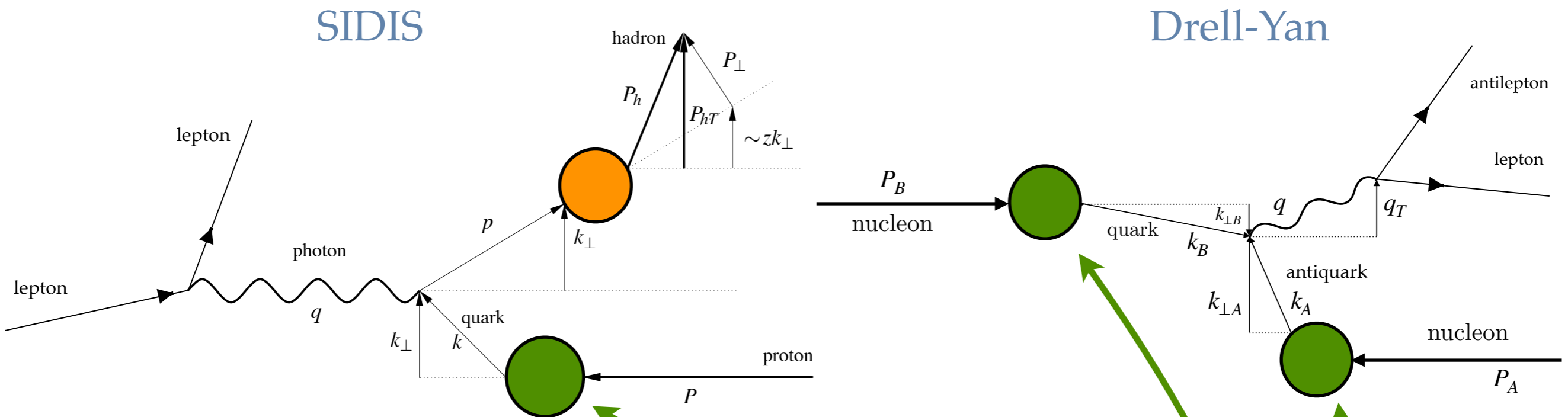


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**GLOBAL FITs**

# 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>



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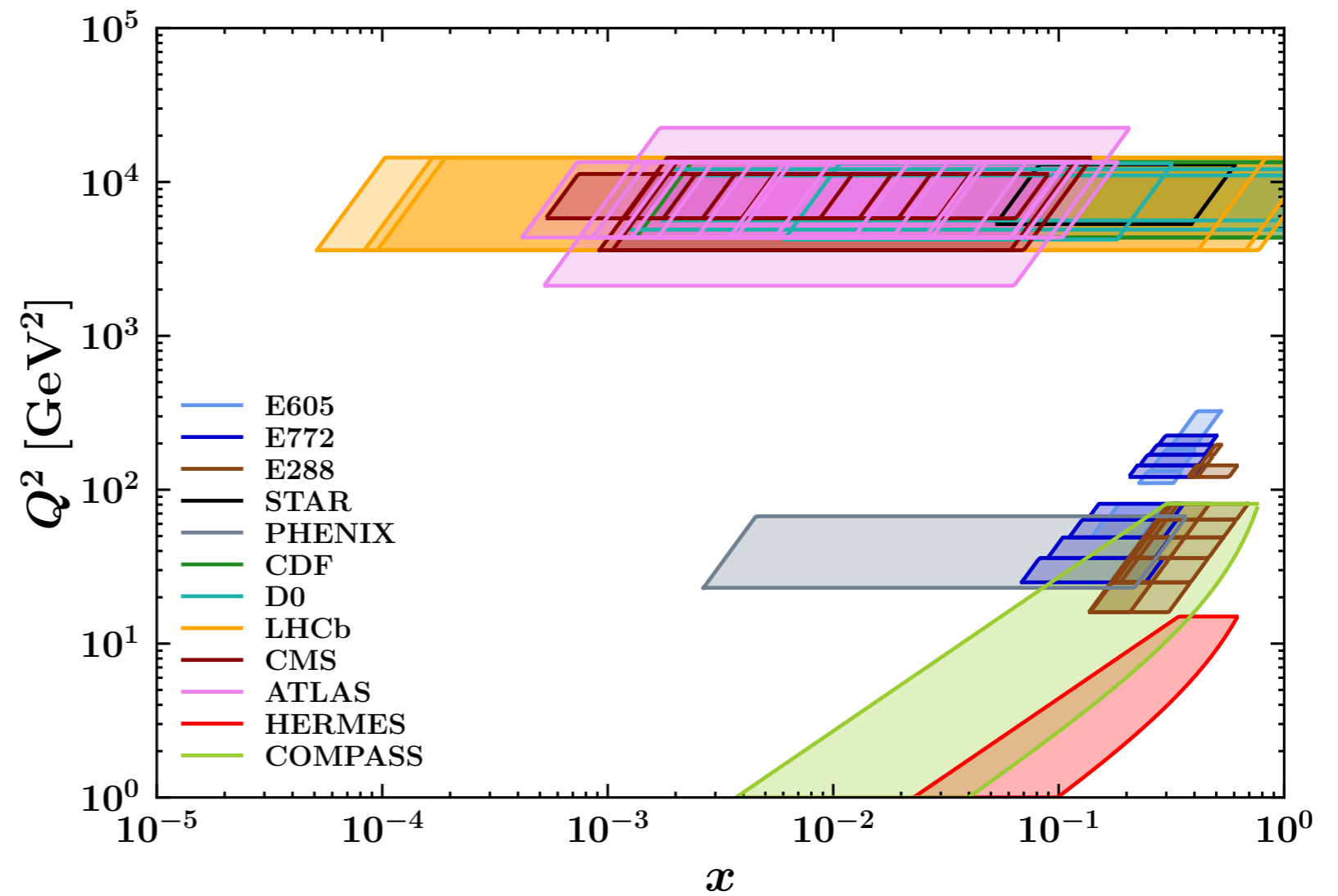
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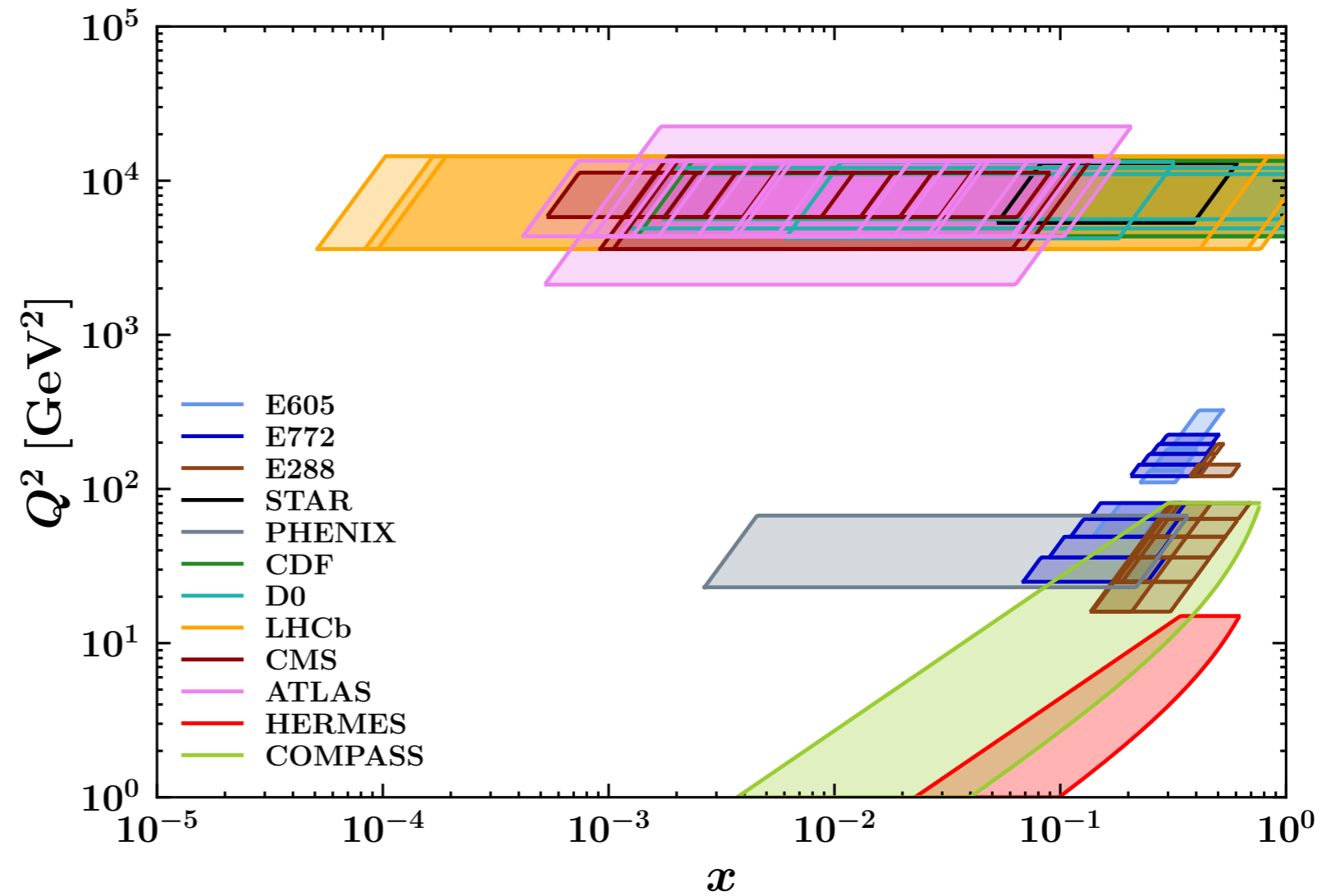
# MAP22: included data sets



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Drell-Yan data

484

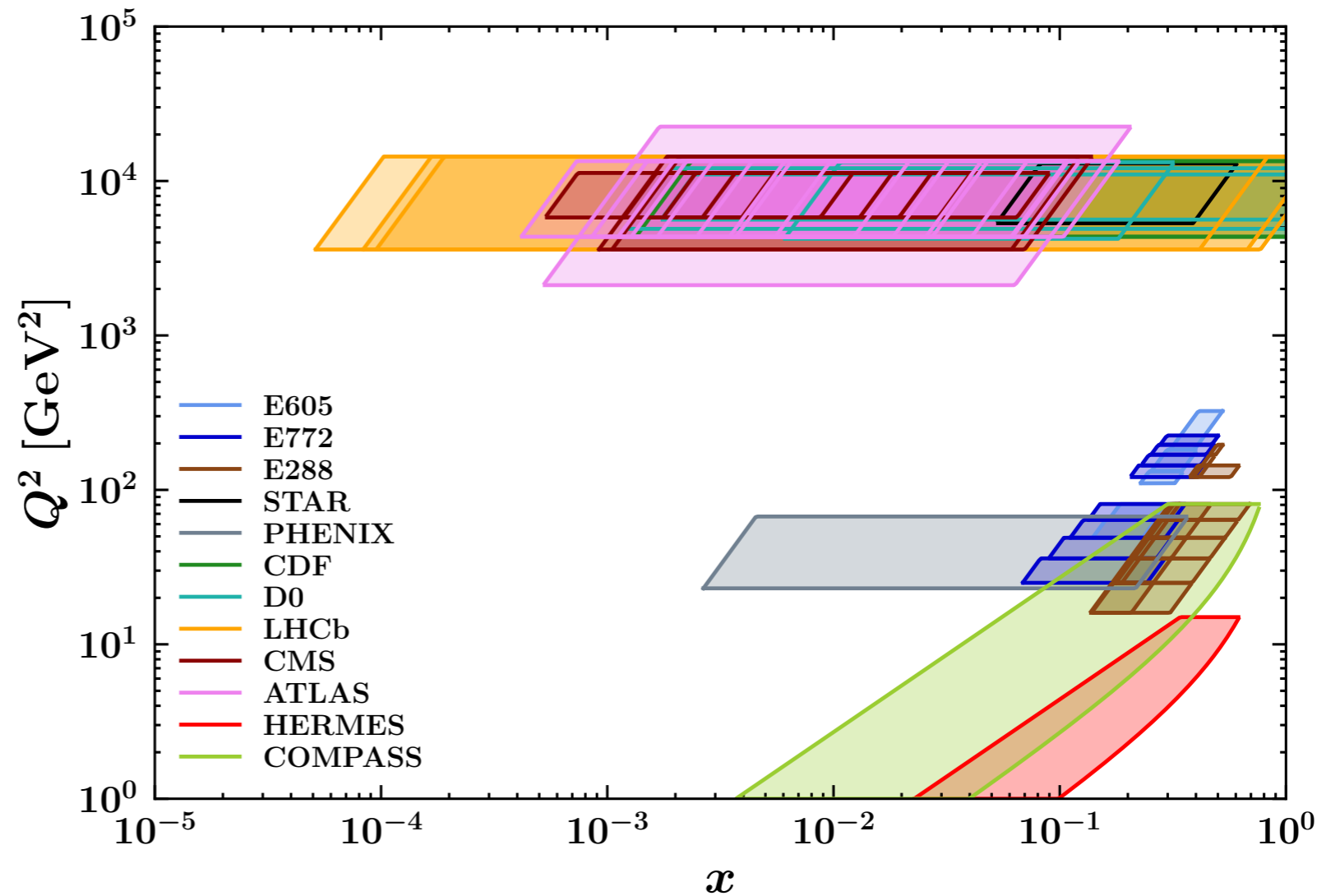


# MAP22: included data sets

**Drell-Yan data**      **484**

Fixed-target:  
E288, E605, E772

Collider mode:  
RHIC, Tevatron, LHC



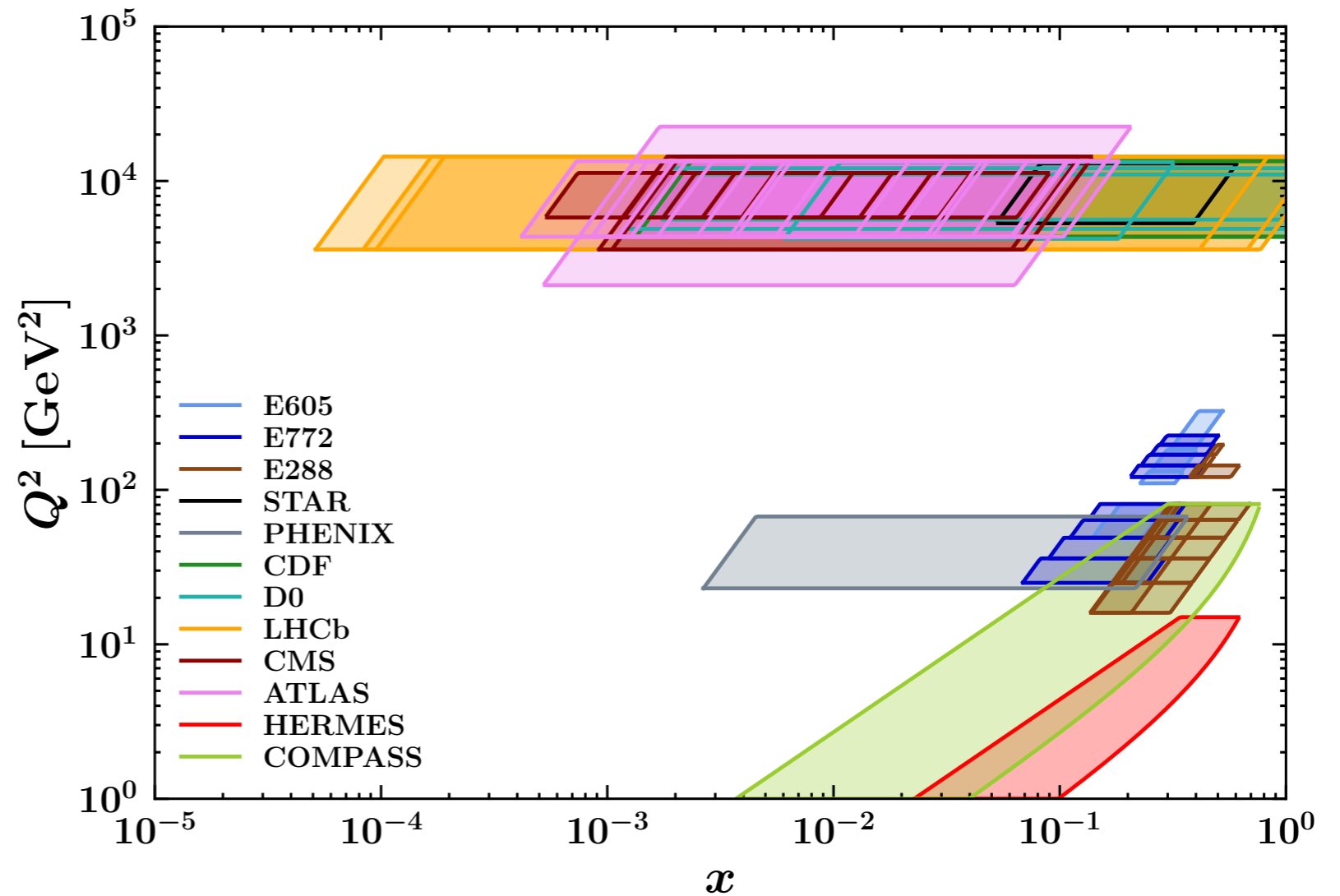
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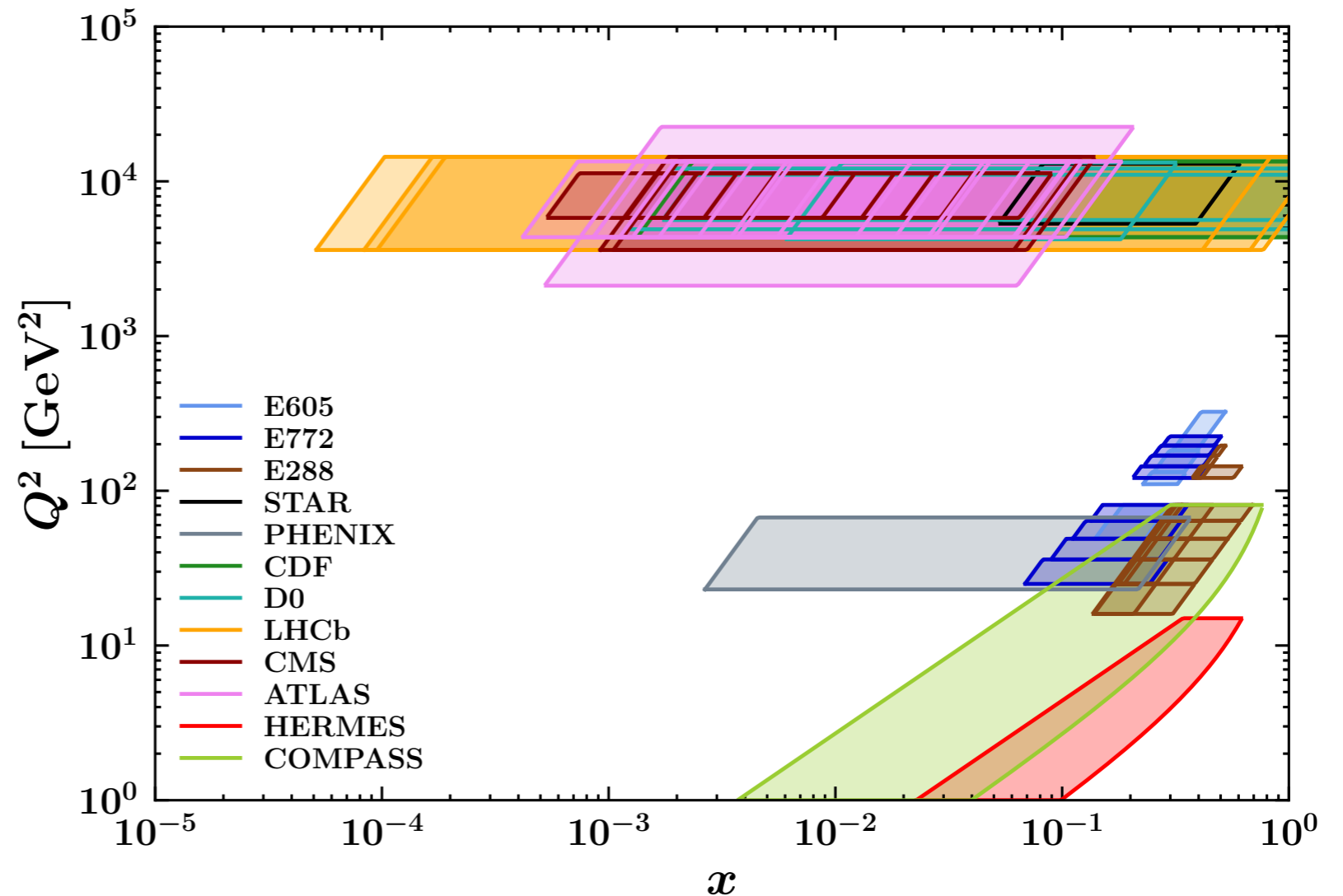
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HERMES, COMPASS



# MAP22: included data sets

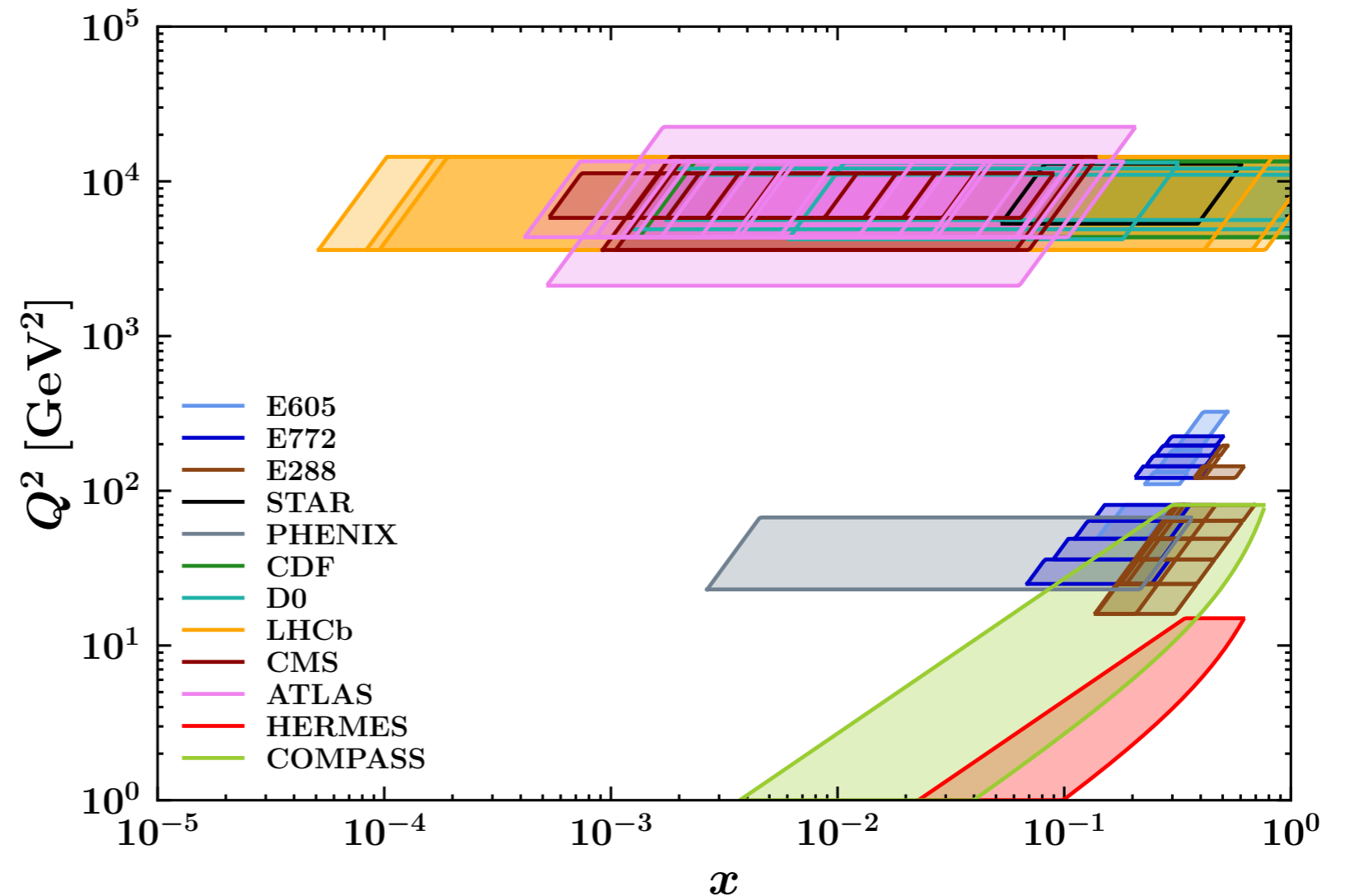
**Drell-Yan data**      **484**

Fixed-target:  
E288, E605, E772

Collider mode:  
RHIC, Tevatron, LHC

**SIDIS data**      **1547**

HERMES, COMPASS



**Total number of data: 2031**



# MAP22: Perturbative accuracy

Resummation of large logs

-

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Resummation of large logs

$$S_{\text{pert}}(\mu_b, \mu) = 1 + \sum_{k=0}^{\infty} R_{N^k \text{LL}}$$

$$S_{\text{pert}}(\mu_b, \mu) = 1 + \sum_{k=0}^{\infty} \sum_{n=1+[k/2]}^{\infty} \left( \frac{\alpha_S(\mu)}{4\pi} \right)^n \sum_{k=1}^{2n} L^{2n-k} R^{(n, 2n-k)}$$

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Accuracy	$H$ and $C$	$K$ and $\gamma_F$	$\gamma_K$	PDF/FF and $\alpha_s$ evol.
LL	0	-	1	-
NLL	0	1	2	LO
NLL'	1	1	2	NLO
NNLL	1	2	3	NLO
NNLL'	2	2	3	NNLO
<b>N<sup>3</sup>LL<sup>-</sup></b>	<b>2</b>	<b>3</b>	<b>4</b>	<b>NNLO/NLO</b>
N <sup>3</sup> LL	2	3	4	NNLO
N <sup>3</sup> LL'	3	3	4	N <sup>3</sup> LO

# MAP22: NP parametrization

$$f_{NP}(x, b_T^2) \exp \left\{ g_K(b_T^2) \ln \frac{\sqrt{\zeta}}{\sqrt{\zeta_0}} \right\}$$

Bacchetta, Gamberg, Goldstein, et al., PLB 659 (2008)

Bacchetta, Conti, Radici, PRD 78 (2008)

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

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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**
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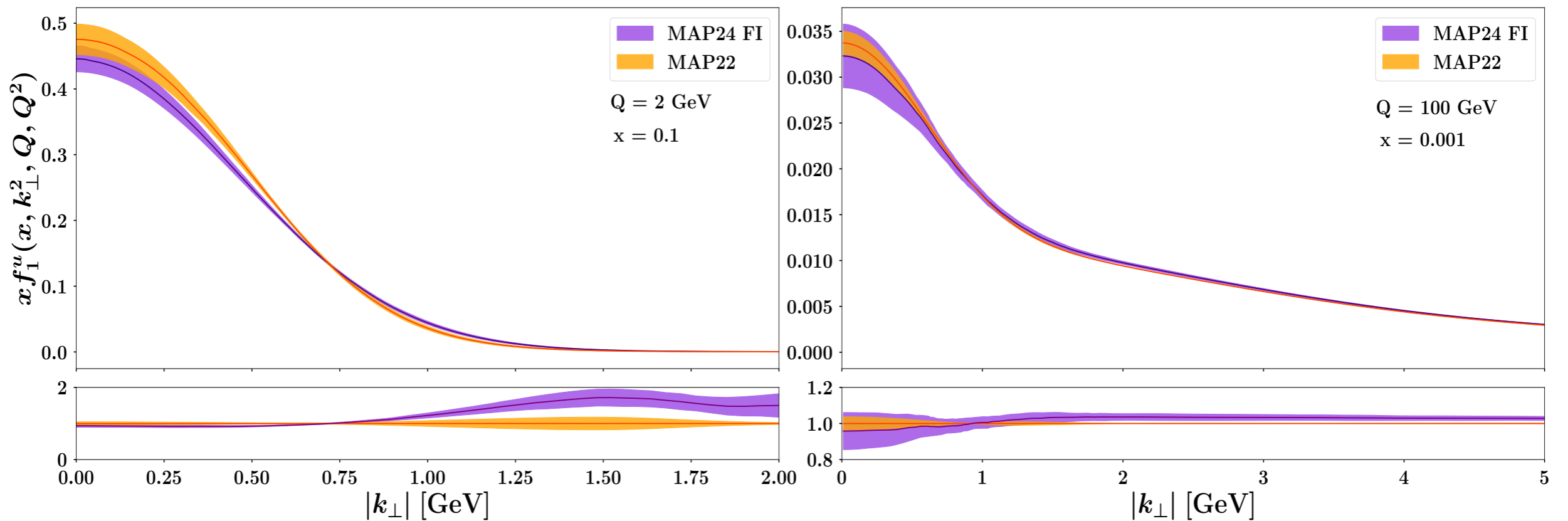


# MAPTMD24

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>

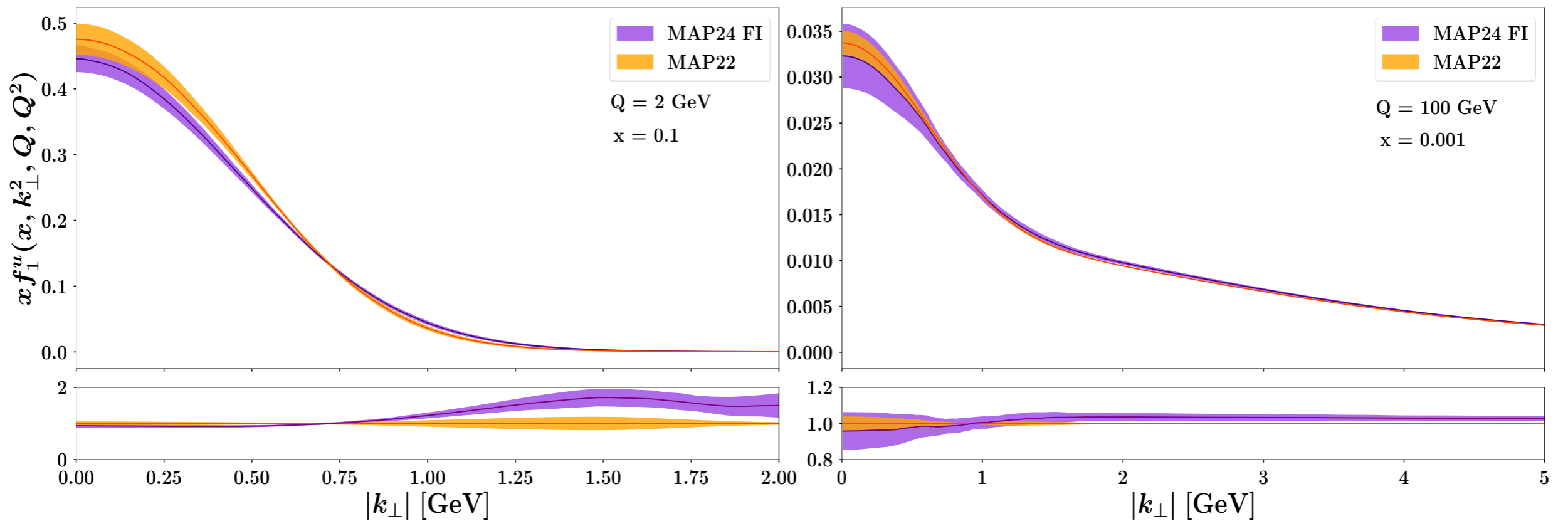
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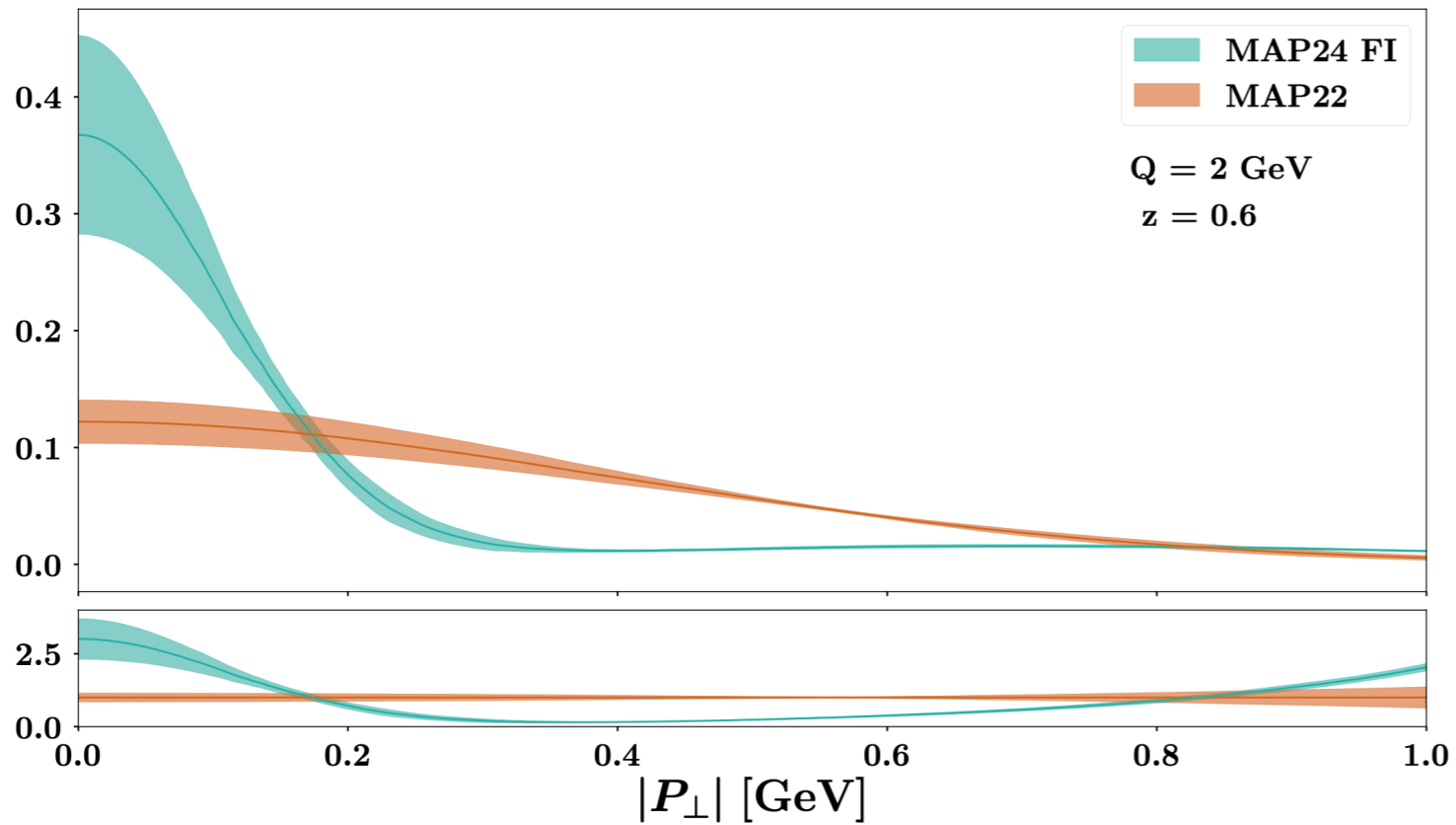
**TMD PDFs are compatible with MAP22**

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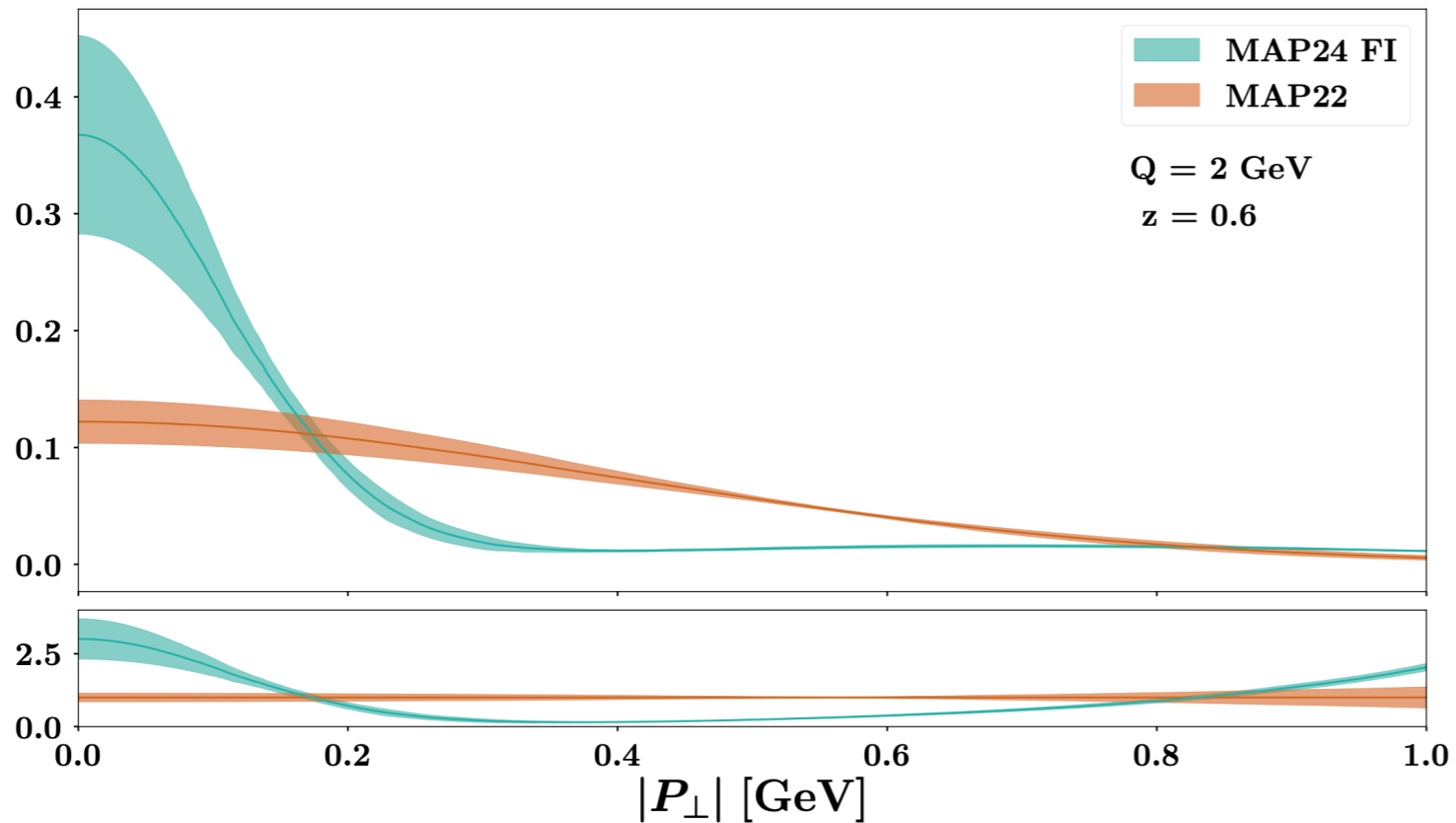
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## MAPFF1.0nnlo

- 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
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Data set	$N_{\text{dat}}$	$\chi_0^2/N_{\text{dat}}$
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## NNPDF + DSS

## MMHT + DSS (MAP22)



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←→  
**COMPATIBILITY**

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## NNPDF + DSS

## MMHT + DSS (MAP22)

# MAPTMD24

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## NNPDF + DSS

## MMHT + DSS (MAP22)

TU QUOQUE,  
BRUTE  
HERMES

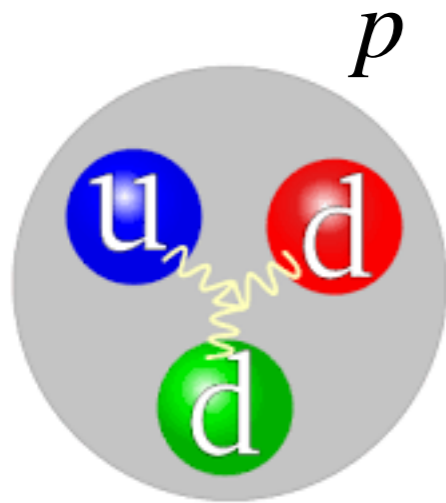


# MAPTMD24: new approach

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

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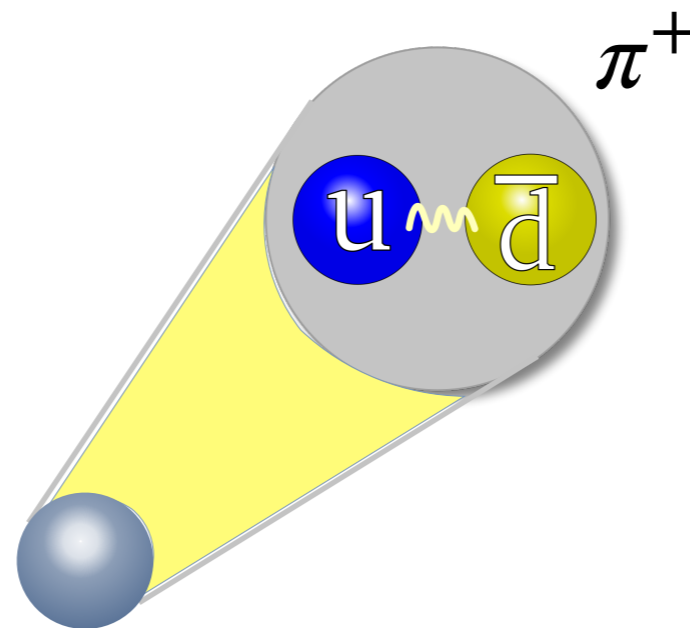
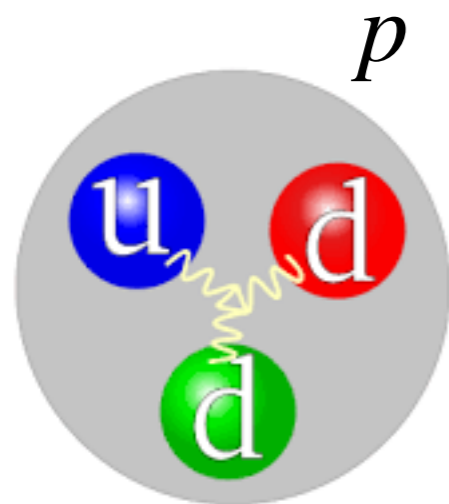
$u, d$

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

$s$  (*sea*)

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$u, d$

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

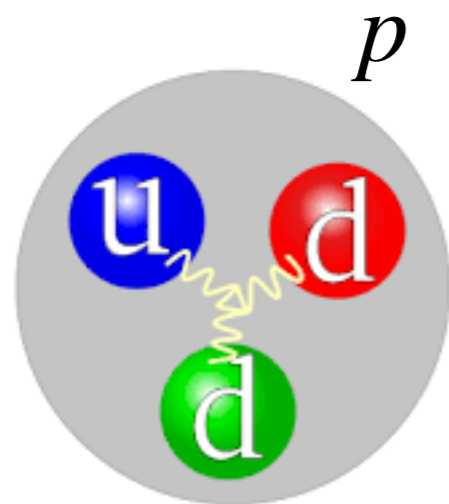
$s$  (*sea*)

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

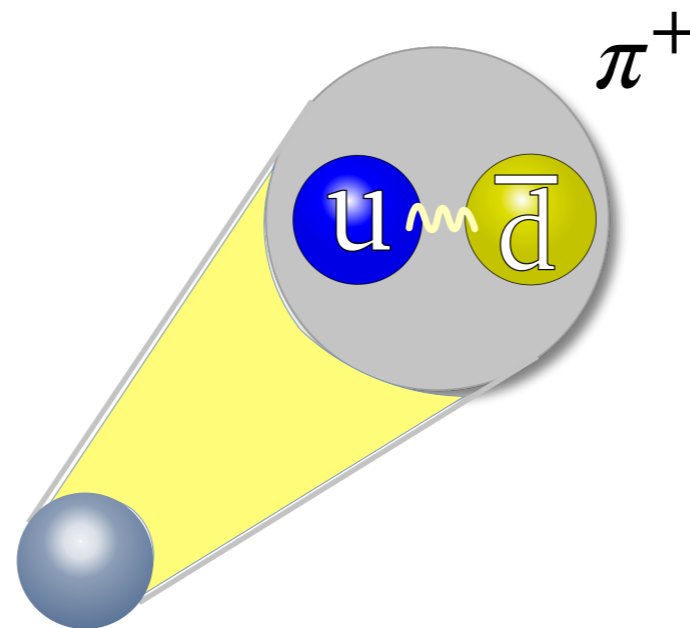
$d \rightarrow \pi^+, \dots$

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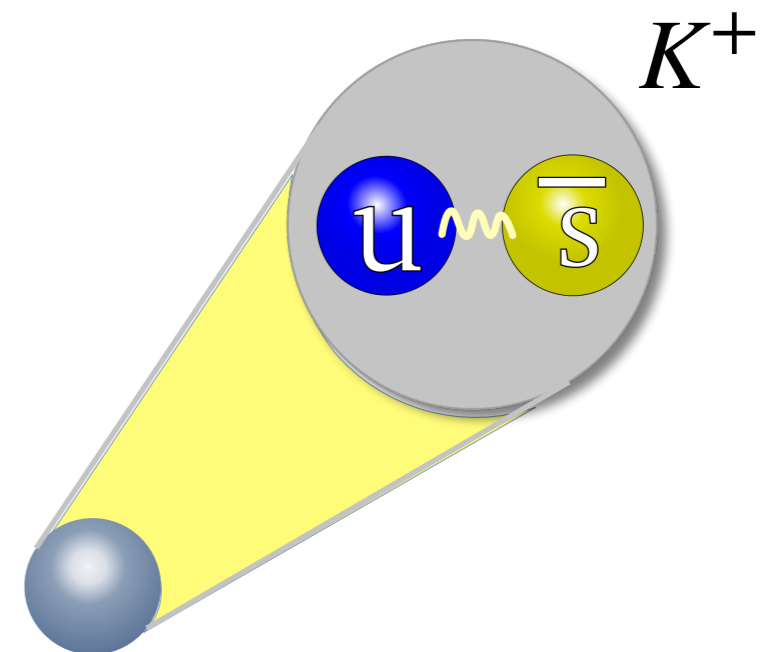
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$u, d$   
 $\bar{u}, \bar{d}$   
 $s$  (*sea*)



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

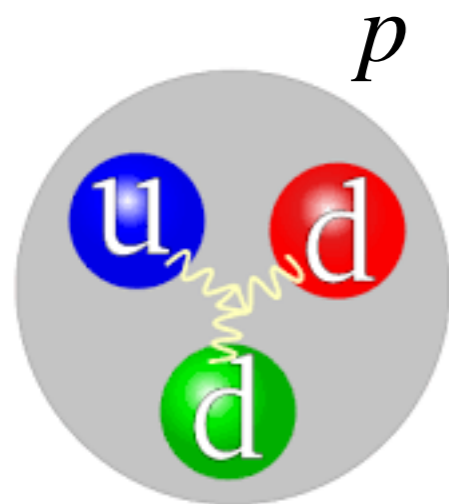


$u \rightarrow K^+, \dots$   
 $\bar{s} \rightarrow K^+, \dots$   
 $d \rightarrow K^+, \dots$

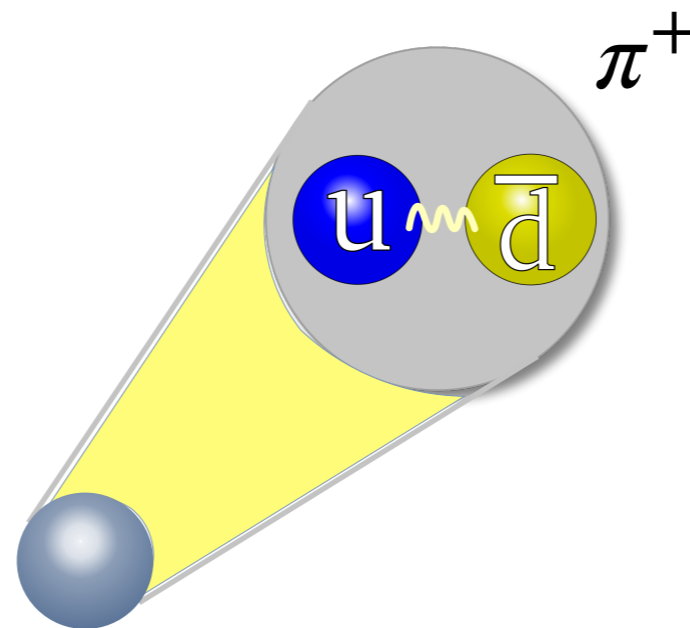


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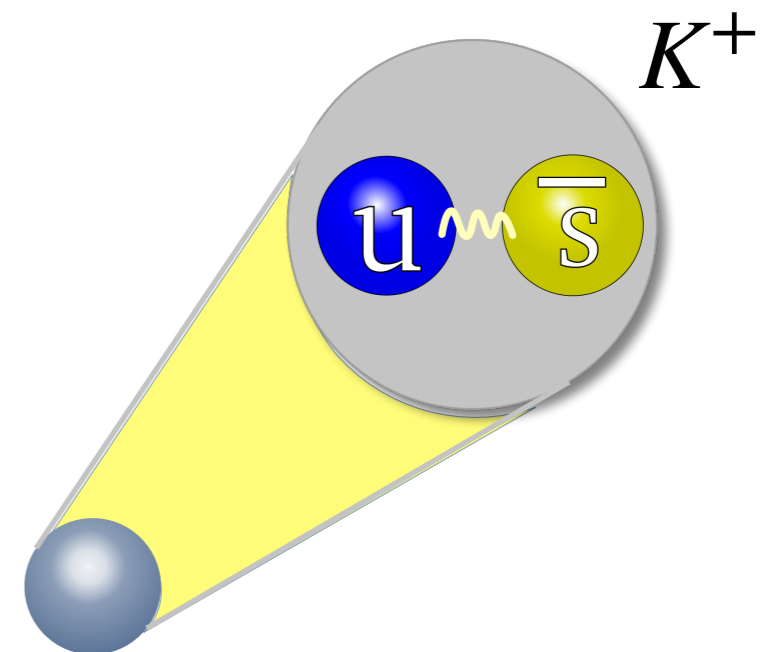
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$u, d$   
 $\bar{u}, \bar{d}$   
 $s$  (*sea*)



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



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

**charge conjugation**

# MAPTMD24: new approach

## HERMES

$$e + p \rightarrow e' + \pi^+ + X$$

$$e + p \rightarrow e' + \pi^- + X$$

$$e + p \rightarrow e' + K^+ + X$$

$$e + p \rightarrow e' + K^- + X$$

# MAPTMD24: new approach

**HERMES**

$$e + p \rightarrow e' + \pi^+ + X$$

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**high sensitivity to flavor dependence**

# MAPTMD24: new approach

## HERMES

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$$e + p \rightarrow e' + K^+ + X$$

$$e + p \rightarrow e' + K^- + X$$

+ deuteron target

**high sensitivity to flavor dependence**

## COMPASS

deuteron target & unidentified final state hadron

## Drell-Yan

$q\bar{q}$  in the initial state

# MAPTMD24: new approach

**HERMES**

$$e + p \rightarrow e' + \pi^+ + X$$

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+ deuteron target

**high sensitivity to flavor dependence**

**COMPASS**

deuteron target & unidentified final state hadron

**Drell-Yan**

$q\bar{q}$  in the initial state

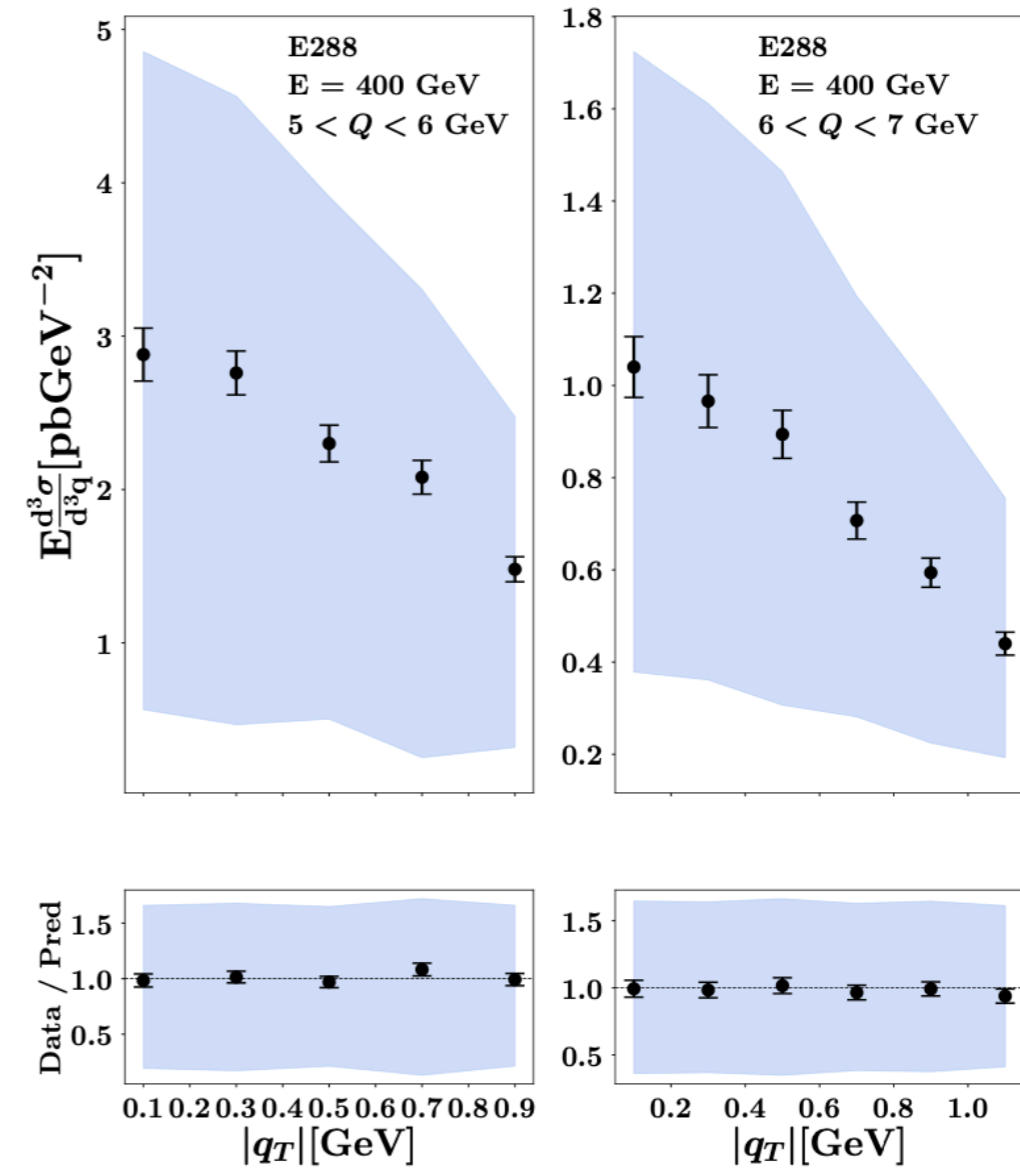
**low sensitivity to flavor dependence**

# 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>

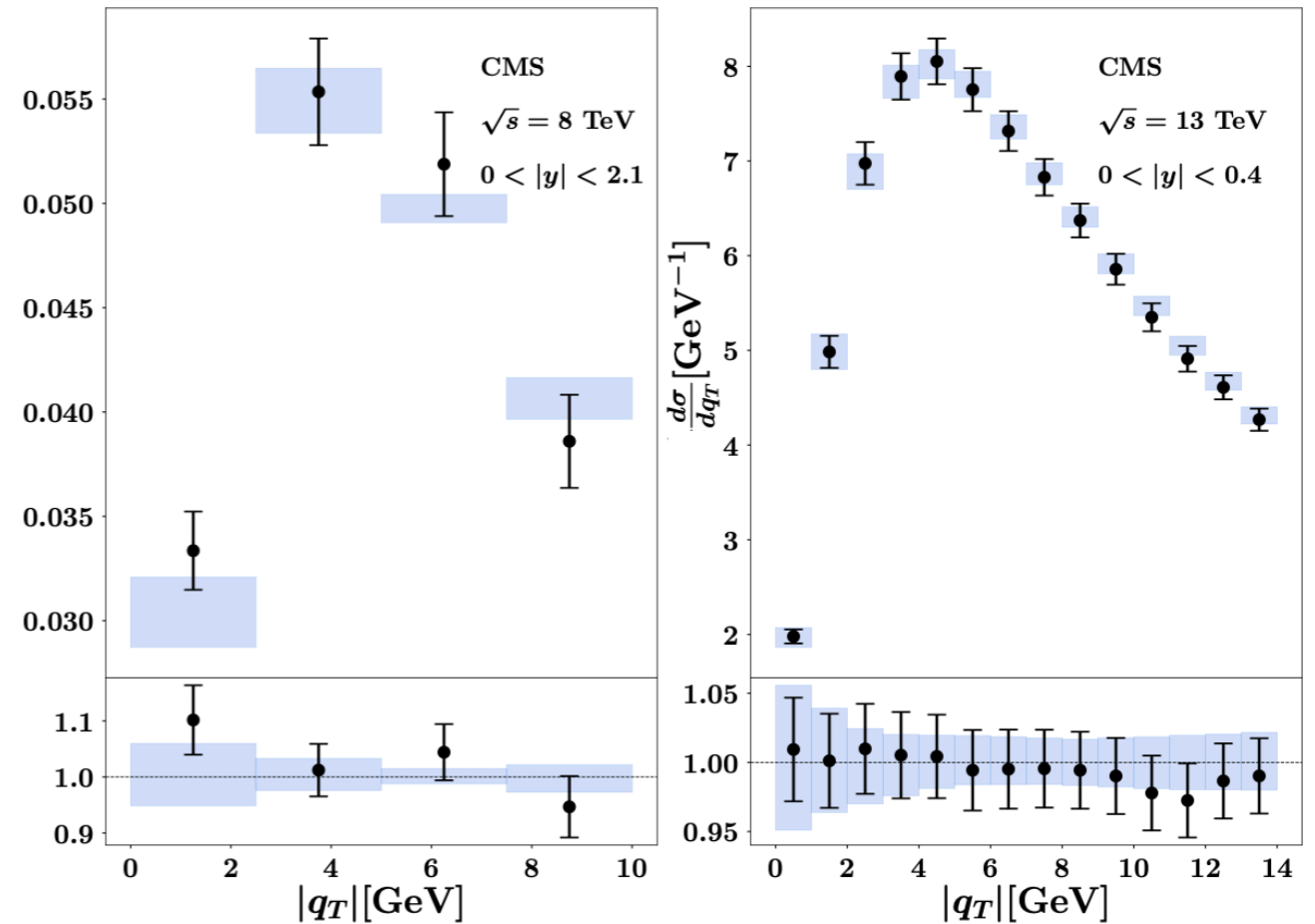
# MAPTMD24: results

Data set	$N^3LL$			
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DY fixed-target total	233	0.63	0.31	0.94
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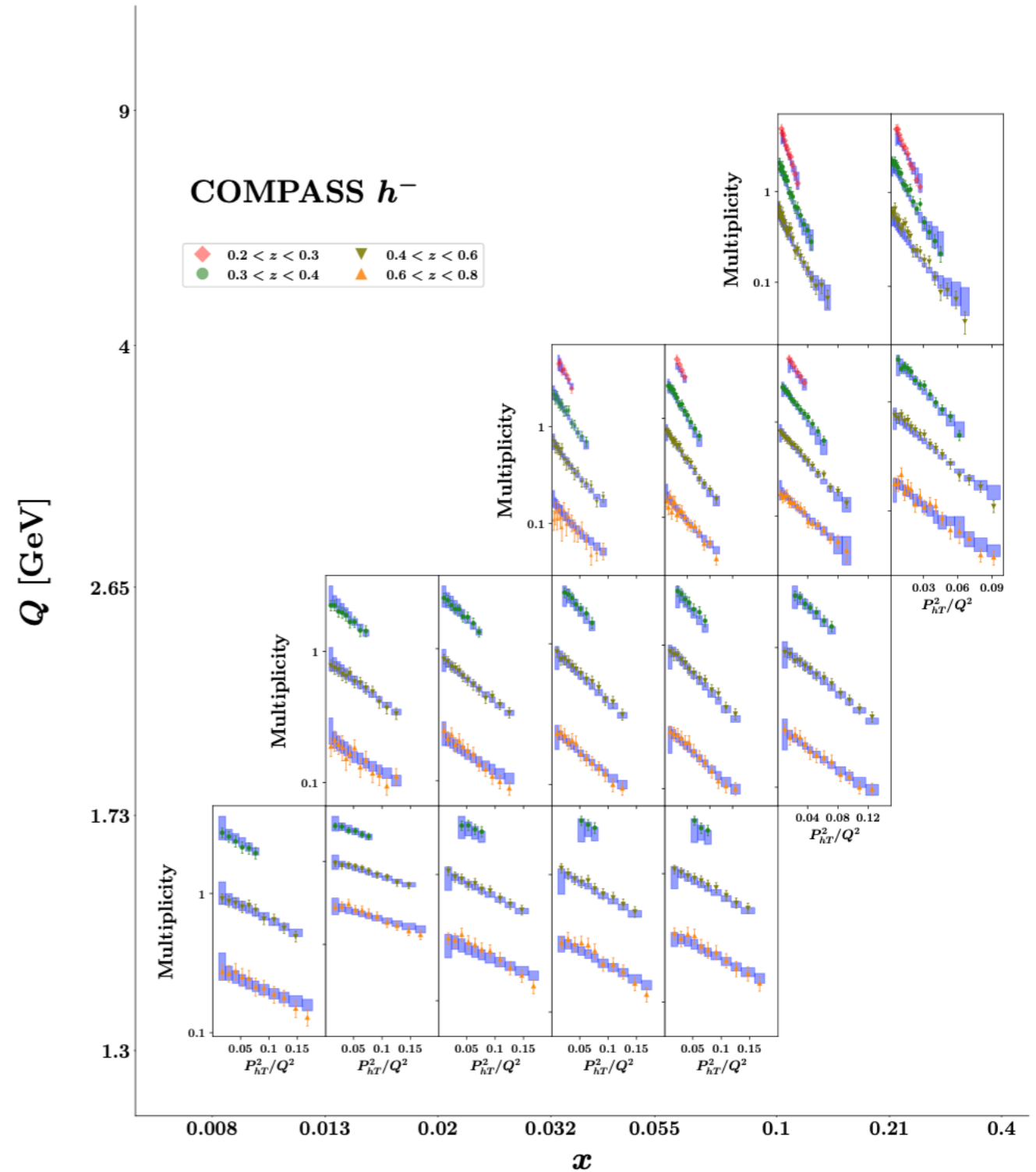
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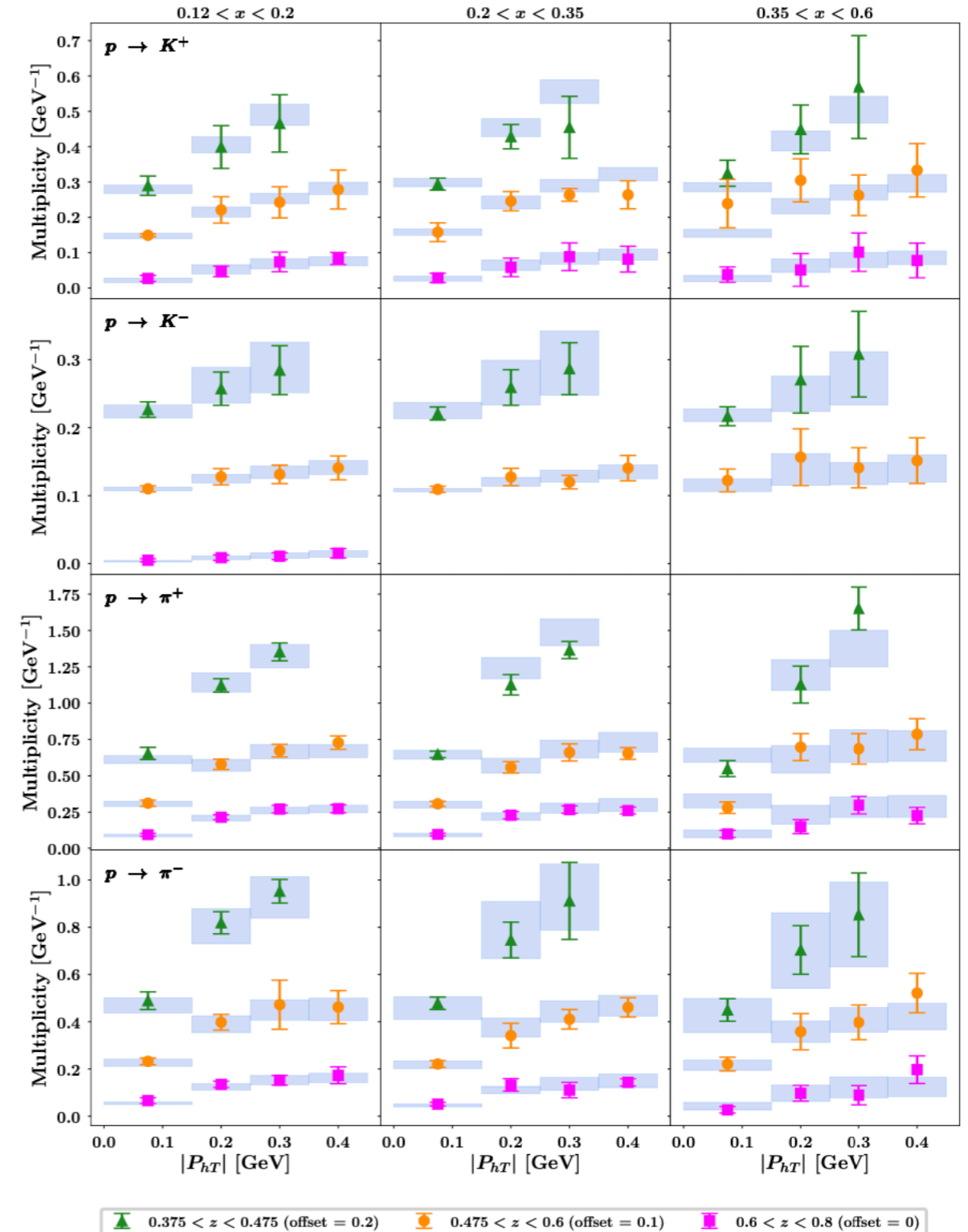
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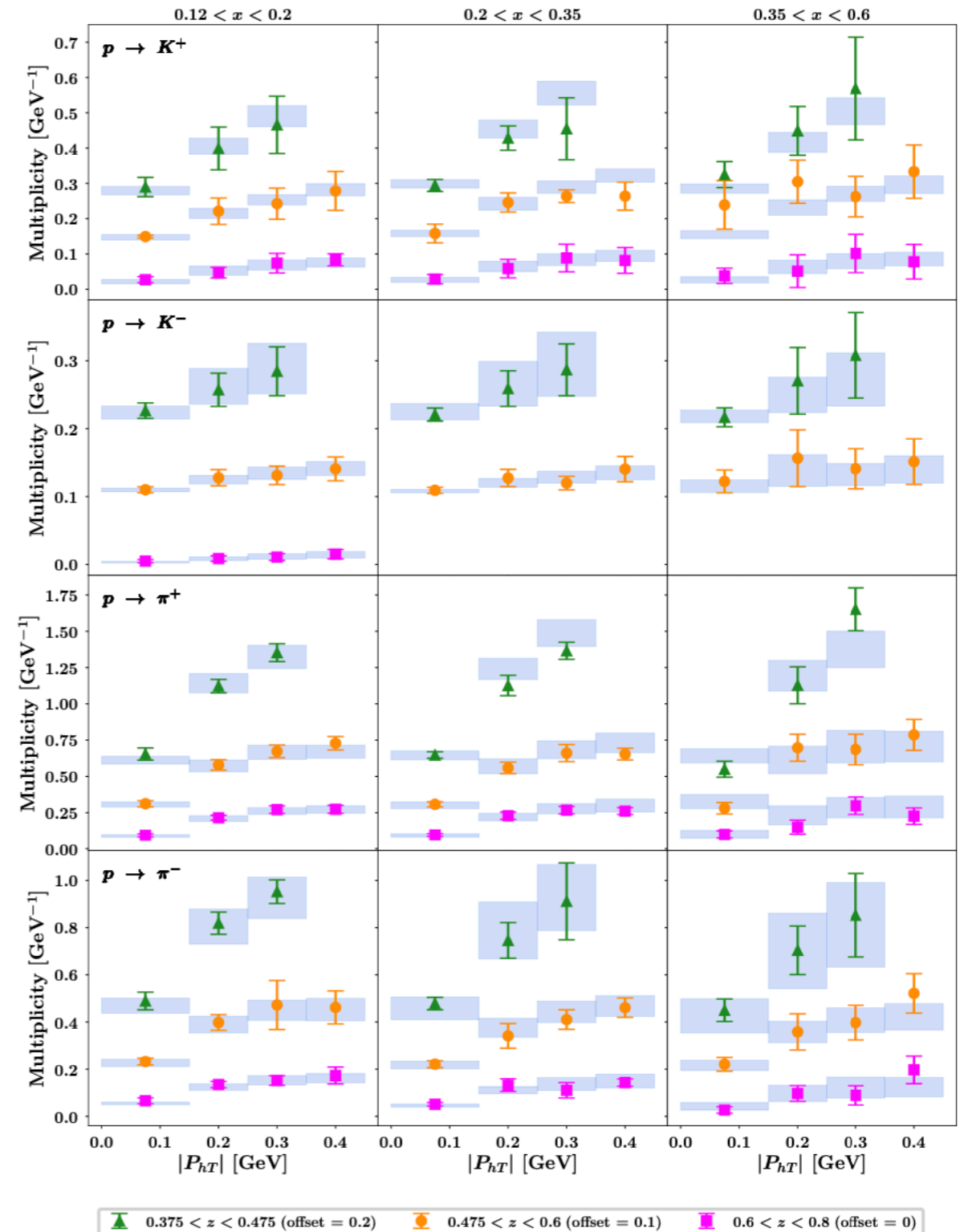
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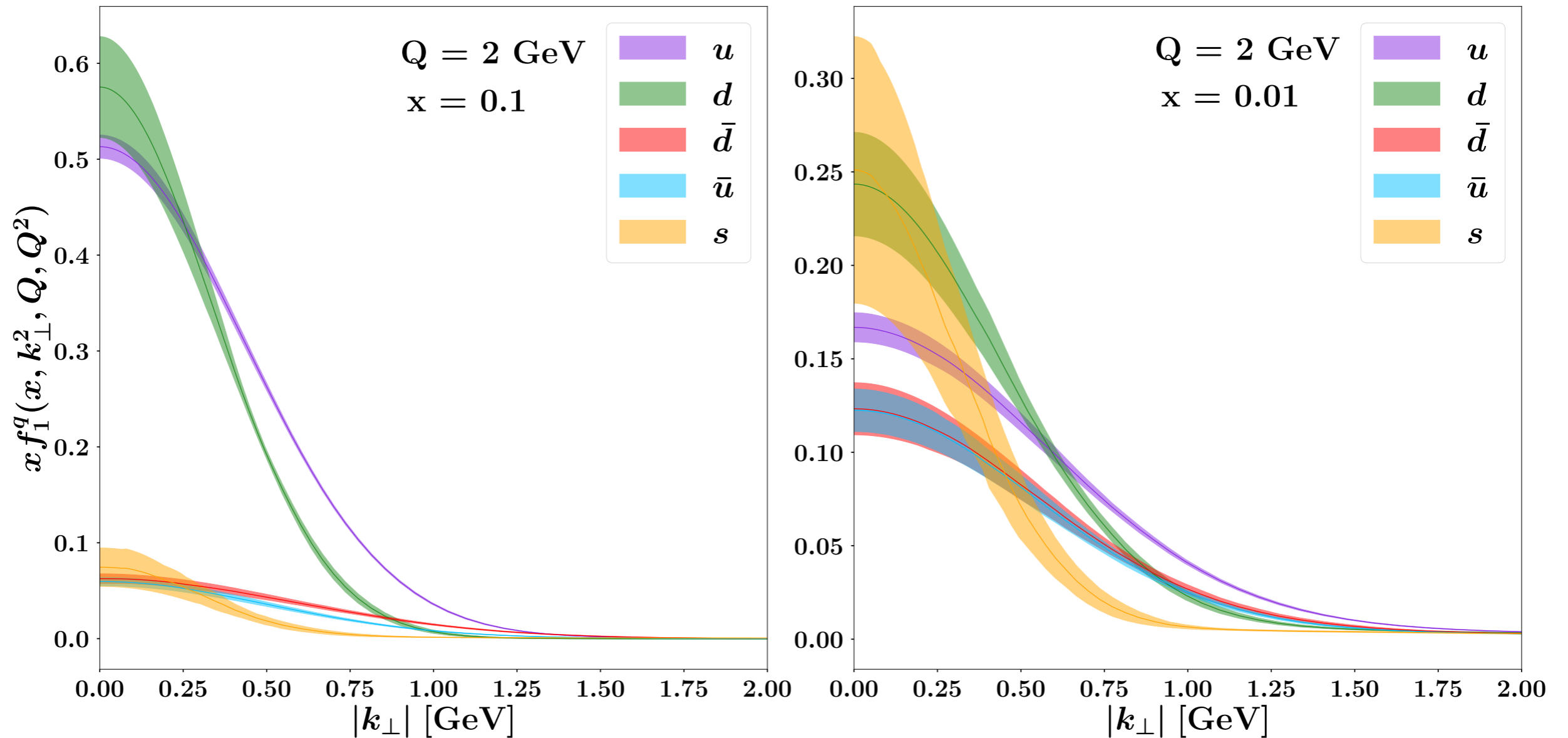
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The agreement between theory and HERMES data has increased a lot!



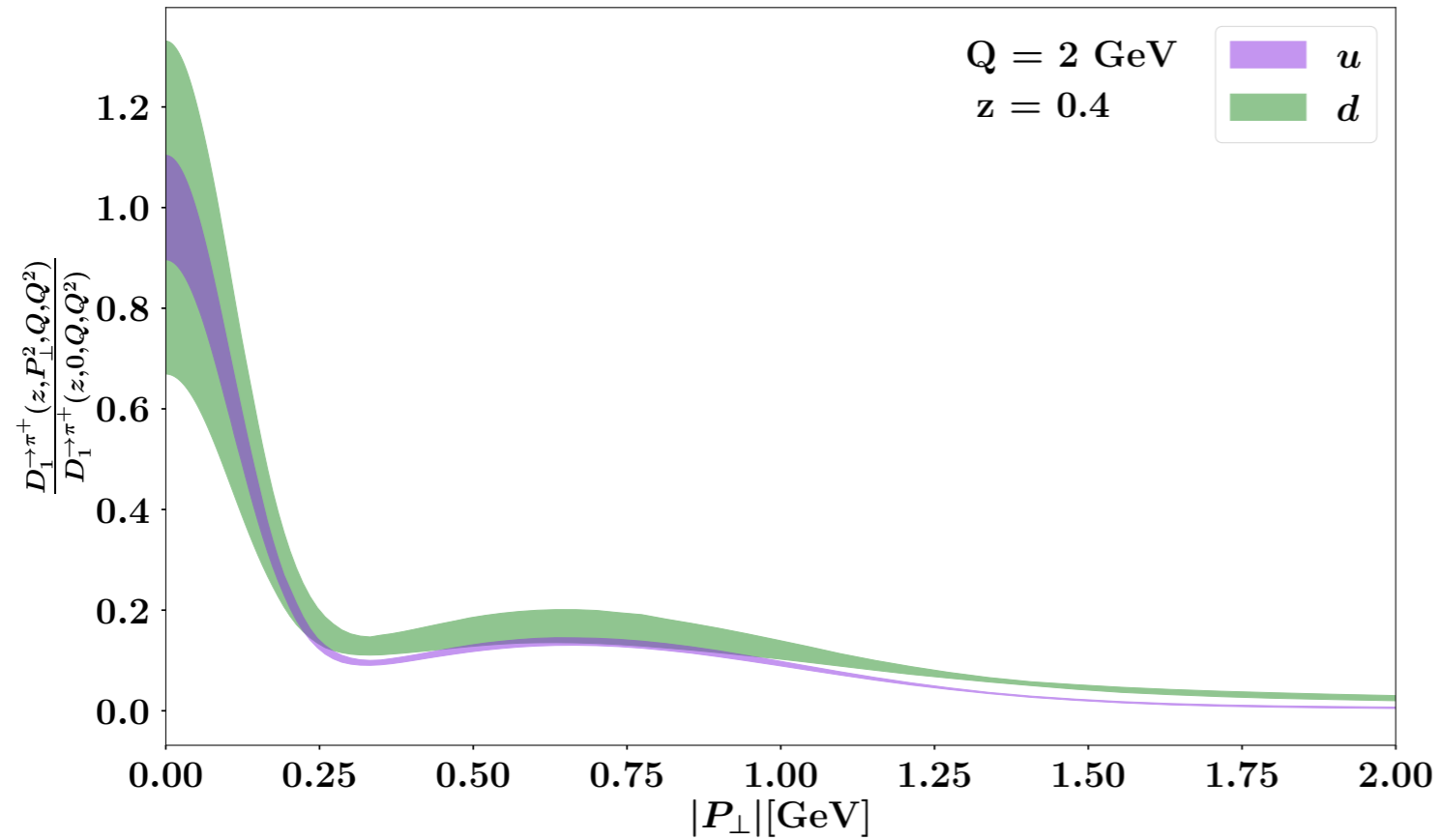
## Flavor-dependent TMD PDFs



Evidence of different behaviors for different flavors

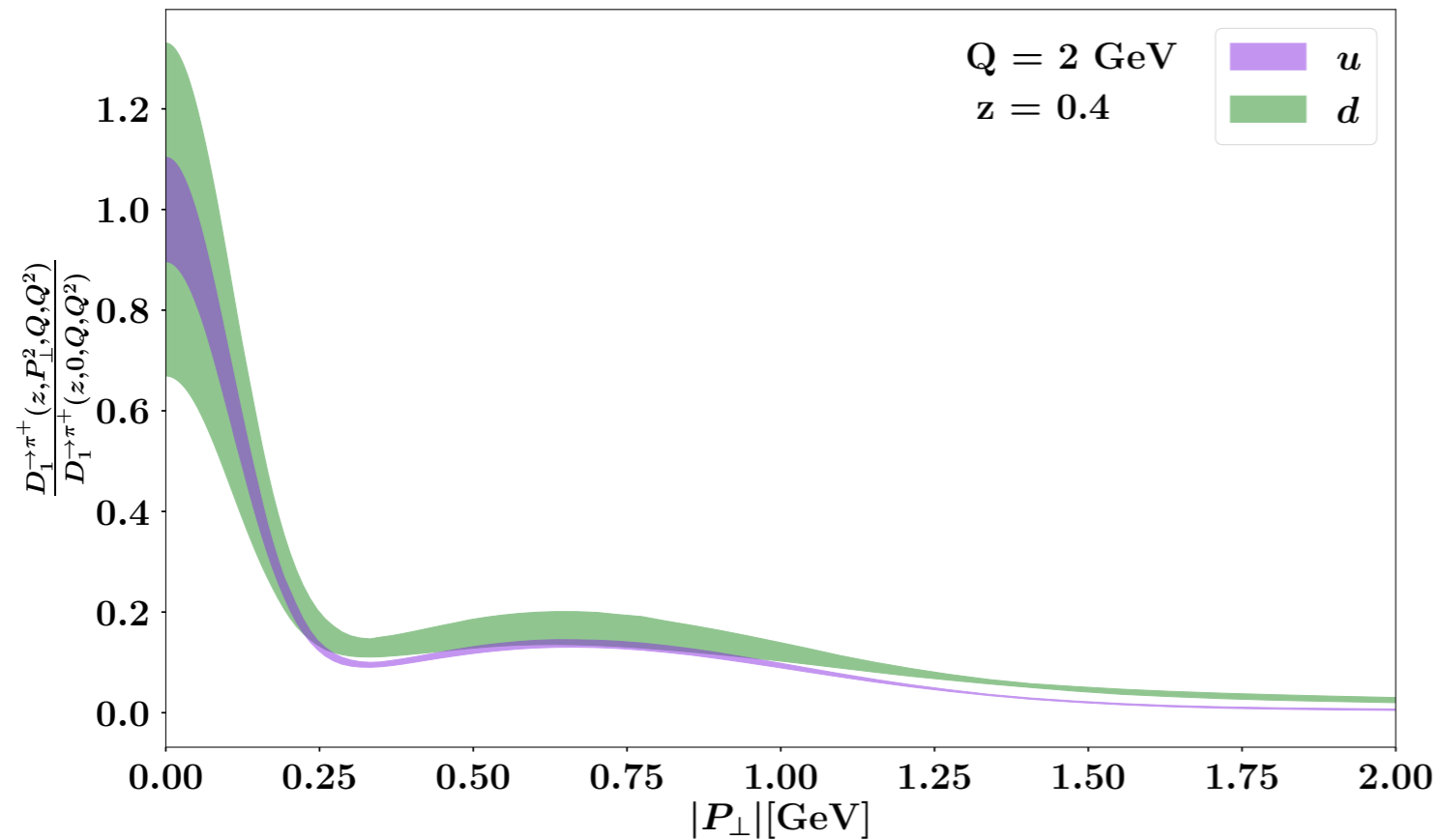
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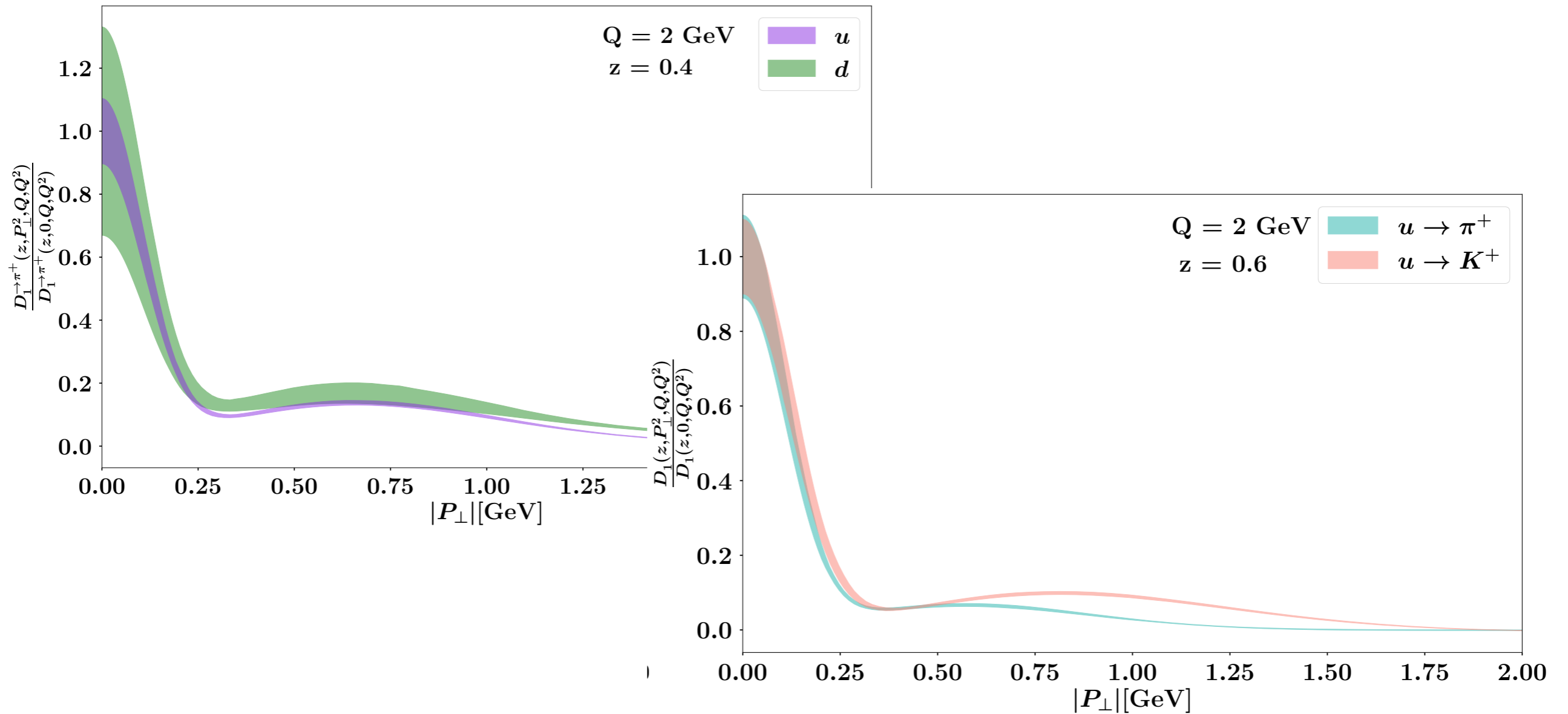
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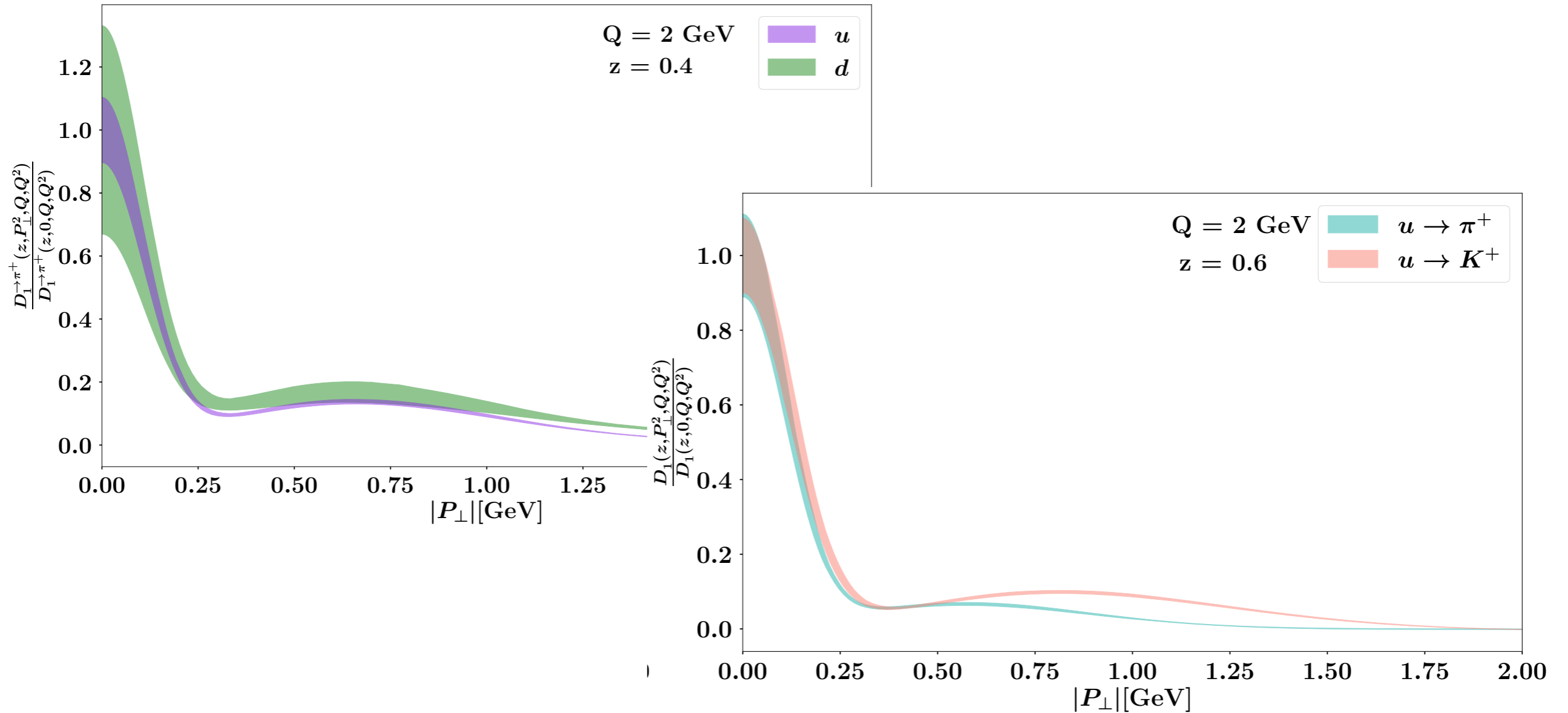
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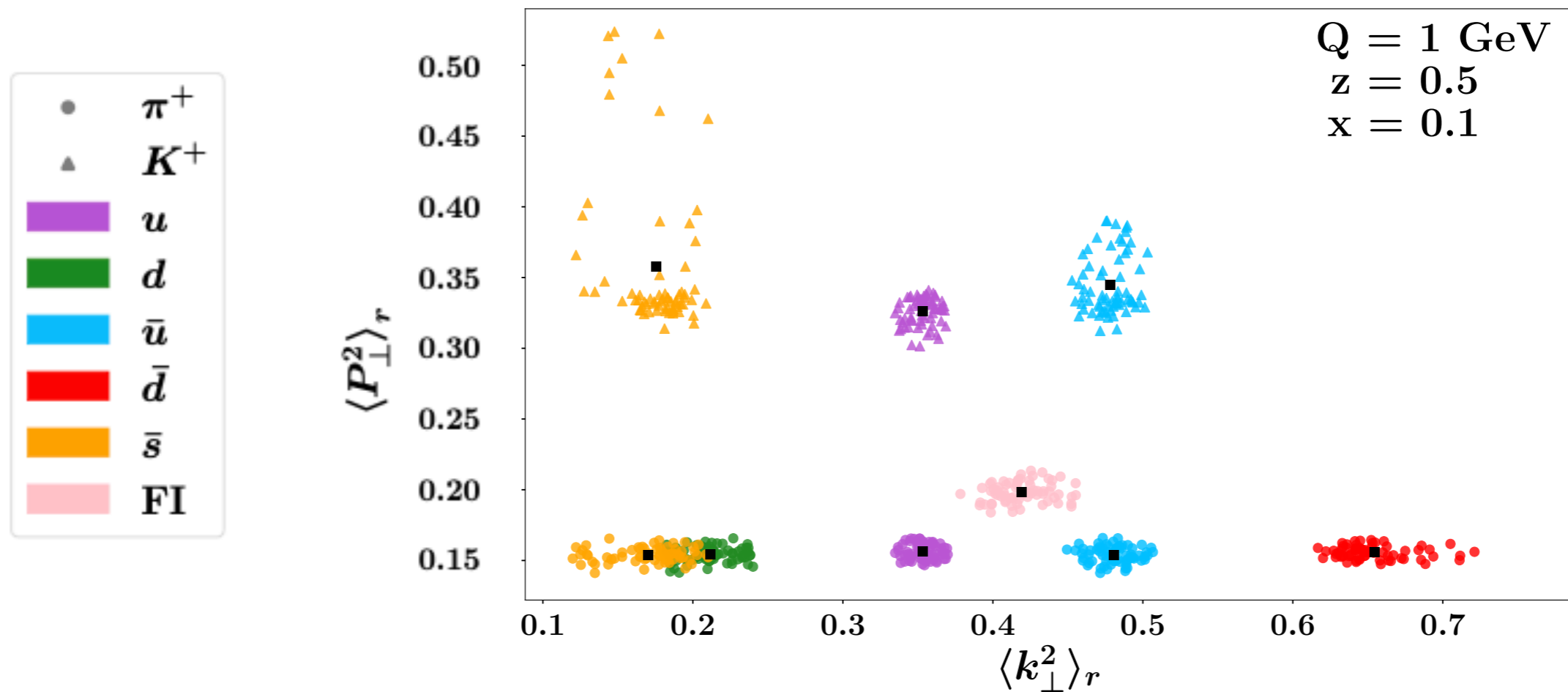
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Some evidence of different behaviors for different measured hadrons



# MAPTMD24: results

## TMD's “effective width”



Evidence of different behaviors for different flavors

Evidence of different behaviors for different measured hadrons

# MAPTMD24: results

Collins-Soper kernel:

# MAPTMD24: results

Collins-Soper kernel: kernel of the rapidity evolution equation

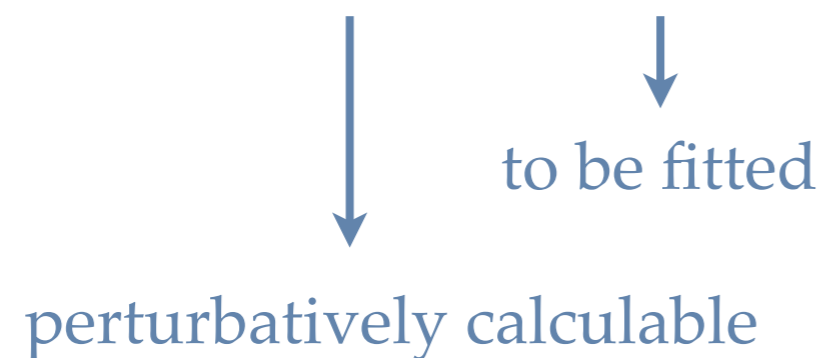
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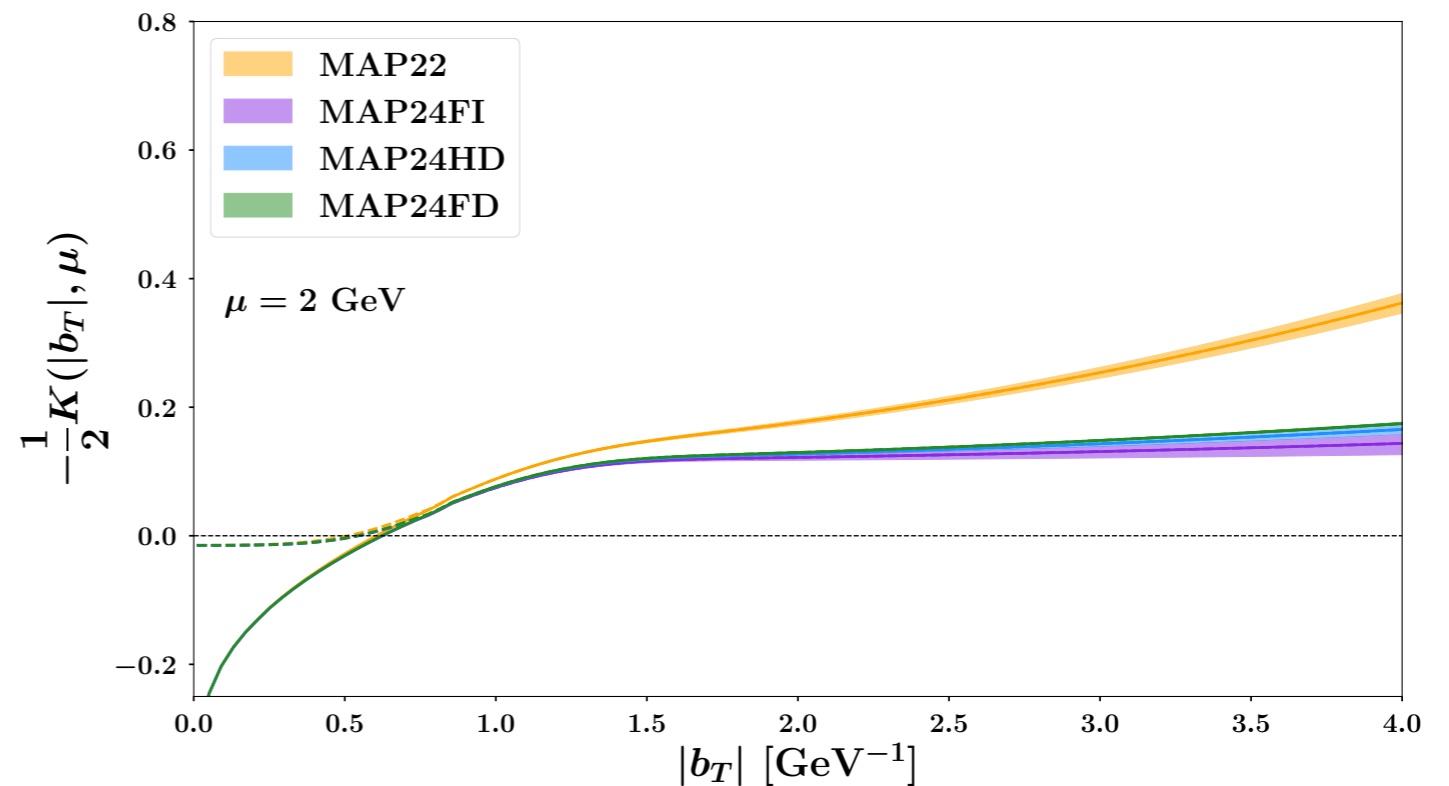
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↓  
perturbatively calculable

↓  
to be fitted



See S. Mukerjee's talk

- New feature: almost-linear behaviour at large  $b_T$

# Conclusions

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- We observe **non-trivial differences** in the transverse momentum distribution of partons inside hadrons

**Backup**

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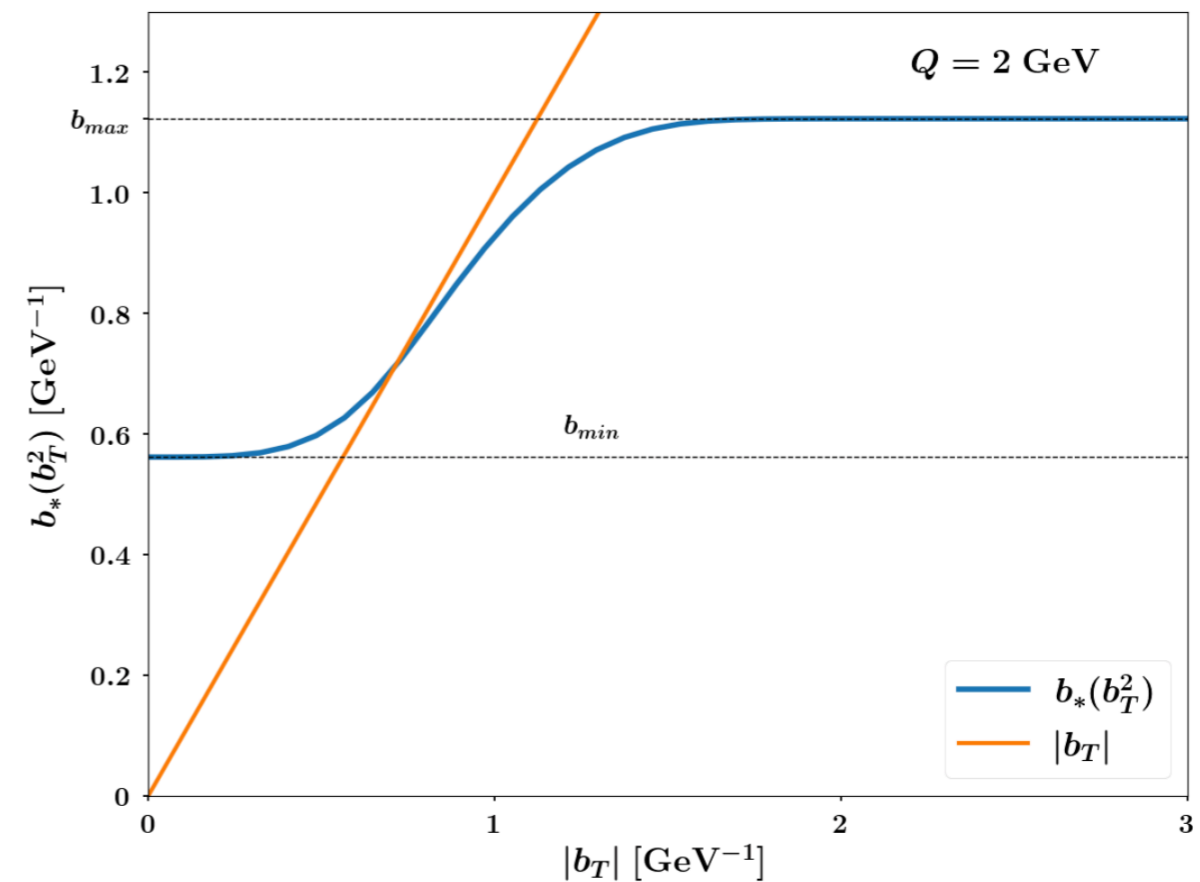
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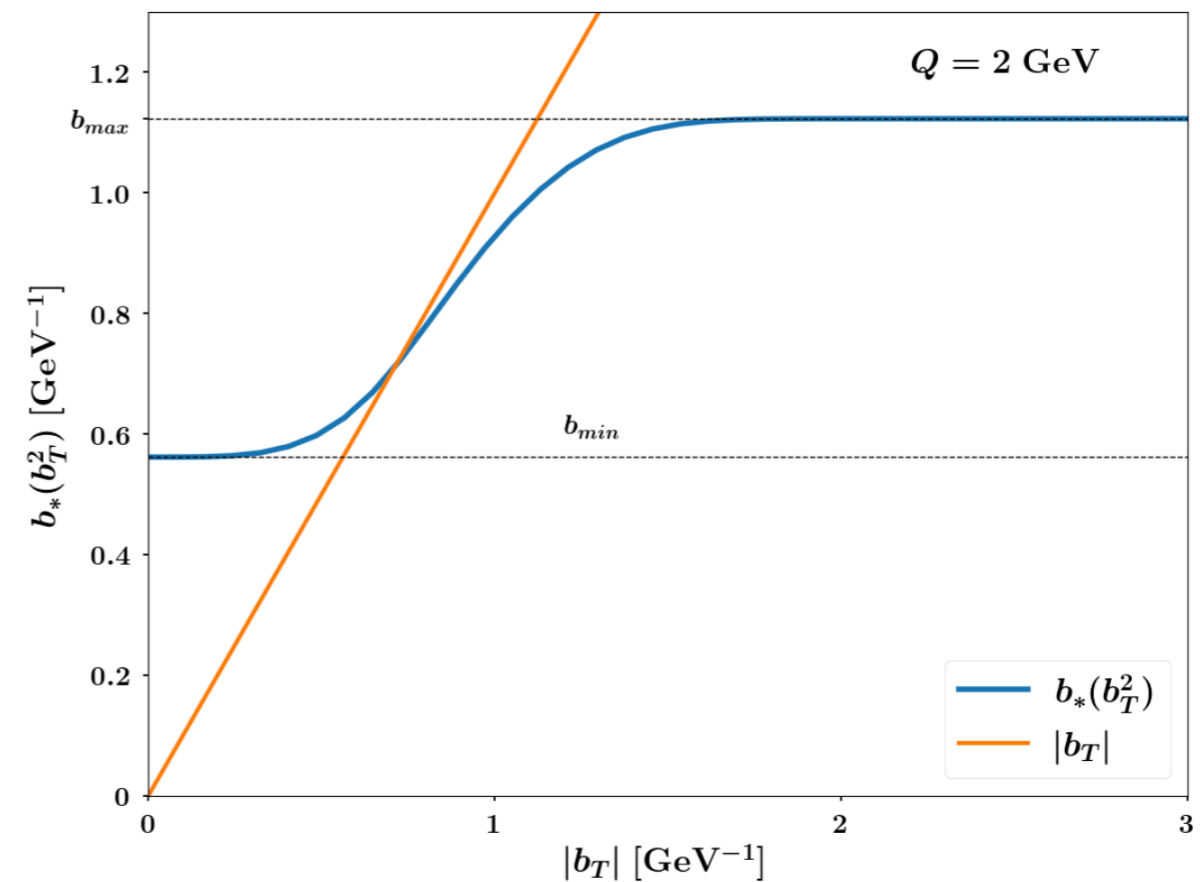
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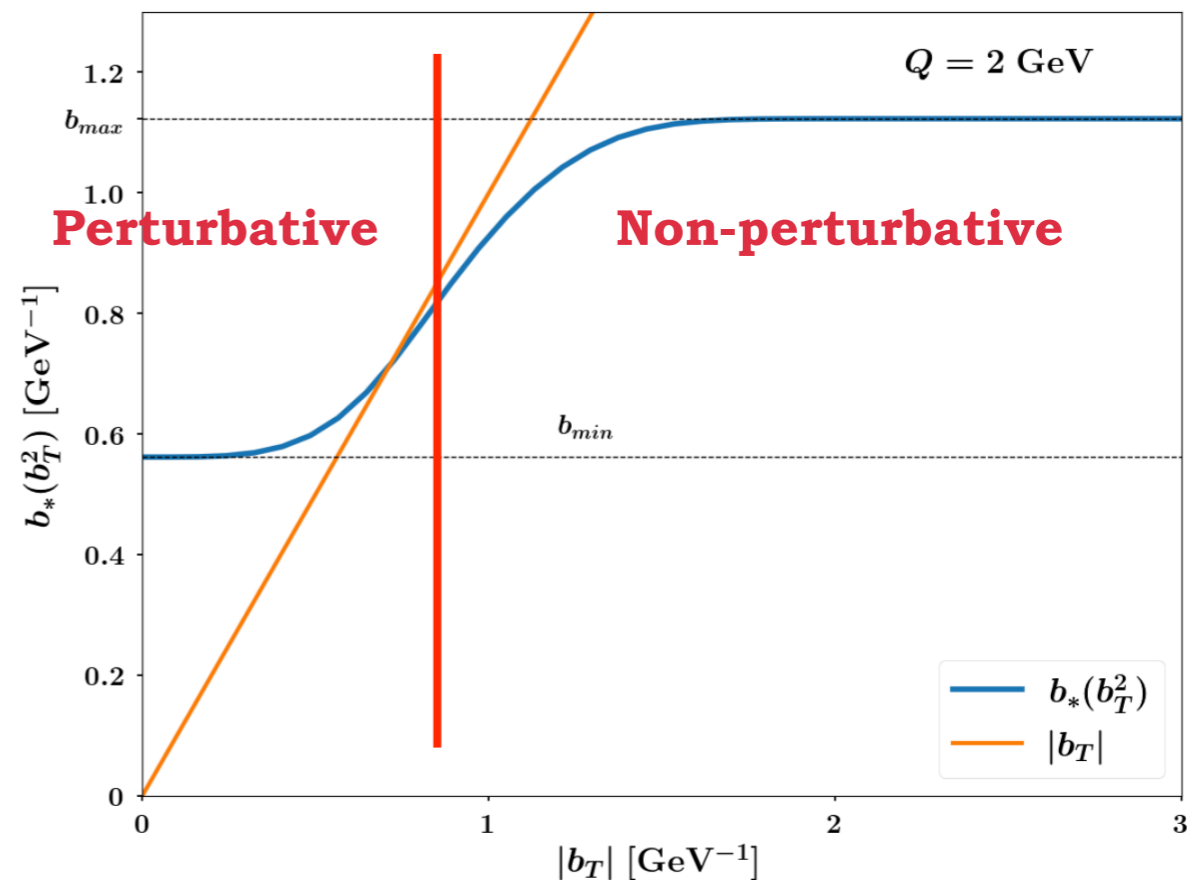
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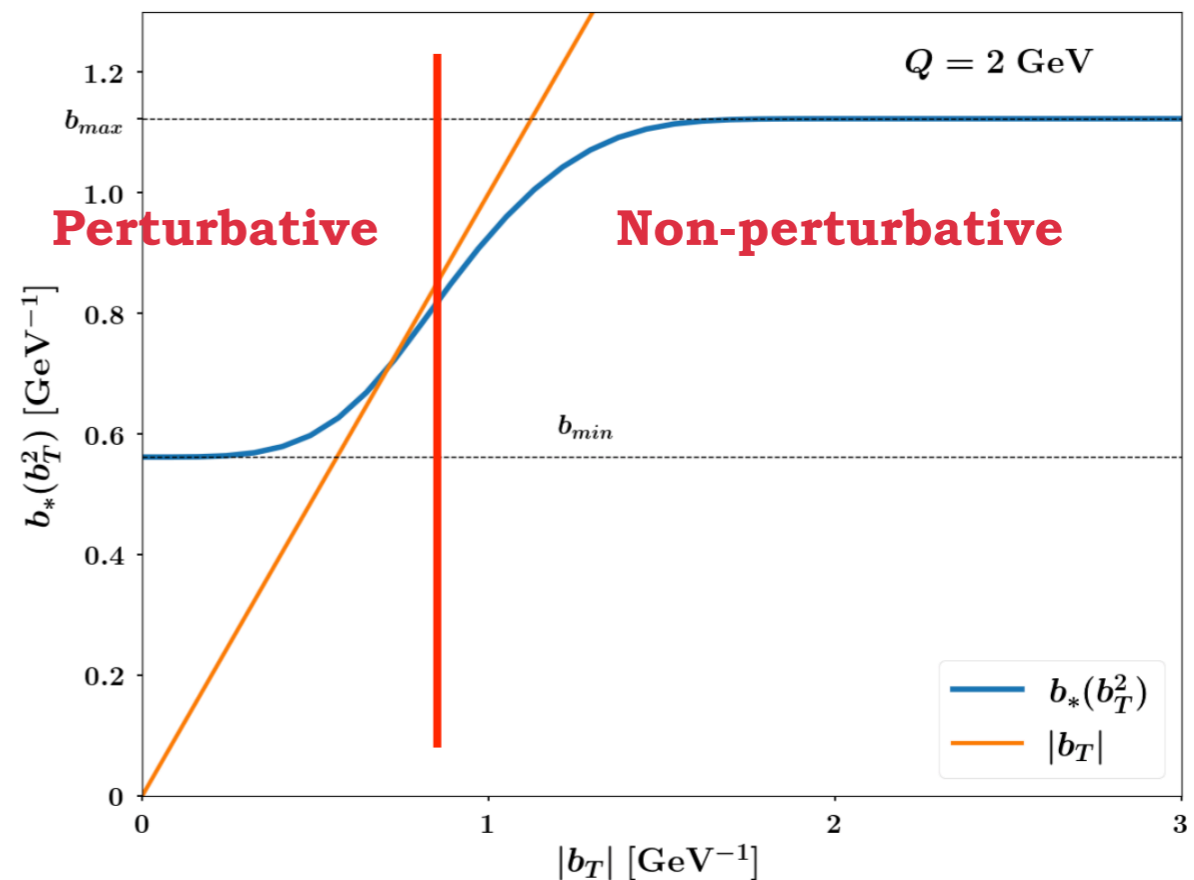
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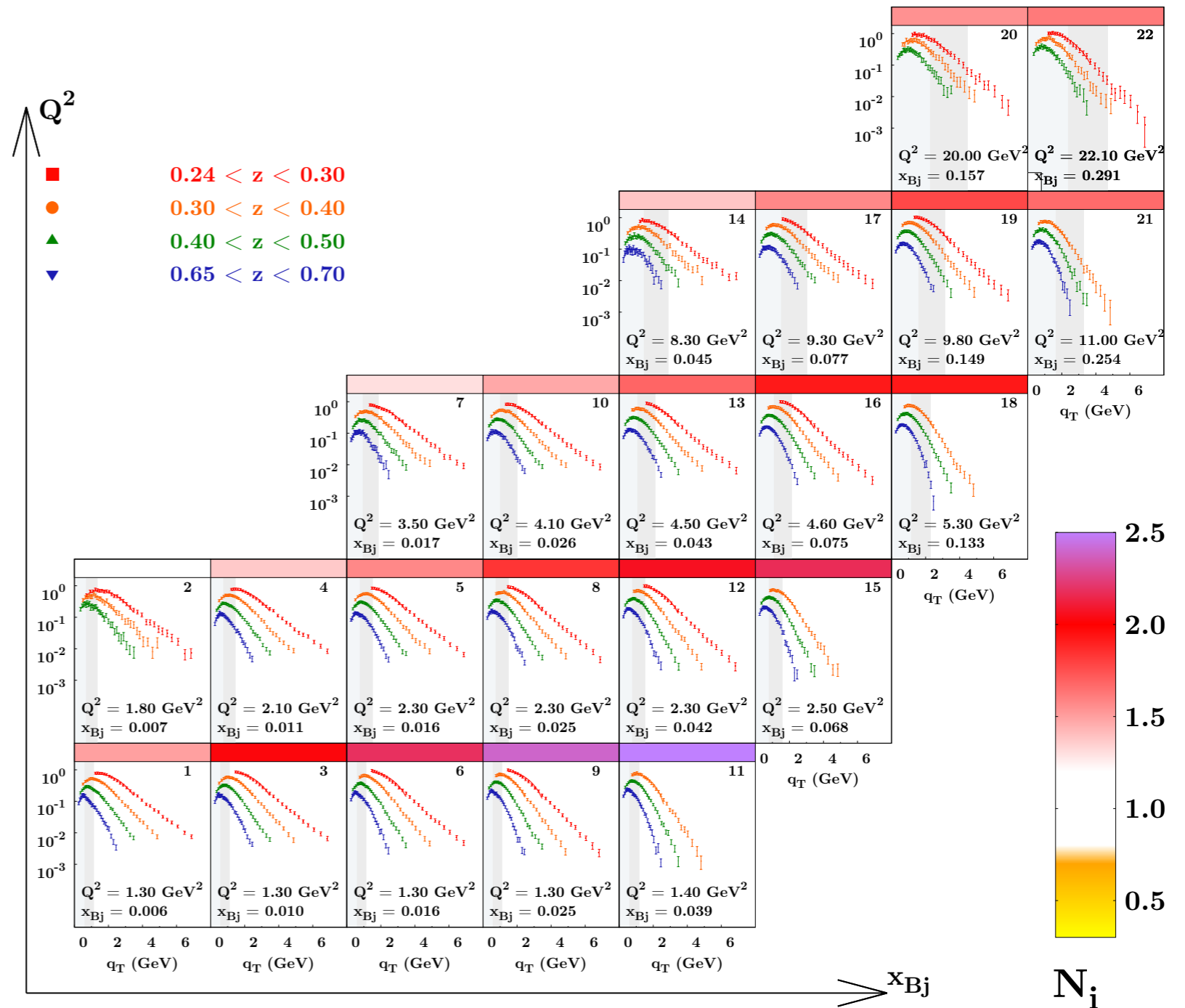
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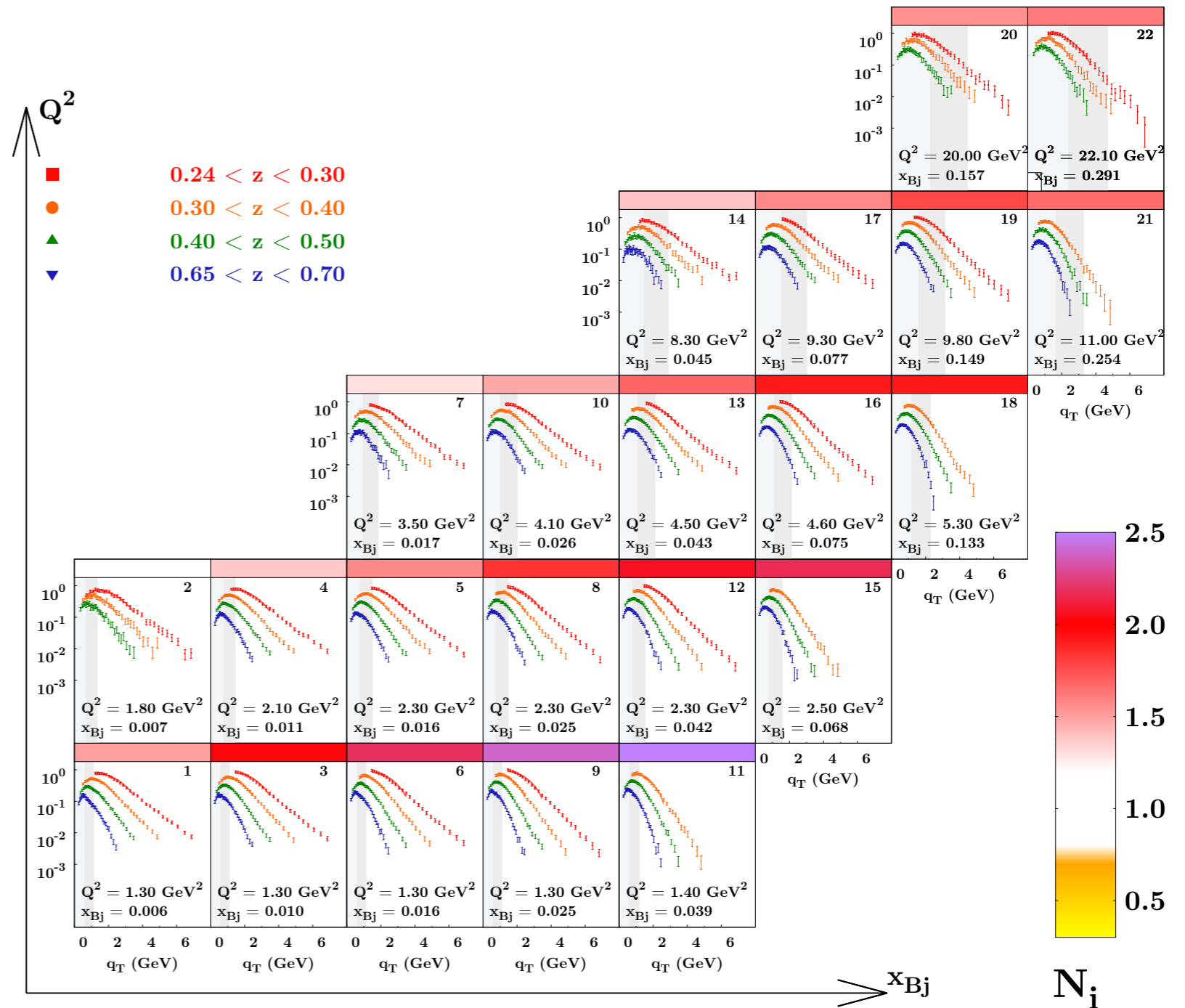
Normalization issue confirmed also in other analyses from different collaborations



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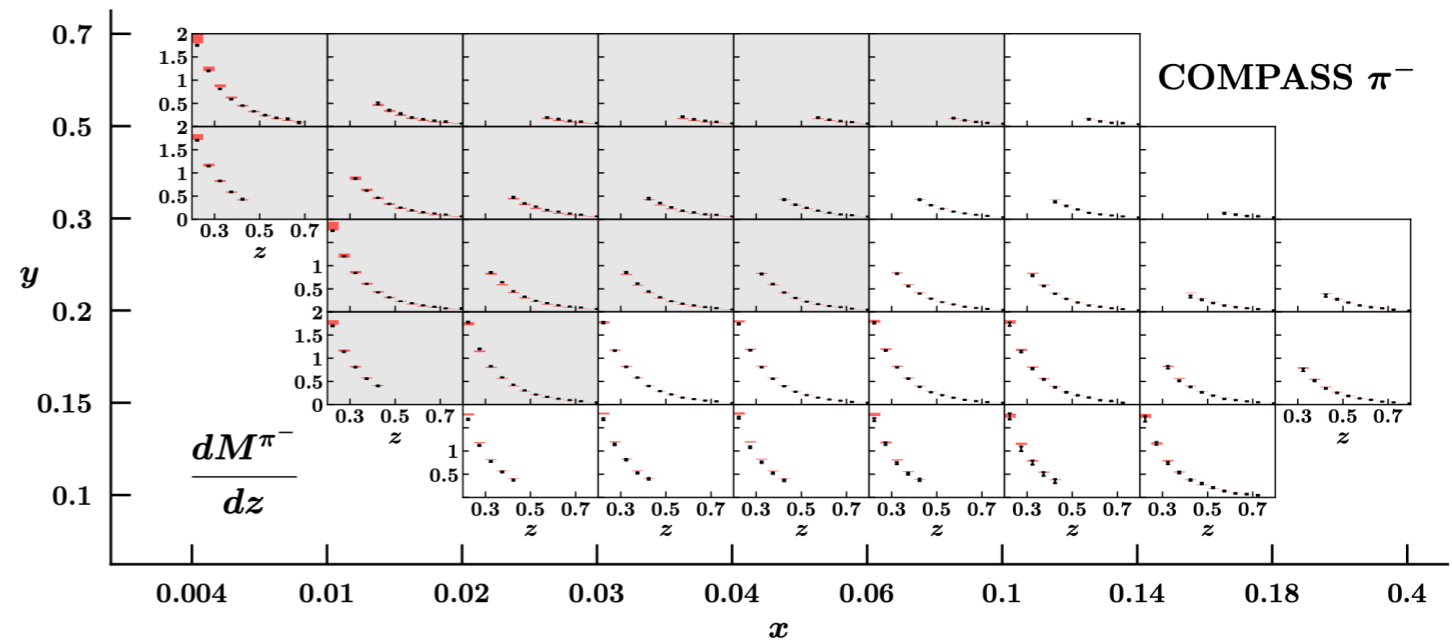
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Khalek, Bertone, Nocera, et al., PRD 104 (2021)

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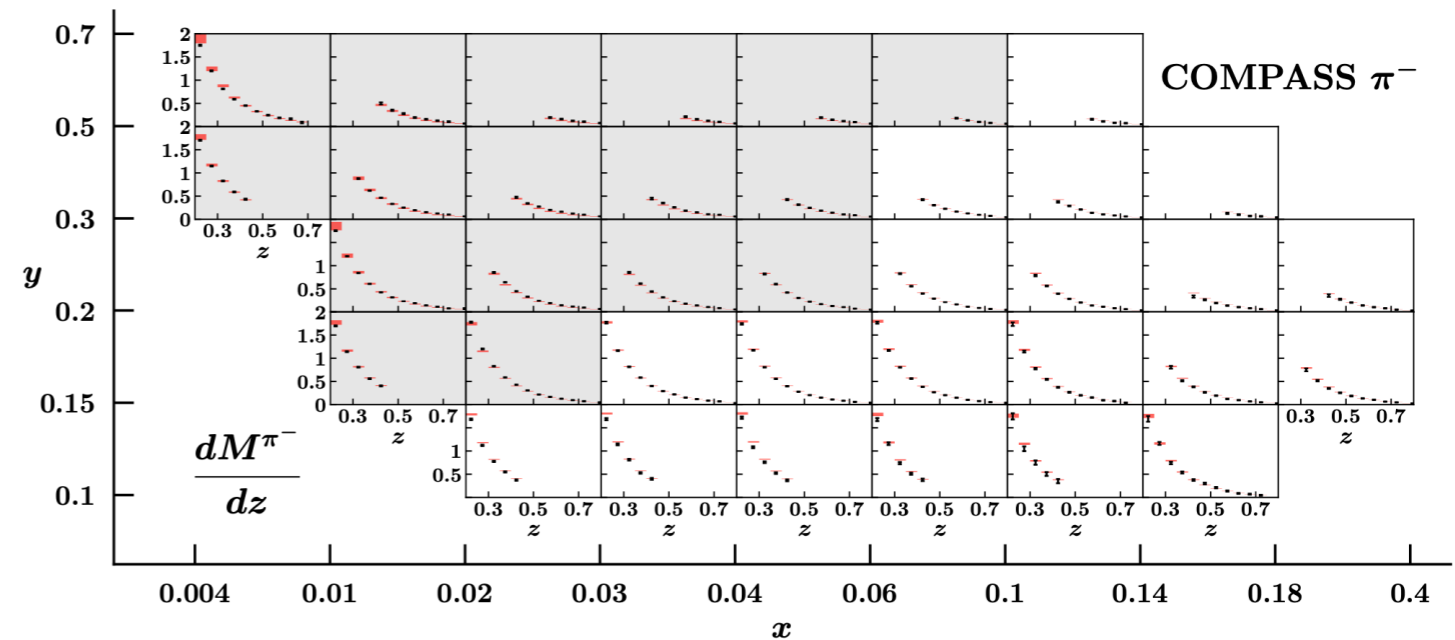
Collinear SIDIS cross section

Normalization of prediction such that

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

Piacenza, PhD thesis (2020)

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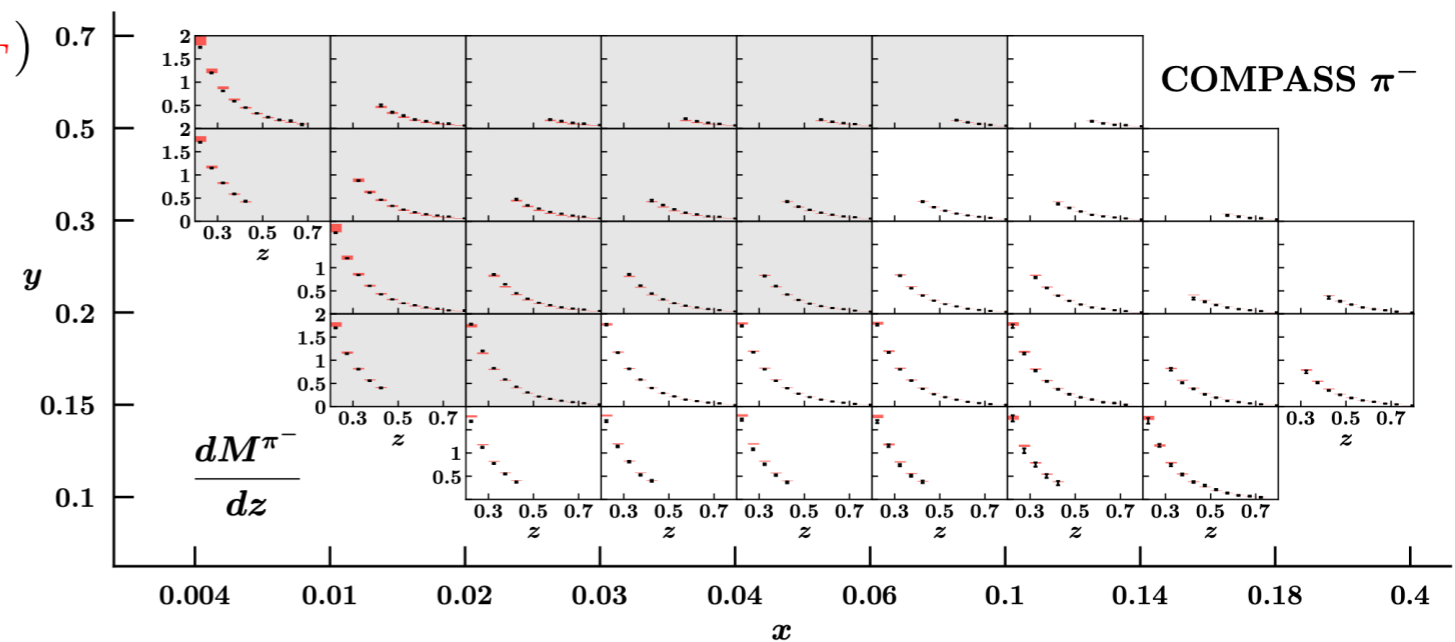
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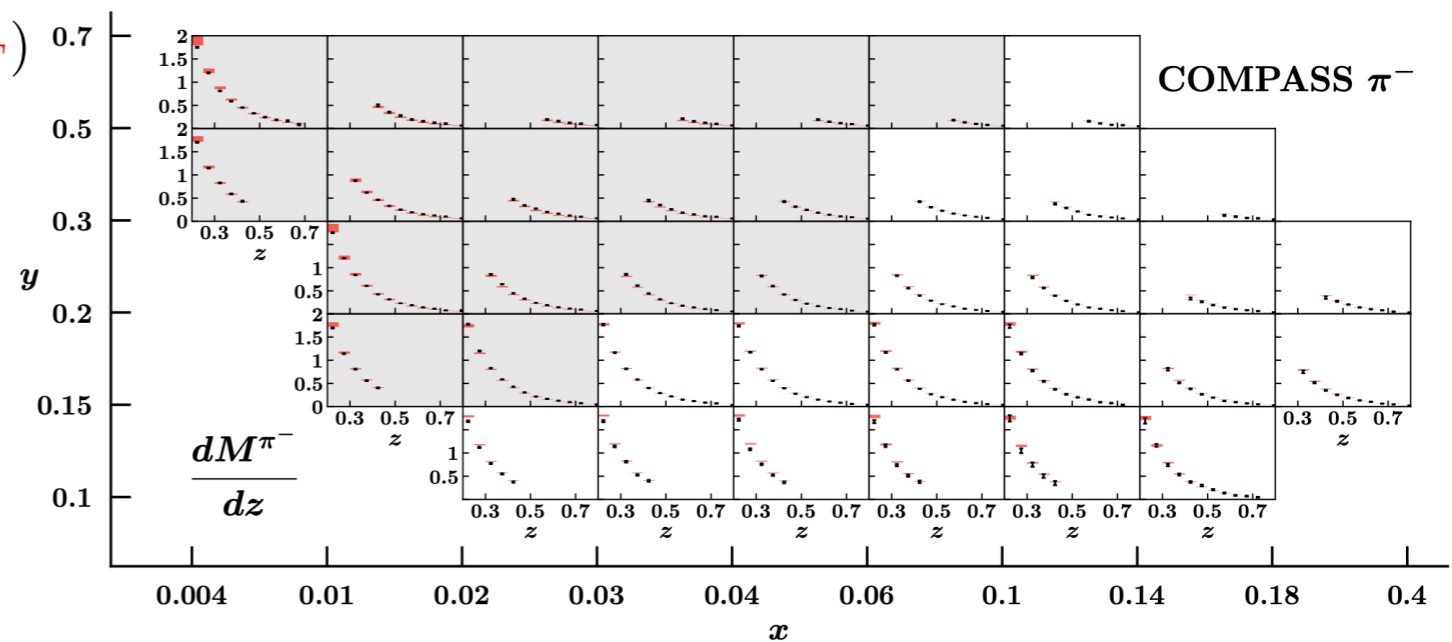
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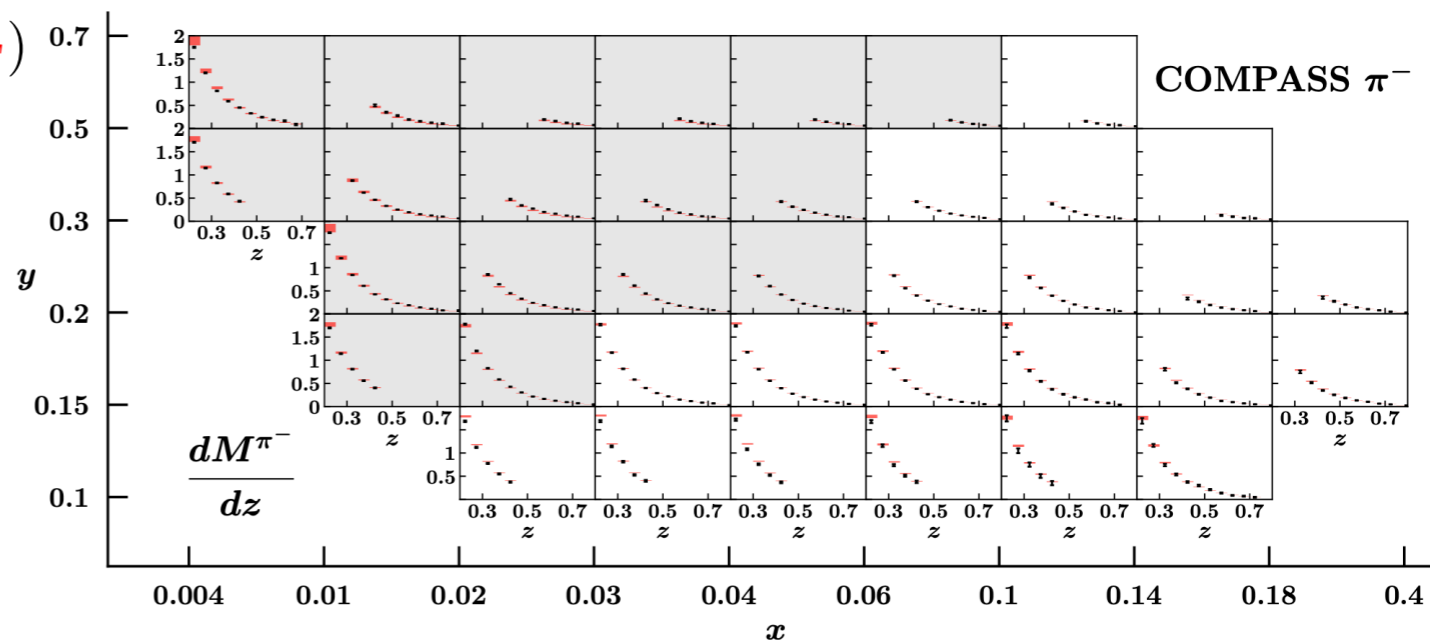
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**Calculable before the fit**

**Good agreement for almost all bins**

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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

