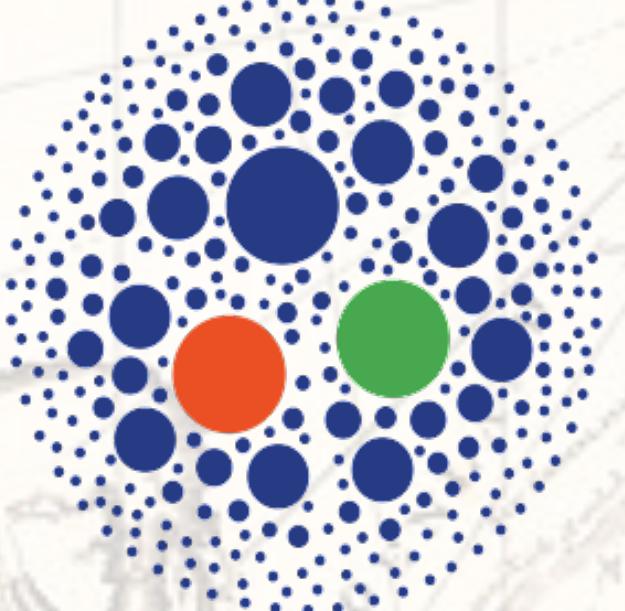




Istituto Nazionale di Fisica Nucleare



**HAS QCD**

HADRONIC STRUCTURE AND  
QUANTUM CHROMODYNAMICS



UNIVERSITÀ  
DI PAVIA

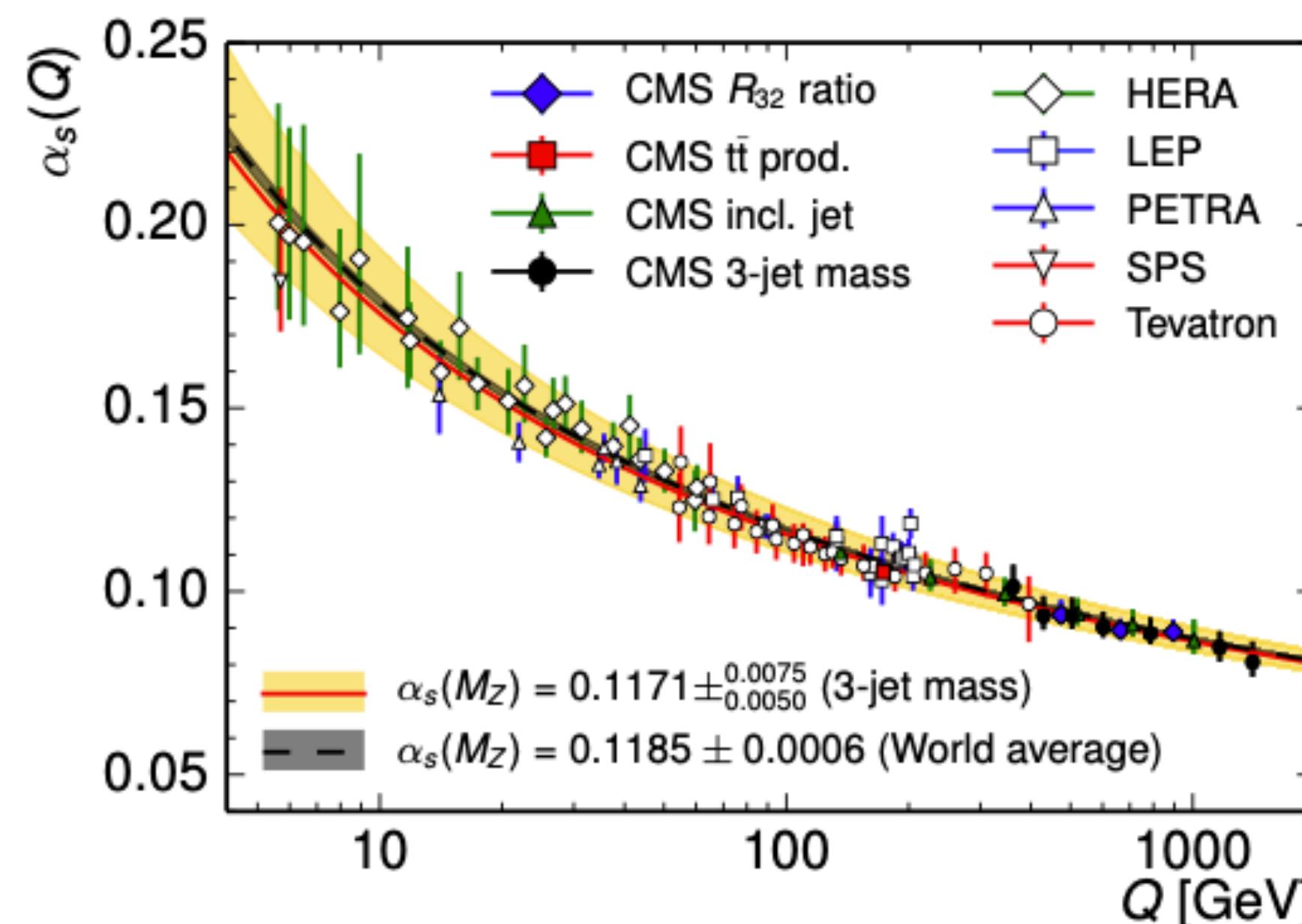
# MAPTMD22:

## A new extraction of unpolarized TMDs through global fits

MAP Collaboration  
Matteo Cerutti

# Running coupling - QCD

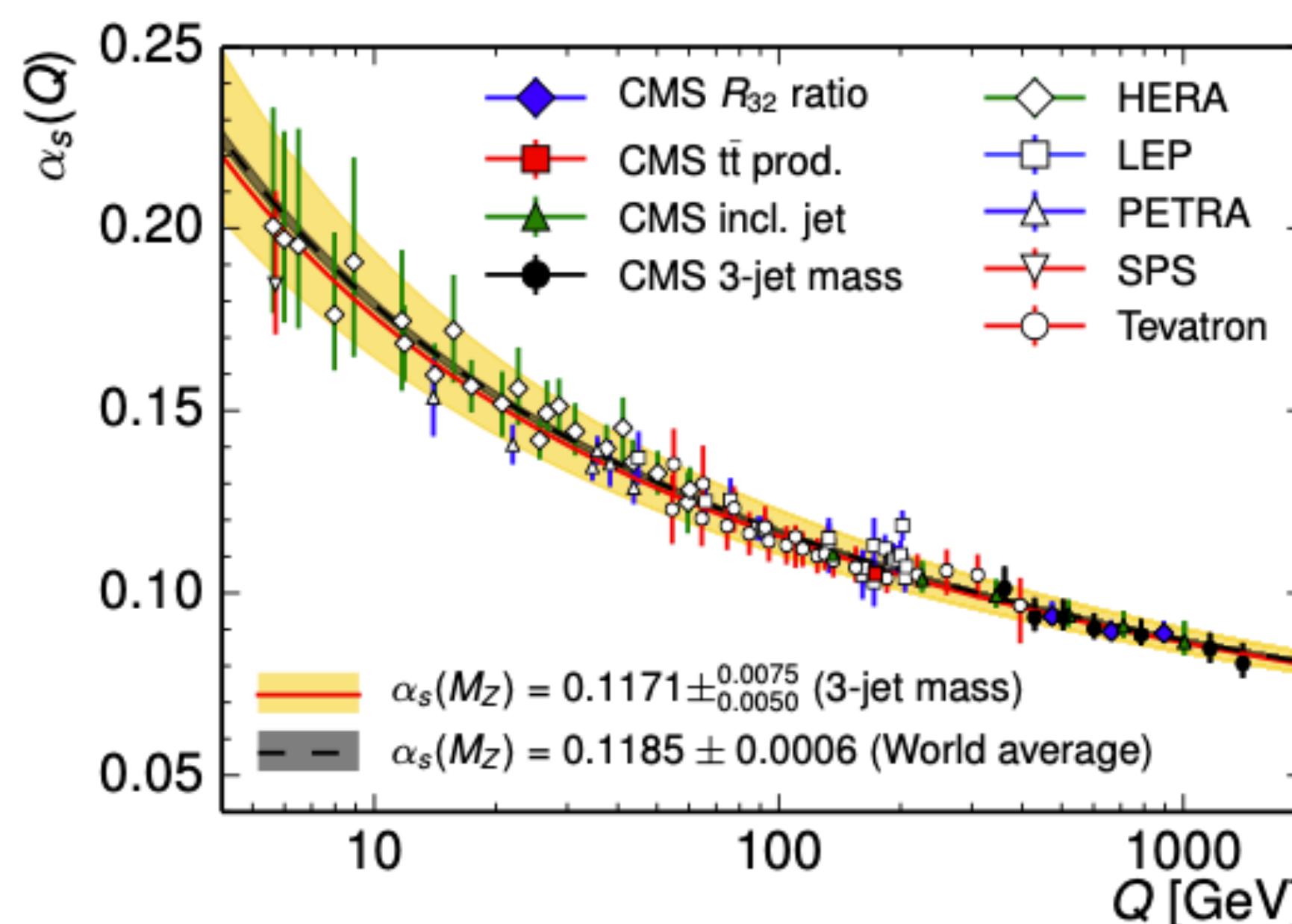
It depends on the energy scale of the process



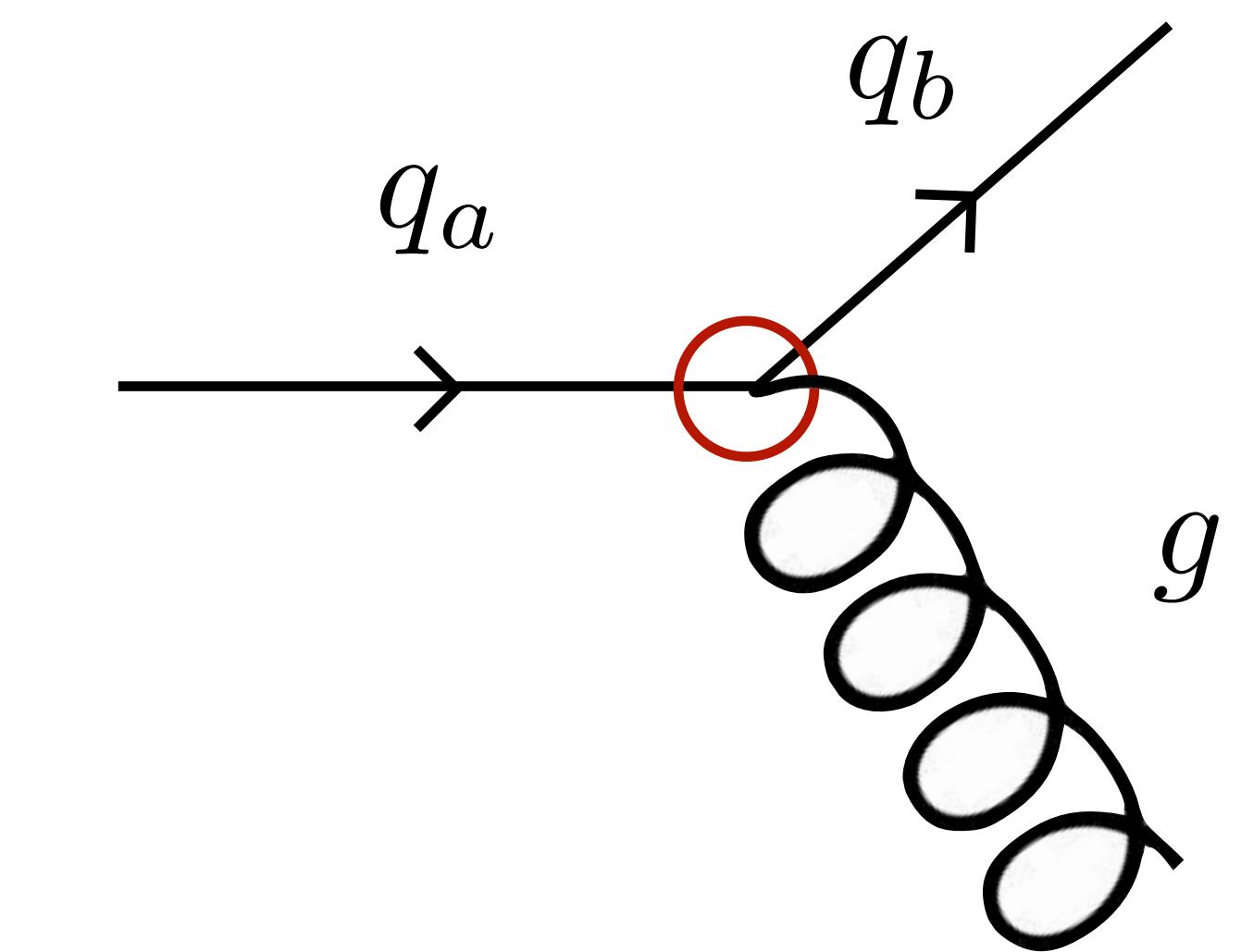
CMS Collaboration, Eur.Phys.J. C 75 (2015)

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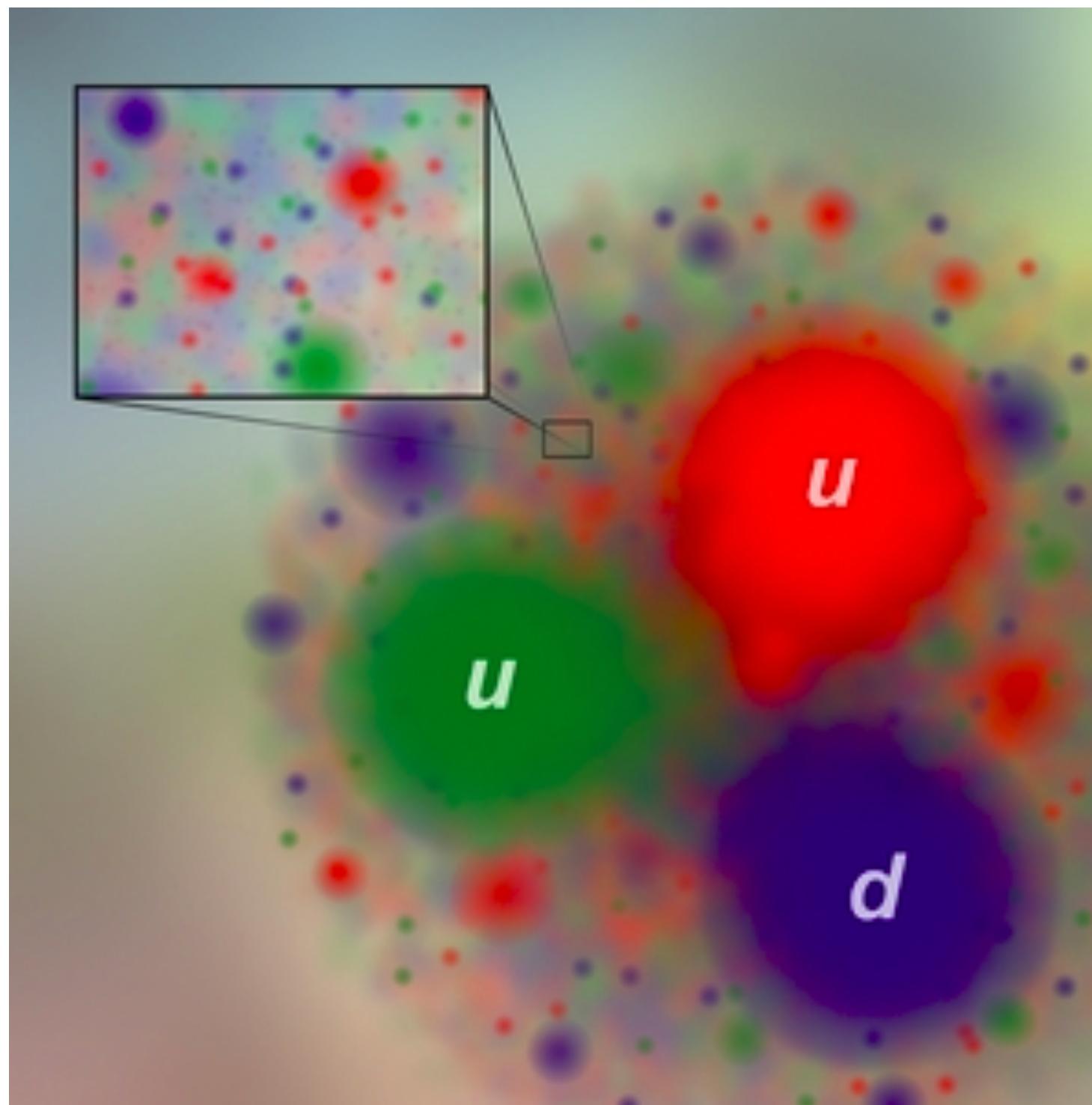
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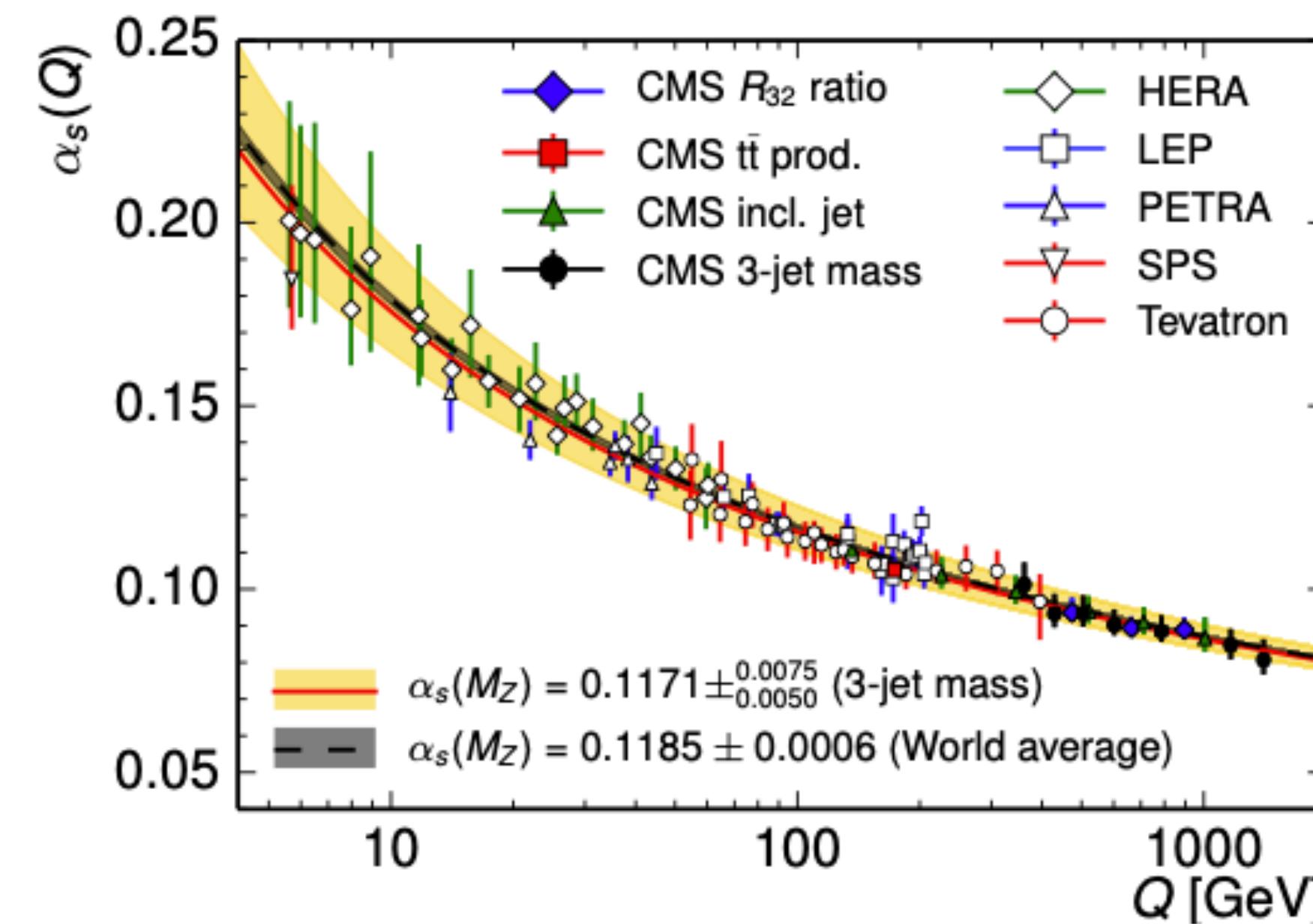
Perturbative Physics

$$d\sigma(Q) \sim d\sigma^{(0)} + \alpha_S^1(Q)d\sigma^{(1)} + \dots$$

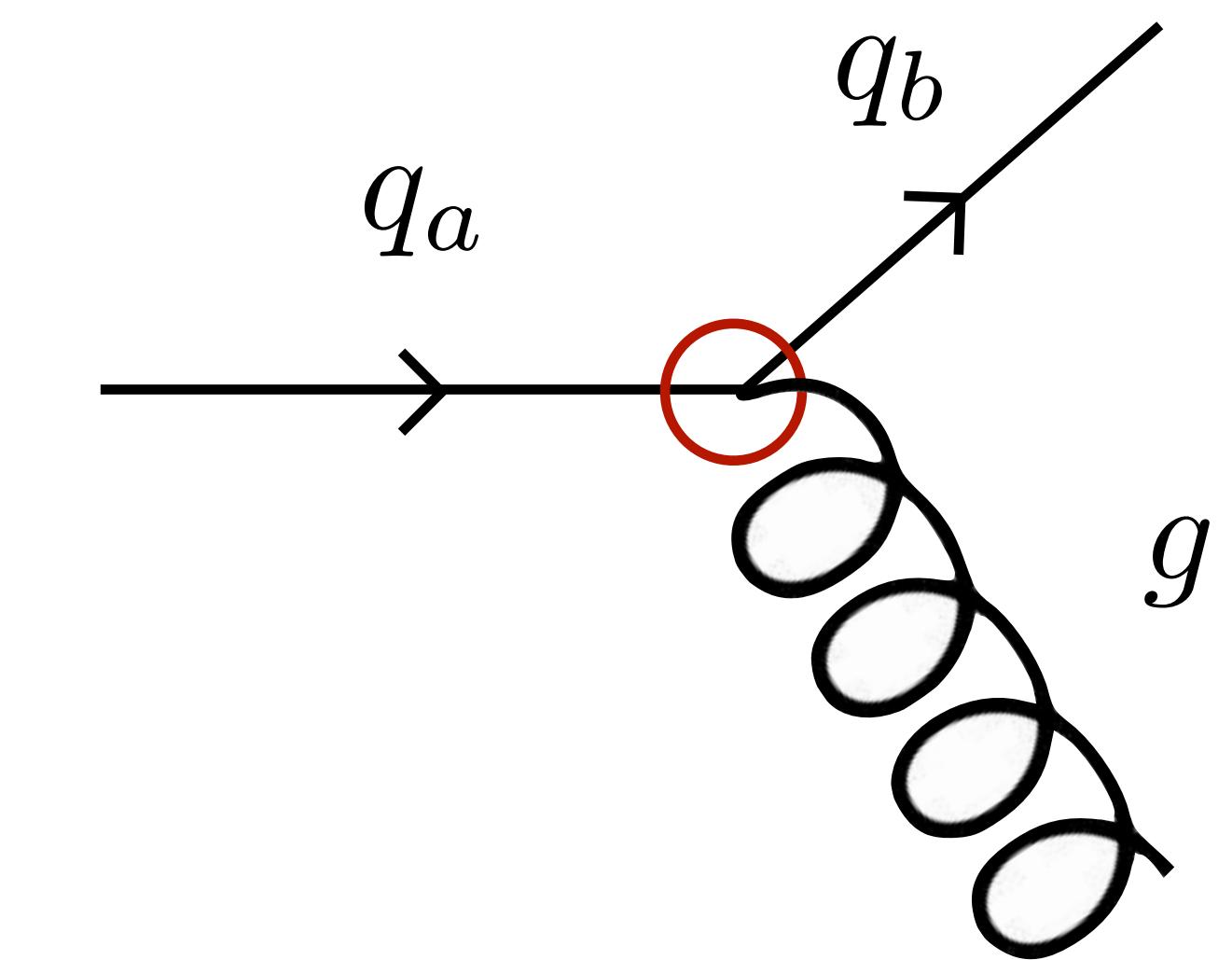
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It depends on the energy scale of the process



CMS Collaboration, Eur.Phys.J. C 75 (2015)



Non-Perturbative Physics

Color confinement

Perturbative Physics

$$d\sigma(Q) \sim d\sigma^{(0)} + \alpha_S^1(Q)d\sigma^{(1)} + \dots$$

# Understanding the color confinement

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Map of the internal  
structure of hadrons



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Map of the internal  
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Parton Distribution  
Functions (PDFs)

# Understanding the color confinement

Map of the internal structure of hadrons

Hadronization process

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Fragmentation Functions (FFs)

# Parton Distribution Functions (PDFs)

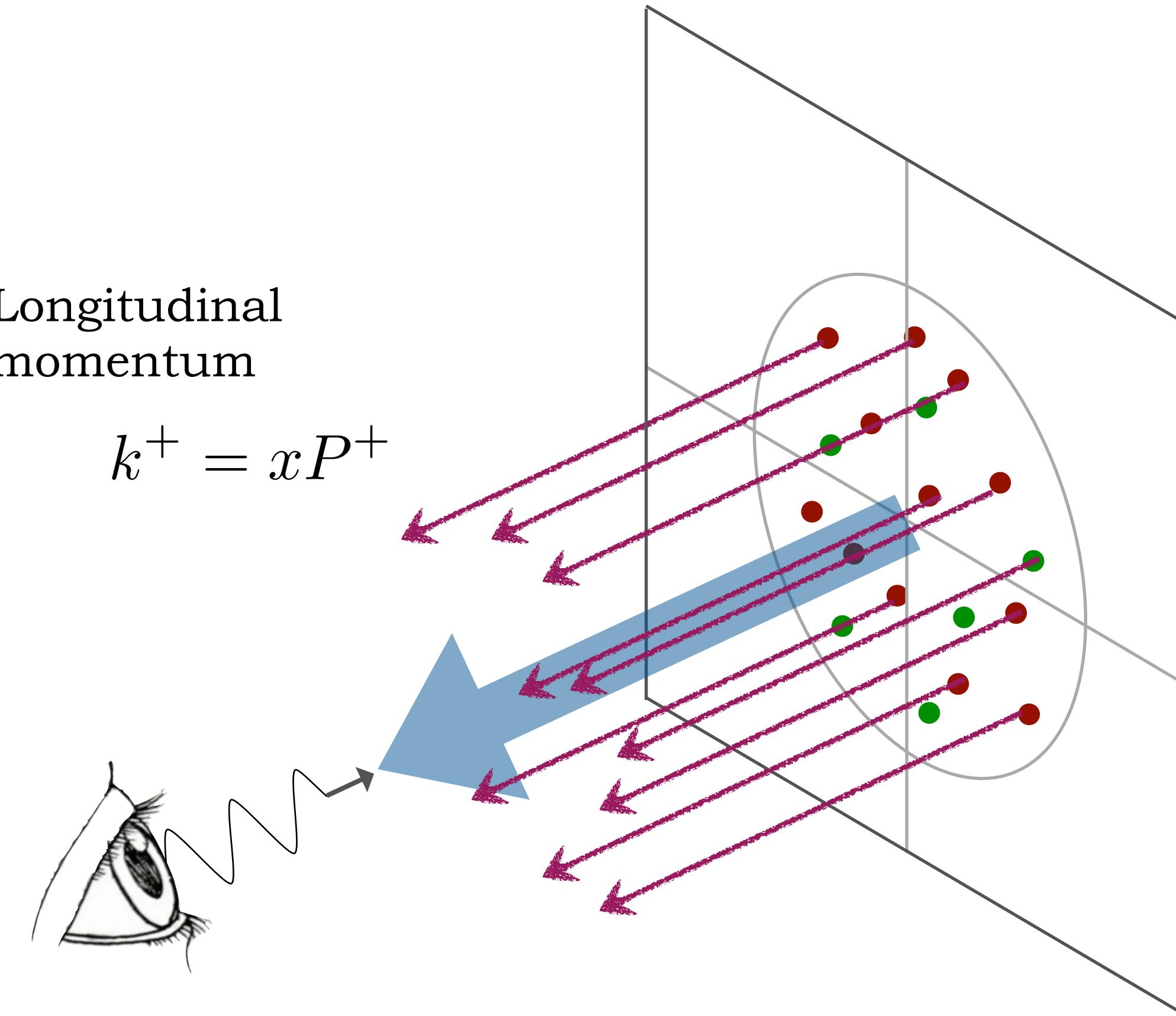
## 1D maps

### Collinear framework

The only nonzero component of the quark momentum is the one in the same direction of the parent hadron

Longitudinal momentum

$$k^+ = xP^+$$



$x$  : fraction of longitudinal momentum of the parent hadron carried by the internal quark

# Parton Distribution Functions (PDFs)

## 1D maps

### Collinear framework

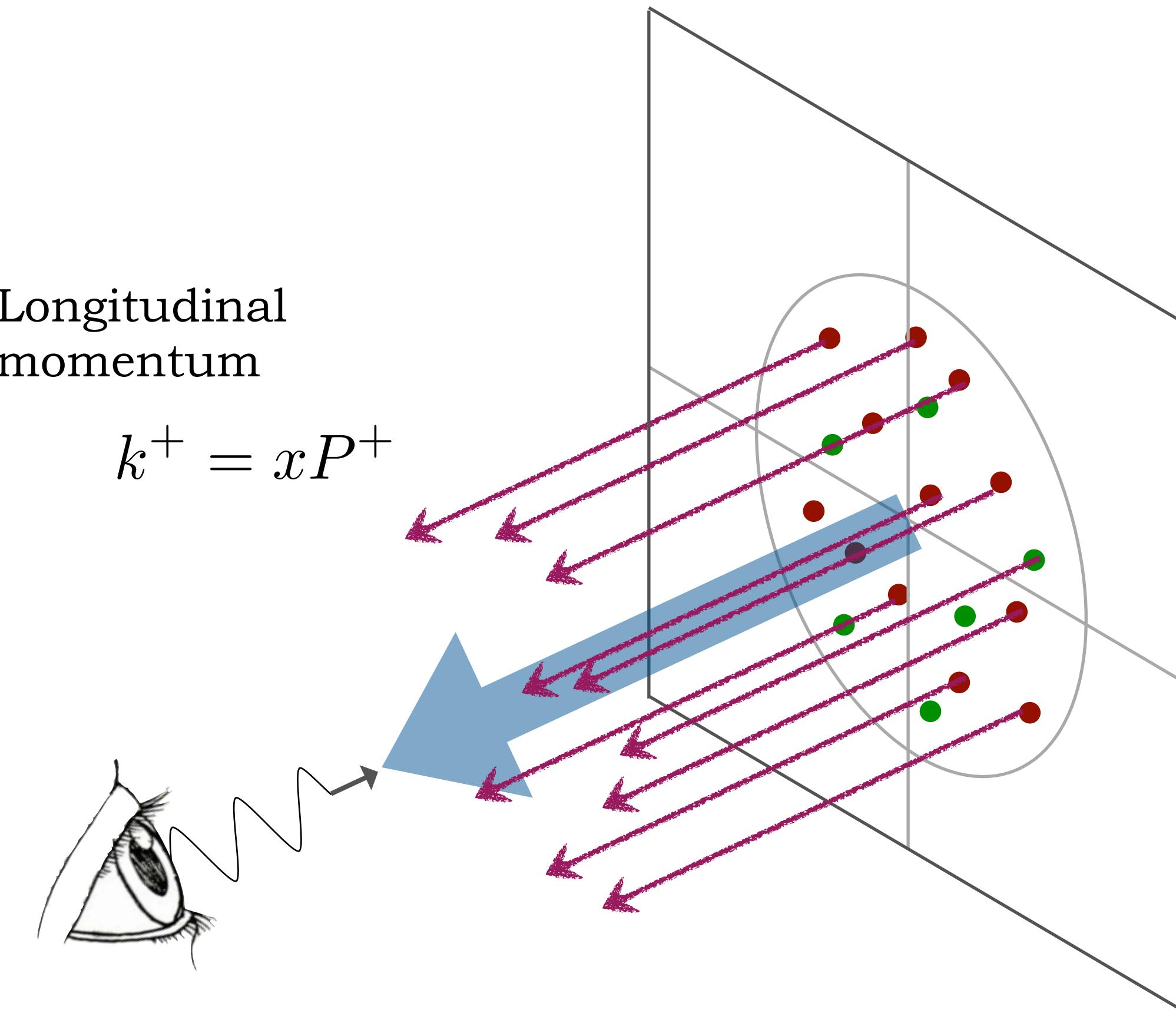
Quark Polarisation

Nucleon Pol.

	U	L	T
U	$f_1(x)$		
L			
T			

Longitudinal  
momentum

$$k^+ = xP^+$$



$$f_1(x)$$

probability density of finding an unpolarised quark (gluon) carrying a fraction  $x$  of the unpolarised hadron momentum

# Parton Distribution Functions (PDFs)

## 1D maps

### Collinear framework

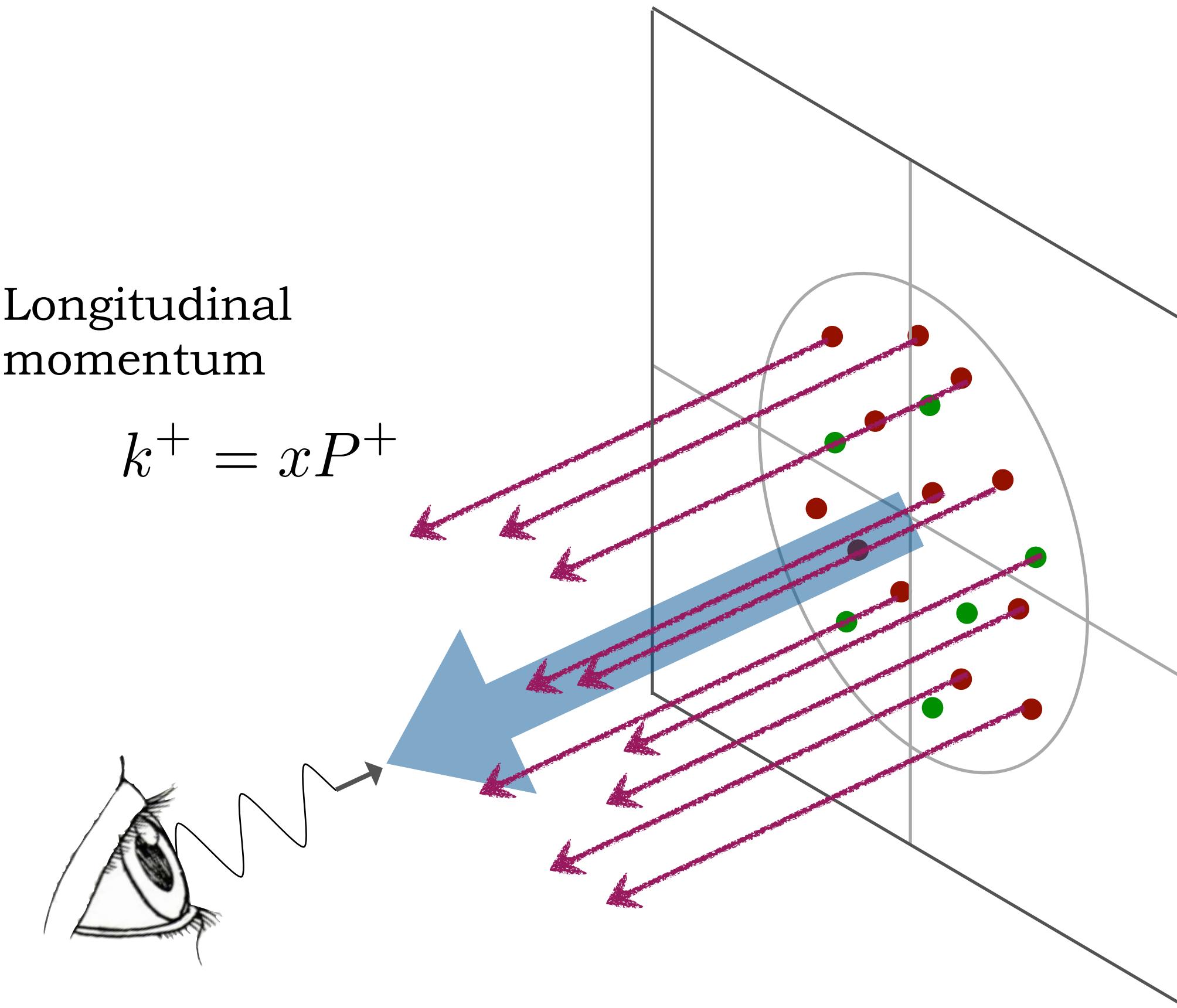
#### Quark Polarisation

Nucleon Pol.

	U	L	T
U	$f_1(x)$		
L		$g_1(x)$	
T			$h_1(x)$

Longitudinal  
momentum

$$k^+ = xP^+$$



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# Unexplained observations

## 1D maps

We cannot explain, for e.g. :

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- ✖ Single Spin Asymmetries (SSA)

J. Adams et al., P.R.L. 92 (2004) 171801

- ✖ Violation of the Lam-Tung sum rule

J. S. Conway et al., P.R. D39 (1989)

- ✖ Results of EMC experiment (“spin crisis”)

J. Ashman et al., P.L. B206 (1988)

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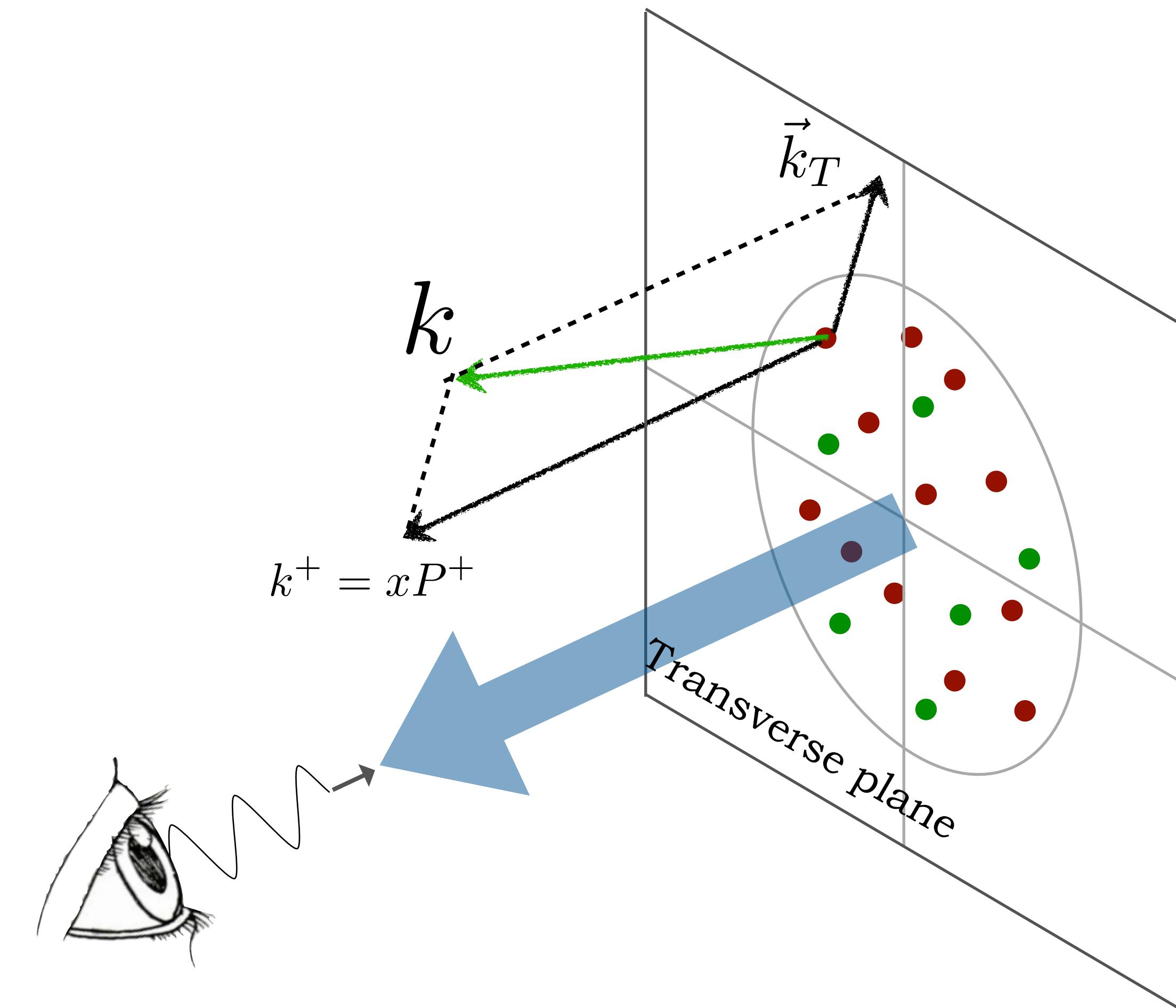


# Transverse-Momentum-Dependent PDFs (TMDs)

## 3D maps

### Non-collinear framework

The quark momentum is characterised by an intrinsic component transverse to the parent hadron momentum



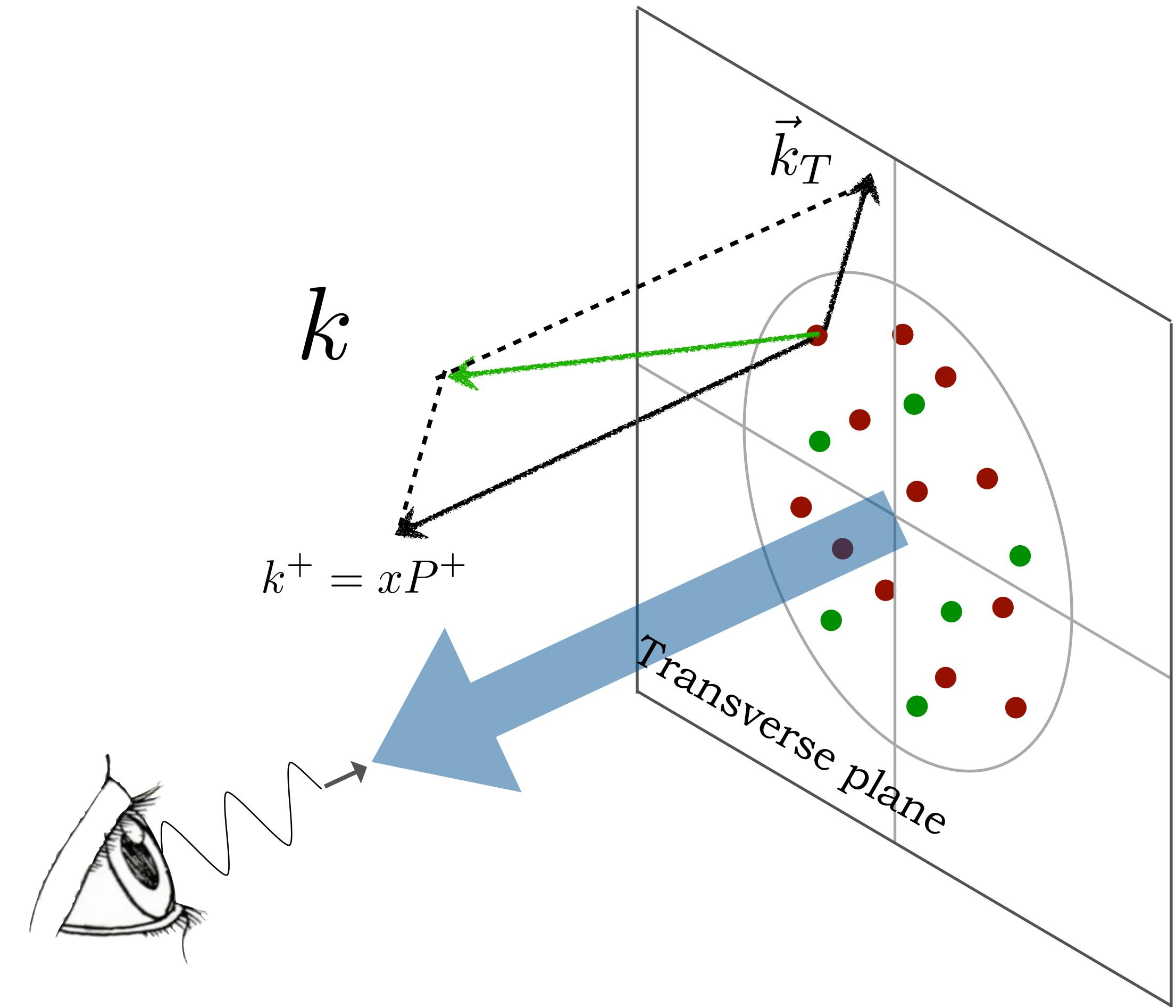
$\vec{k}_T$  = intrinsic (non-perturbative) transverse momentum of the quark

# 3D Maps

Quark Polarization

	U	L	T
U			
L			
T			

Nucleon Pol.

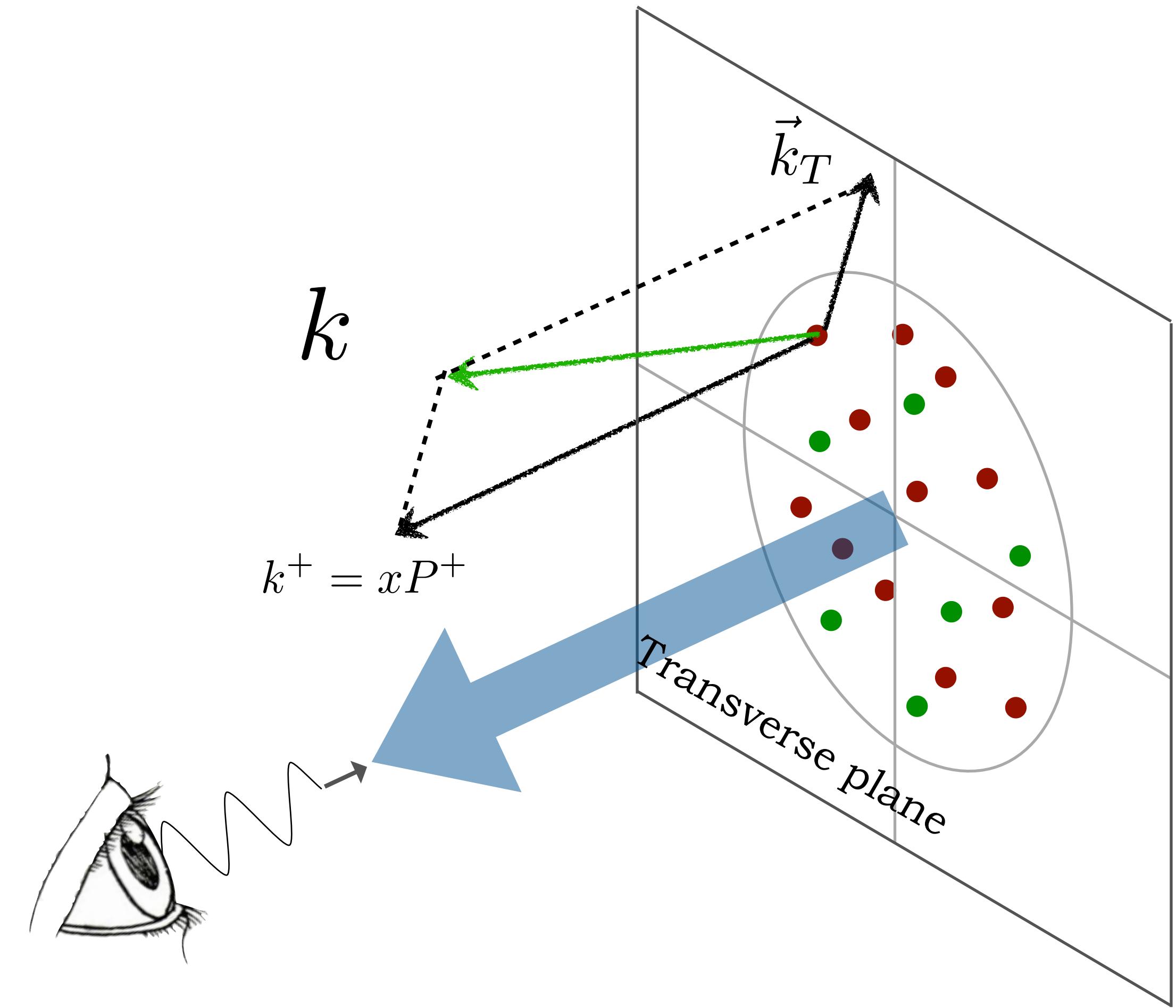


# 3D Maps

Quark Polarization

	U	L	T
U	$f_1$		
L		$g_1$	
T			$h_1$

Nucleon Pol.

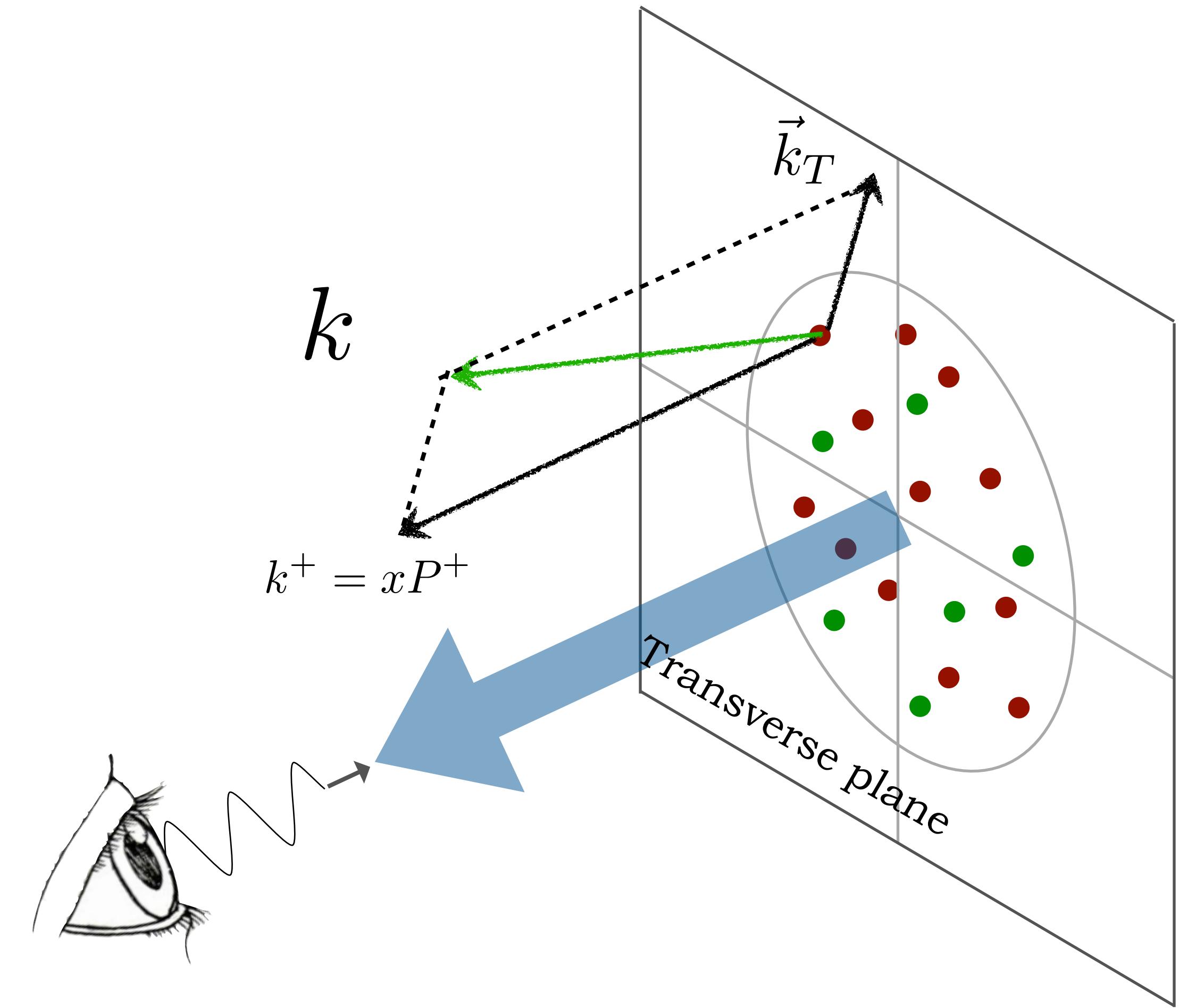


# 3D Maps

Quark Polarization

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

Nucleon Pol.



# 3D Maps

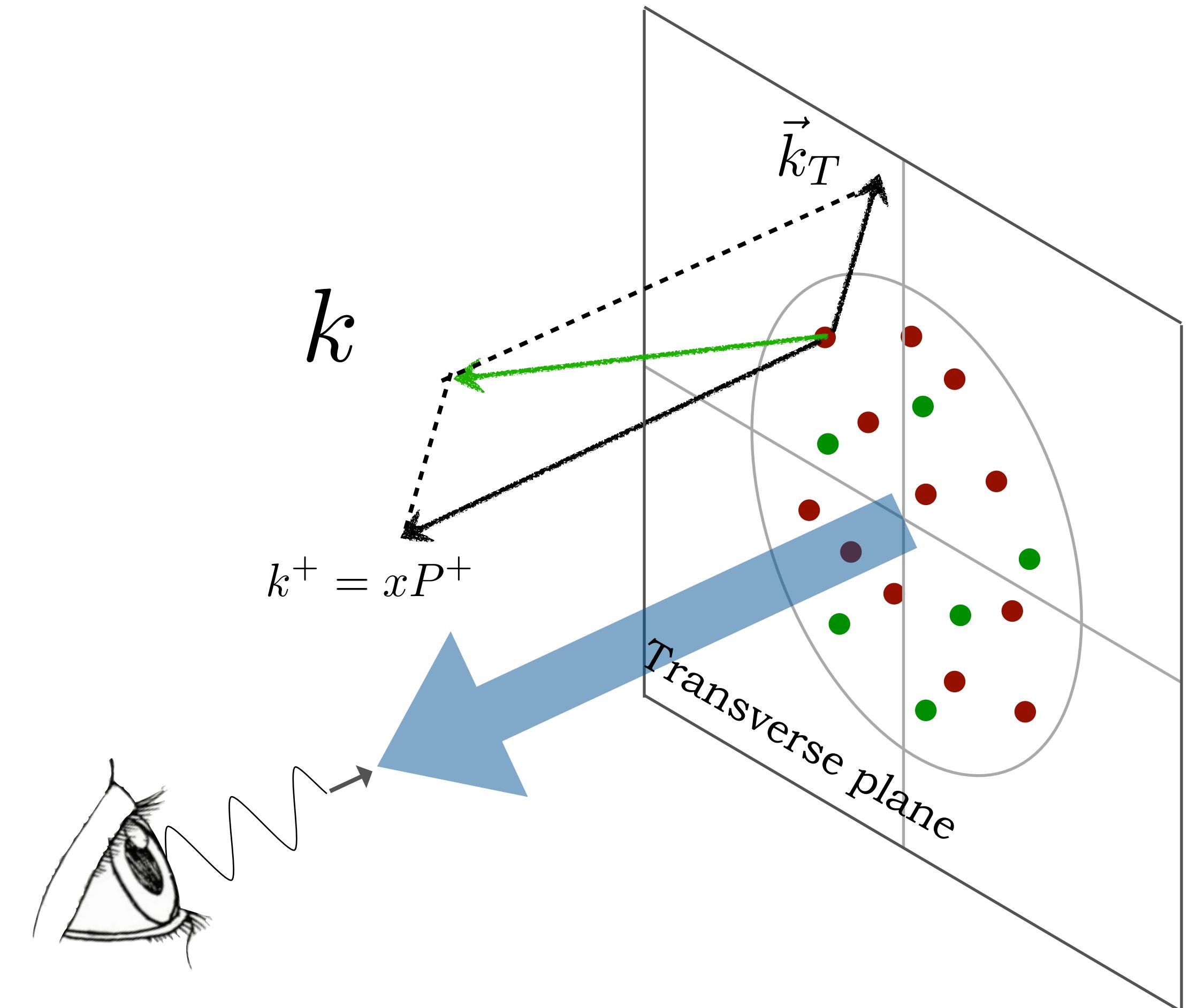
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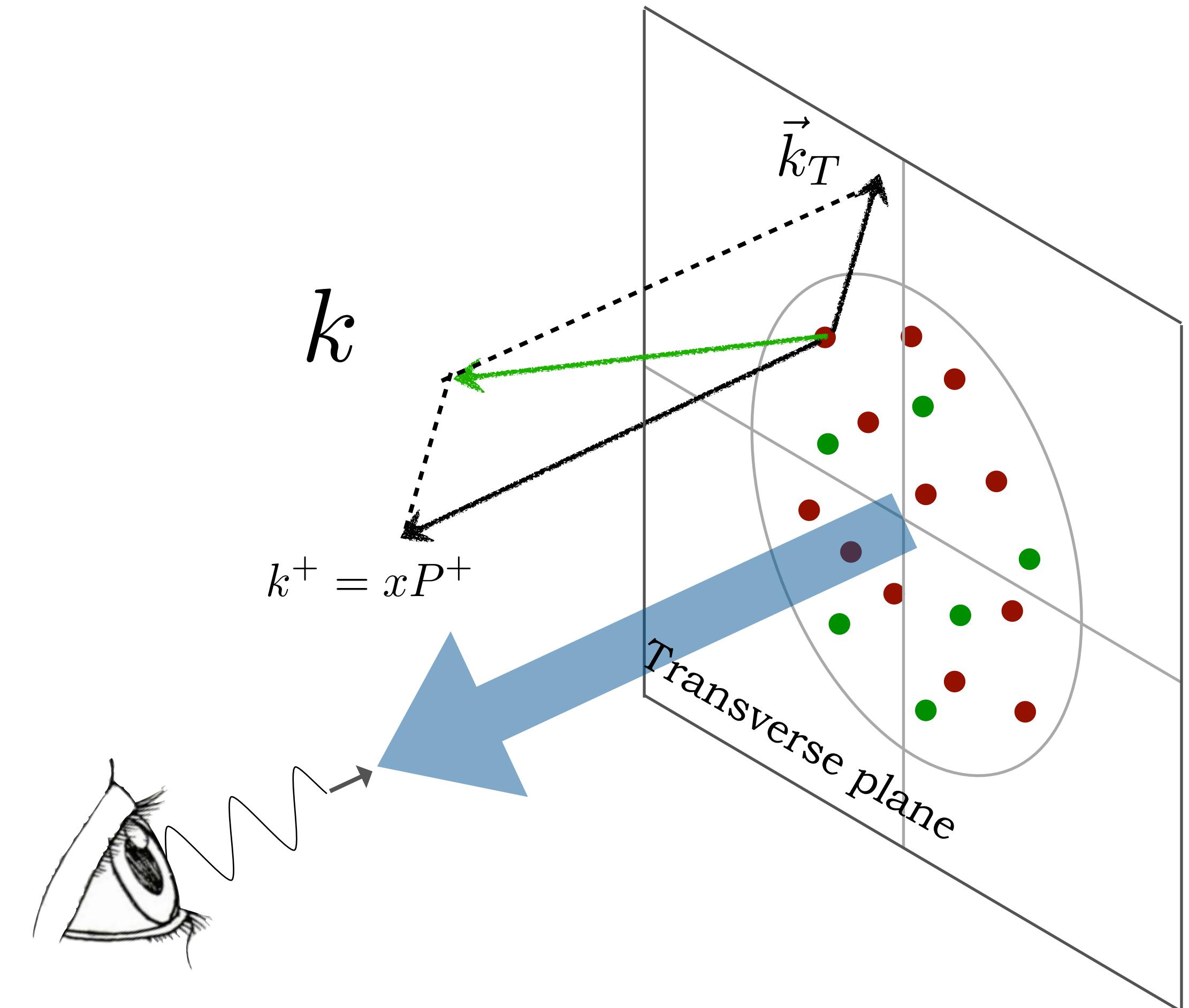
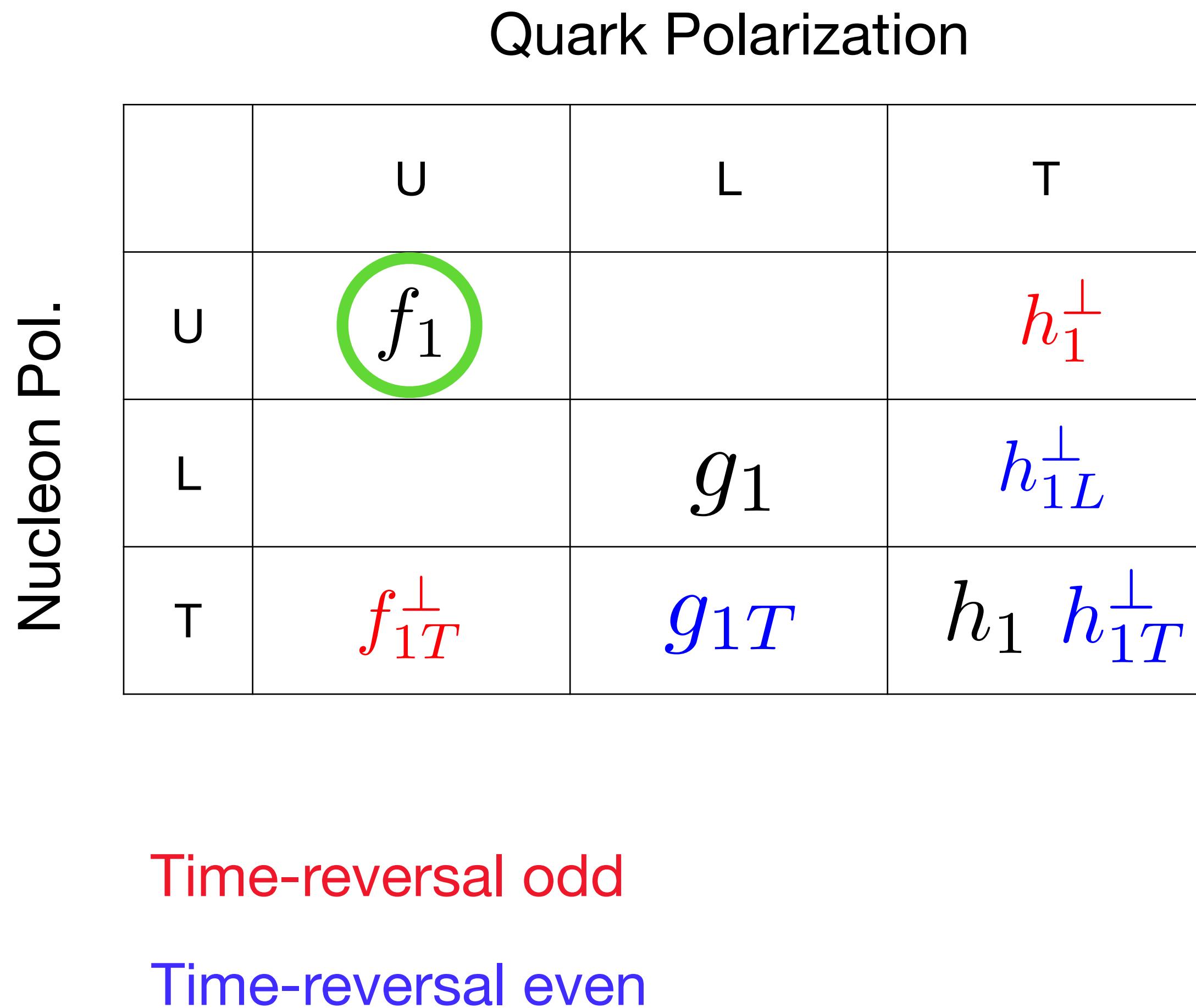
Nucleon Pol.

Time-reversal odd

Time-reversal even



# 3D Maps



# TMD Factorization

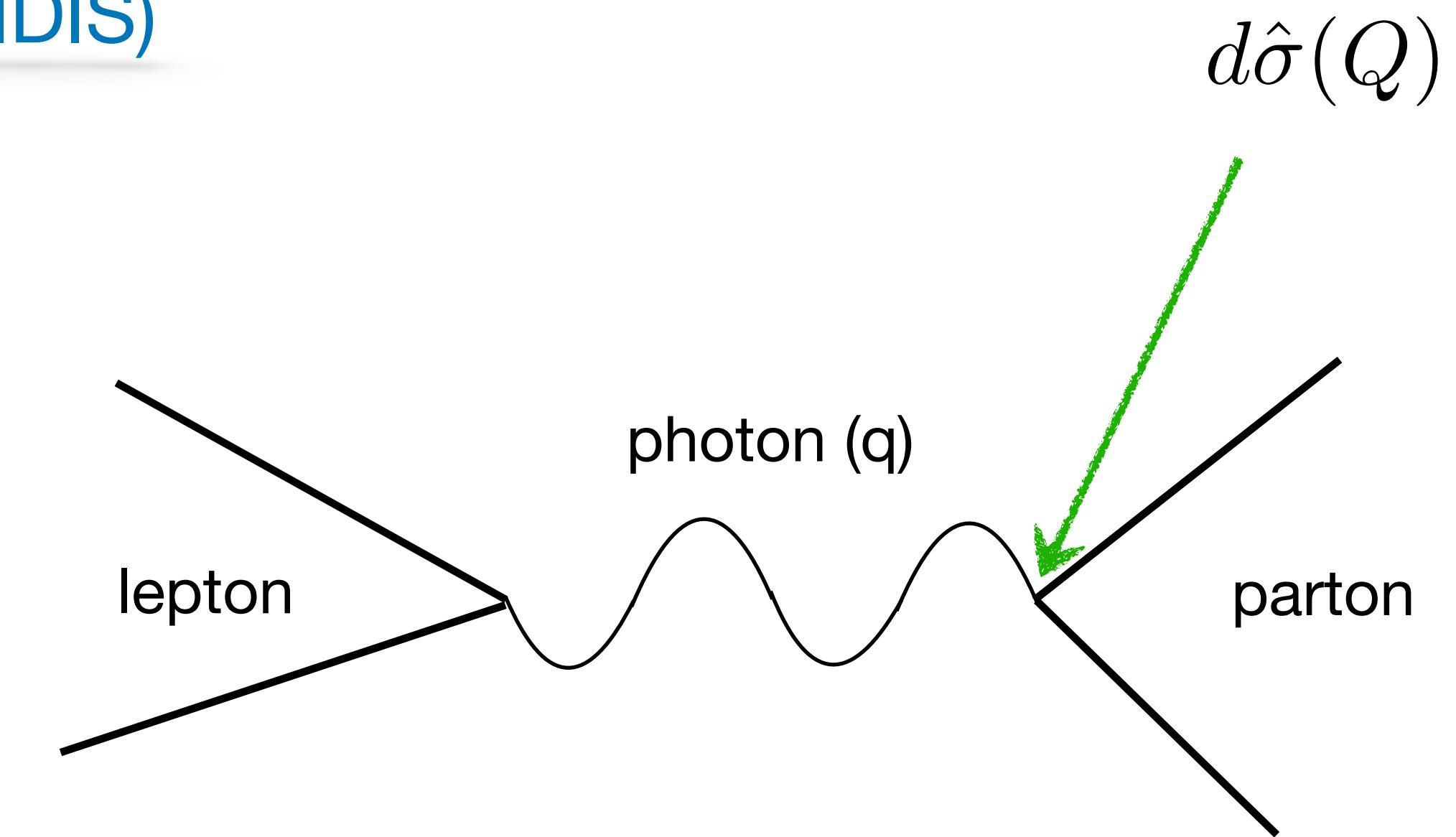
Semi-Inclusive Deep Inelastic Scattering (SIDIS)

If  $Q^2 \gg M^2$

$d\sigma \sim \text{perturbative} \otimes \text{nonperturbative}$



Elementary cross section



# TMD Factorization

Semi-Inclusive Deep Inelastic Scattering (SIDIS)

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

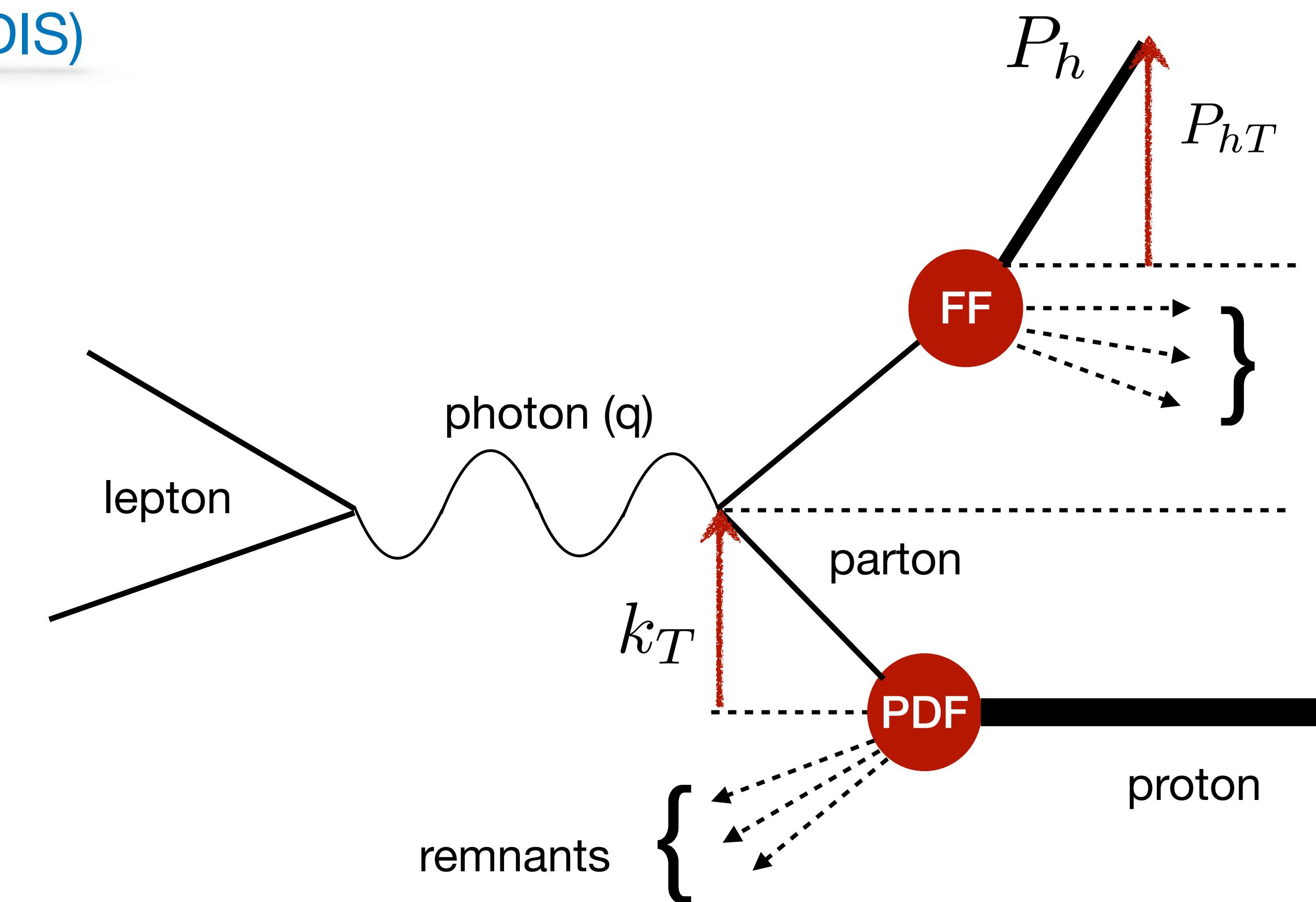
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Elementary cross section



TMD partonic densities



# TMD Factorization

Factorization theorems for several processes:

- $l + N \rightarrow l' + h + X$  SIDIS
- $e^+ + e^- \rightarrow h_1 + h_2 + X$  DIA
- $H_1 + H_2 \rightarrow l^- + l^+ + X$  Drell Yan
- $H_1 + H_2 \rightarrow W/Z + X$  Drell Yan
- $H_1 + H_2 \rightarrow \text{jet} + X$
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Factorization theorems for several processes:

- $l + N \rightarrow l' + h + X$  SIDIS  $\rightarrow$  TMD PDF and FF
- $e^+ + e^- \rightarrow h_1 + h_2 + X$  DIA  $\rightarrow$  TMD FFs
- $H_1 + H_2 \rightarrow l^- + l^+ + X$  Drell Yan  $\rightarrow$  TMD PDFs
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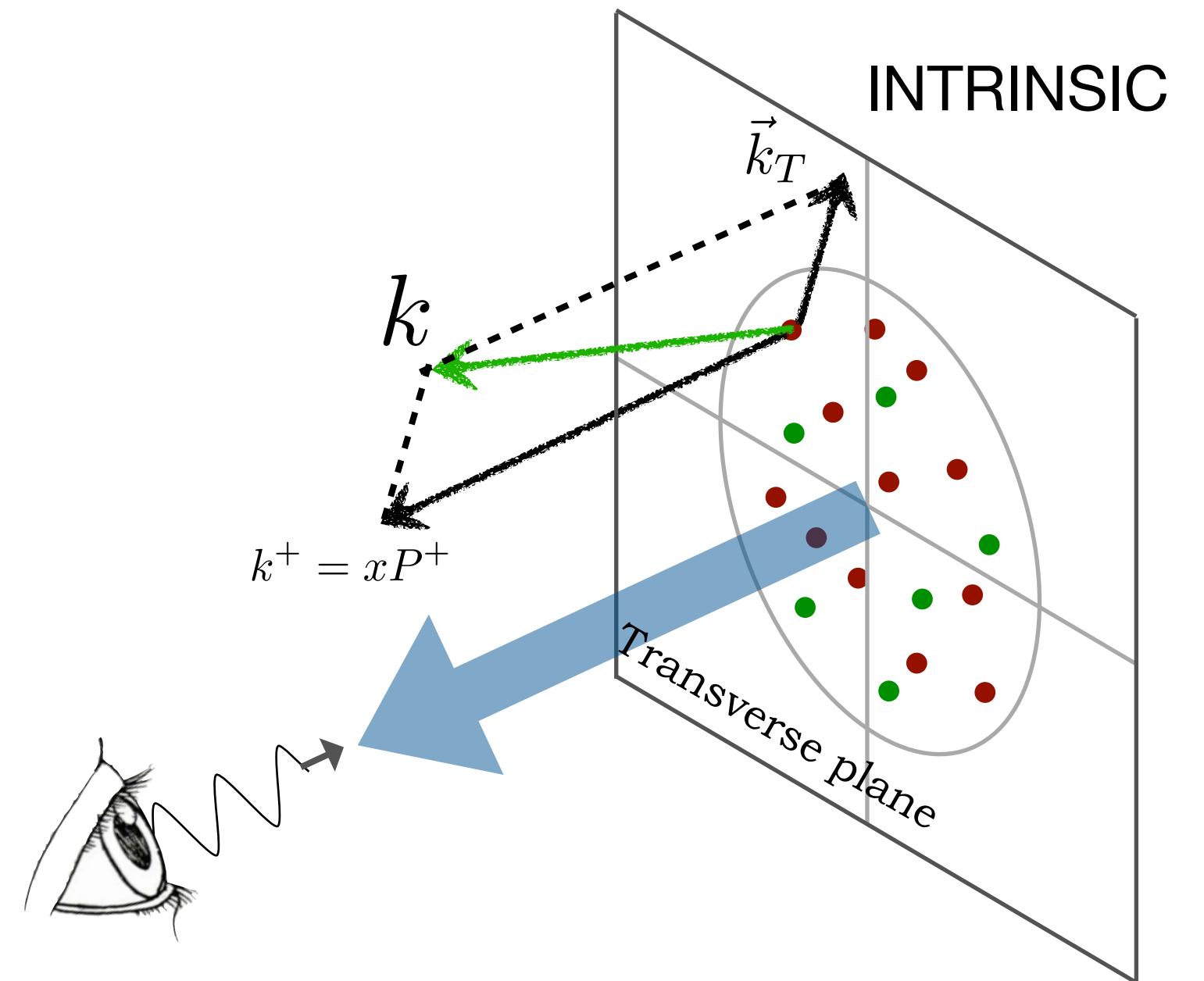
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# Structure of a TMD distribution

Fourier transform in  $b_T$ -space

$$\tilde{F}_a(x, b_T^2; \mu, \zeta) = \int \frac{d^2 k_T}{(2\pi)^2} e^{ib_T \cdot k_T} F_a(x, k_T^2; \mu, \zeta)$$

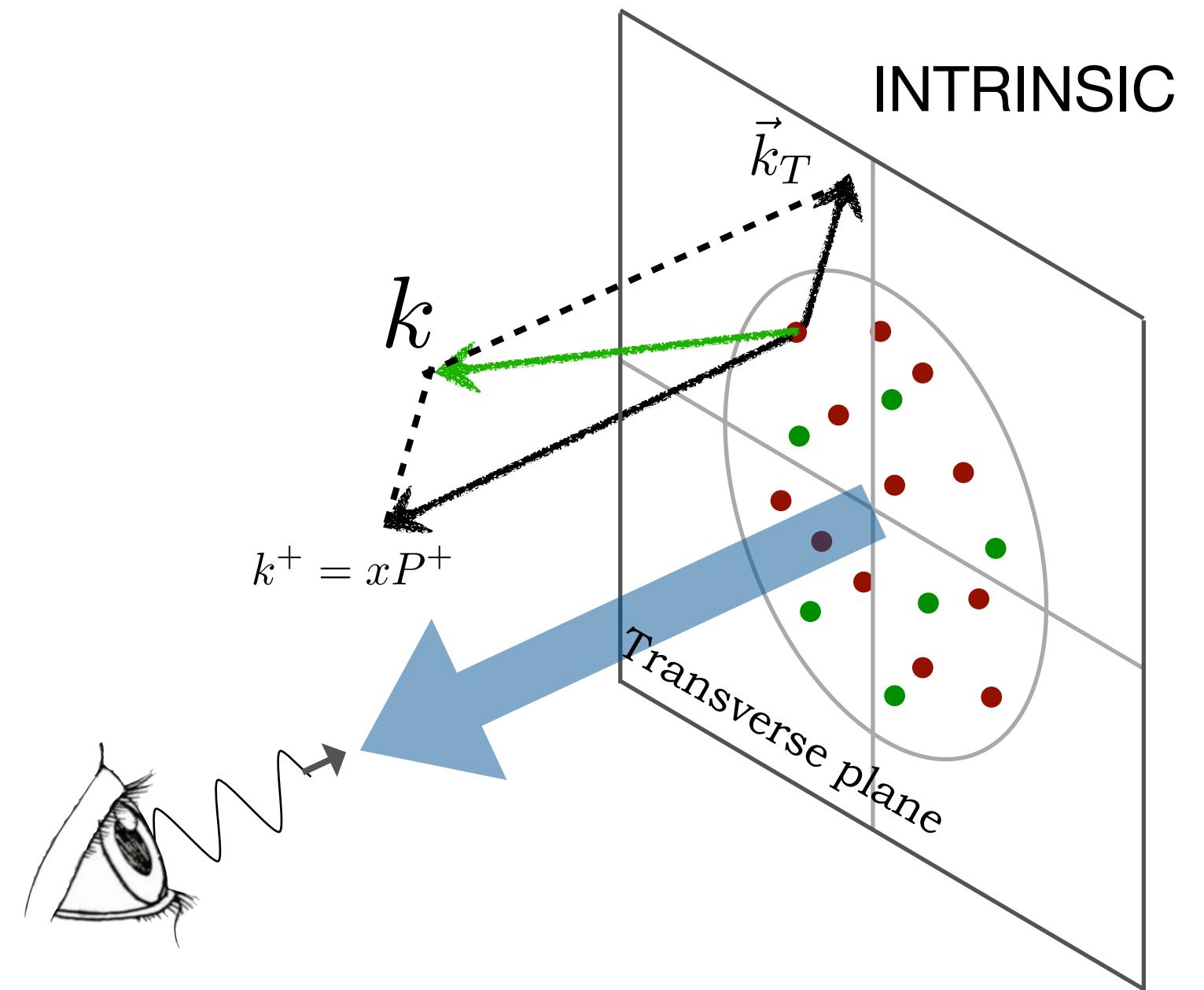


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How to express a TMD distribution?



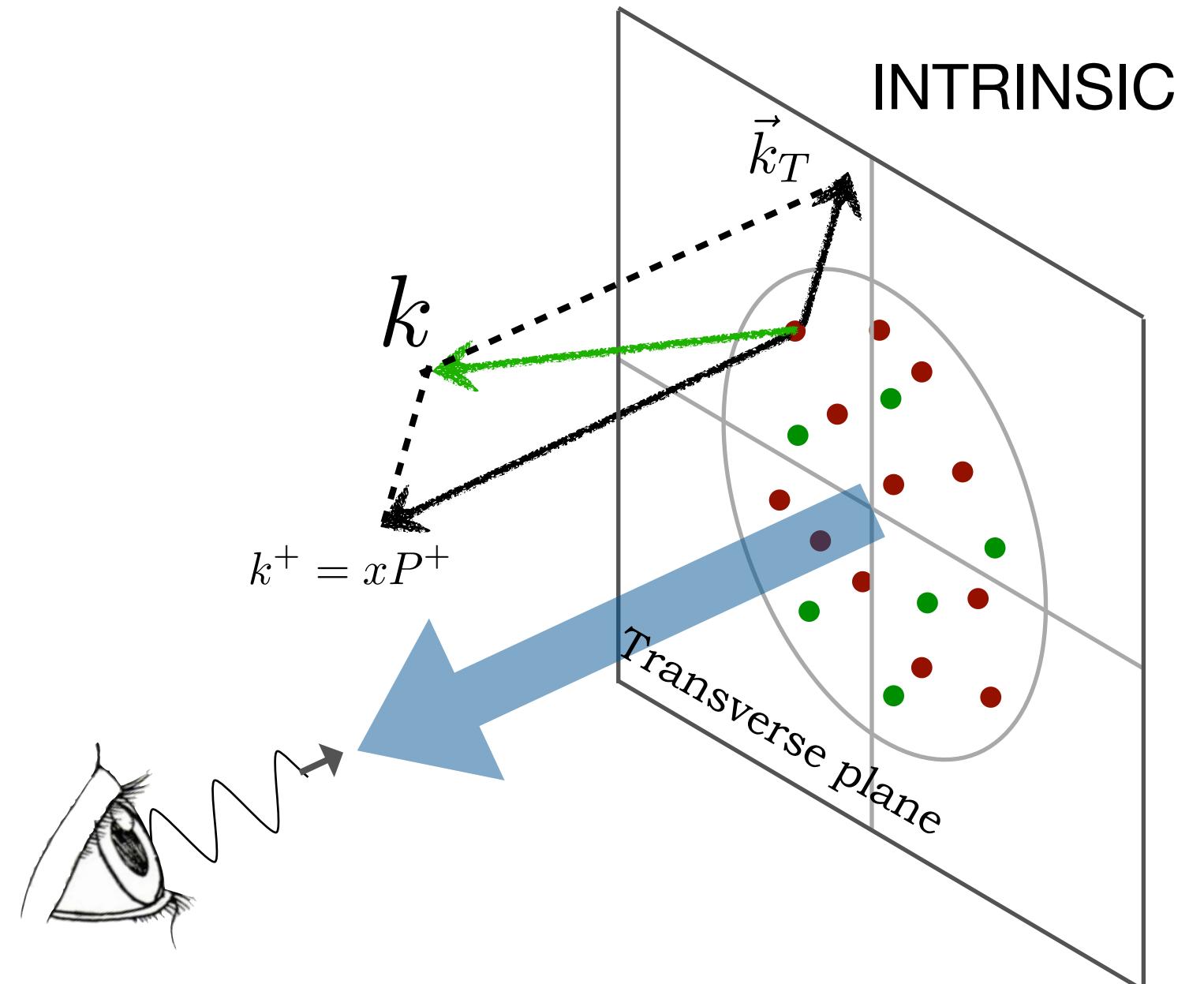
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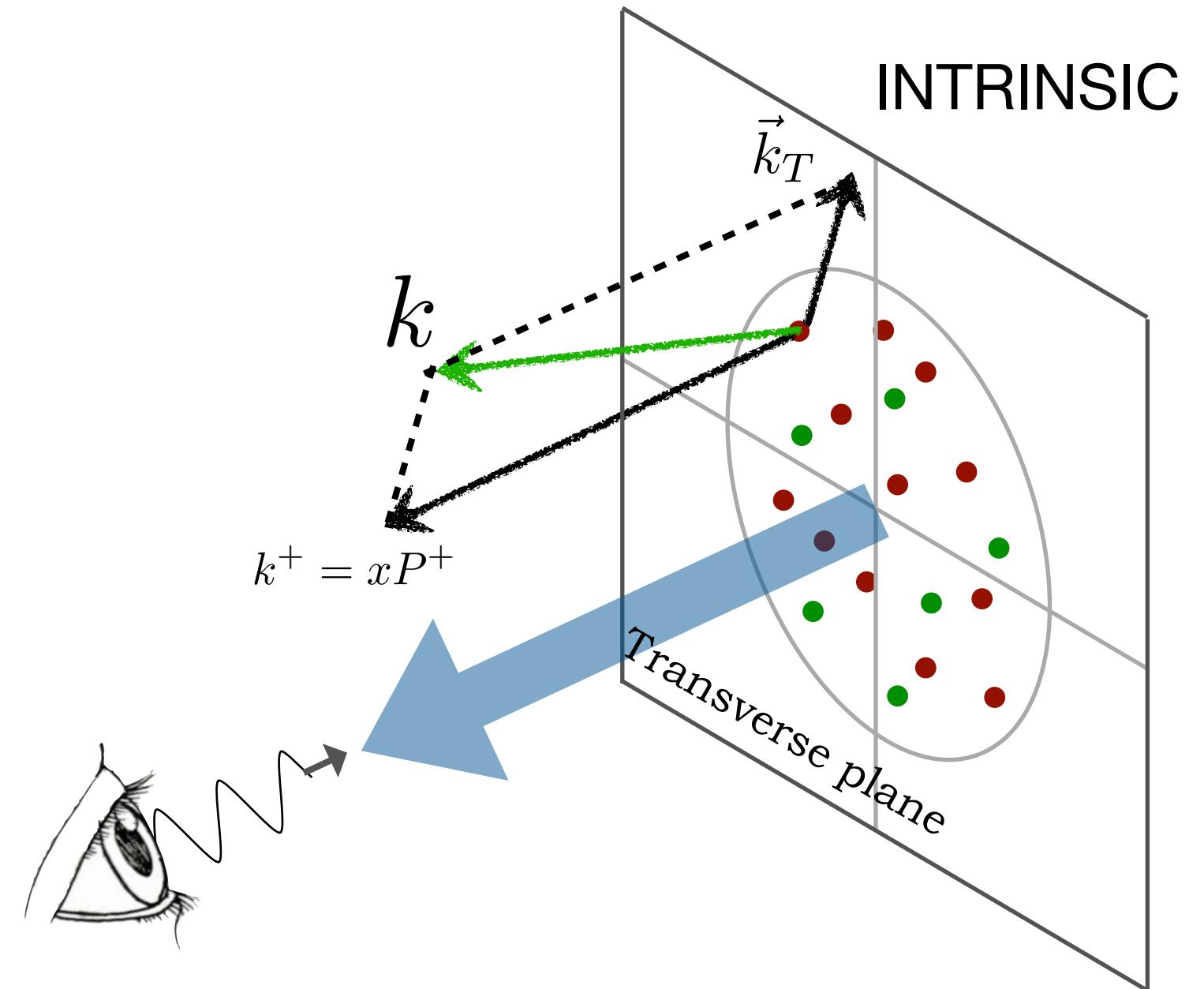
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Evolution  
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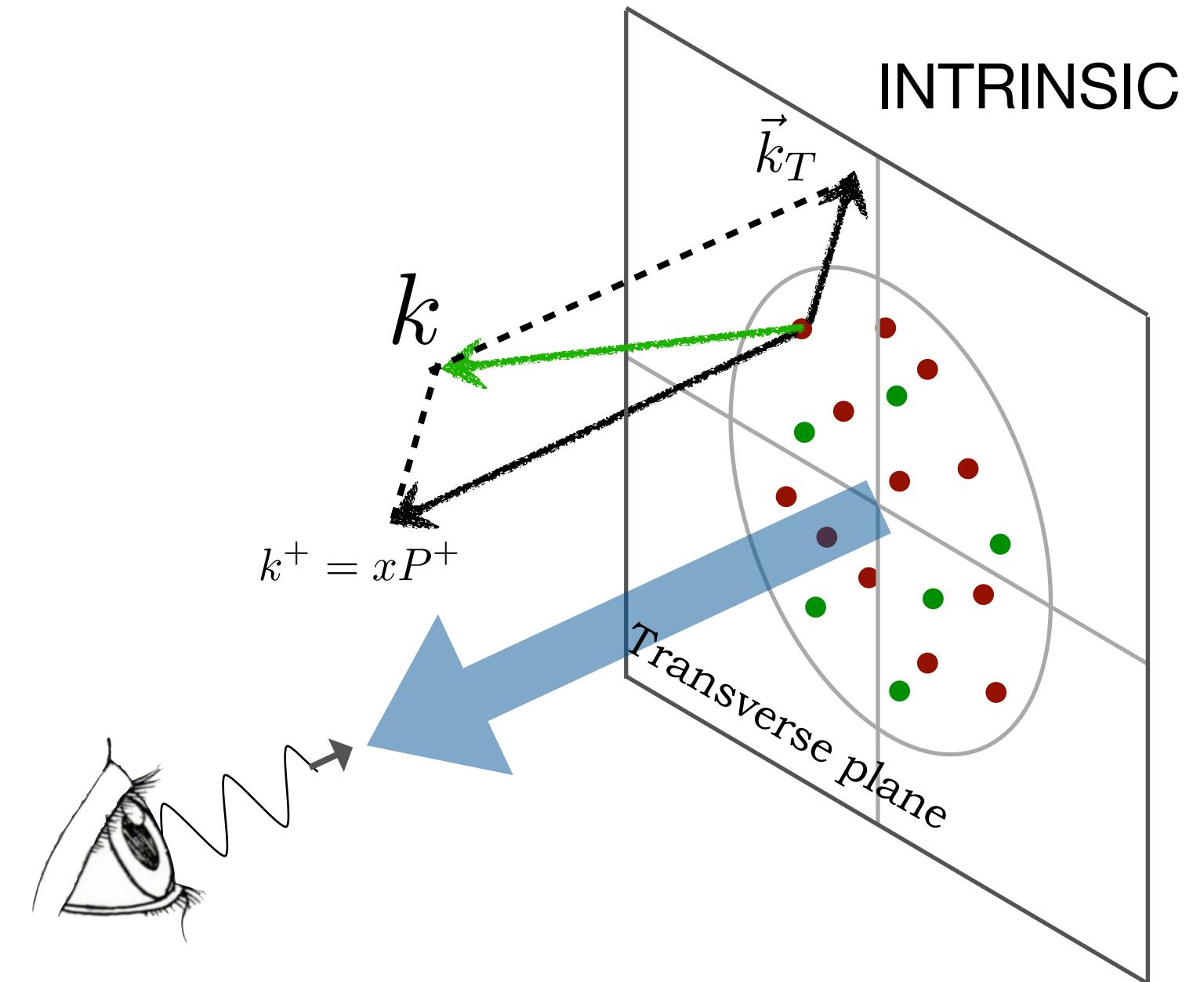
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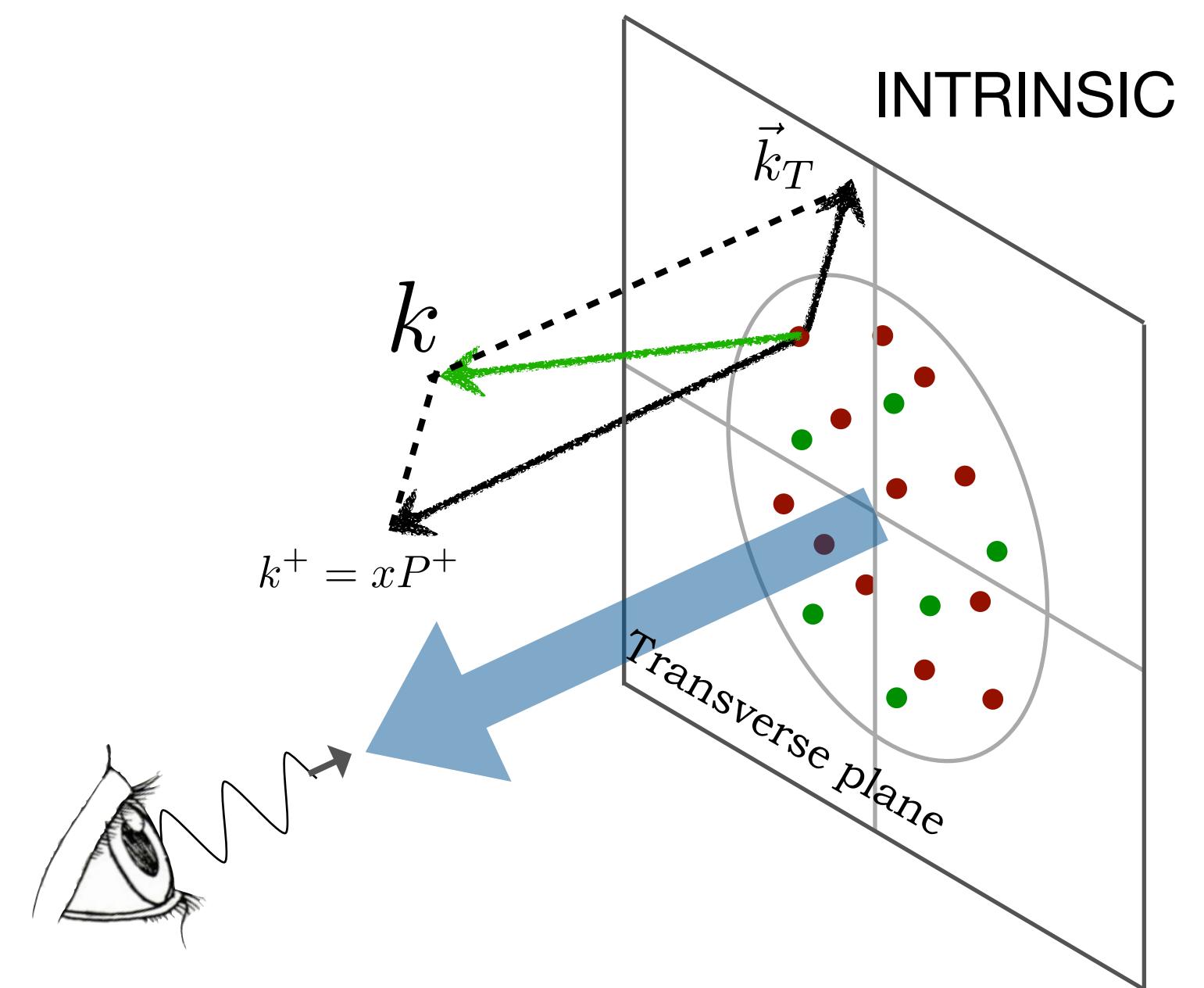
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Evolution  
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TMD part of the parton density  
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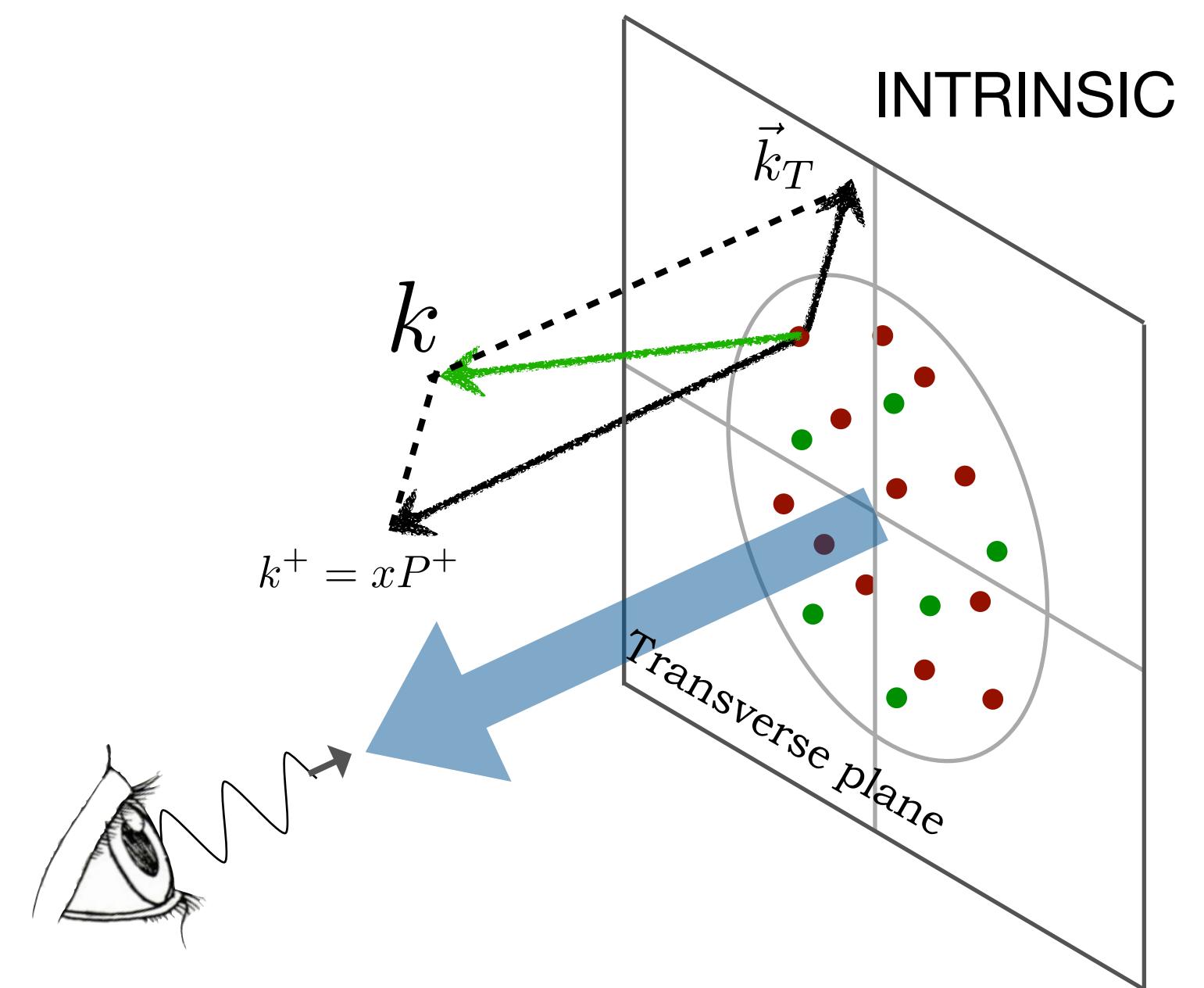
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# Accuracy of calculation

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Accuracy

Sudakov form factor

Matching coefficient

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Accuracy

Sudakov form factor

LL

$$\alpha_S^n \ln^{2n} \left( \frac{Q^2}{\mu_b^2} \right)$$

Matching coefficient

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NLL

$$\alpha_S^n \ln^{2n} \left( \frac{Q^2}{\mu_b^2} \right), \quad \alpha_S^n \ln^{2n-1} \left( \frac{Q^2}{\mu_b^2} \right)$$

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# Recent Global Analyses

	Accuracy	SIDIS	DY	Z production	N of points	$\chi^2/N_{\text{points}}$
Pavia 2017 arXiv:1703.10157	NLL	✓	✓	✓	8059	1.55
SV 2019 arXiv:1912.06532	$N^3LL^-$	✓	✓	✓	1039	1.06
Pavia 2019 arXiv:1912.07550	$N^3LL$	✗	✓	✓	353	1.02

# Our work in the last two years

New Global Fit

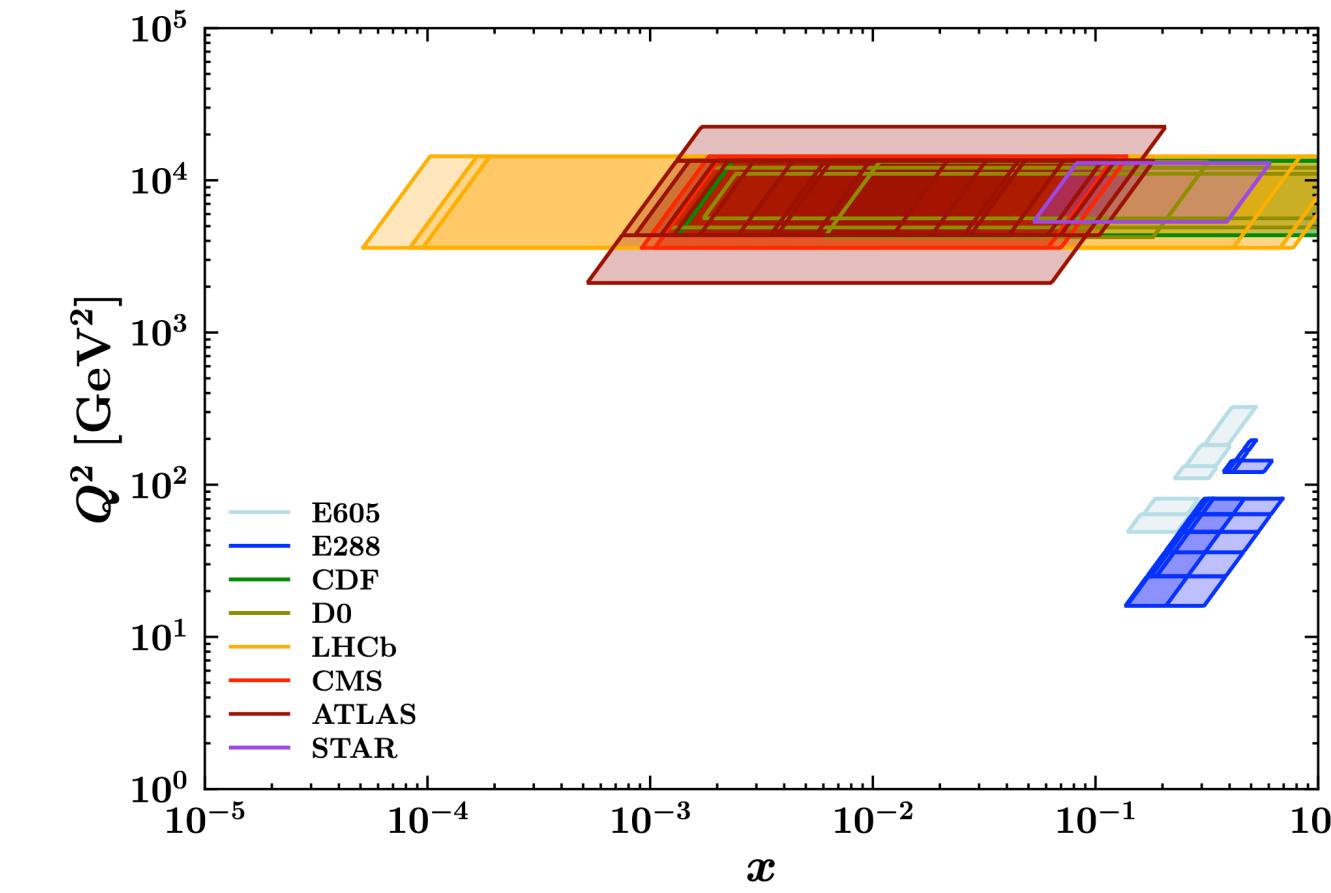
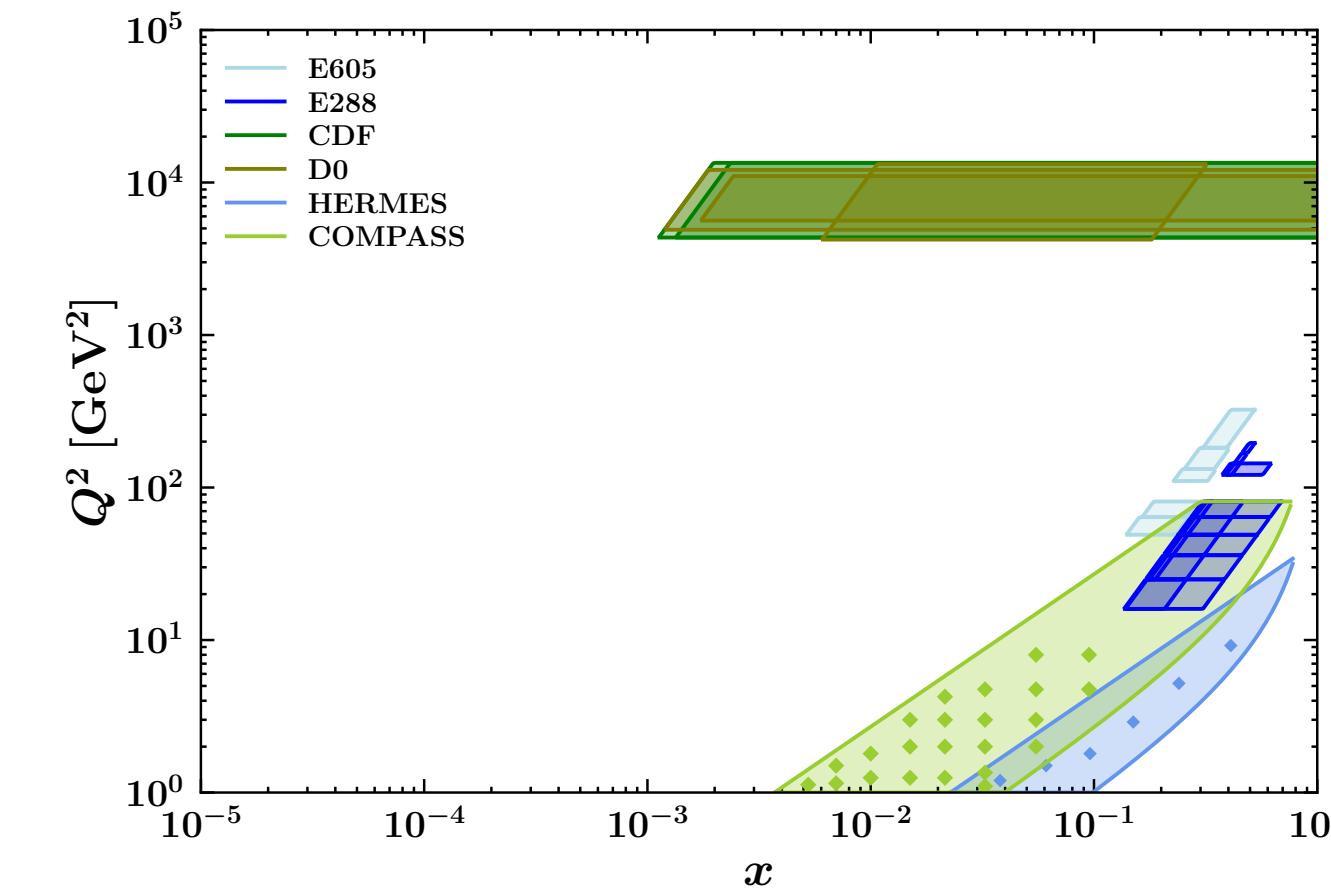
Simultaneously extraction of unpolarized TMD PDFs and FFs

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New Global Fit

✓ SIDIS + Drell Yan

Simultaneously extraction of unpolarized TMD PDFs and FFs



# Our work in the last two years

New Global Fit

✓ SIDIS + Drell Yan

✓ Integrated variables

Simultaneously extraction of unpolarized TMD PDFs and FFs



## Nanga Parbat: a TMD fitting framework

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/vbertone/NangaParbat/releases>

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

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

If you instead want to download a specific tag:

<https://github.com/MapCollaboration>

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New Global Fit

✓ SIDIS + Drell Yan

✓ Integrated variables

○ Up to  $N^2LL/N^3LL$

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# A new Global Fit: MAPTMD22

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Pavia 2017 <a href="https://arxiv.org/abs/1703.10157">arXiv:1703.10157</a>	NLL	✓	✓	✓	8059	1.55
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Pavia 2019 <a href="https://arxiv.org/abs/1912.07550">arXiv:1912.07550</a>	$N^3LL$	✗	✓	✓	353	1.02
<b>MAPTMD22</b>	$N^3LL^-$	✓	✓	✓	<b>2031</b>	<b>1.06</b>

# MAPTMD22 – Included Dataset

Drell-Yan

Fixed-target low-energy DY

RHIC data

LHC and Tevatron data

$9 \lesssim Q \lesssim 11$  GeV excluded ( $\Upsilon$  resonance)

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

484 experimental points

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SIDIS

HERMES data

COMPASS data

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

$$0.2 < z < 0.7$$

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

**484 experimental points**

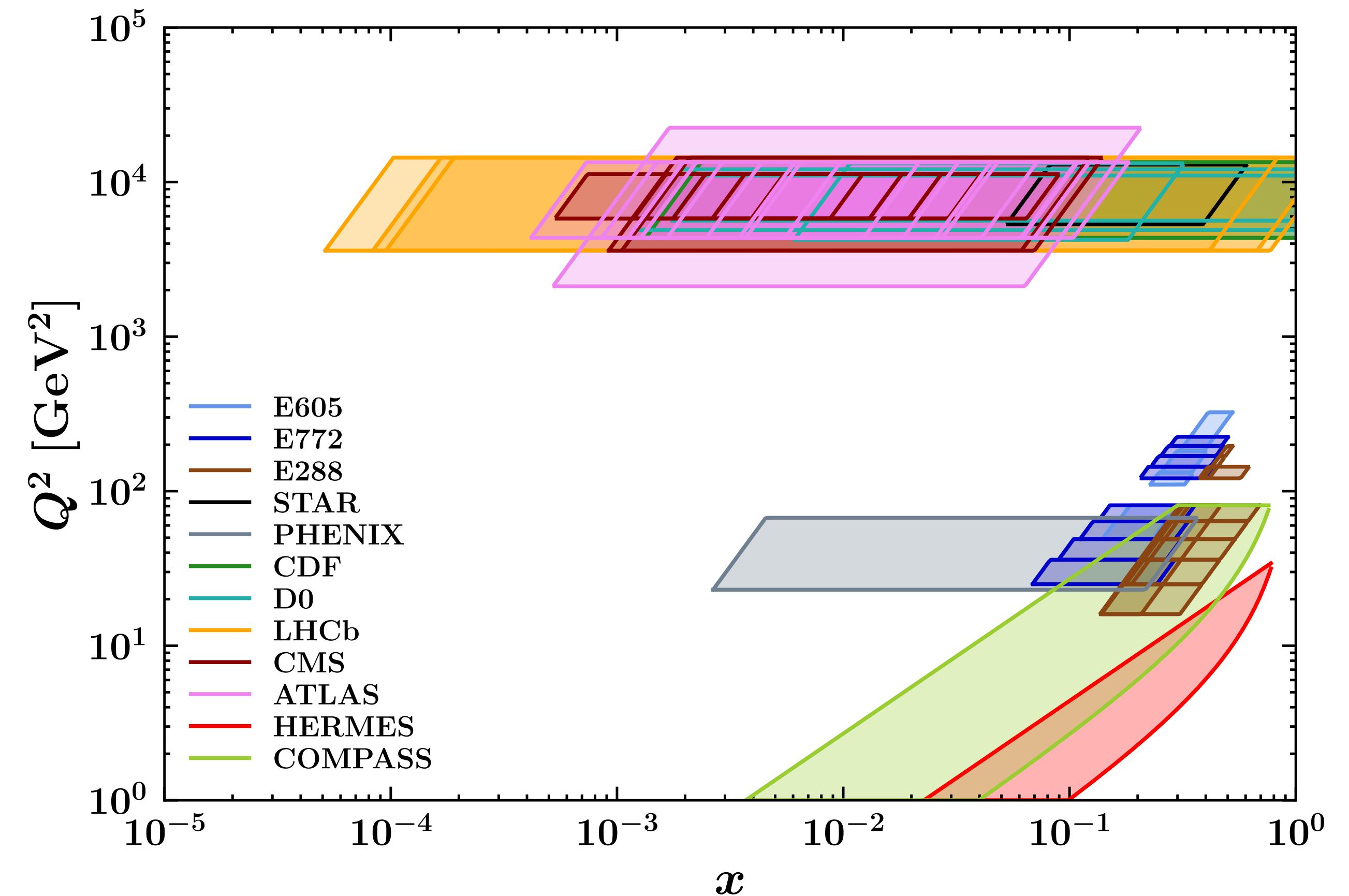
**1547 experimental points**

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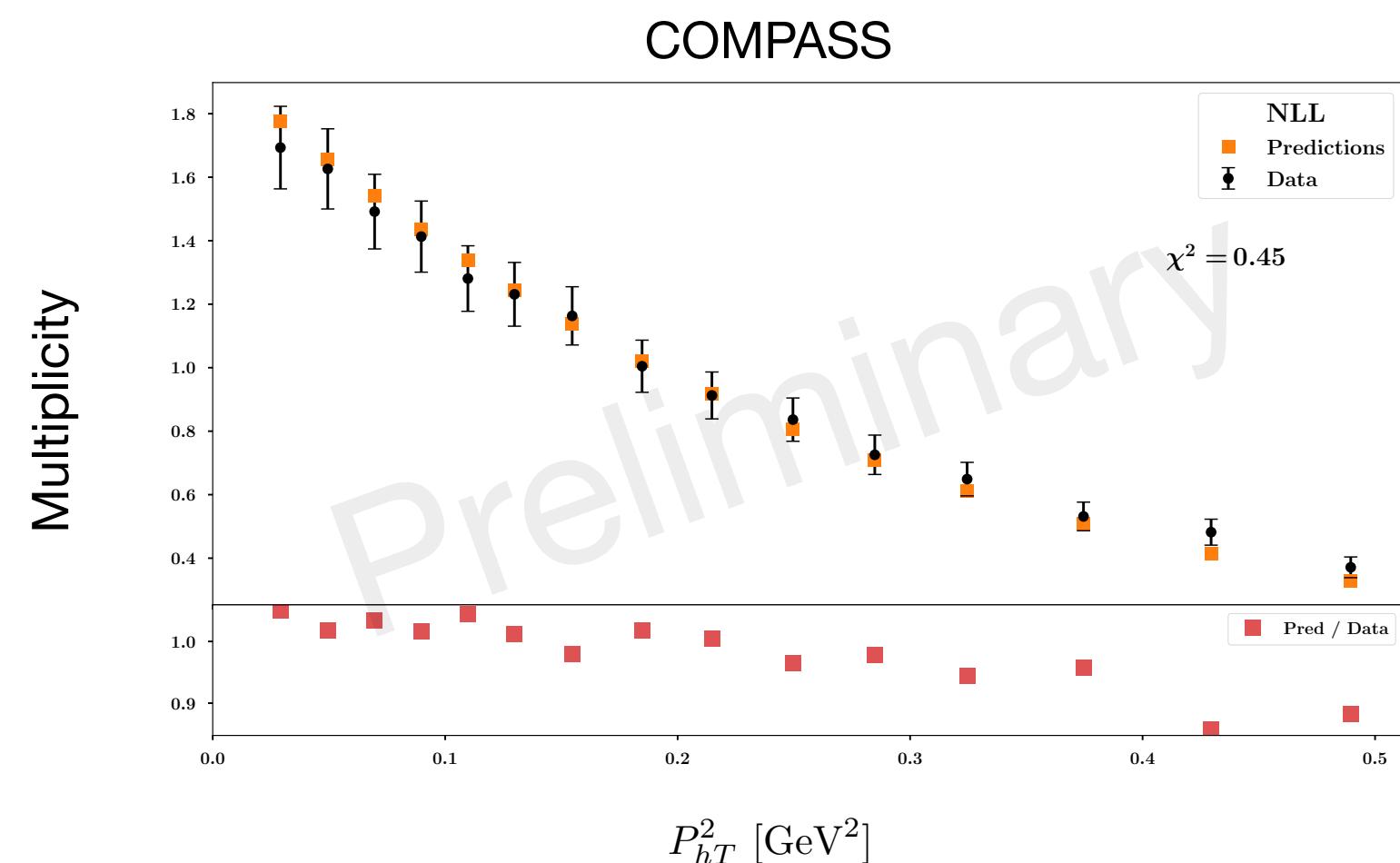
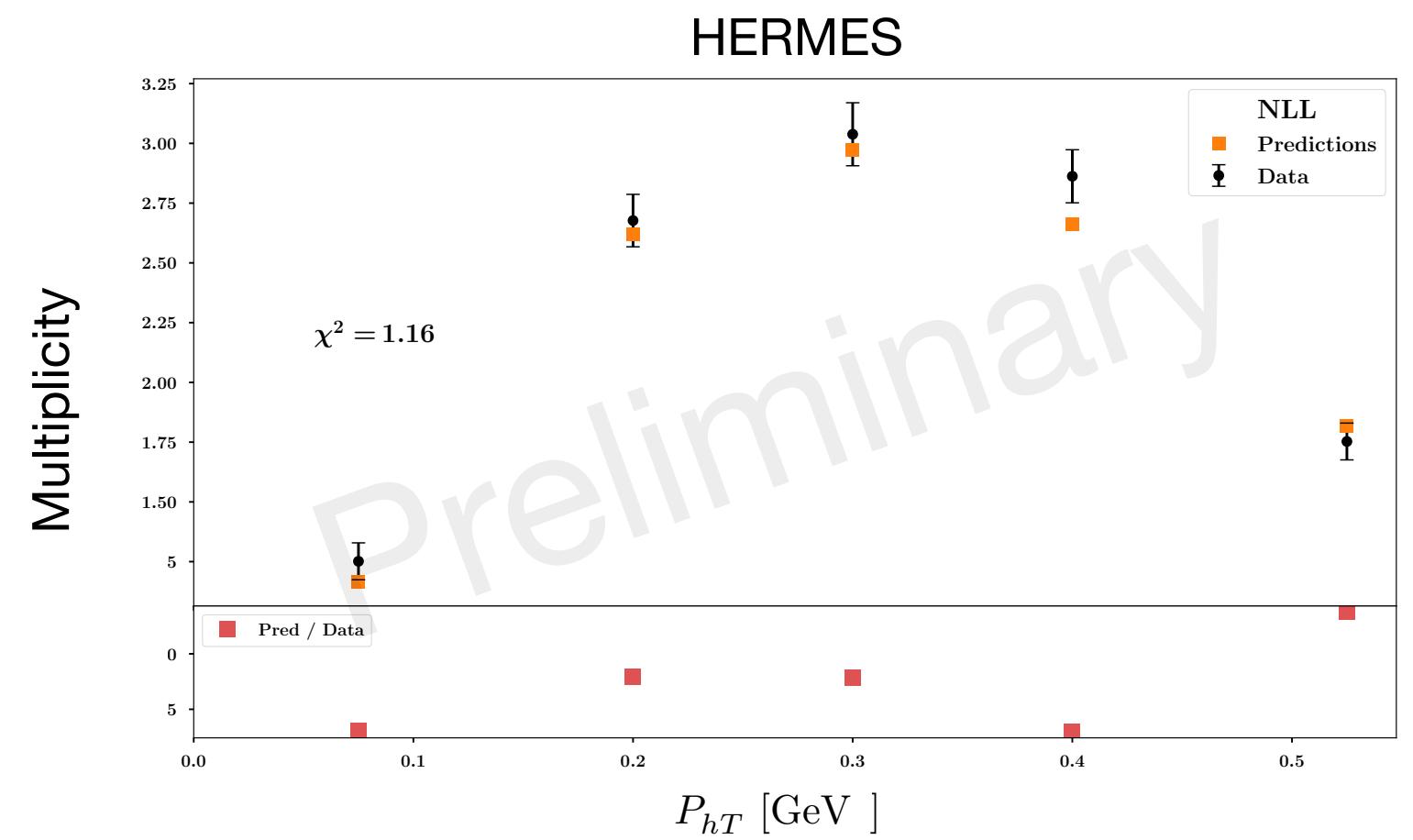
1547 experimental points

Total: 2031 fitted experimental points



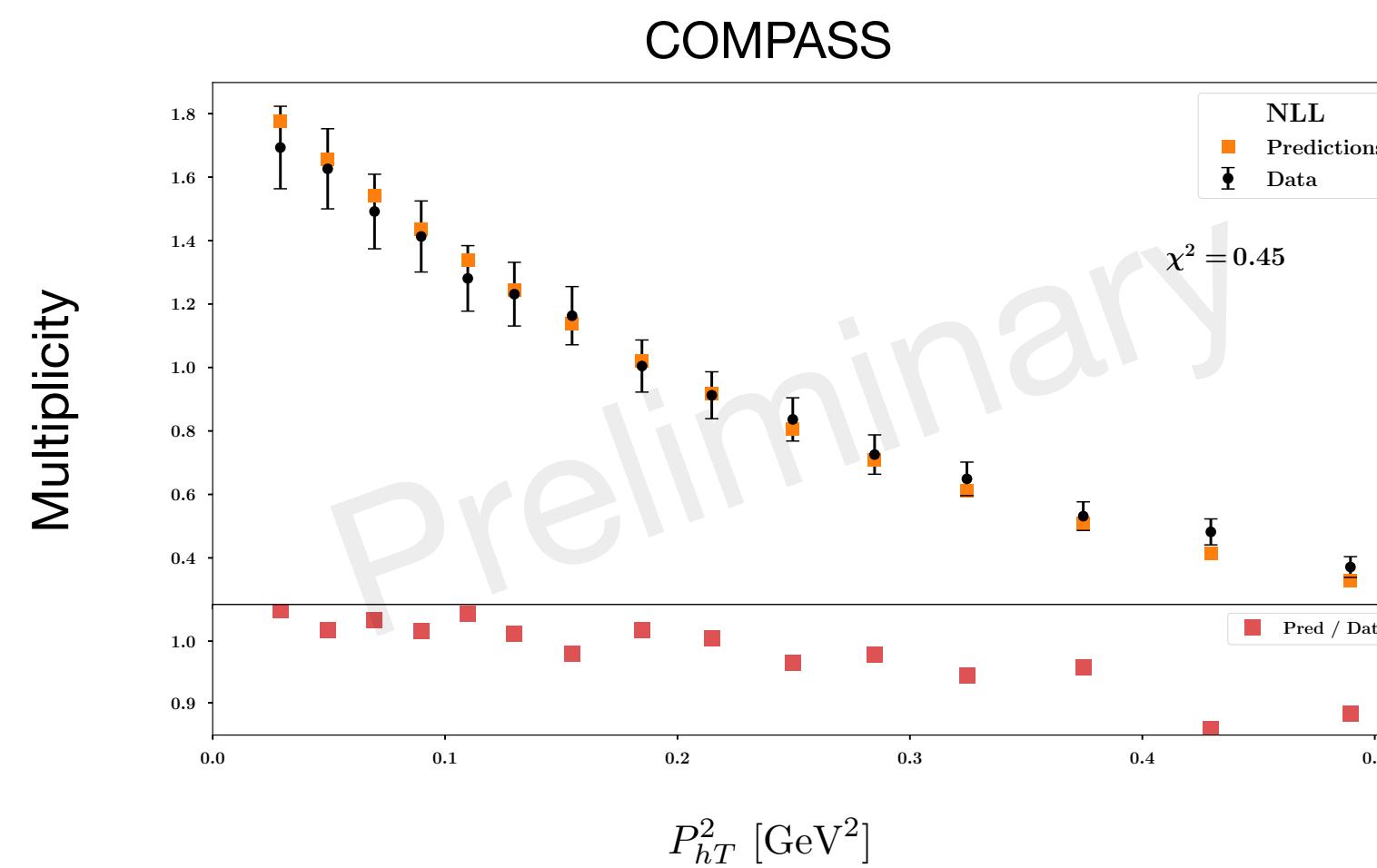
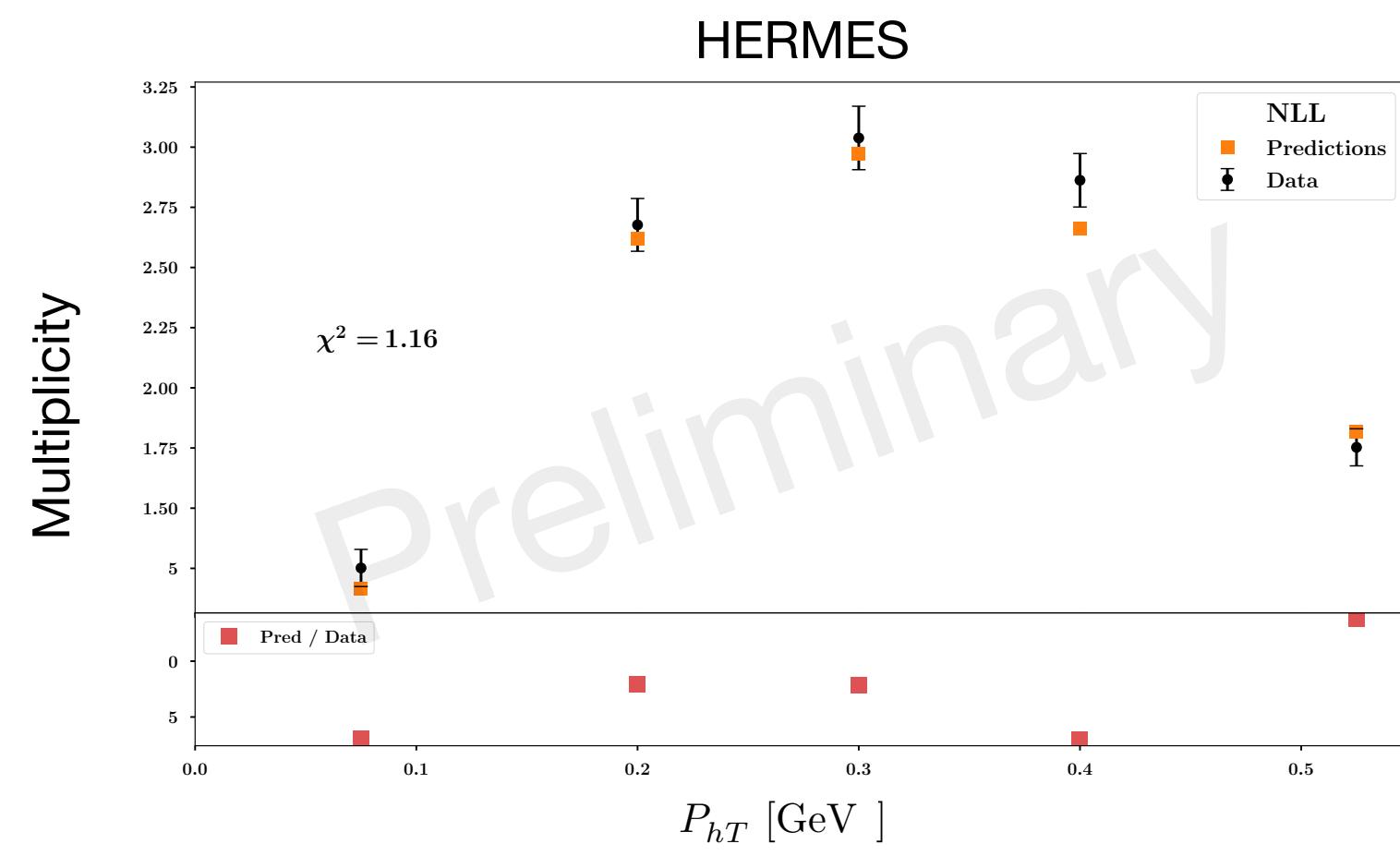
# MAPTMD22 – Normalization of SIDIS

## SIDIS multiplicities at NLL

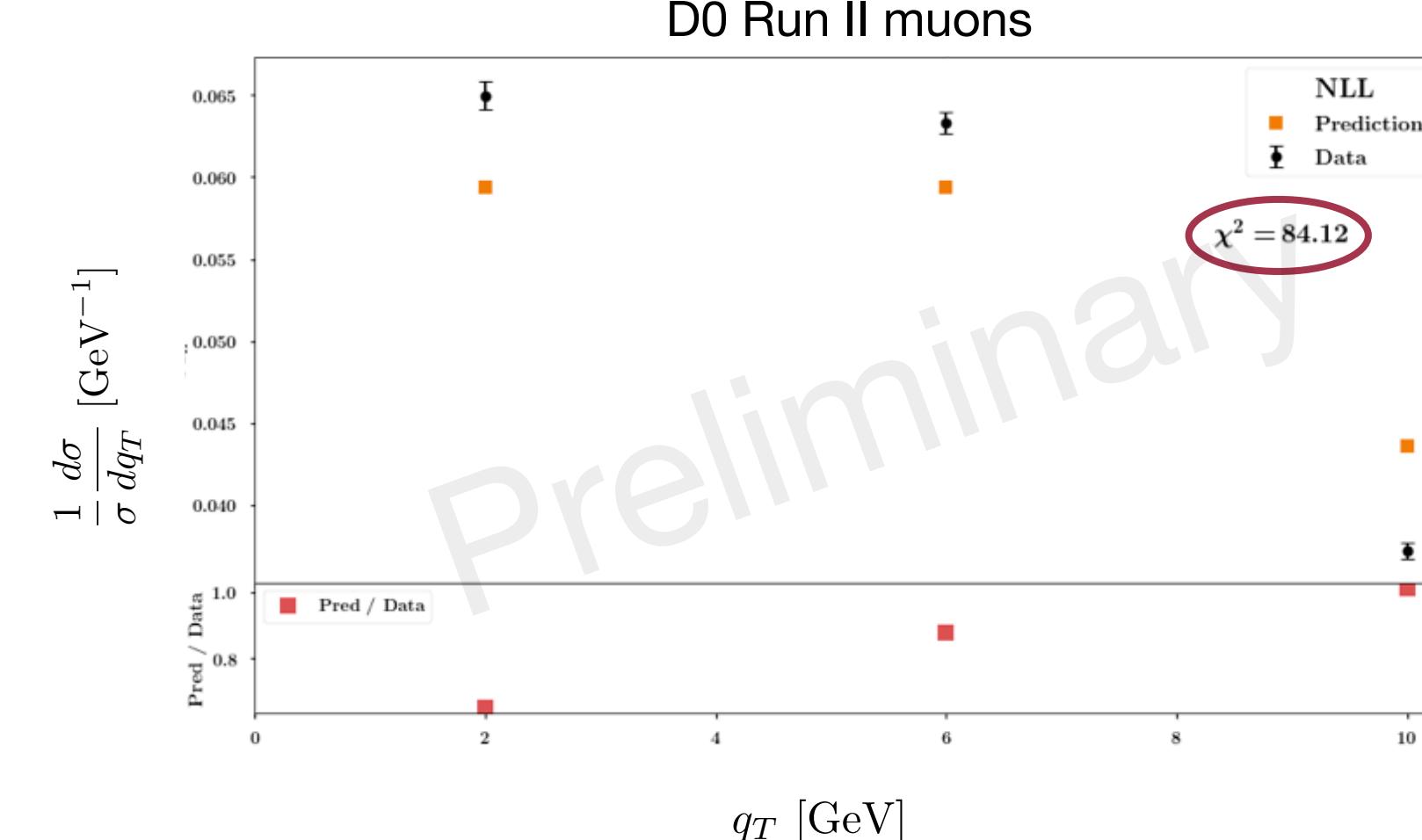
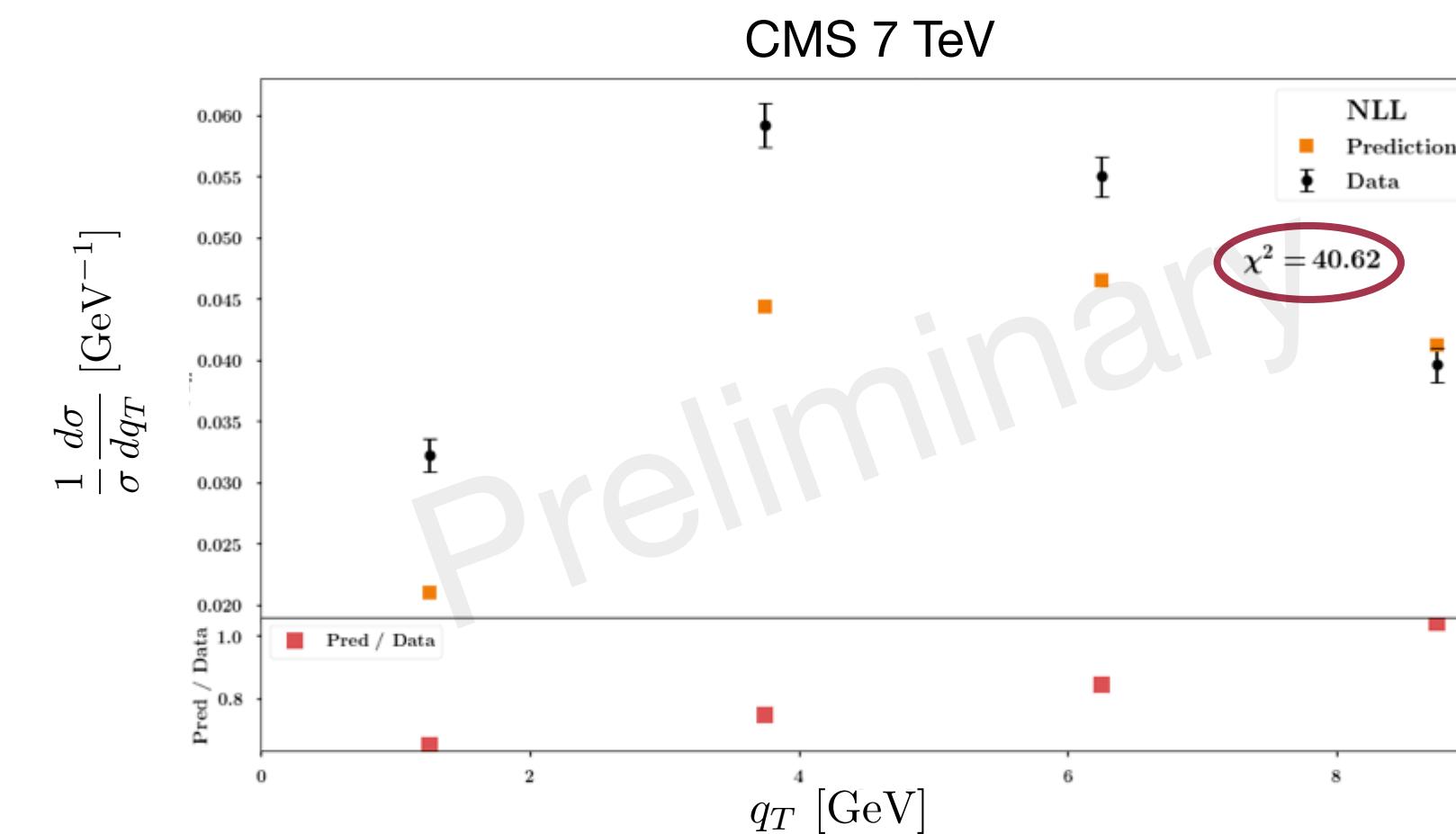


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## SIDIS multiplicities at NLL



## High-Energy Drell-Yan at NLL

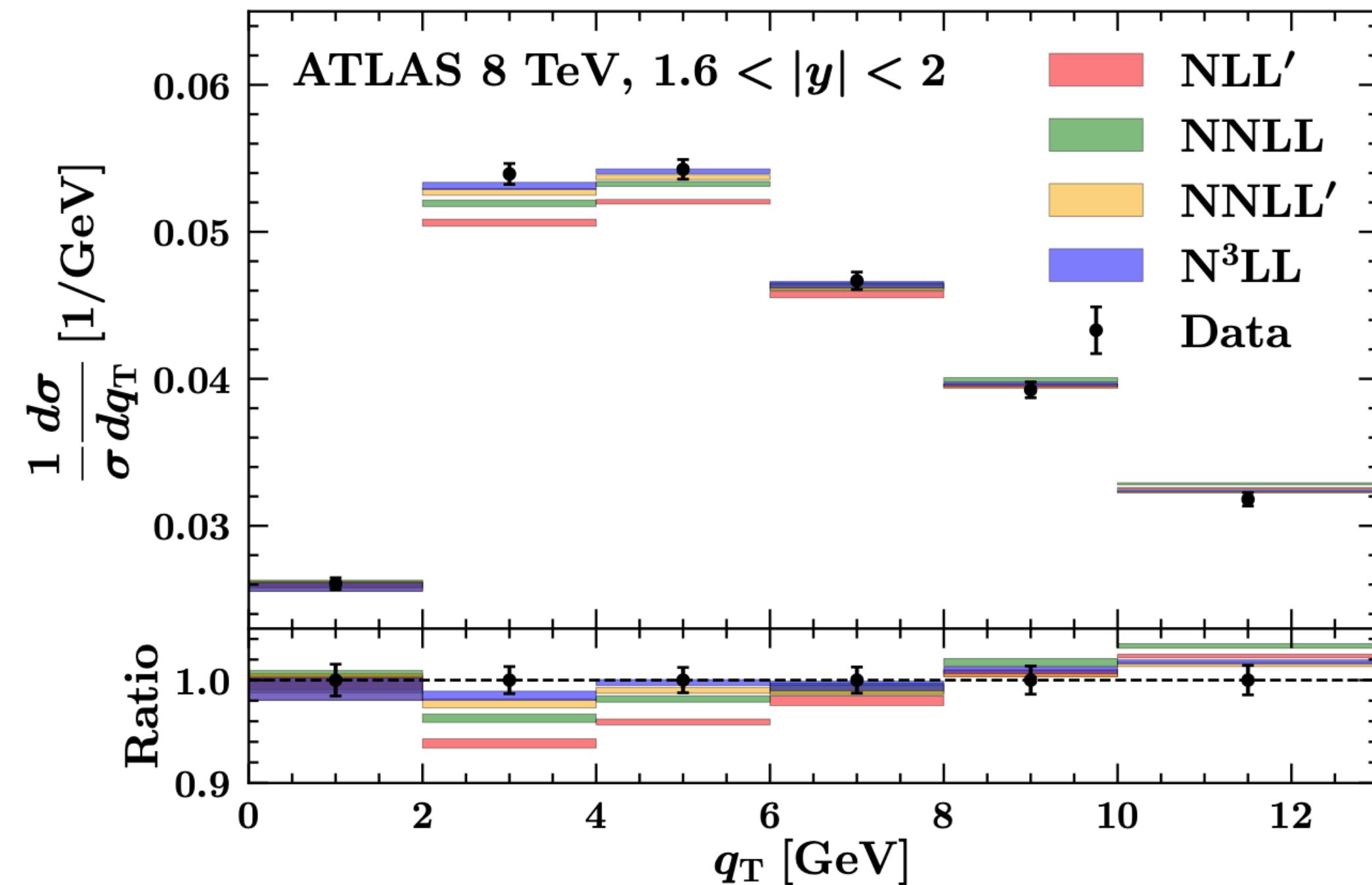


# MAPTMD22 – Normalization of SIDIS

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High-Energy Drell-Yan beyond NLL

$$Q \sim 100 \text{ GeV}$$

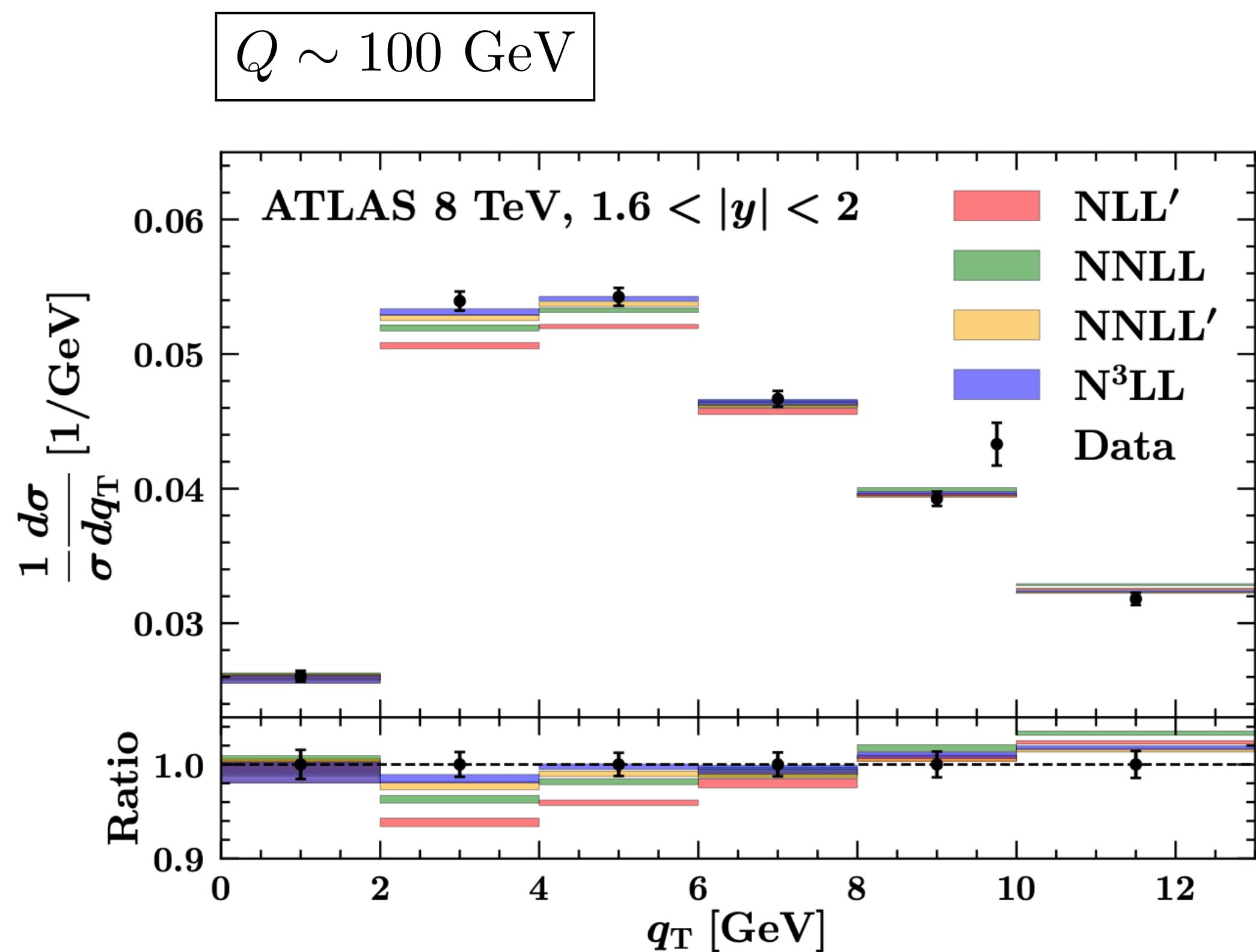


Bacchetta, Bertone, Bissolotti, Bozzi, Delcarro, Piacenza, Radici, [arXiv:1912.07550](https://arxiv.org/abs/1912.07550)

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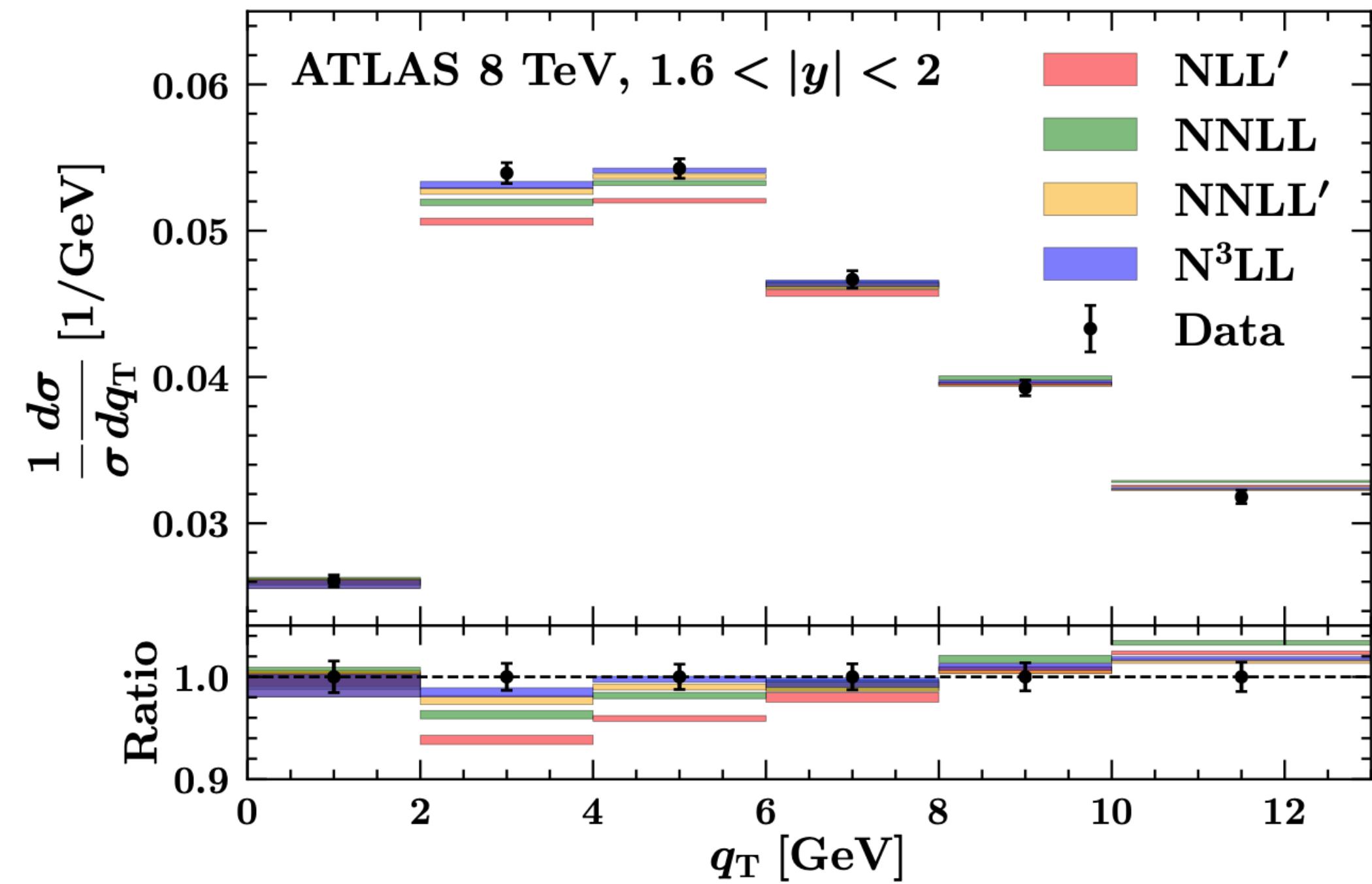
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SIDIS multiplicities beyond NLL

$$Q \sim 2 \text{ GeV}$$

High-Energy Drell-Yan beyond NLL

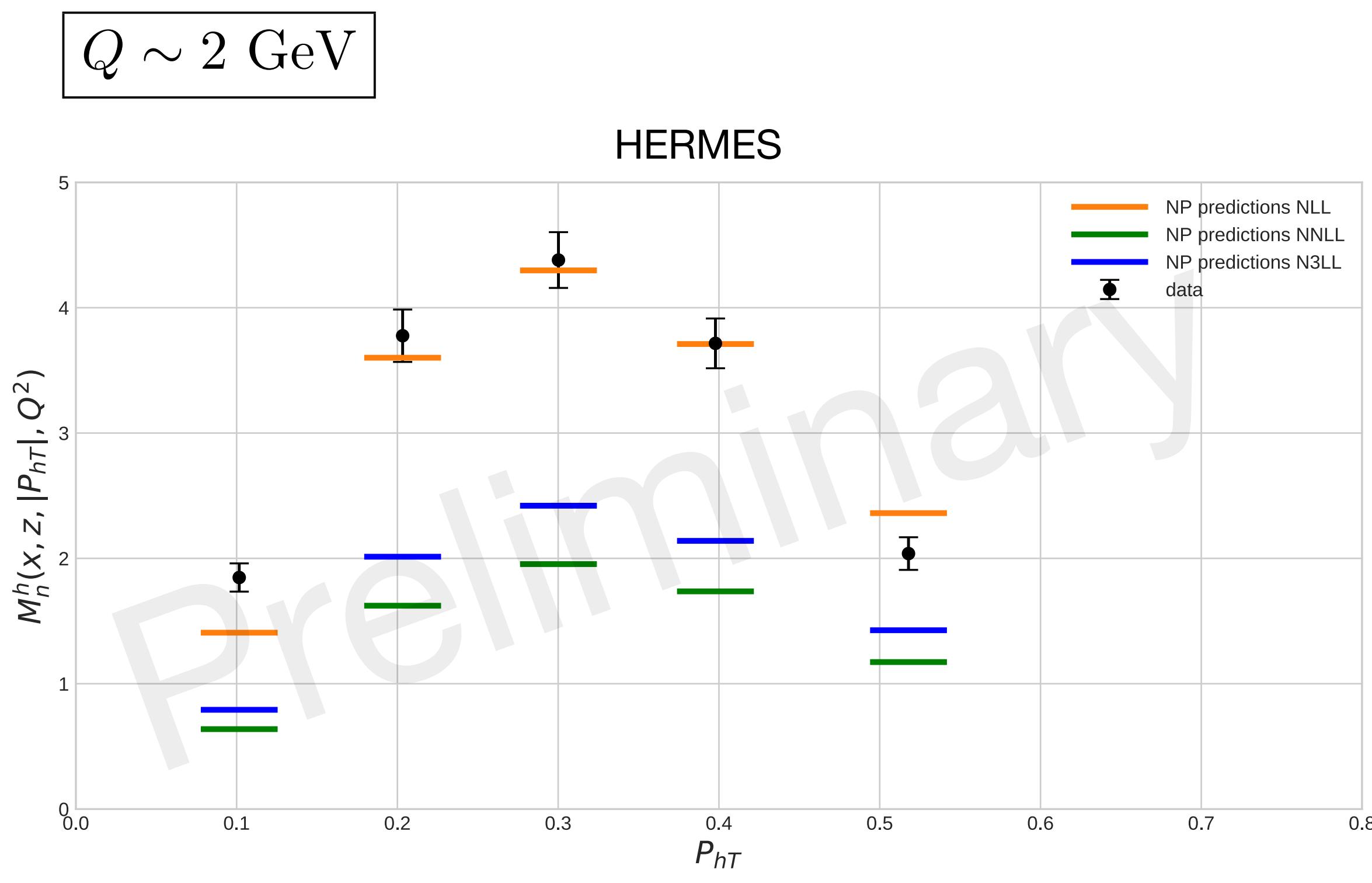
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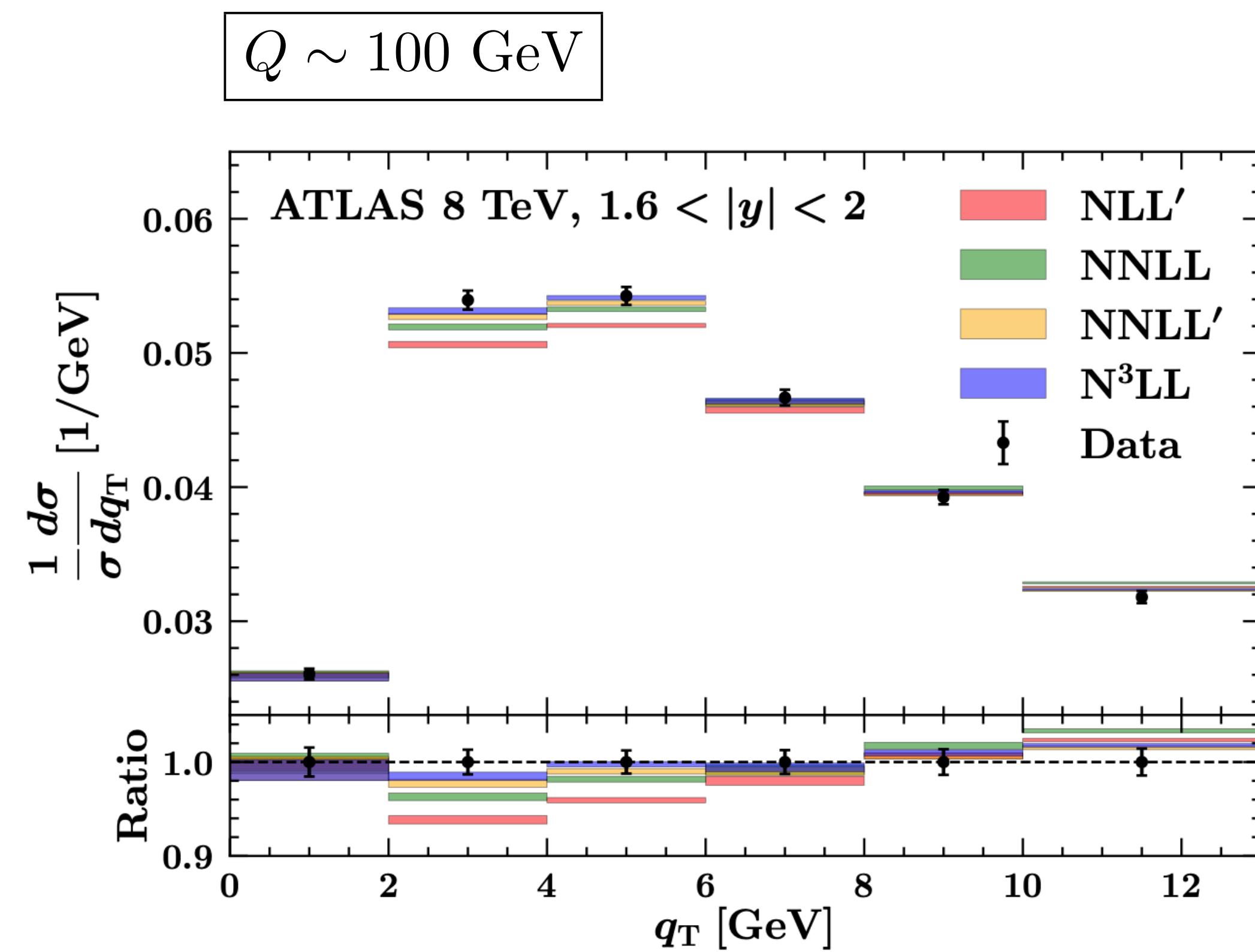
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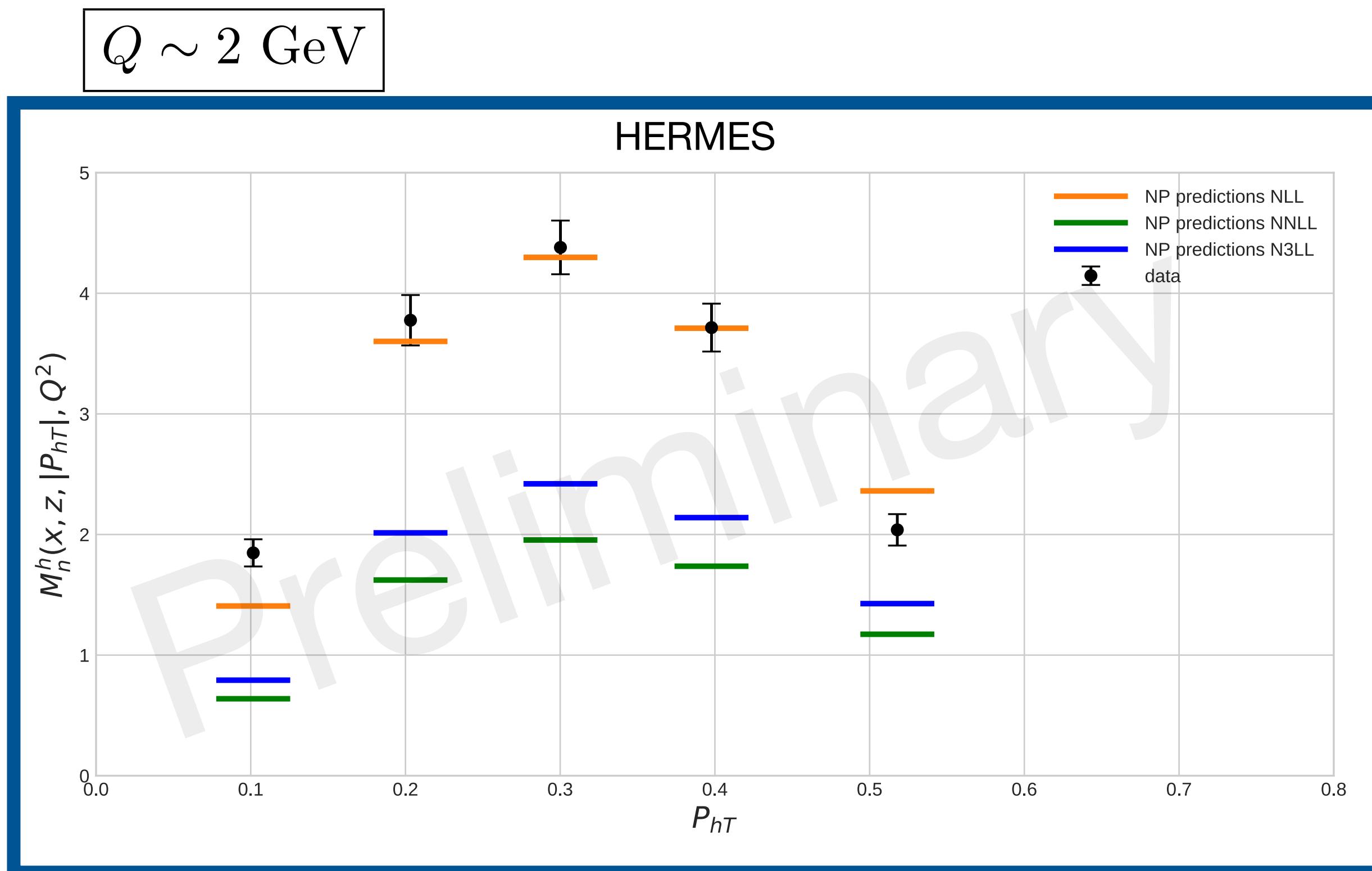
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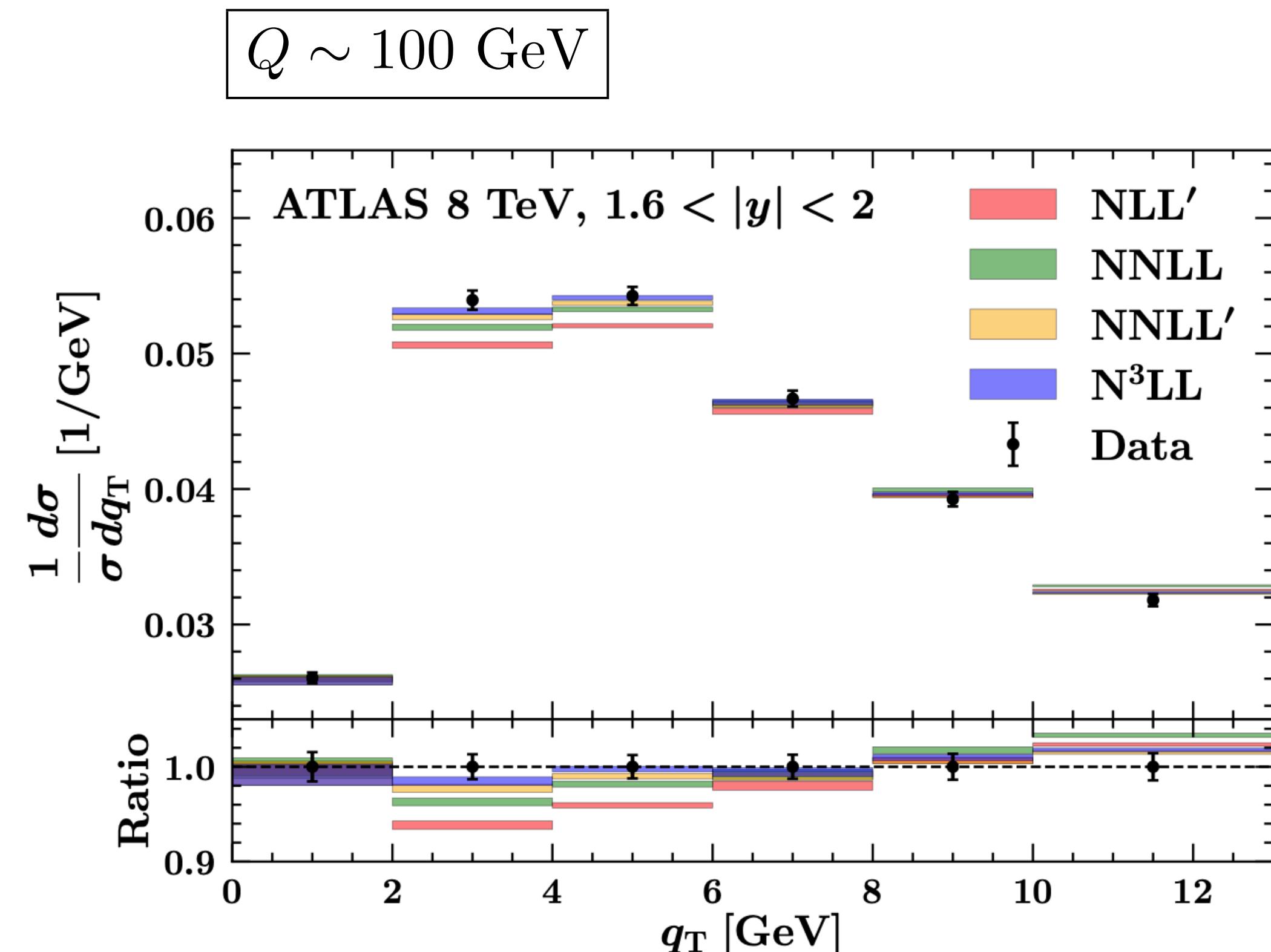
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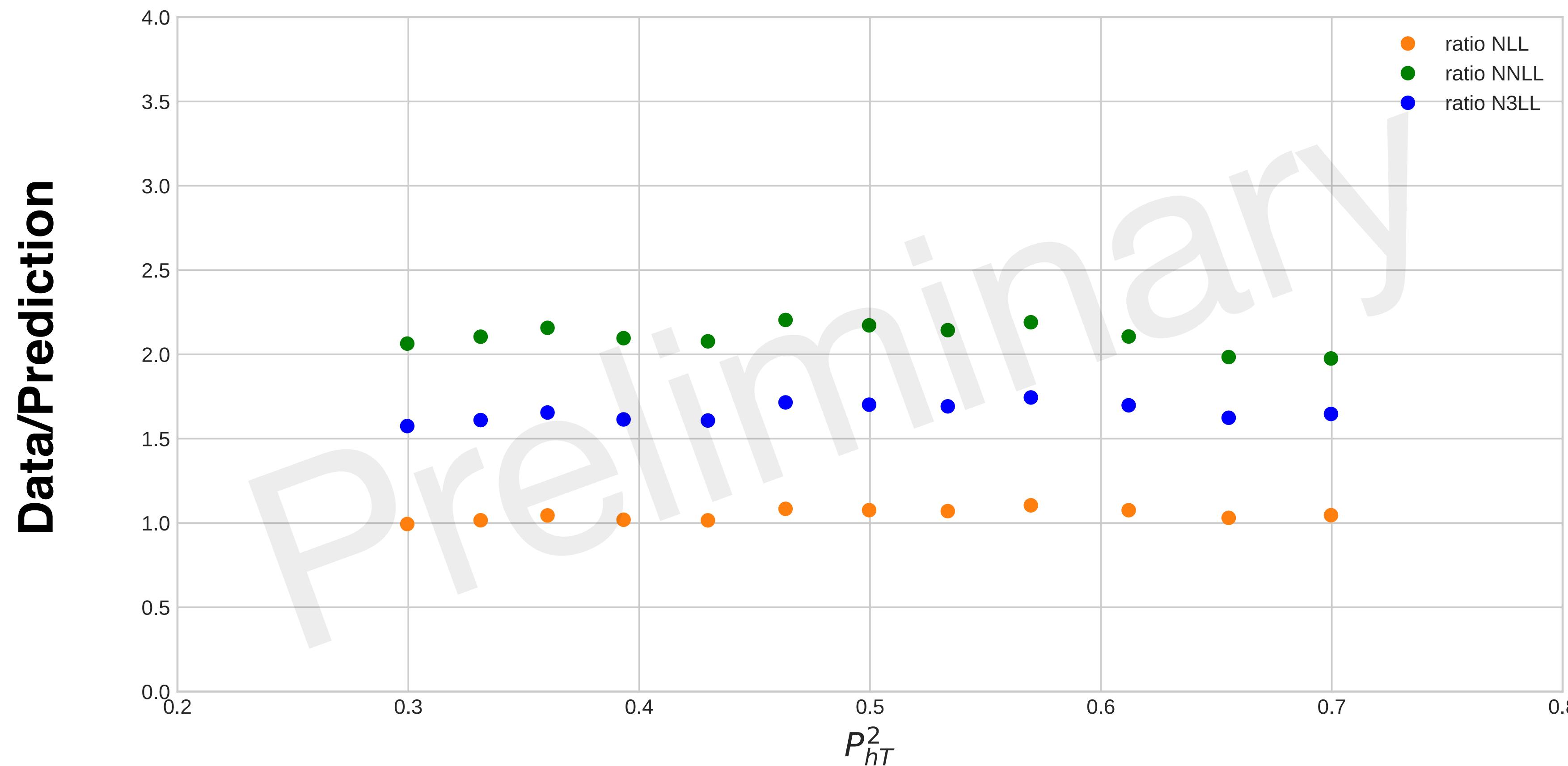
The description considerably worsens at higher orders!!

Bacchetta, Bertone, Bissolotti, Bozzi, Delcarro, Piacenza, Radici, [arXiv:1912.07550](https://arxiv.org/abs/1912.07550)

# MAPTMD22 – Normalization of SIDIS

COMPASS multiplicities (one of many bins)

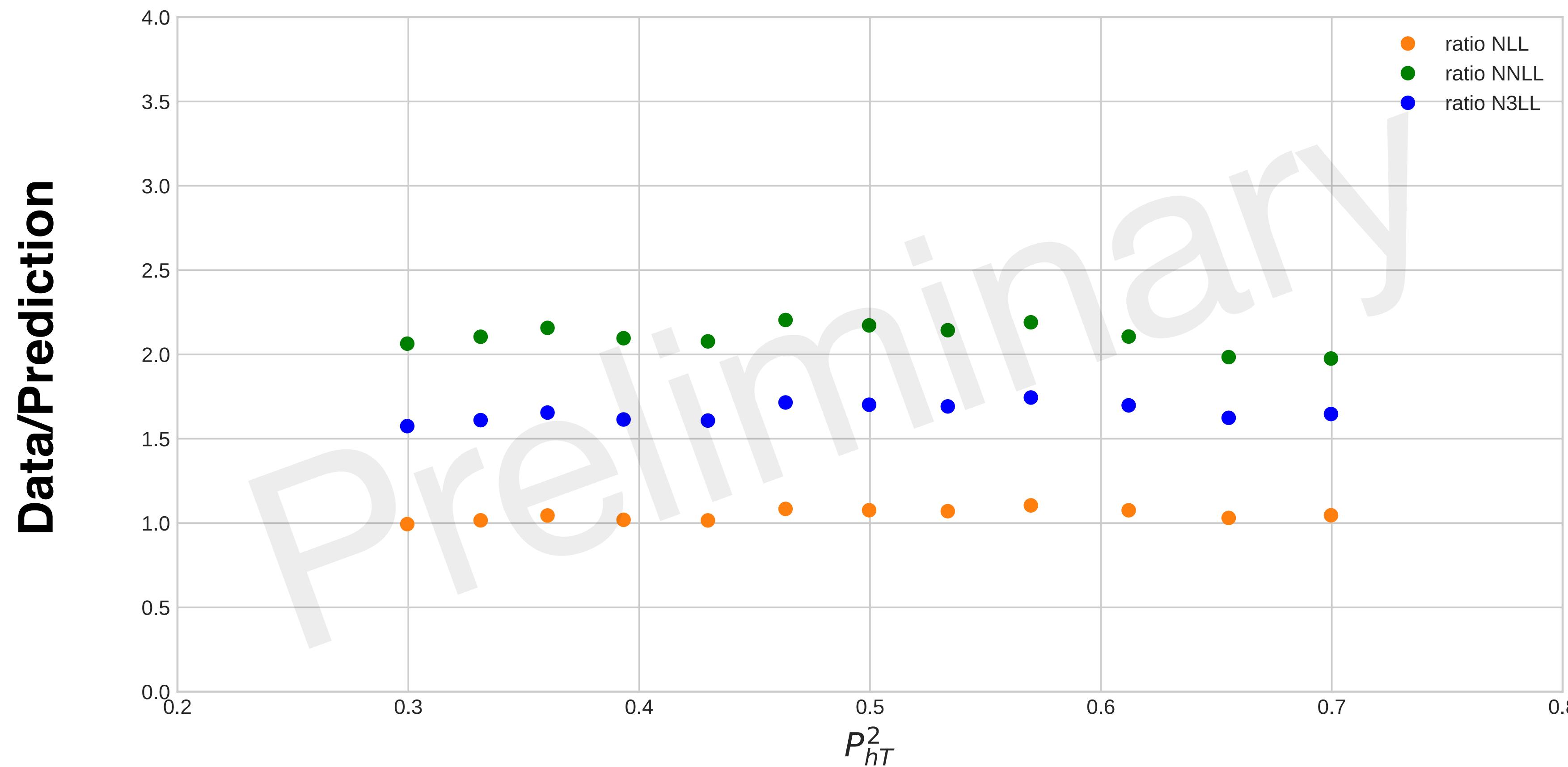
J.O. Gonzalez-Hernandez, PoS DIS2019 (2019) 176



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

# MAPTMD22 – Normalization of SIDIS

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Collinear SIDIS cross section

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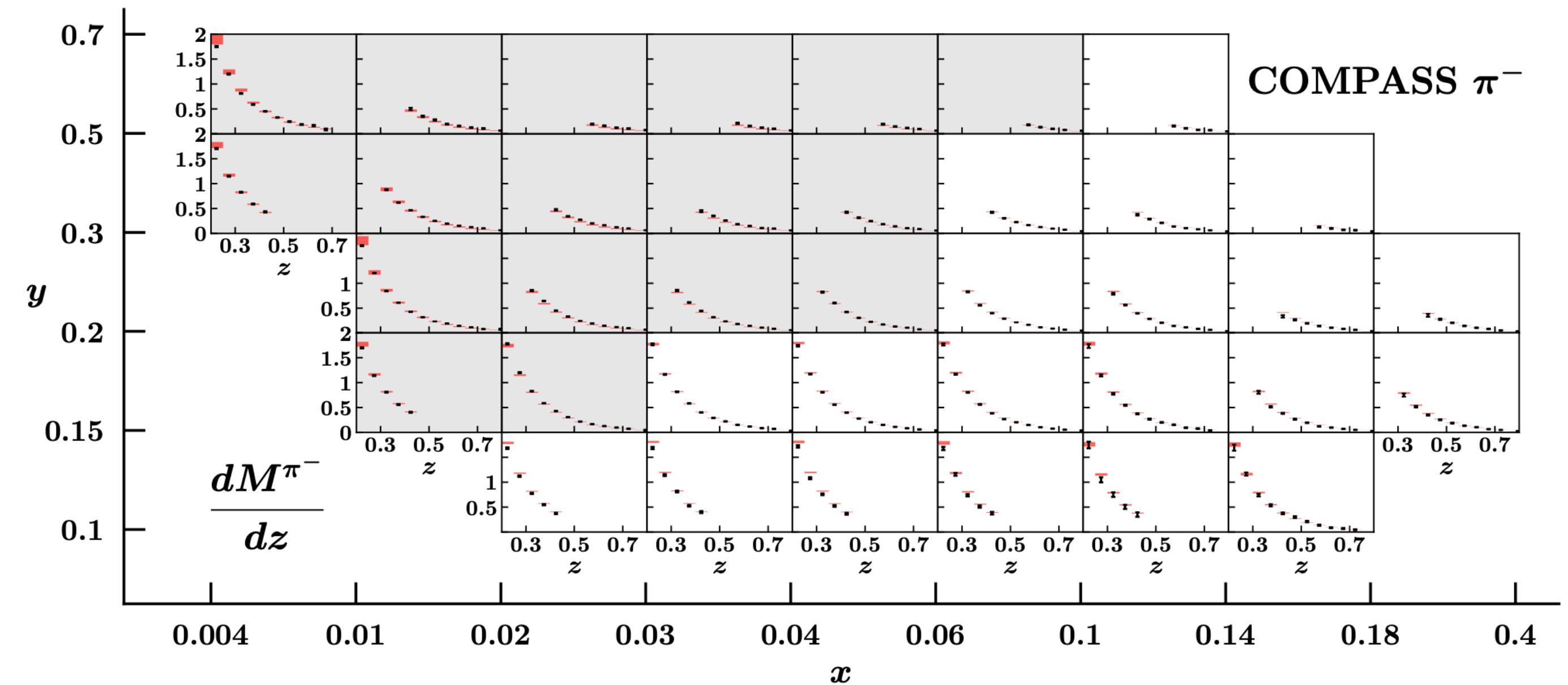
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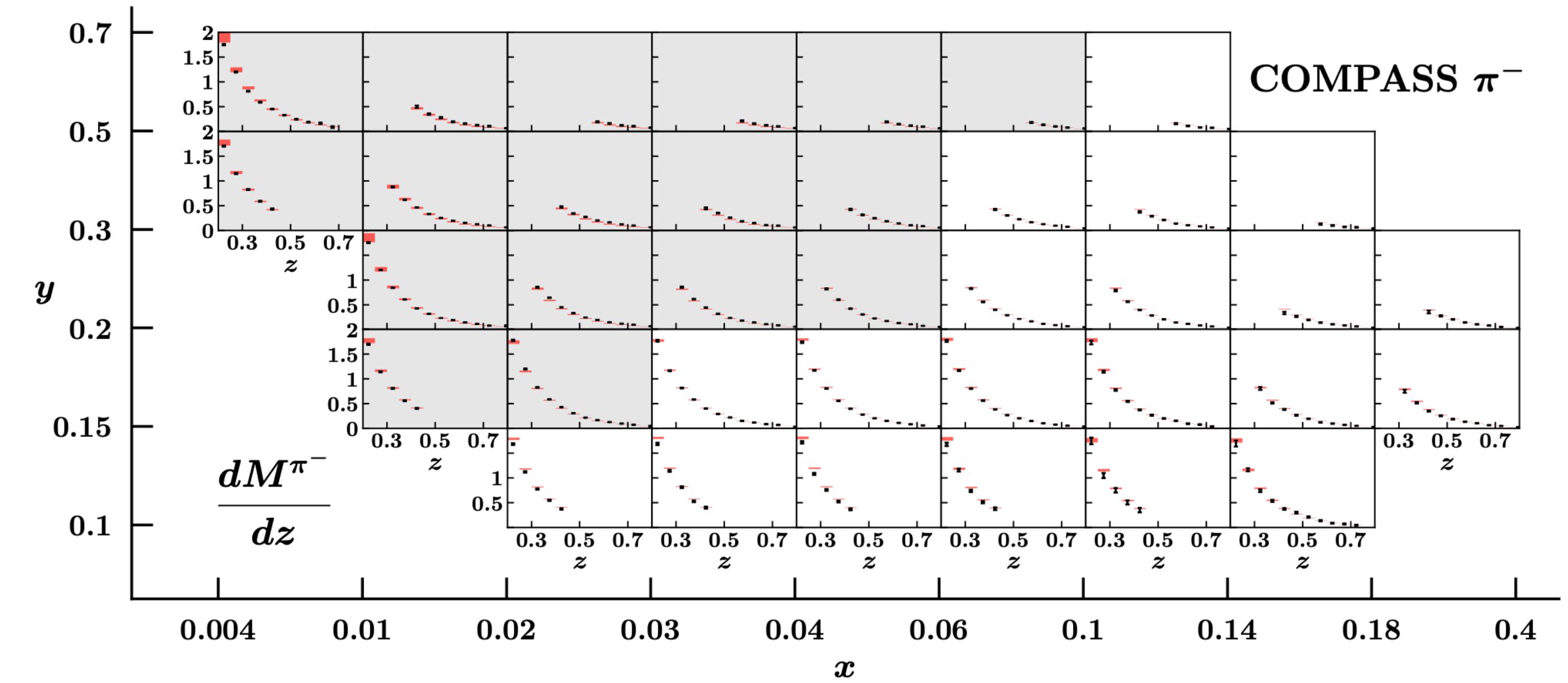
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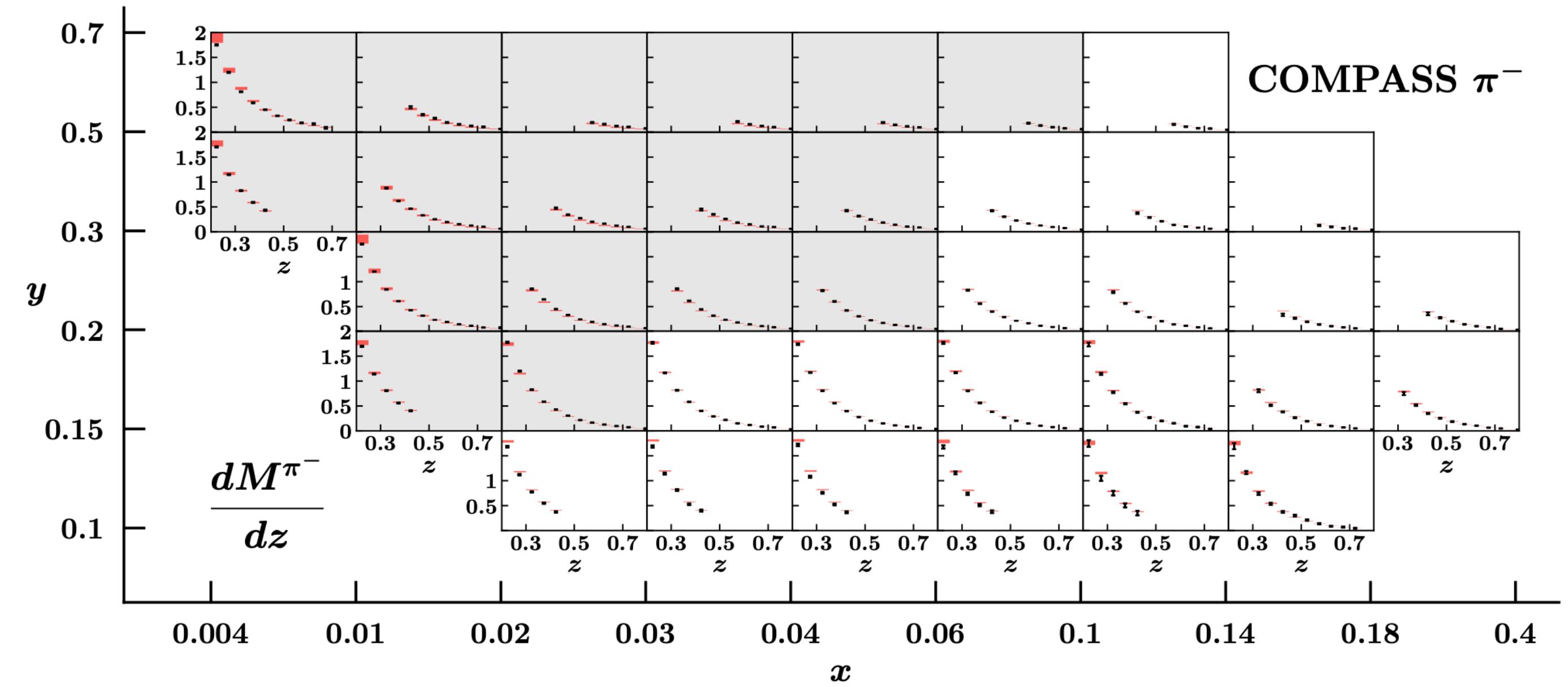
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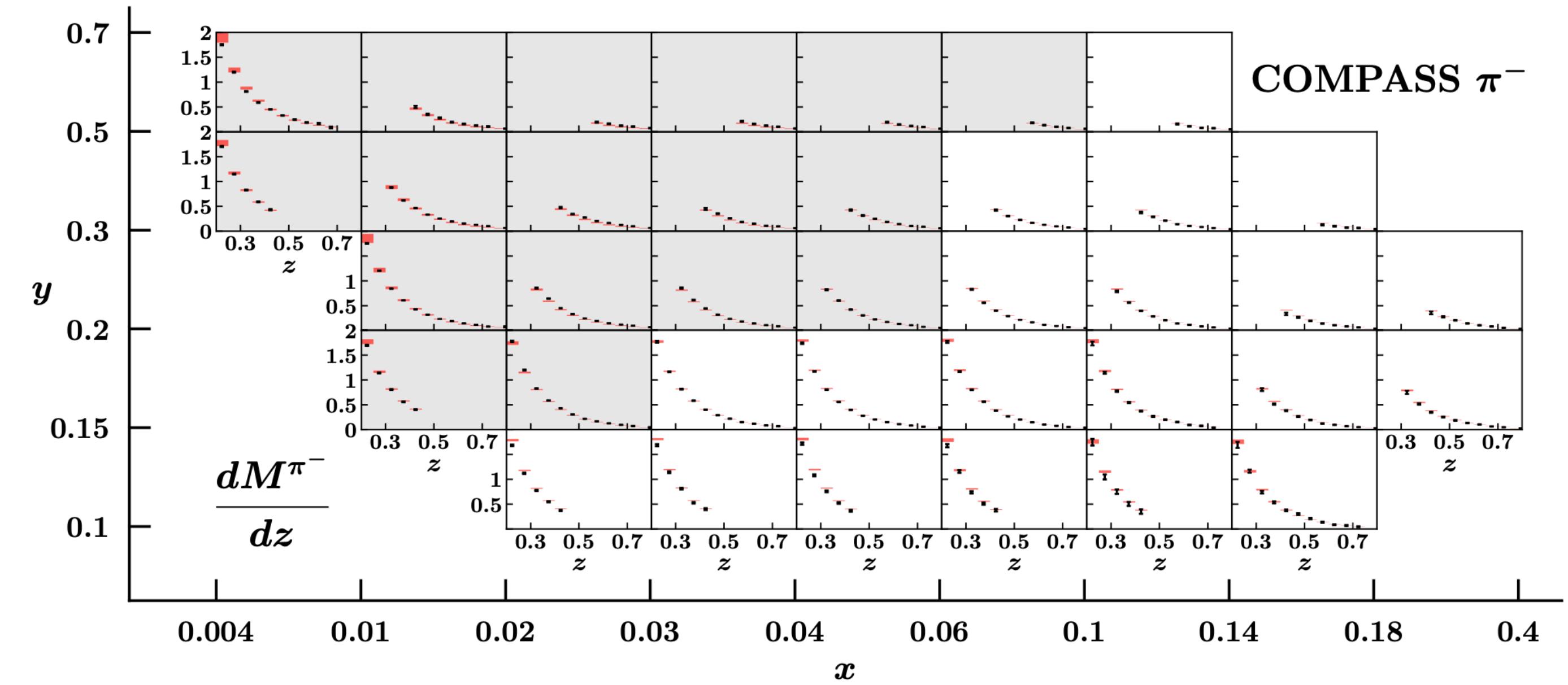
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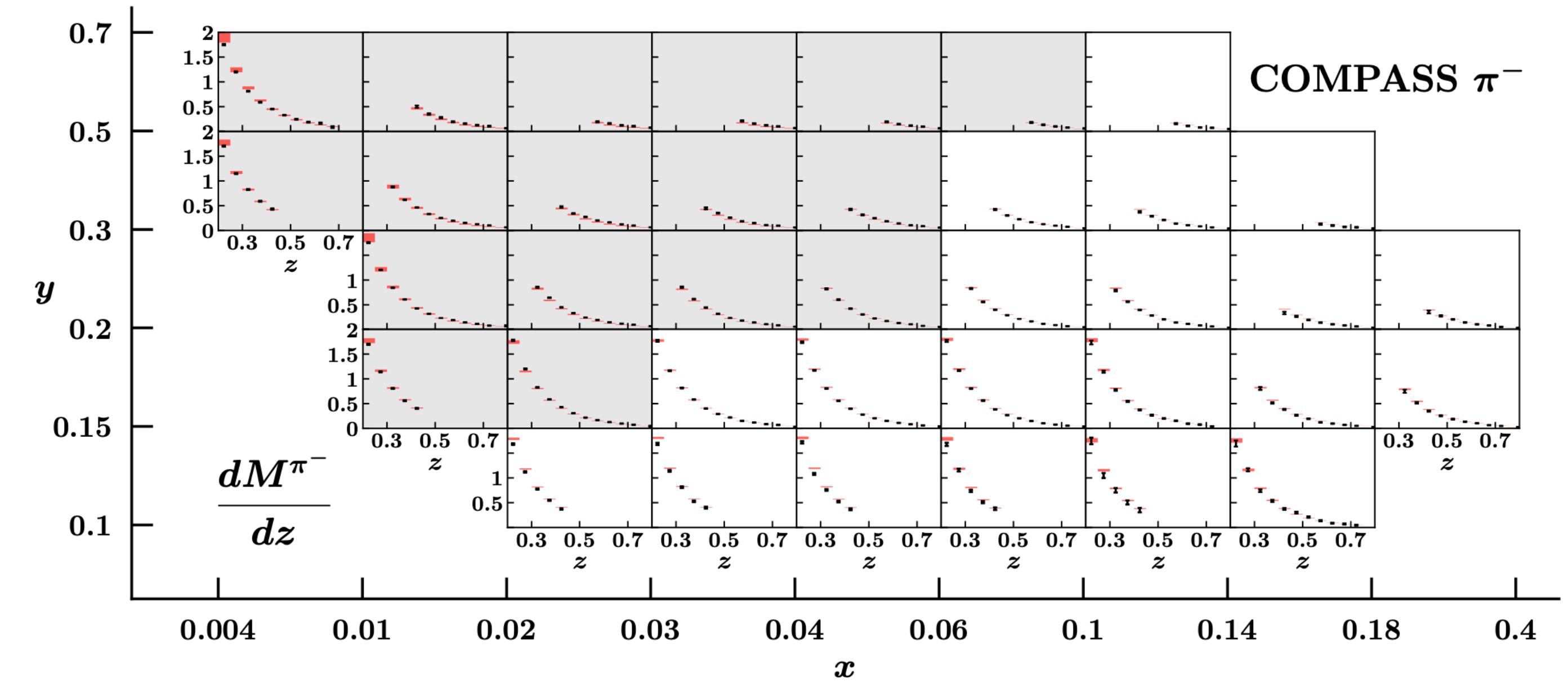
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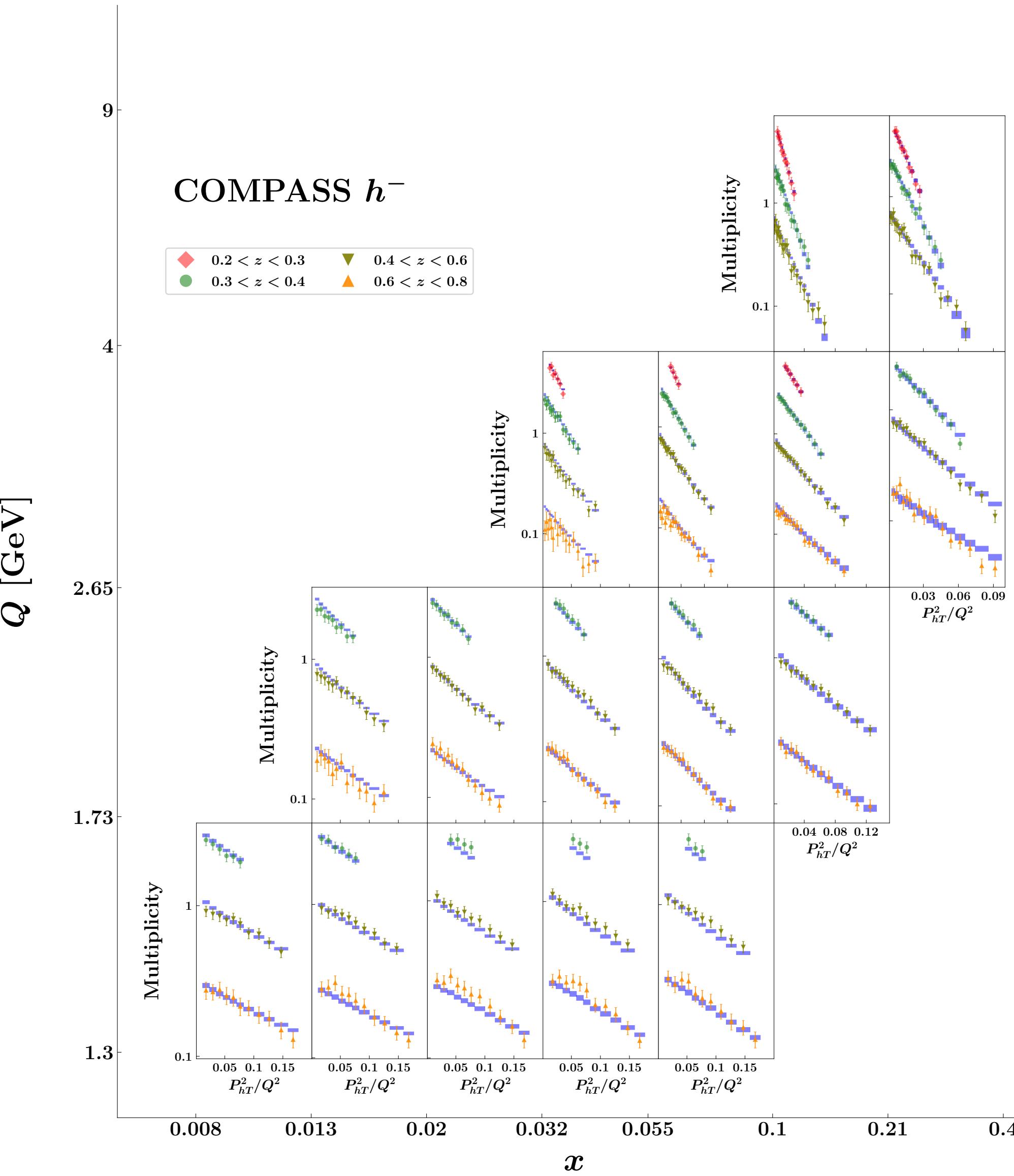
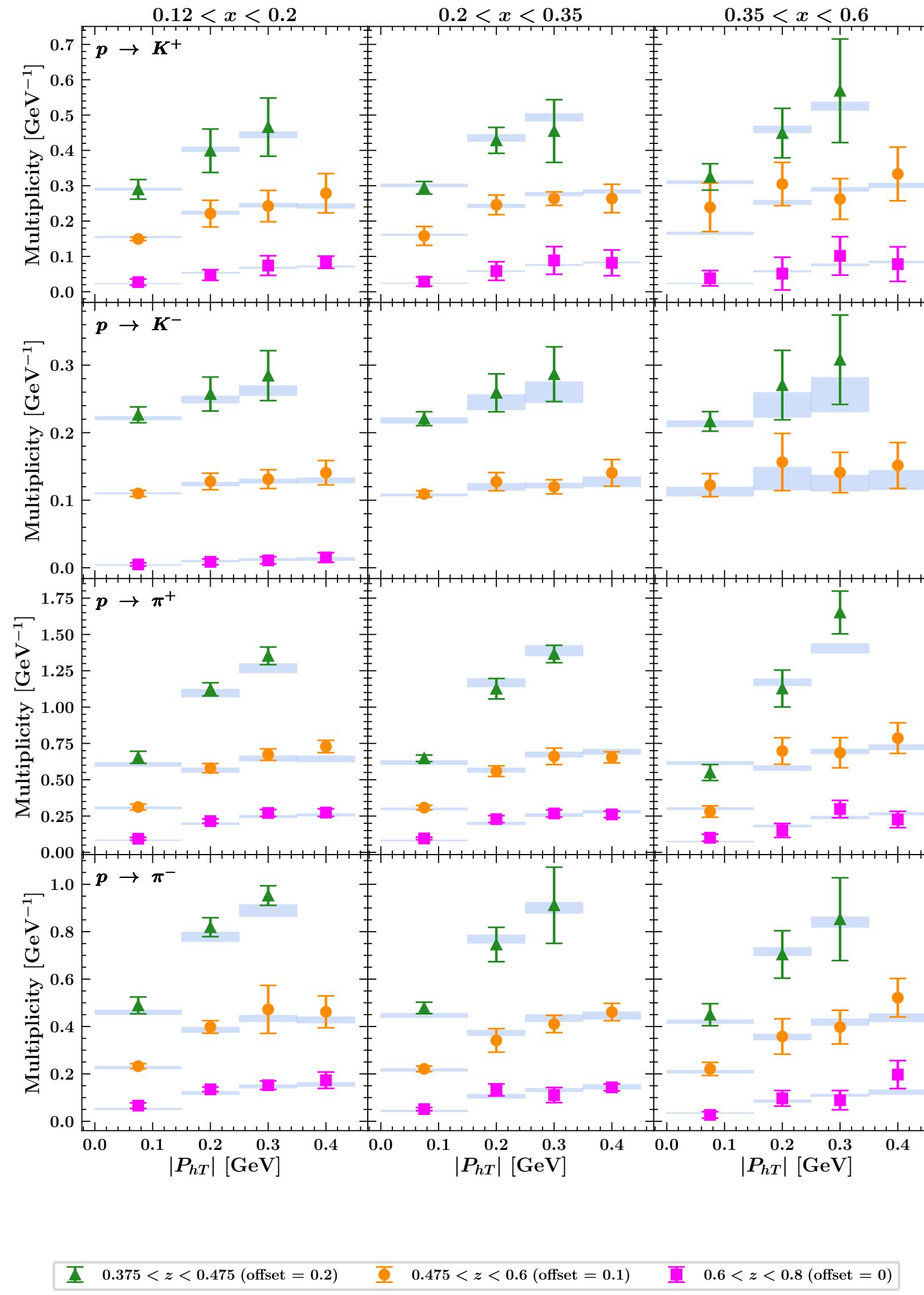
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**Independent of the fitting parameters!!**

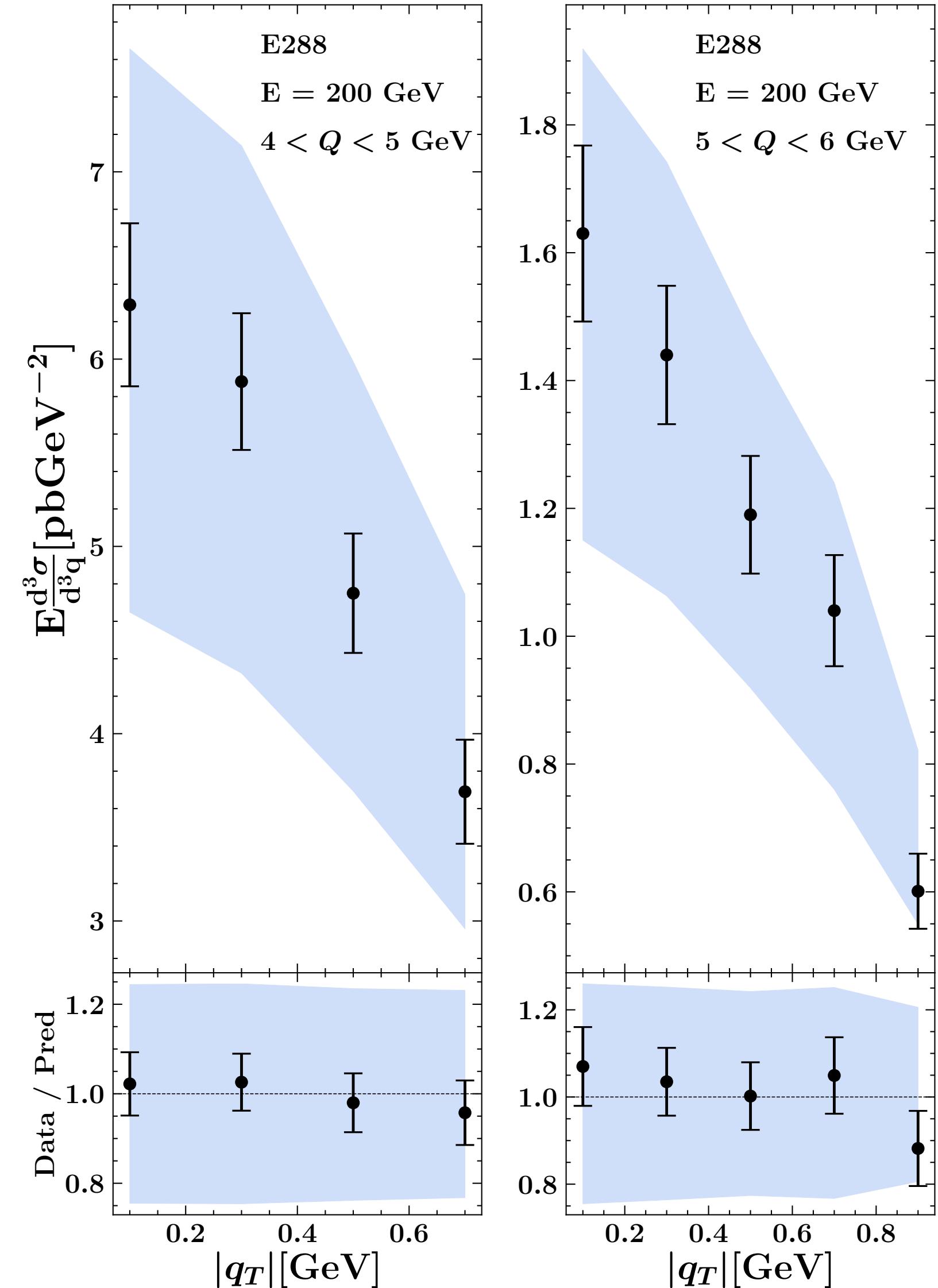
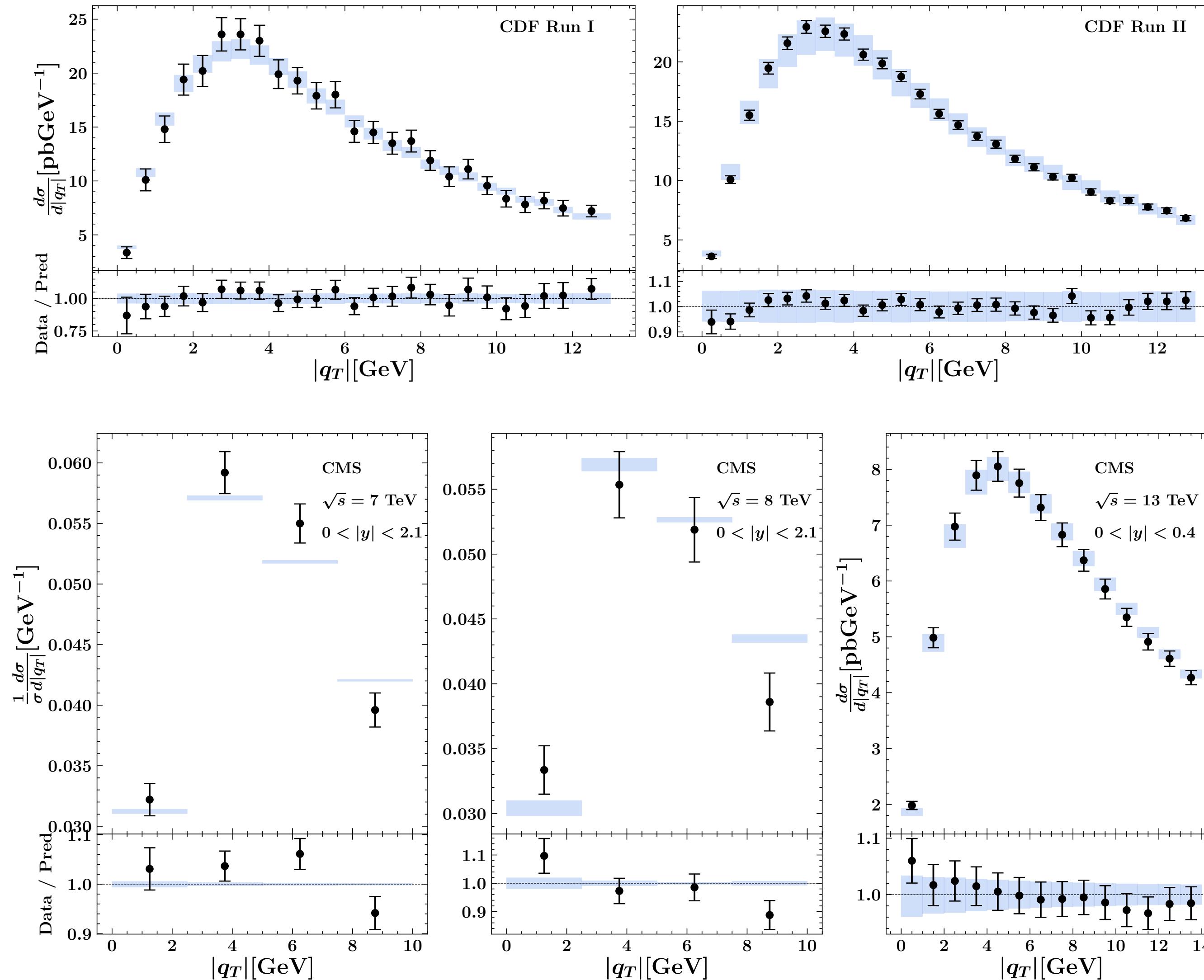


# MAPTMD22 – Results of the fit $\chi^2/N_{\text{data}} = 1.06$

HERMES

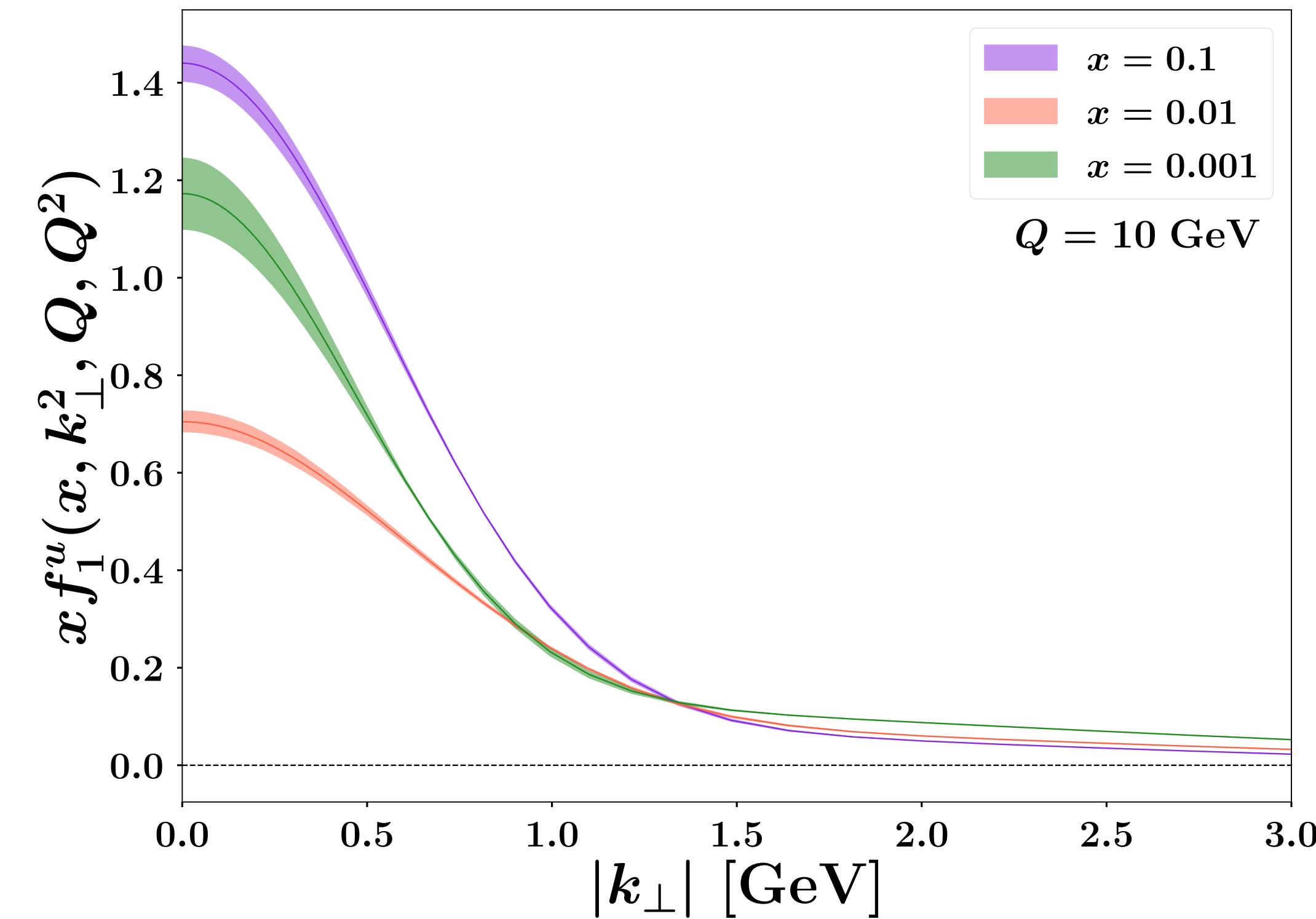
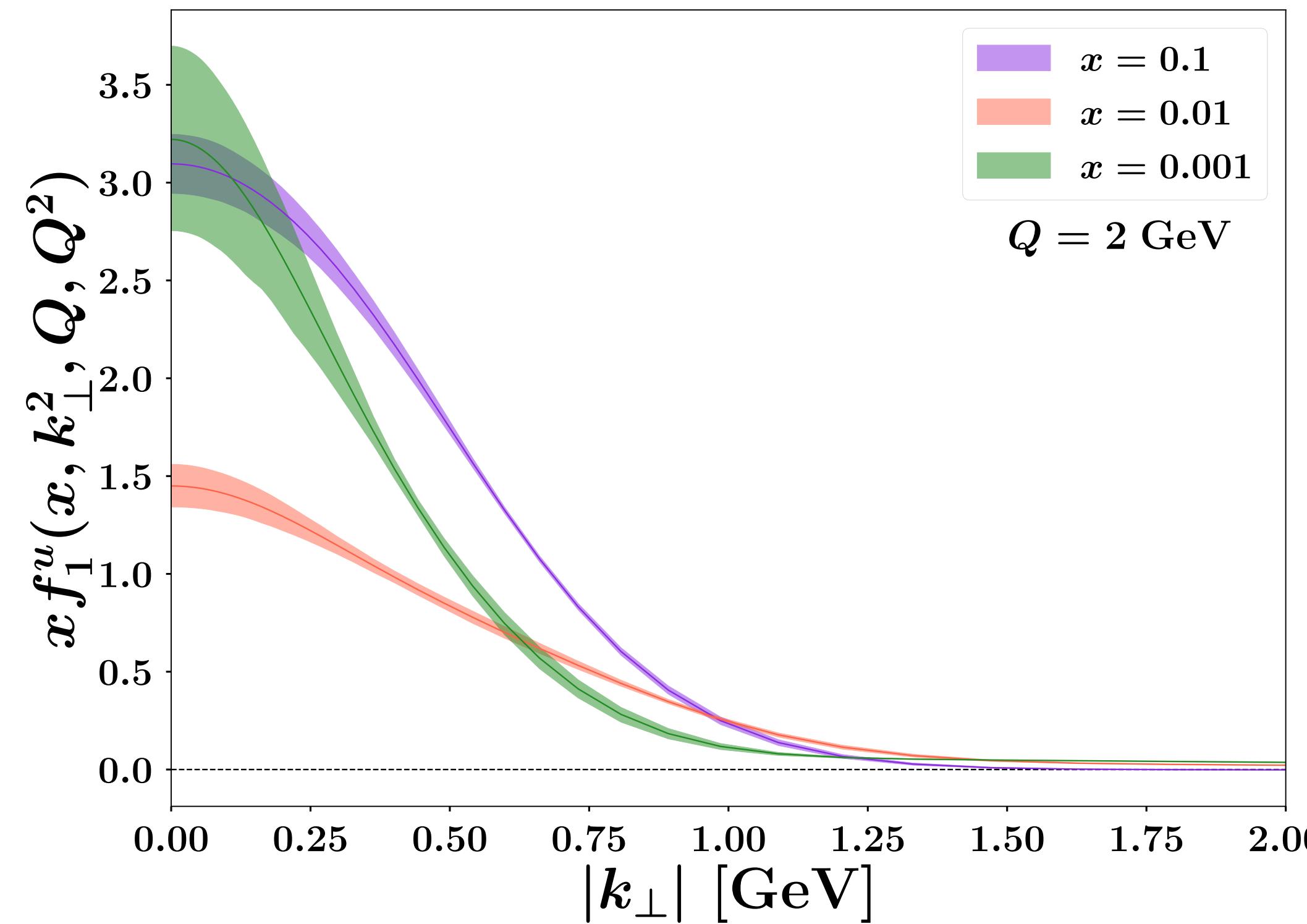


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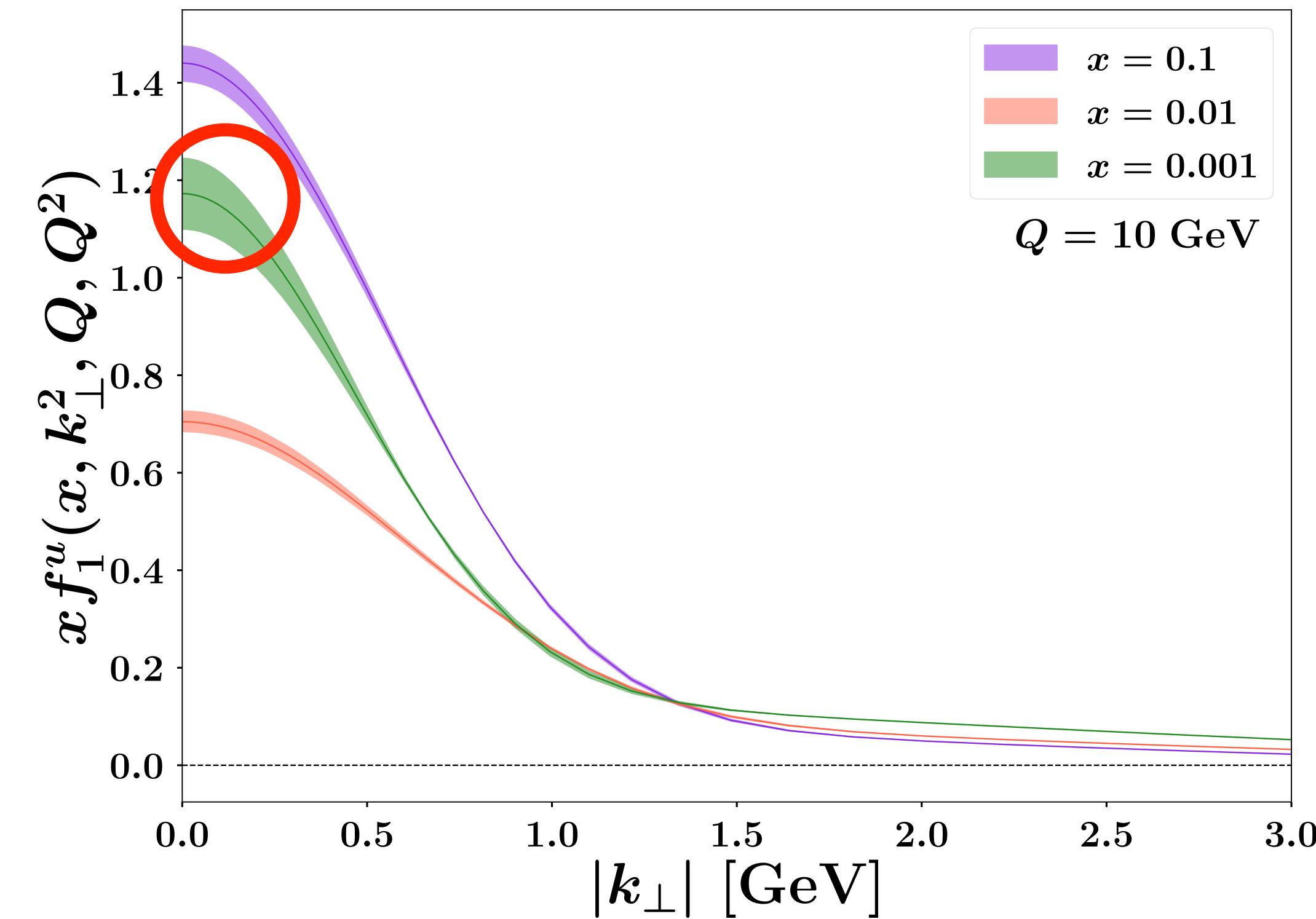
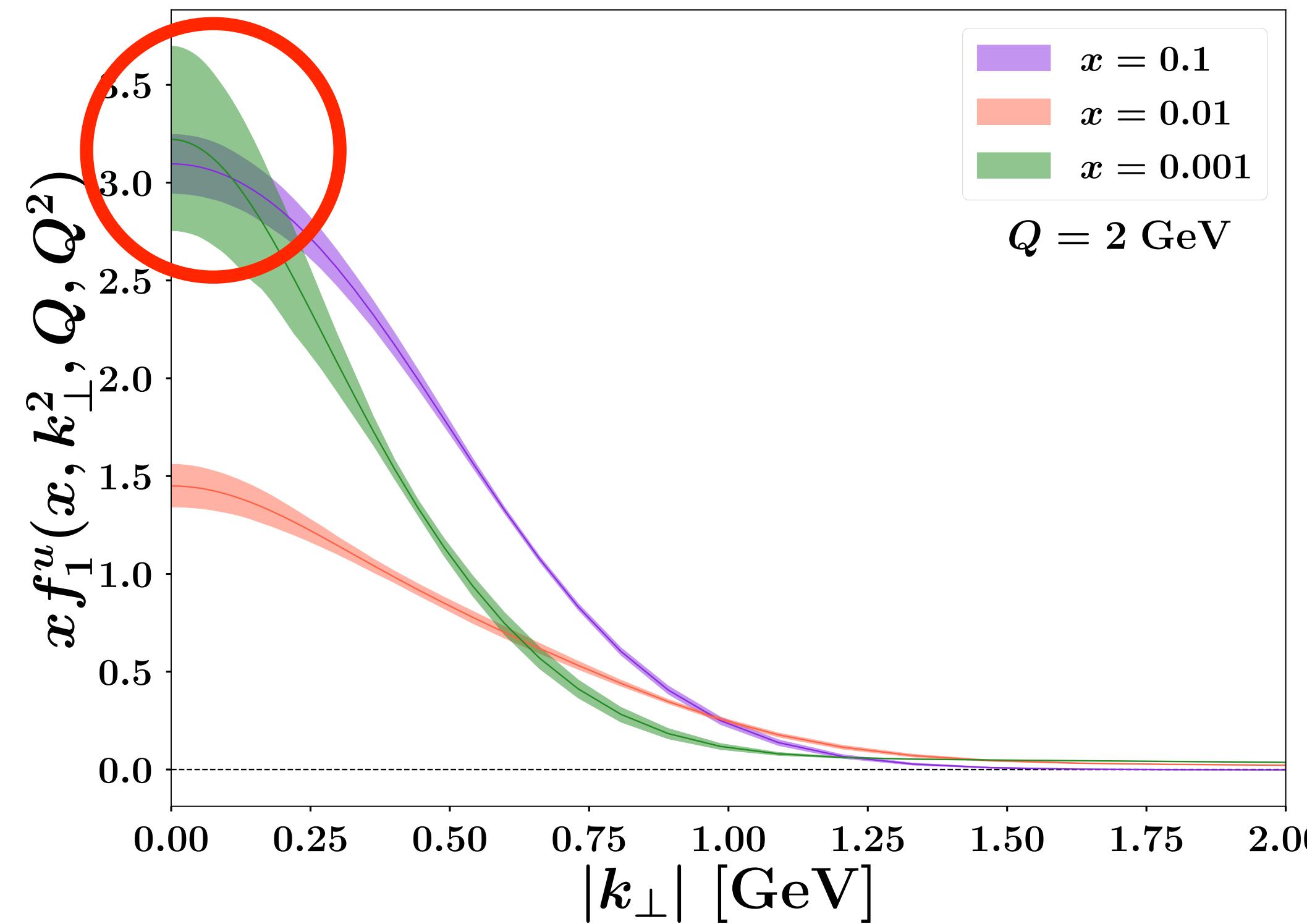
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Visualisation of TMD PDFs



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Kernel of the rapidity evolution equation

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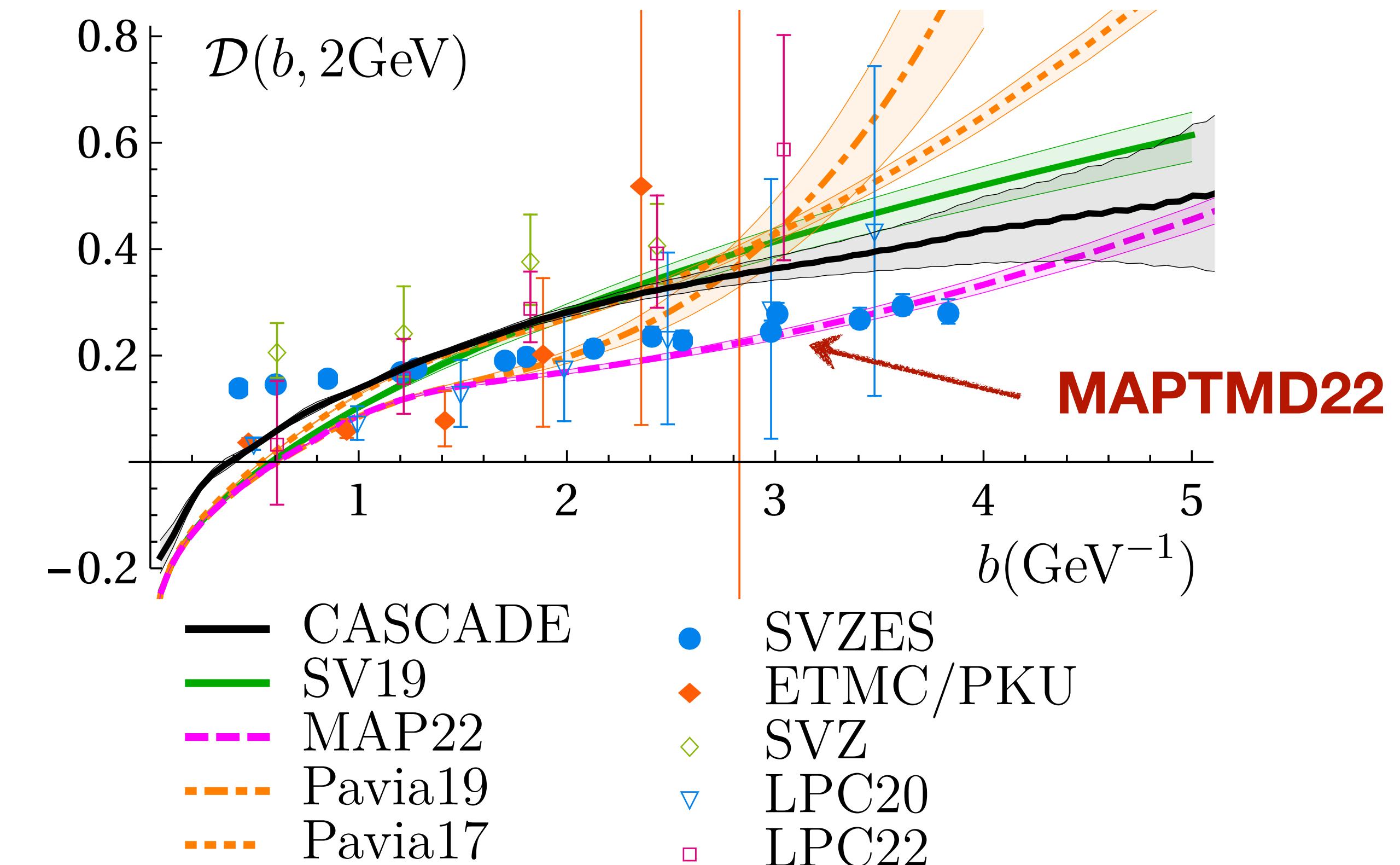
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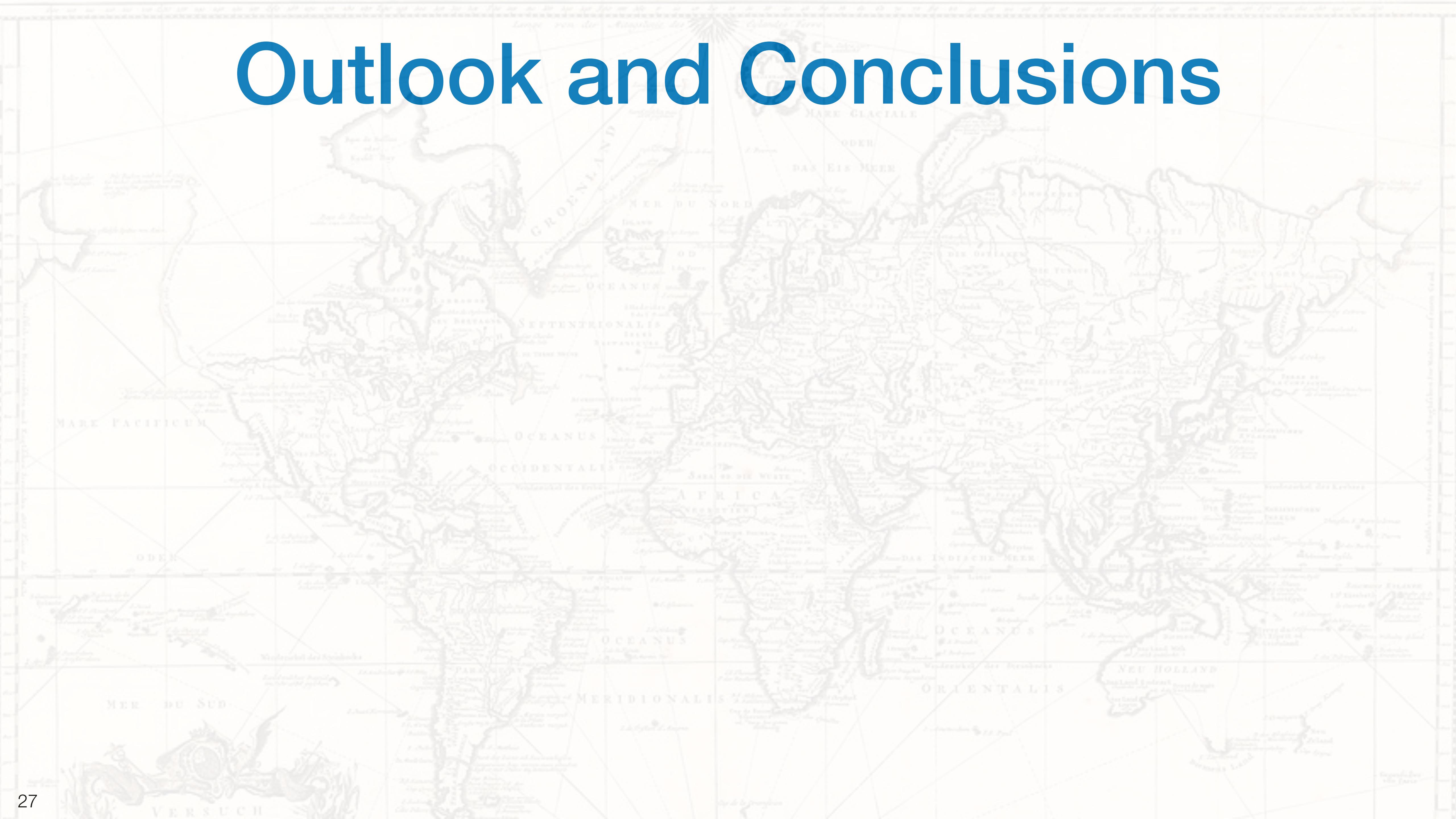


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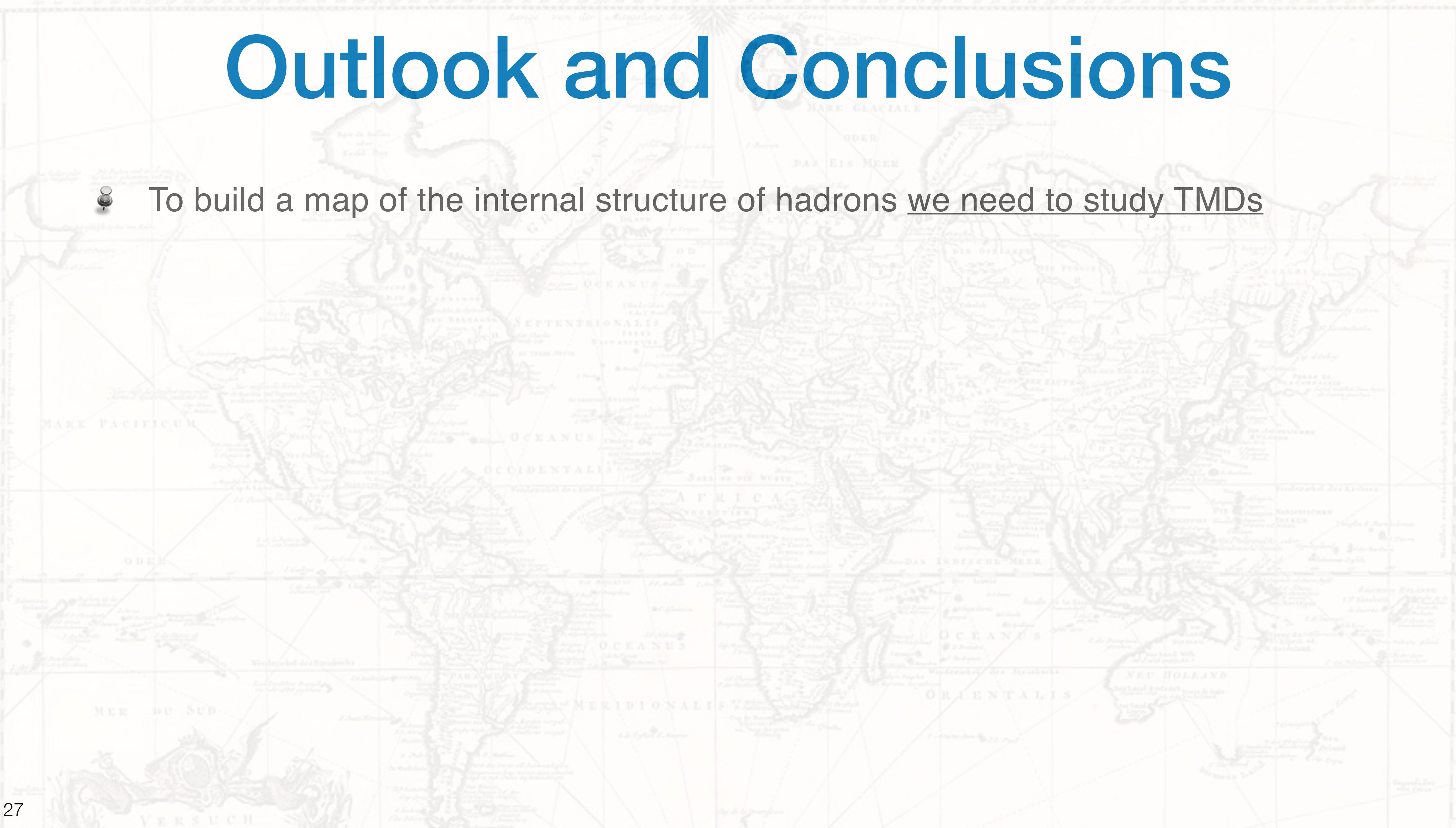


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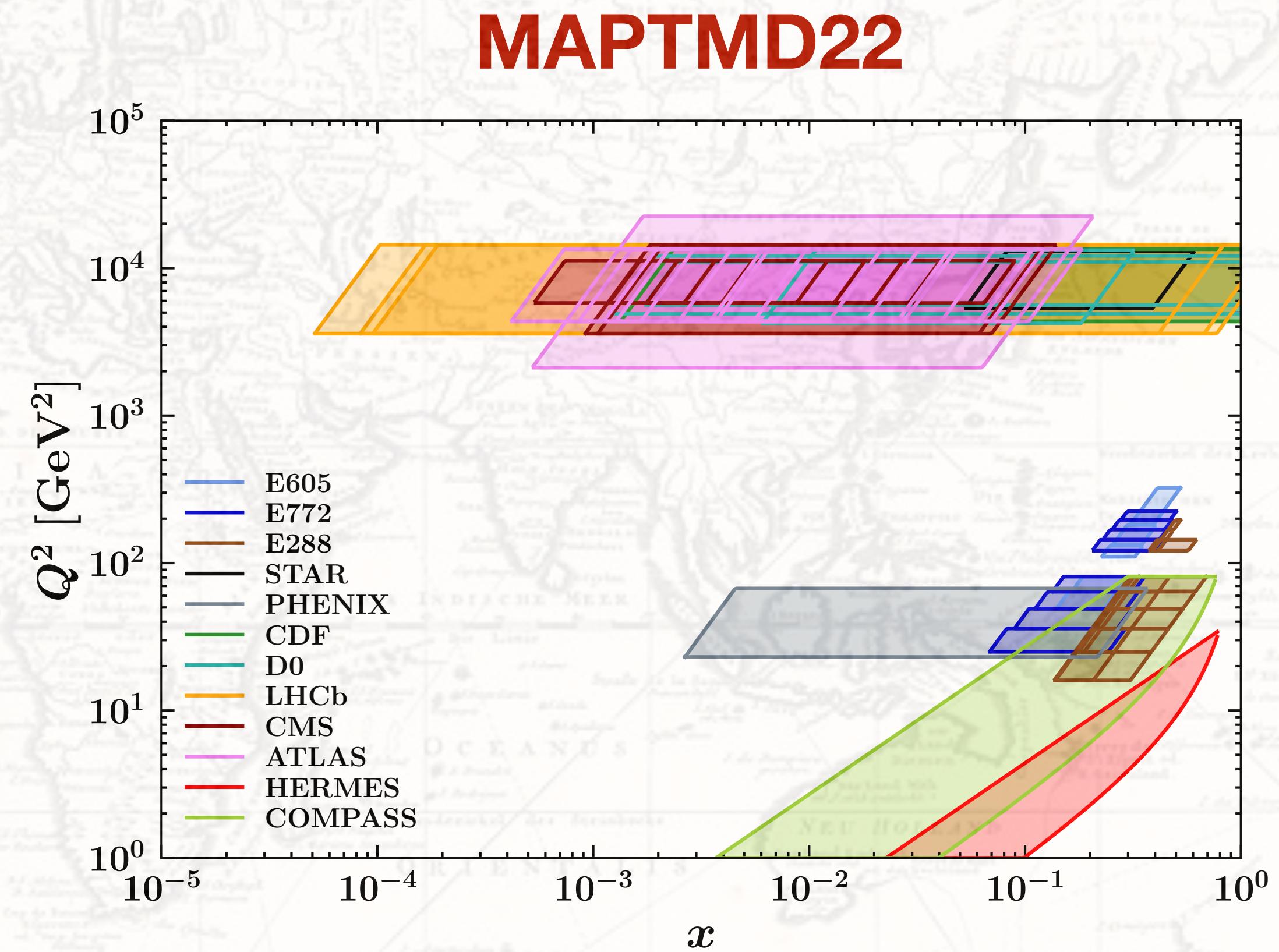


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Present...

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## Present...

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## Future...

- The EIC machine will provide us a large amount of experimental data which will allow us to better constrain TMDs

