



IWHSS 2018

UNIVERSITÄT BONN

19-21 March

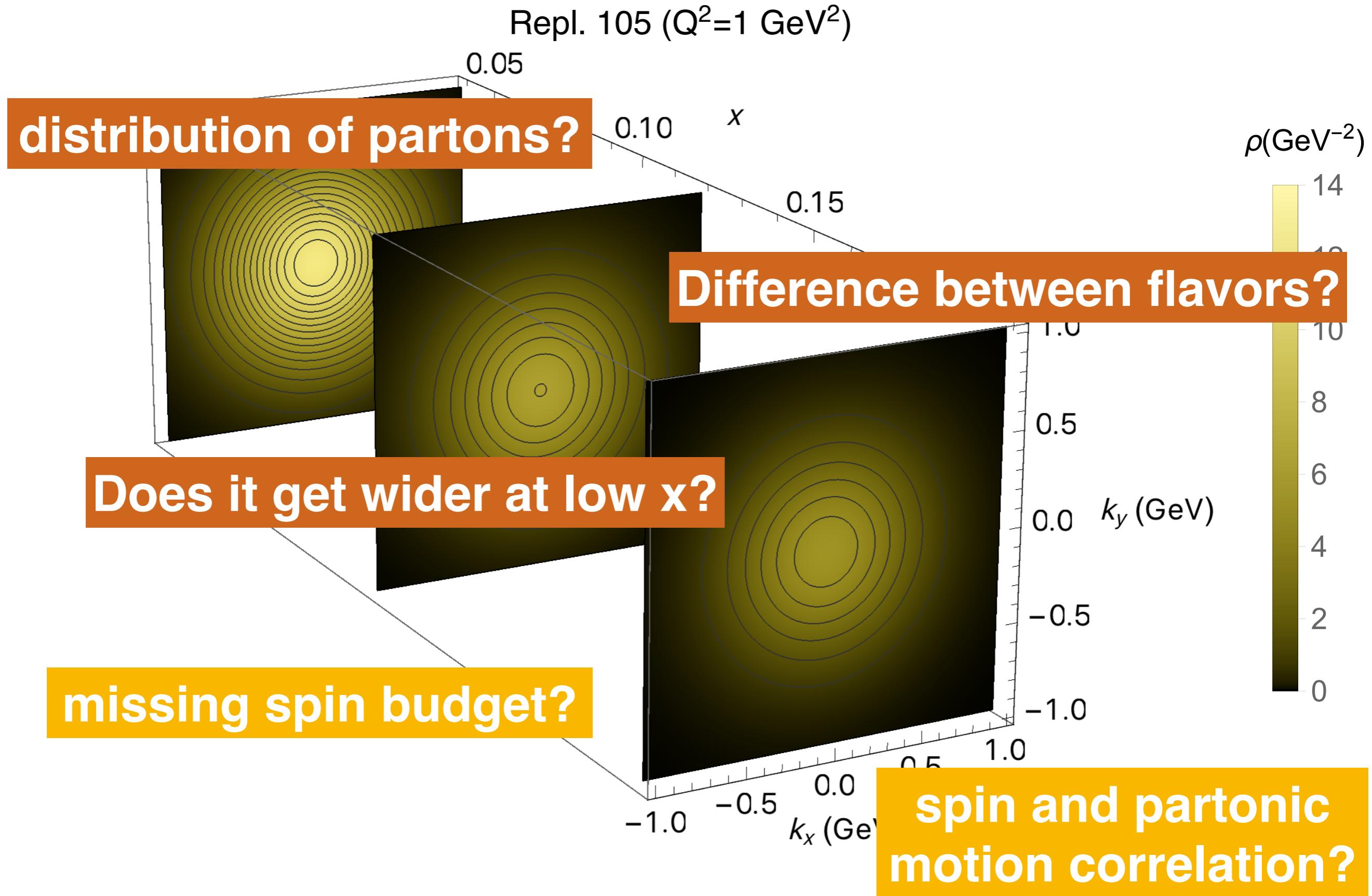
A global fit of partonic Transverse Momentum Dependent distributions

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In collaboration with A. Bacchetta, C. Pisano, M. Radici, A. Signori



3DSPIN: structure of the nucleon



Transverse Momentum Distributions: TMD PDF

quark pol.

Unpolarized

nucleon pol.

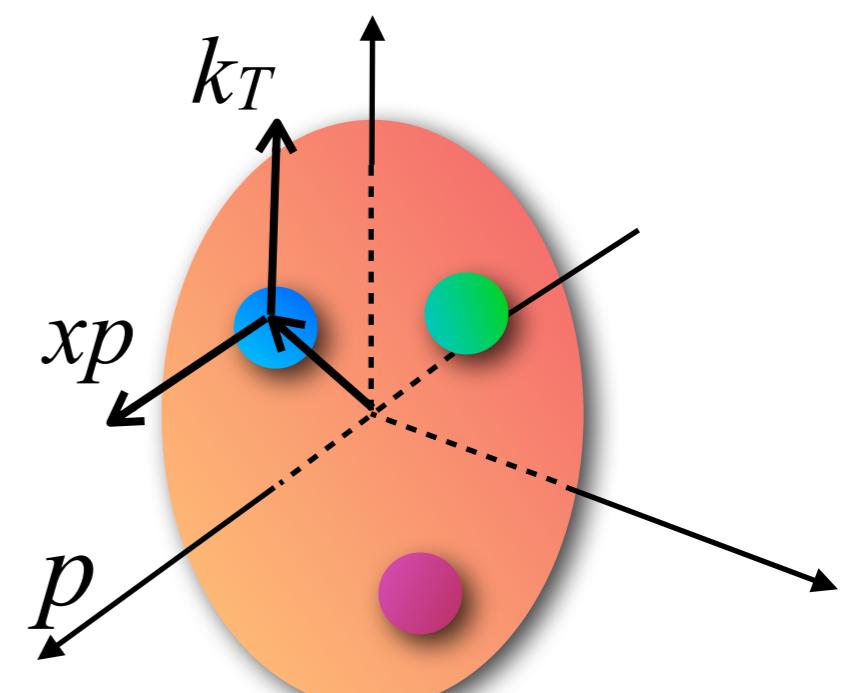
	U	L	T
U	f_1		h_1^\perp
L		g_{1L}	h_{1L}^\perp
T	f_{1T}^\perp	g_{1T}	h_1, h_{1T}^\perp

dependence on:

longitudinal momentum fraction x

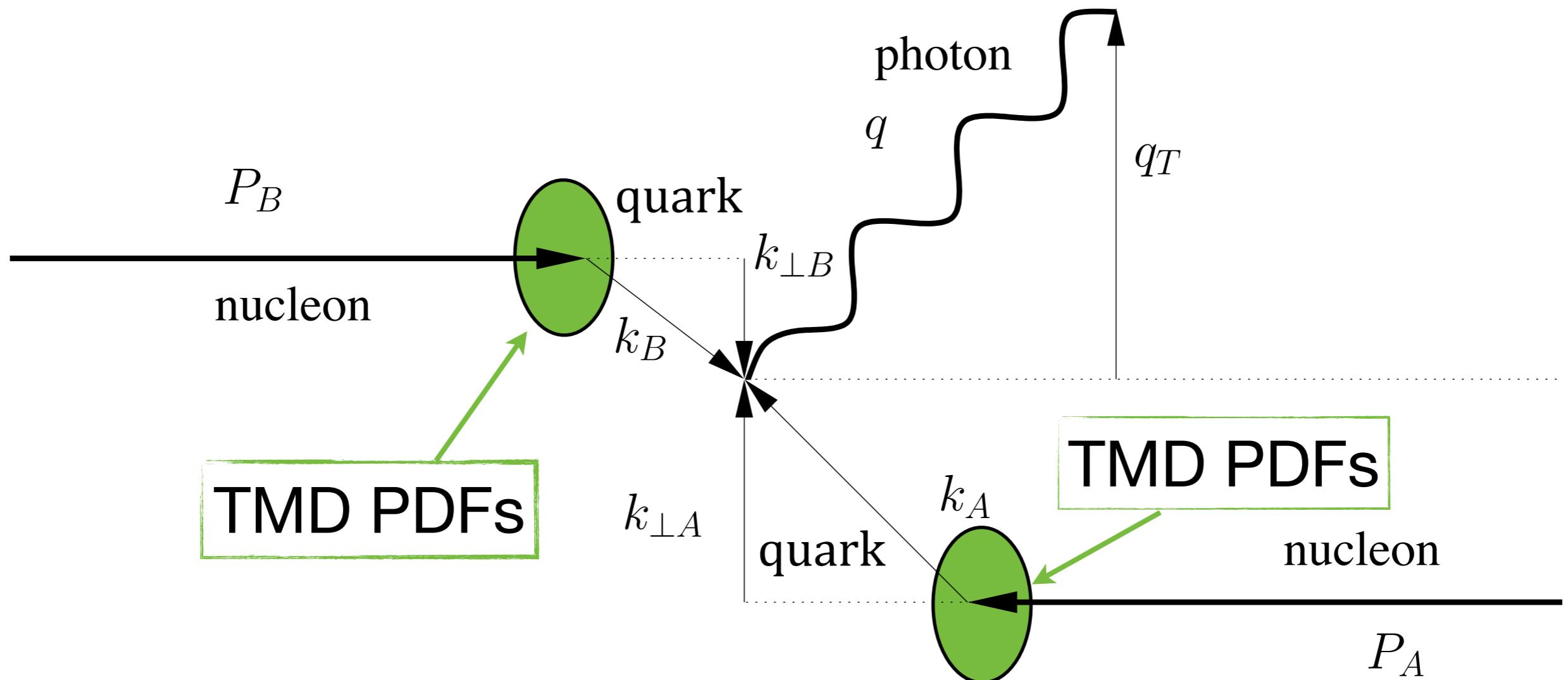
transverse momentum k_\perp

energy scale



Extraction from SIDIS & Drell-Yan

Drell-Yan \ Z production

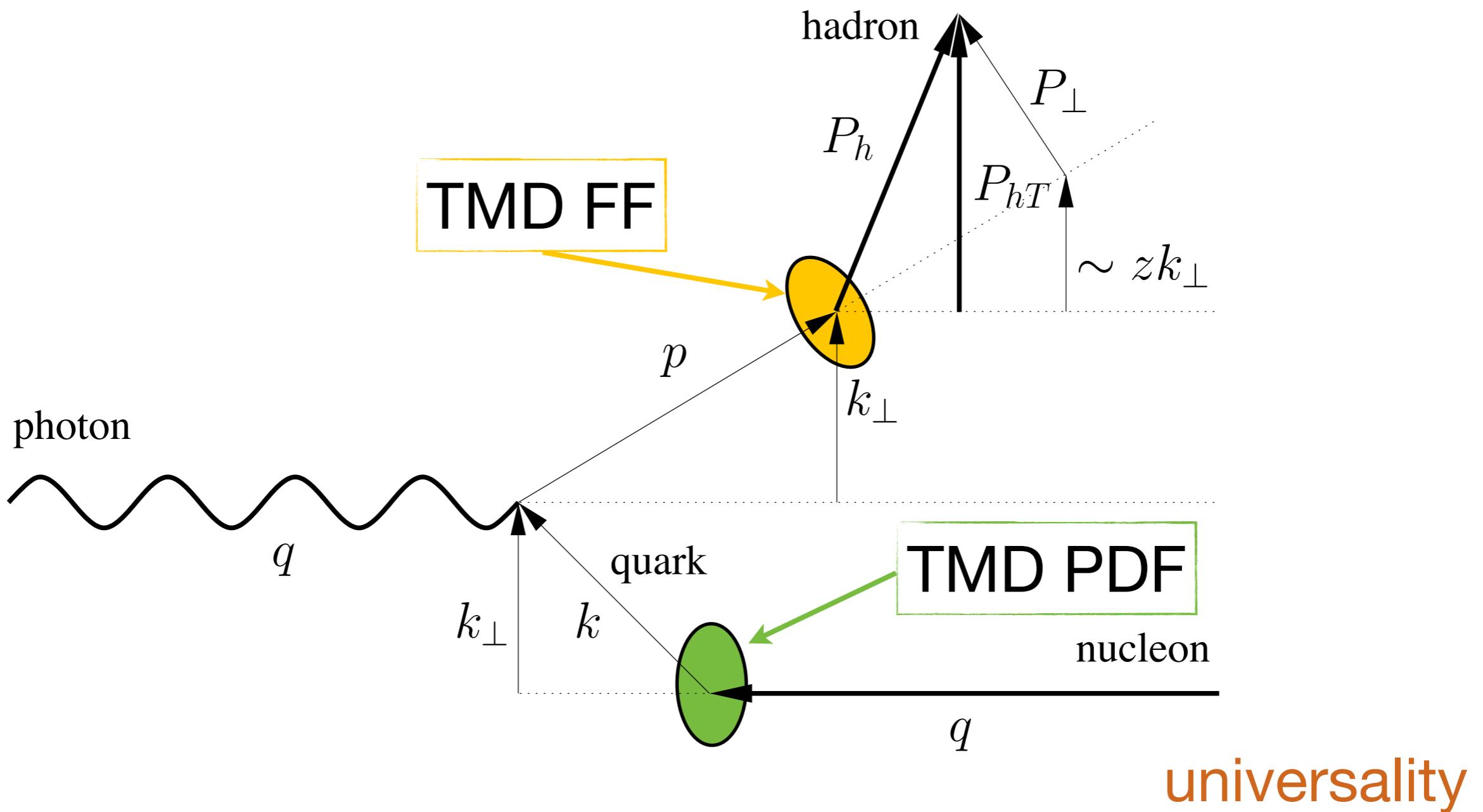


$$A + B \rightarrow \gamma^* \rightarrow l^+ l^-$$

$$A + B \rightarrow Z \rightarrow l^+ l^-$$

Extraction from SIDIS & Drell-Yan

Semi-inclusive Deep Inelastic Scattering



$$l(\ell) + N(\mathcal{P}) \rightarrow l(\ell') + h(\mathcal{P}_h) + X$$

TMDs: Fragmentation Function

quark pol.

Unpolarized

U	L	T
D_1		H_1^\perp

**TMD Fragmentation Functions
(TMD FFs)**

dependence on:

longitudinal momentum fraction z

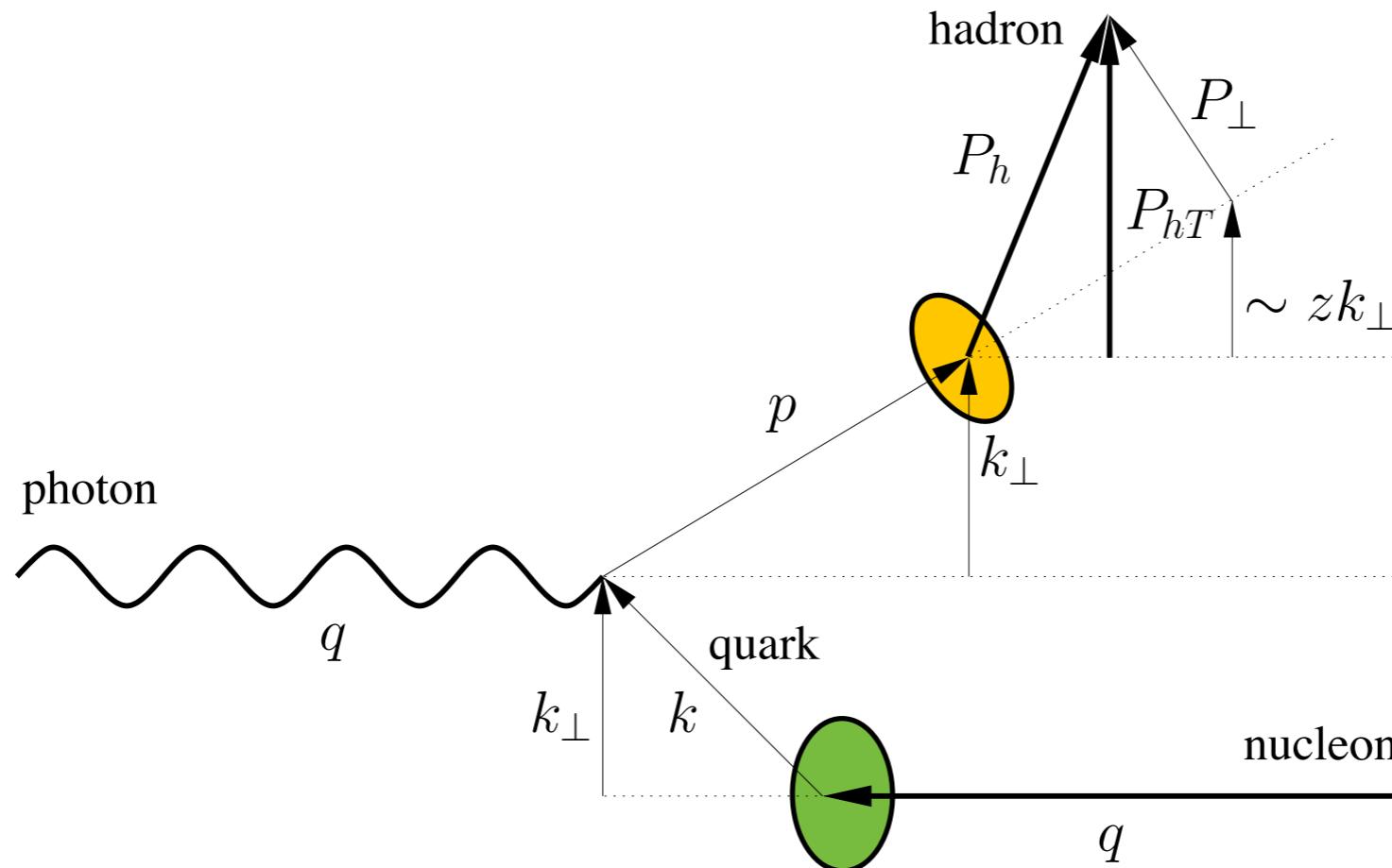
transverse momentum P_\perp

energy scale

Structure functions and TMDs

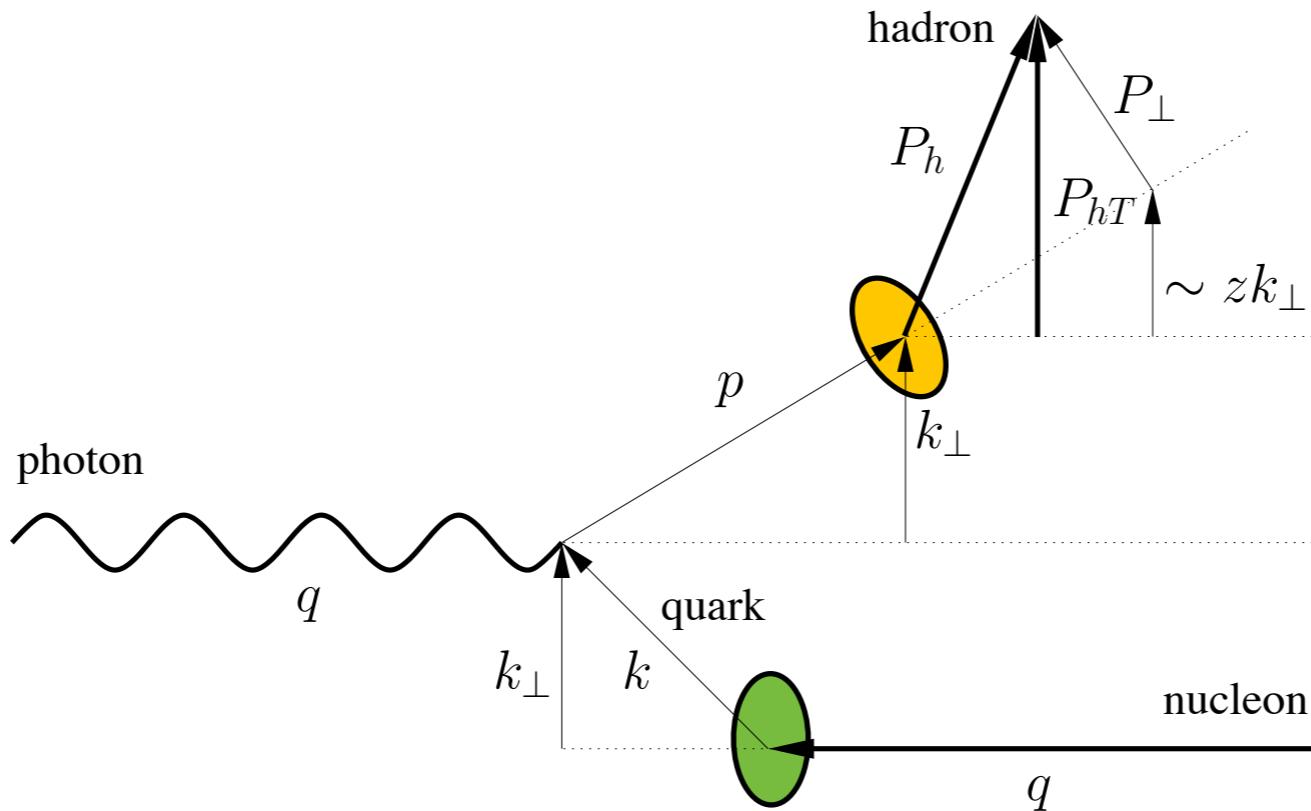
multiplicities

$$m_N^h(x, z, \mathbf{P}_{hT}^2, Q^2) = \frac{d\sigma_N^h / (dx dz d\mathbf{P}_{hT}^2 dQ^2)}{d\sigma_{DIS} / (dx dQ^2)} \approx \frac{\pi F_{UU,T}(x, z, \mathbf{P}_{hT}^2, Q^2)}{F_T(x, Q^2)}$$



$$\begin{aligned} F_{UU,T}(x, z, P_{hT}^2, Q^2) &= \sum_a \mathcal{H}_{UU,T}^a(Q^2; \mu^2) \int d^2 k_T d^2 P_T f_1^a(x, k_T^2; \mu^2) D_1^{h/a}(z, P_T^2; \mu^2) \\ &\quad \cdot \delta^2(z k_T - P_{hT} + P_T) + Y_{UU,T}(Q^2, P_{hT}^2) + \mathcal{O}(M^2/Q^2) \end{aligned}$$

Structure functions and TMDs



$$F_{UU,T}(x, z, P_{hT}^2, Q^2) = \sum_a \mathcal{H}_{UU,T}^a(Q^2; \mu^2) \int d^2 k_T d^2 P_T f_1^a(x, k_T^2; \mu^2) D_1^{h/a}(z, P_T^2; \mu^2)$$

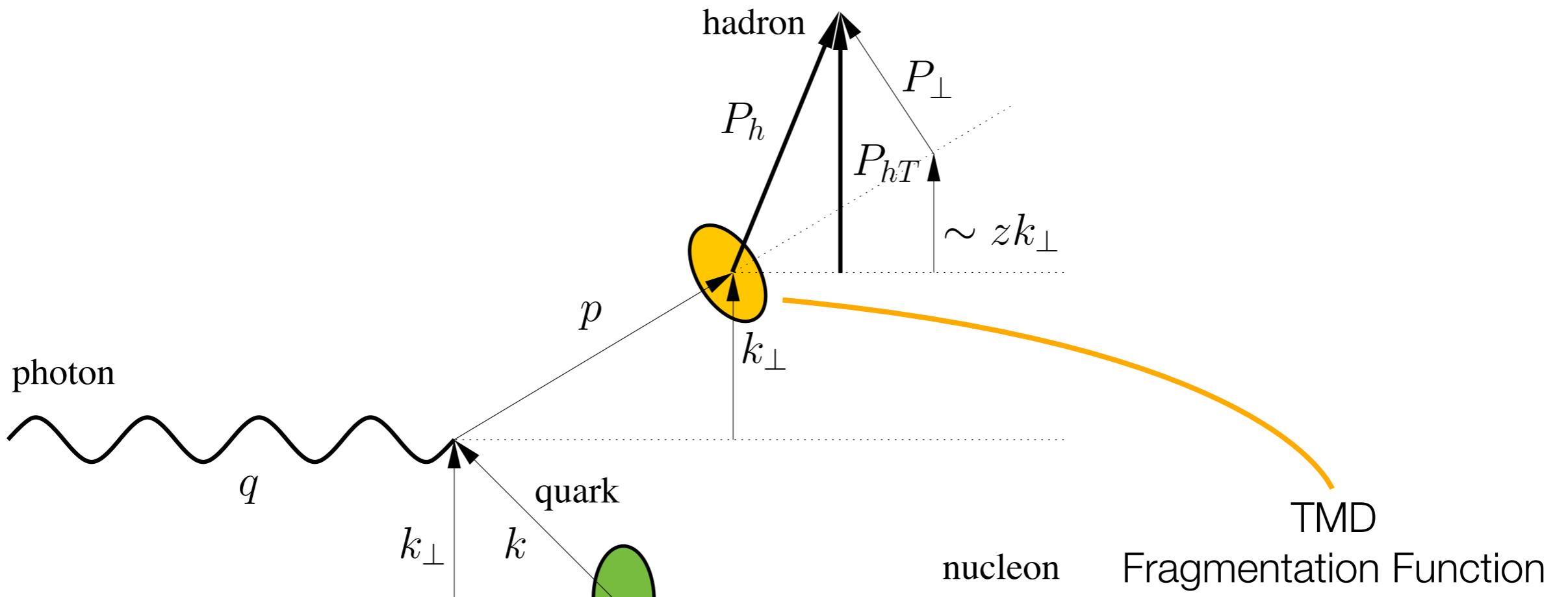
$$\cdot \delta^2(zk_T - P_{hT} + P_T) + Y_{UU,T}(Q^2, P_{hT}^2) + \mathcal{O}(M^2/Q^2)$$

At our accuracy level (LO-NLL):

$$\mathcal{H}_{UU,T} \simeq \mathcal{O}(\alpha_s^0)$$

$$Y_{UU,T}(Q^2, P_h^2 T) \simeq 0$$

Structure functions and TMDs



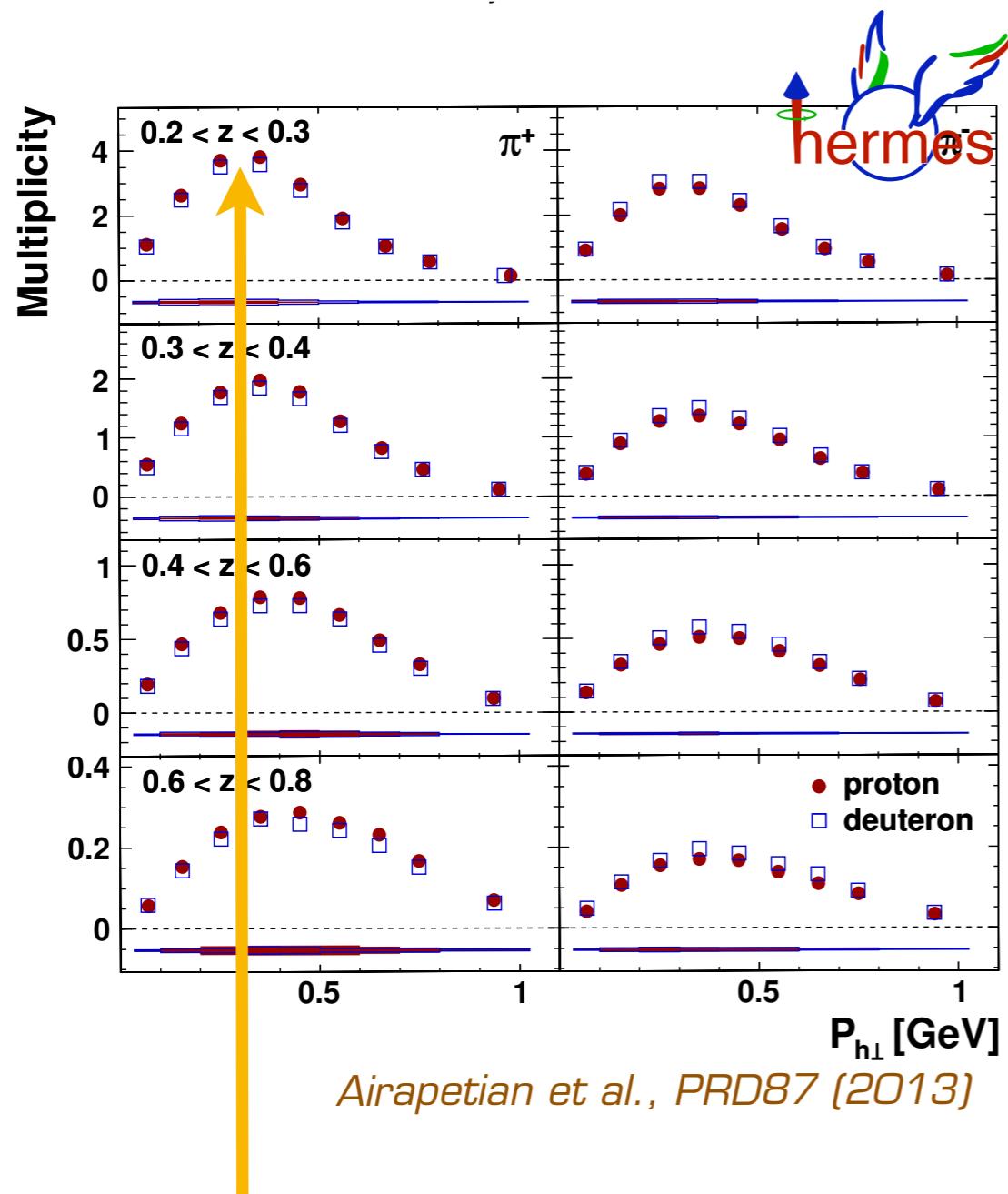
TMD Parton
Distribution Function

TMD
Fragmentation Function

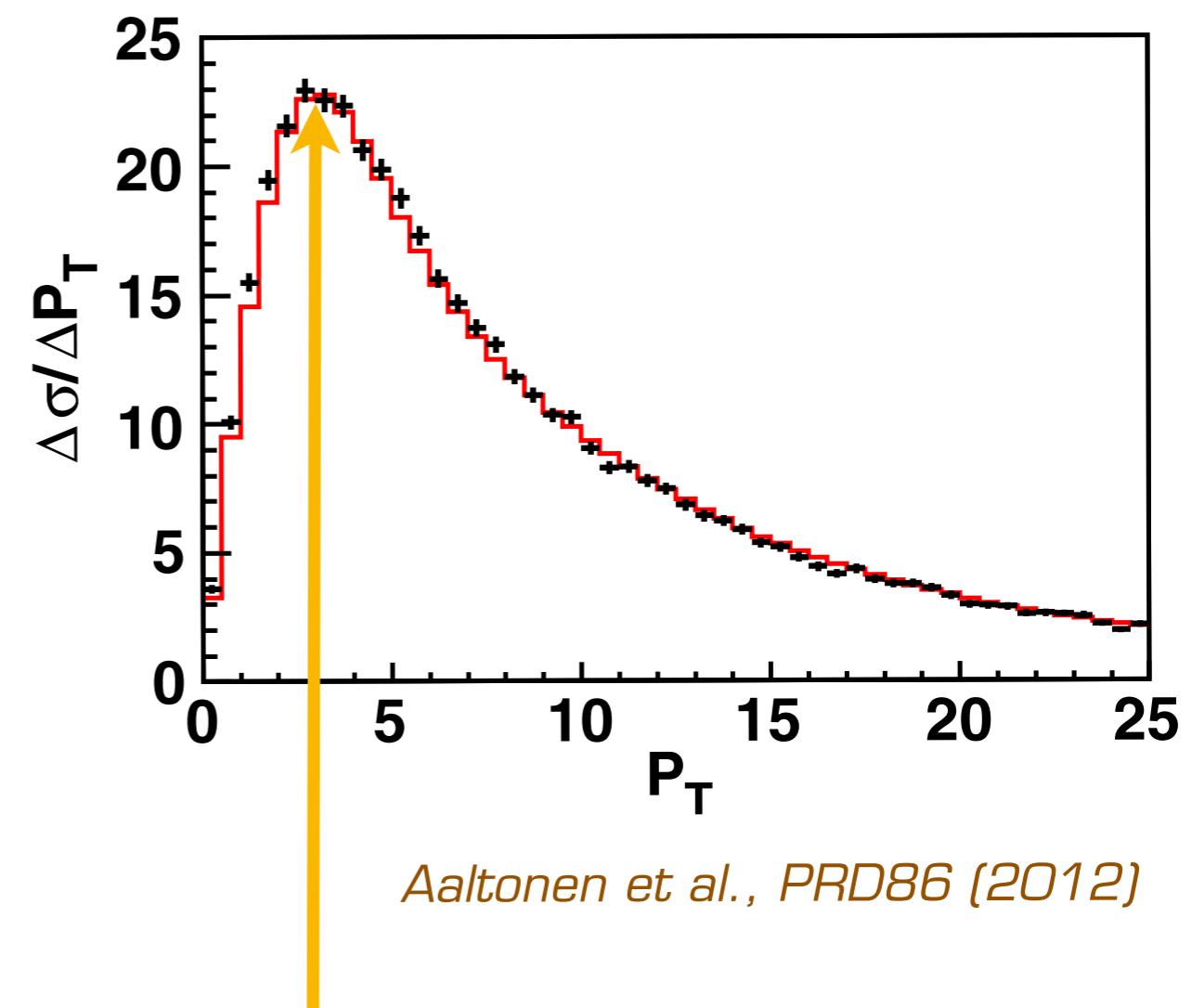
$$F_{UU,T} (x, z, P_{hT}^2, Q^2) \simeq \sum_a \int d^2 k_T d^2 P_T f_1^a (x, k_T^2; \mu^2) D_1^{h/a} (z, P_T^2; \mu^2) \cdot \delta^2 (z k_T - P_{hT} + P_T)$$

TMD Evolution

HERMES, $Q \approx 1.5$ GeV



CDF, $Q \approx 91$ GeV



Width of TMDs changes of one order of magnitude → Evolution

Evolved TMDs

Fourier transform: ξ_T space

$$\begin{aligned}\tilde{f}_1^a(x, \xi_T; \mu^2) &= \\ &= \sum_i \left(\tilde{C}_{a/i} \otimes f_1^i \right) (x, \bar{\xi}_*; \mu_b) e^{\tilde{S}(\bar{\xi}_*; \mu_b, \mu)} e^{g_K(\xi_T) \ln(\mu/\mu_0)} \hat{f}_{NP}^a(x, \xi_T)\end{aligned}$$

Diagram illustrating the decomposition of the evolved TMD function:

- The **collinear PDF** (bottom left) is the starting point.
- An arrow labeled **(Wilson Coefficient)** points to the term $\tilde{C}_{a/i} \otimes f_1^i$.
- An arrow labeled **pQCD** points to the Sudakov form factor $e^{\tilde{S}(\bar{\xi}_*; \mu_b, \mu)}$.
- An arrow labeled **(Sudakov form factor)** points to the evolution factor $e^{g_K(\xi_T) \ln(\mu/\mu_0)}$.
- A curved arrow labeled **non-perturbative part of TMD** points to the **nonperturbative part of evolution**, which is enclosed in a blue oval.
- A curved arrow labeled **non-perturbative part of evolution** points to the **non-perturbative part of TMD**, which is enclosed in a blue oval.

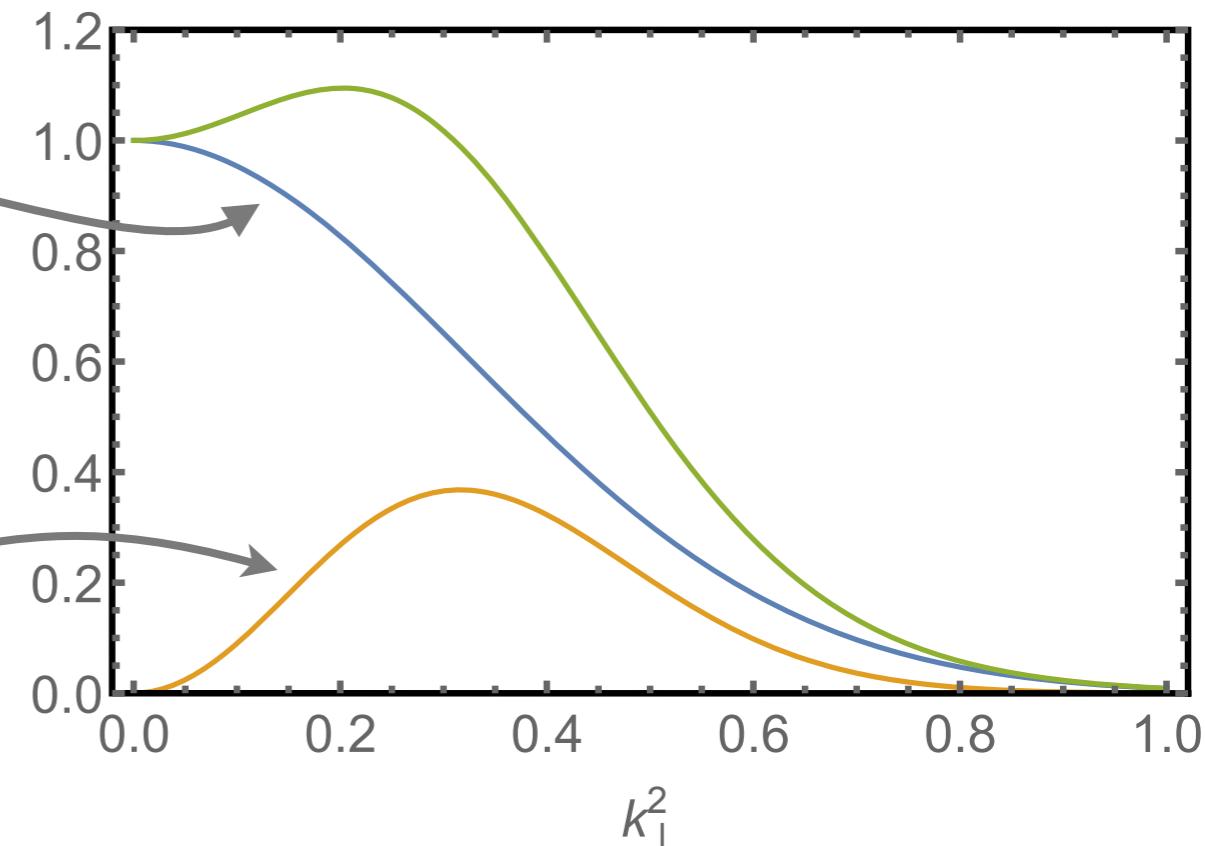
Non-perturbative contributions have to be **extracted** from experimental data, after **parametrization**

Model: non perturbative elements

input TMD PDF ($Q^2 = 1 \text{ GeV}^2$)

$$\hat{f}_{NP}^a = \mathcal{F.T.} \text{ of}$$

$$\left(e^{-\frac{k_T^2}{g_1 a}} + \lambda k_T^2 e^{-\frac{k_T^2}{g_1 a}} \right)$$



sum of two different gaussians
with kinematic dependence on transverse momenta

width x-dependence

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

where

$$N_1 \equiv g_1(\hat{x})$$
$$\hat{x} = 0.1$$

Model: non perturbative elements

Free parameters

$$N_1, \alpha, \sigma, \lambda$$

4 for TMD PDF

$$N_3, N_4, \beta, \delta, \gamma, \lambda_F$$

6 for TMD FF

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

1 for NP contribution to
TMD evolution

In total we have **11 parameters**, for intrinsic transverse momentum
(4 PDFs, 6 FFs) and evolution (g_2)

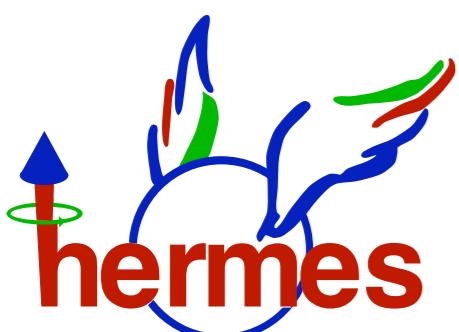
Experimental data



SIDIS μN
6252
data points

The logo consists of two blue stylized letters, "e" and "f", forming a larger "ef" shape. To the right of the logo, "E288" is written above "E605", both in bold black font.

Drell-Yan
203
data points



SIDIS $e N$
1514
data points



Z Production
90
data points

Data selection and analysis

$Q^2 > 1.4 \text{ GeV}^2$

$0.2 < z < 0.7$

$P_{hT}, q_T < \text{Min}[0.2Q, 0.7Qz] + 0.5 \text{ GeV}$

Motivations behind kinematical cuts

TMD factorization ($P_{hT}/z \ll Q^2$)

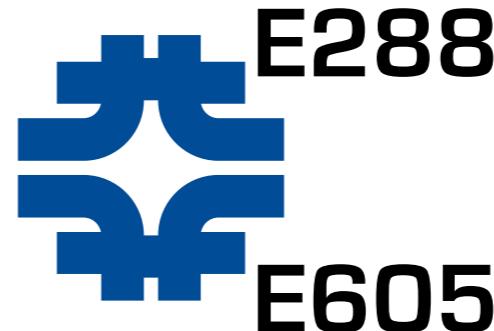
Avoid target fragmentation (low z)
and exclusive contributions (high z)

Experimental data



SIDIS μN

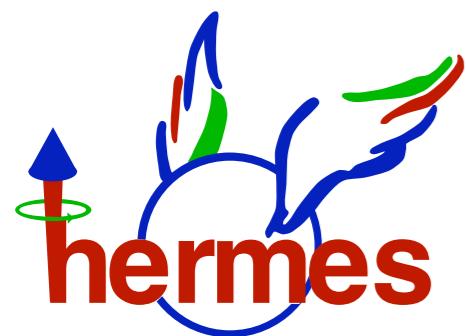
6252
data points



Drell-Yan

203
data points

Total: 8059 data



SIDIS $e N$

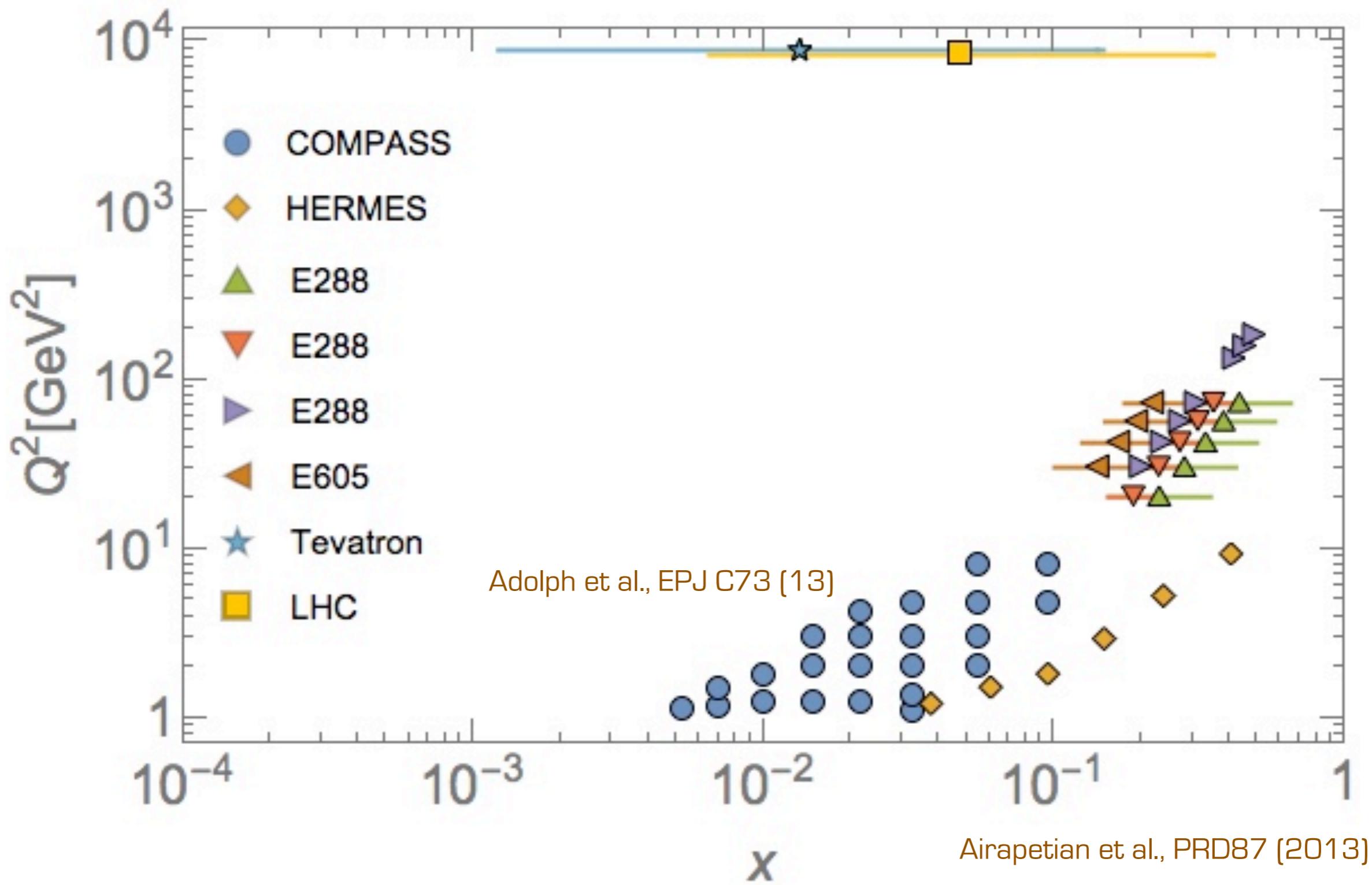
1514
data points



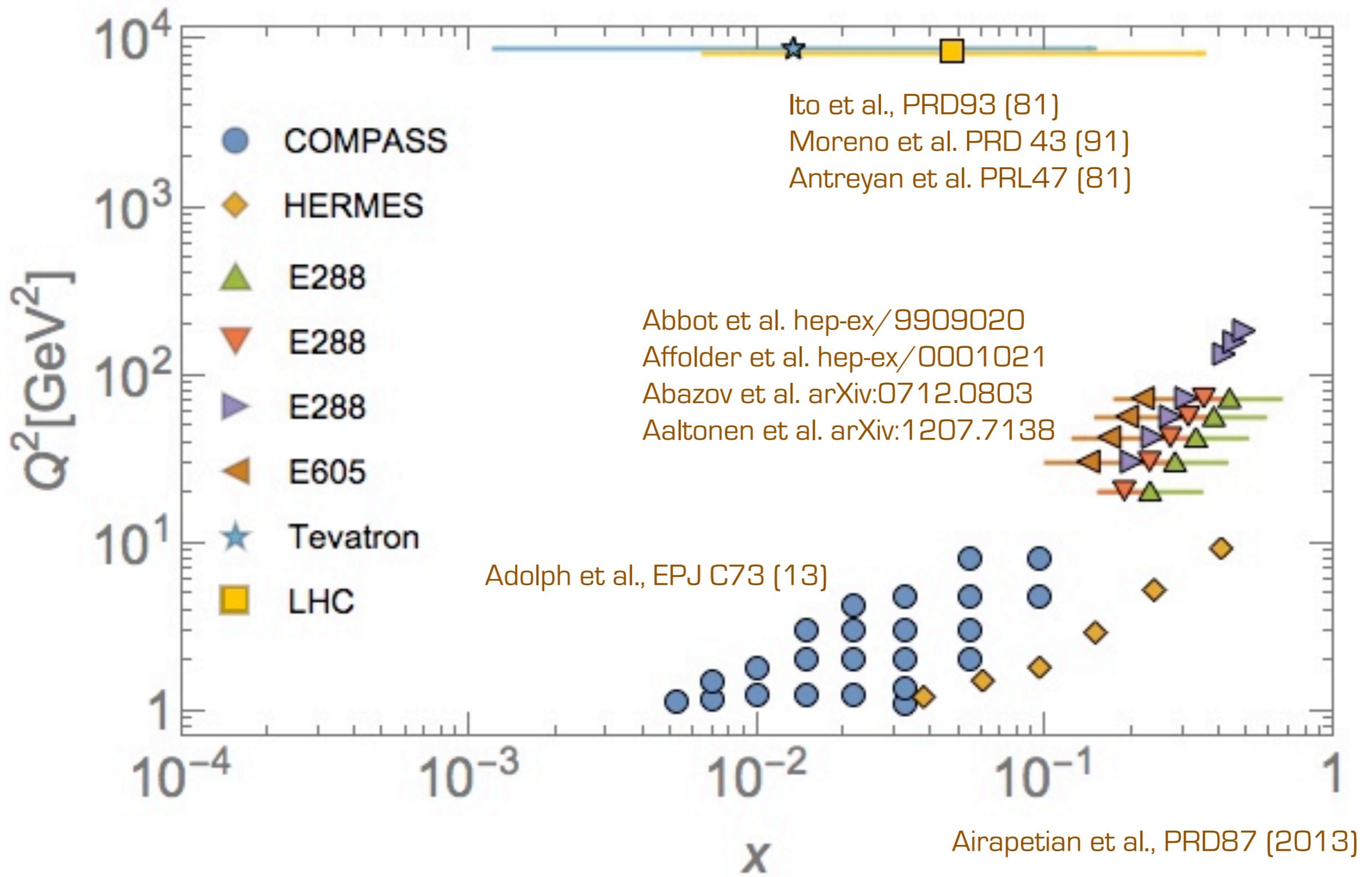
Z Production

90
data points

Data region



Data region



An almost global fit

	Framework	HERMES	COMPASS	DY	Z production	N of points
Pavia 2017 (+ JLab)	LO-NLL	✓	✓	✓	✓	8059

[JHEP06(2017)081]

Summary of results

Total number of data points: 8059

Total number of free parameters: 11

→ 4 for TMD PDFs → 6 for TMD FFs
→ 1 for TMD evolution

$$\chi^2/d.o.f. = 1.55 \pm 0.05$$

χ^2/dof

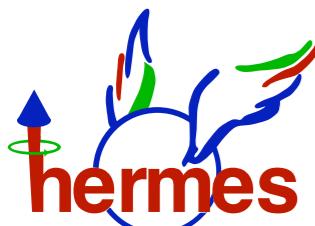
4.83

2.47

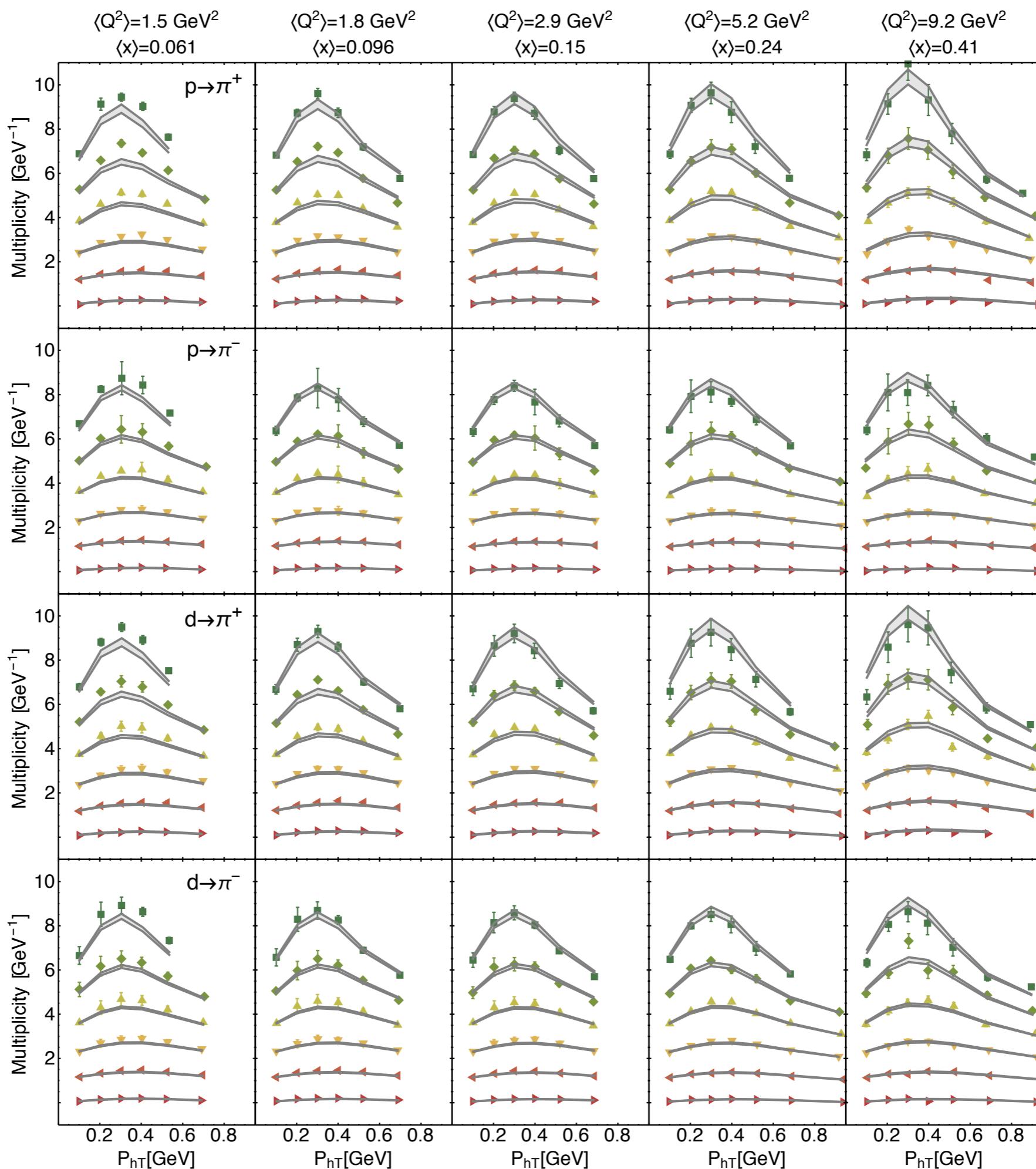
3.46

2.00

Hermes data pion production

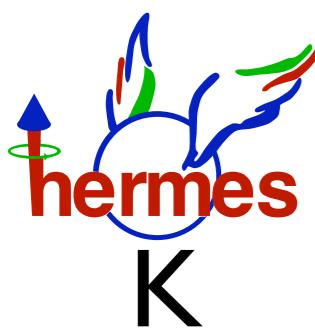
 π

- $\langle z \rangle = 0.24$ (offset=5)
- $\langle z \rangle = 0.28$ (offset=4)
- $\langle z \rangle = 0.34$ (offset=3)
- $\langle z \rangle = 0.43$ (offset=2)
- $\langle z \rangle = 0.54$ (offset=1)
- $\langle z \rangle = 0.70$ (offset=0)



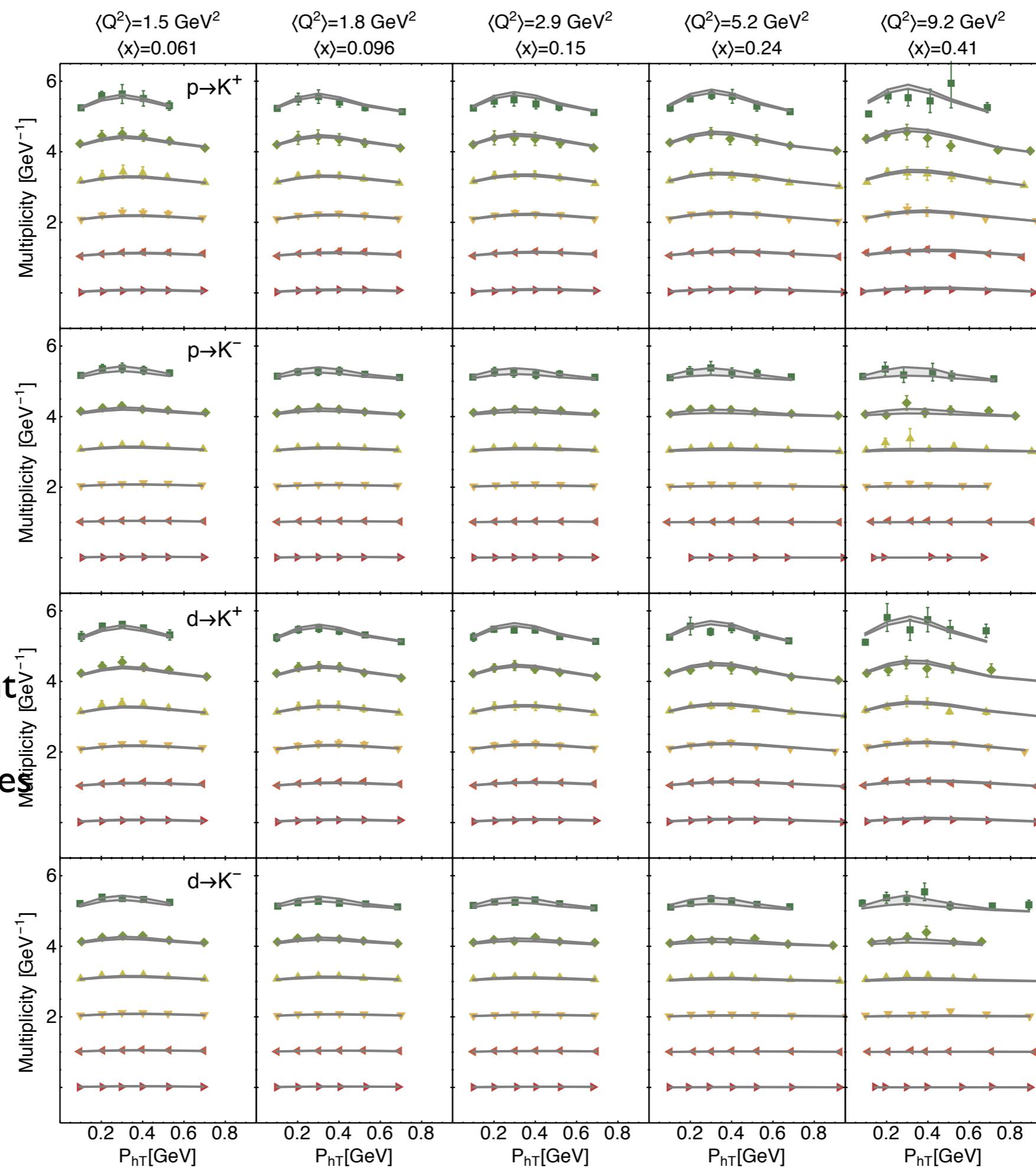
19

Hermes data kaon production



better agreement
than pions:
larger uncertainties
from FFs

- $\langle z \rangle = 0.24$ (offset=5)
- ◆ $\langle z \rangle = 0.28$ (offset=4)
- ▲ $\langle z \rangle = 0.34$ (offset=3)
- ▼ $\langle z \rangle = 0.43$ (offset=2)
- ◀ $\langle z \rangle = 0.54$ (offset=1)
- ▶ $\langle z \rangle = 0.70$ (offset=0)



χ^2/dof

0.91

0.82

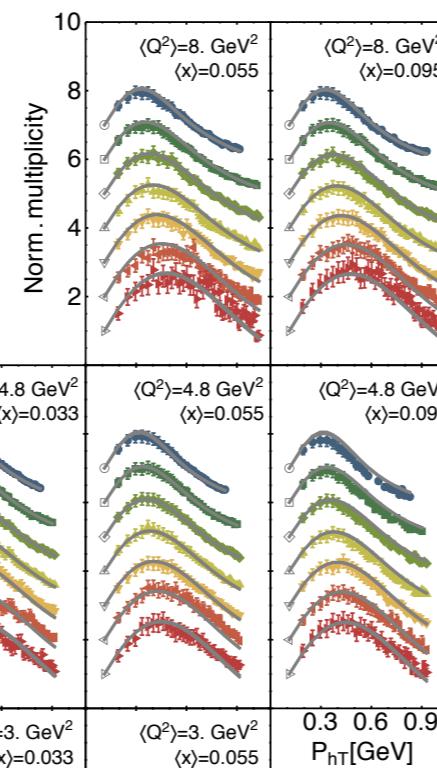
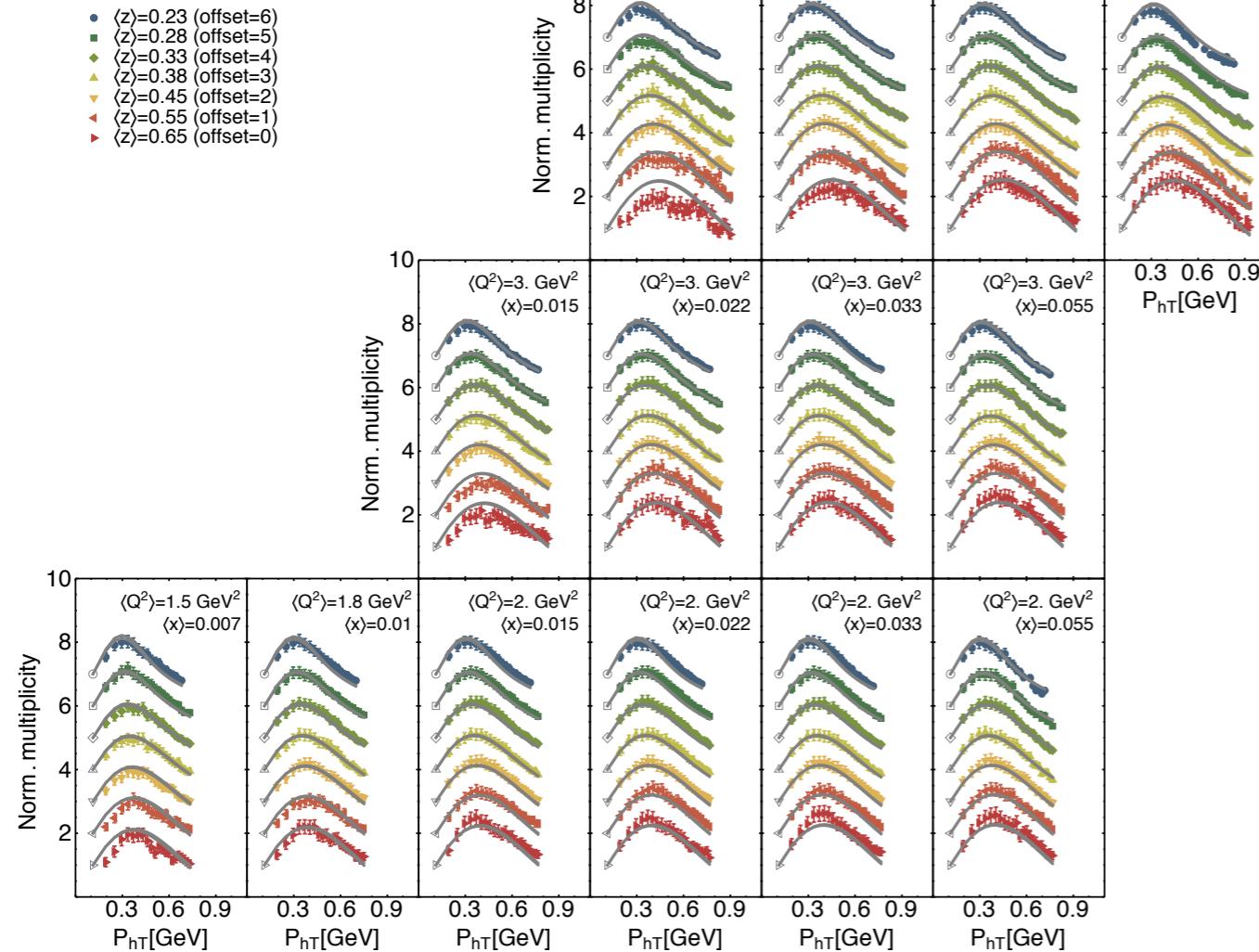
1.31

2.54

20

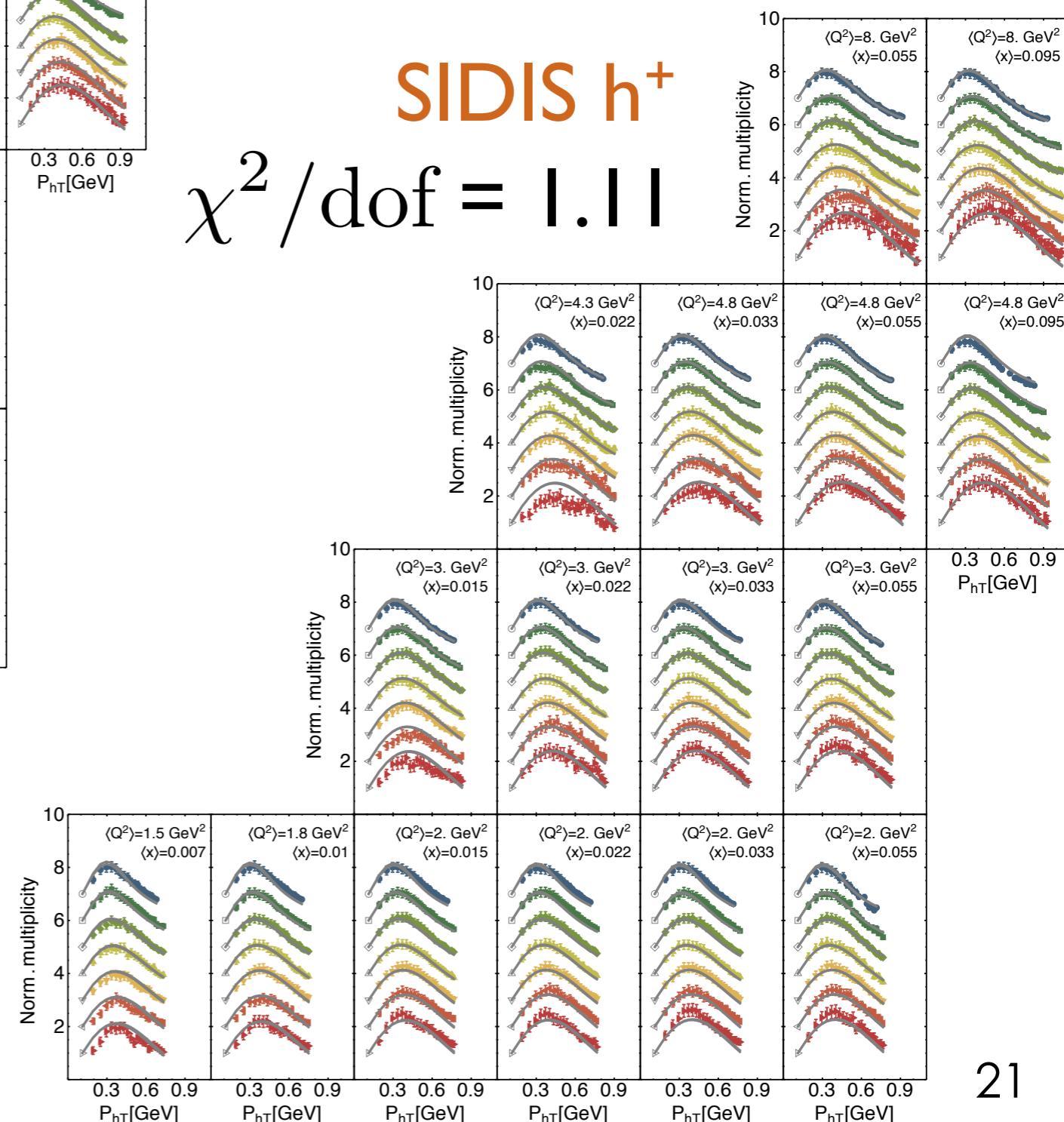
SIDIS h-

$$\chi^2/\text{dof} = 1.61$$



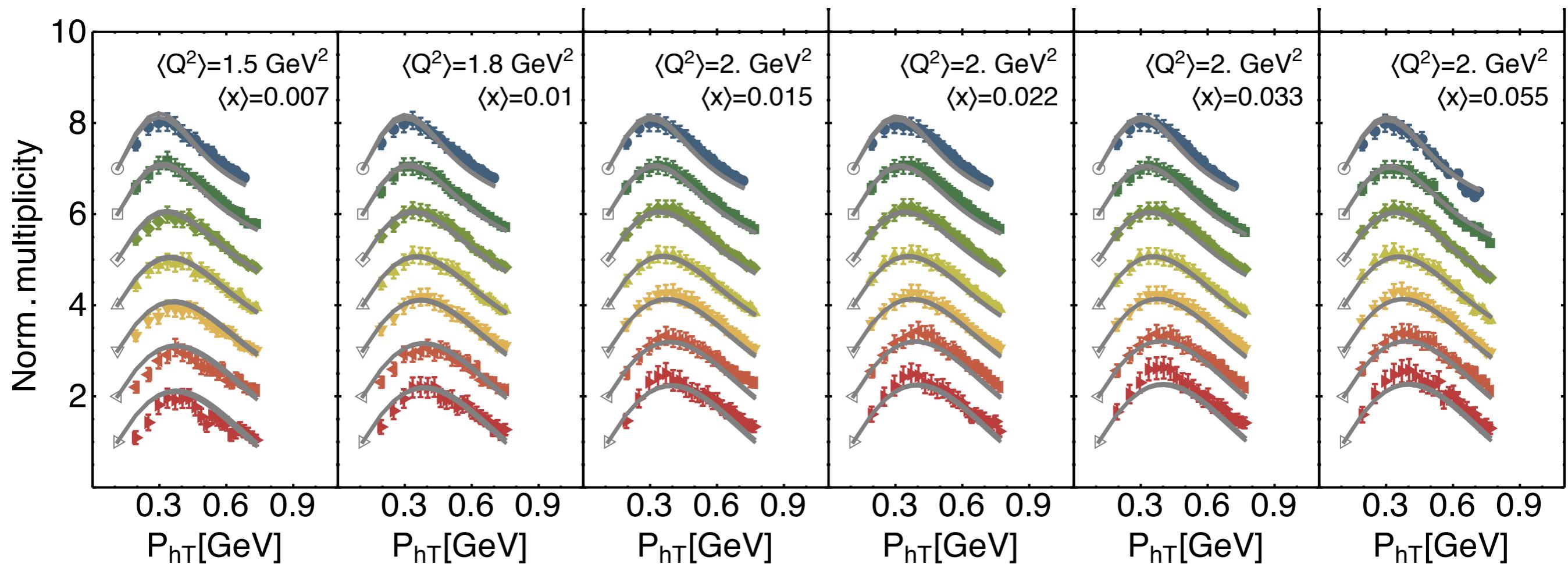
SIDIS h+

$$\chi^2/\text{dof} = 1.11$$



COMPASS data

SIDIS h^+



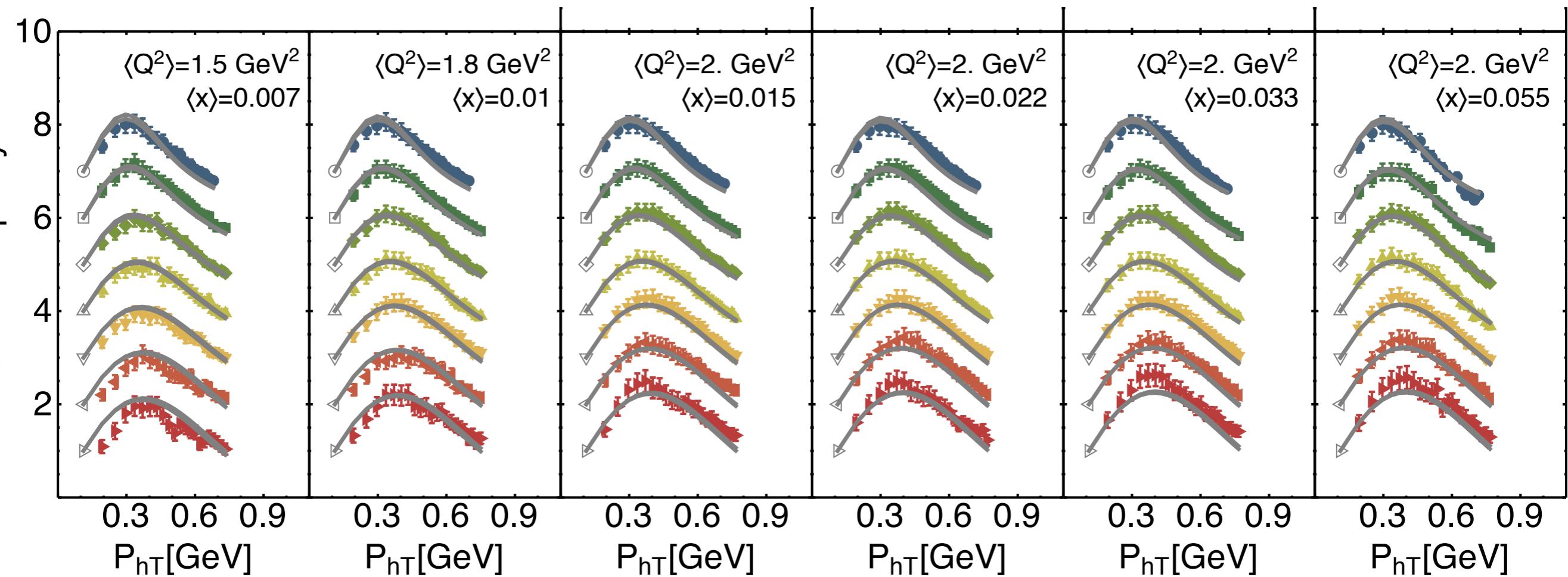
to avoid known problems
with Compass data normalization:

Observable

$$\frac{m_N^h(x, z, P_{hT}^2, Q^2)}{m_N^h(x, z, \min[P_{hT}^2], Q^2)}$$

COMPASS data

SIDIS h^+



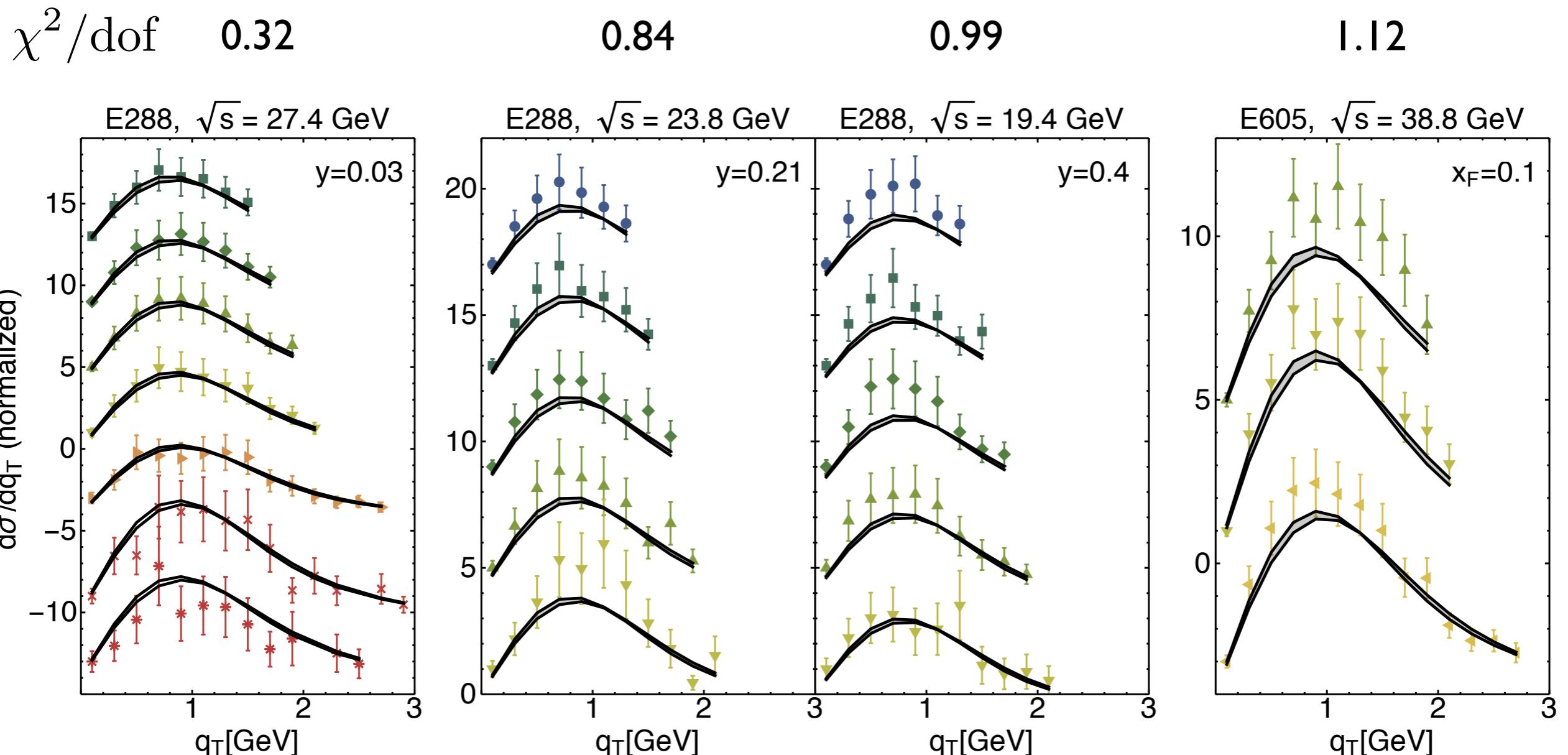
Revised Data:

[Phys.Rev. D97 (2018)
no.3, 032006]

Observable:

$$\frac{m_N^h(x, z, P_{hT}^2, Q^2)}{m_N^h(x, z, \min[P_{hT}^2], Q^2)}$$

Drell-Yan data



Q^2 Evolution: The peak is now at about 1 GeV, it was at 0.4 GeV for SIDIS

Z-boson production data

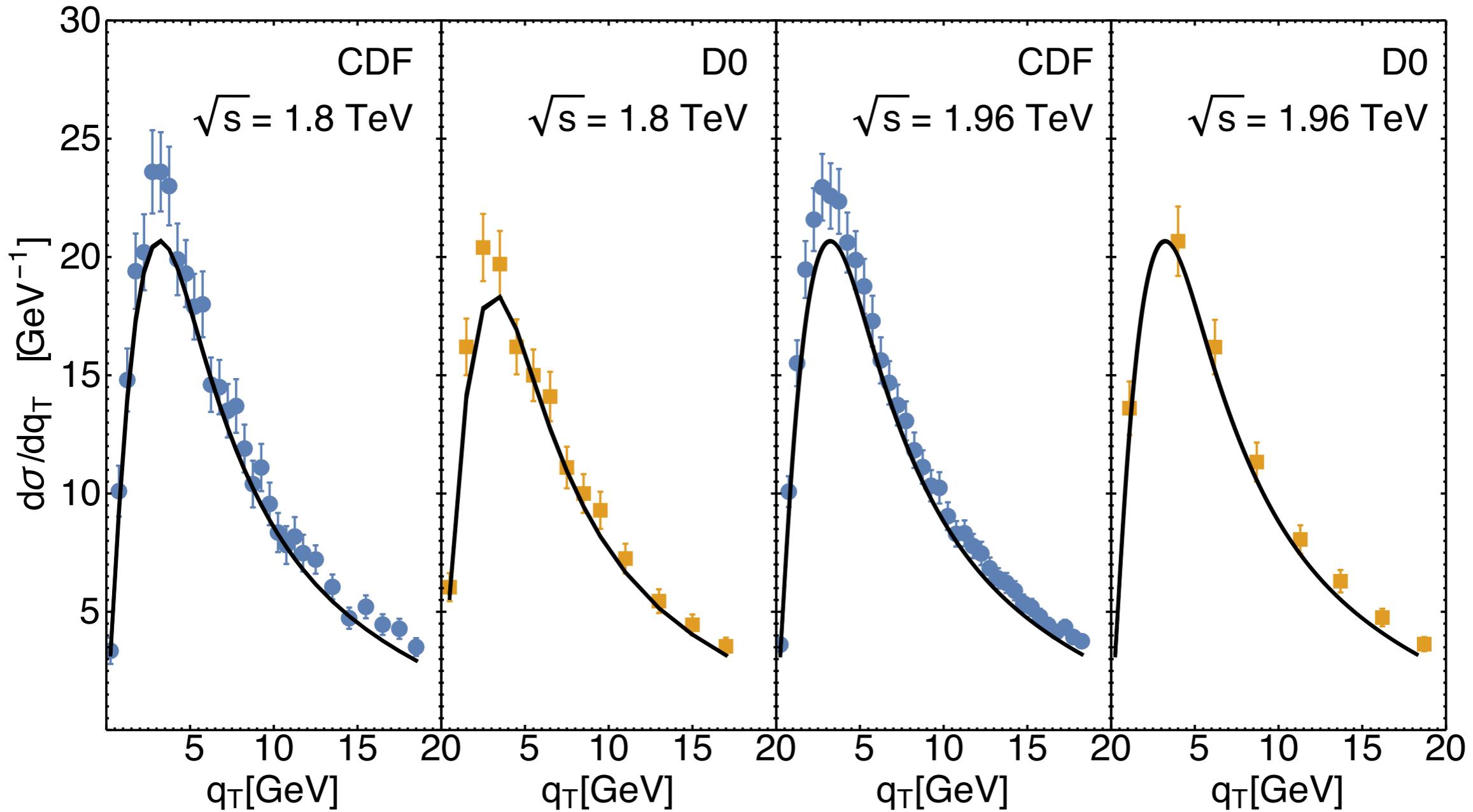
normalization : fixed from DEMS fit, different from exp.
(not really relevant for TMD parametrizations)

χ^2/dof 1.36

1.11

2.00

1.73



Q^2 Evolution: The peak is now at about 4 GeV



Stability of our results

Test of our default choices

How does the χ^2 of a single replica change if we modify them?

Original $\chi^2/\text{dof} = 1.51$

Normalization of HERMES data as done for COMPASS:

$\chi^2/\text{dof} = 1.27$

Parametrizations for collinear PDFs (NLO GJR 2008 default choice):
NLO MSTW 2008 (1.84), NLO CJ12 (1.85)

More stringent cuts (TMD factorization better under control)

$\chi^2/\text{dof} \rightarrow 1$

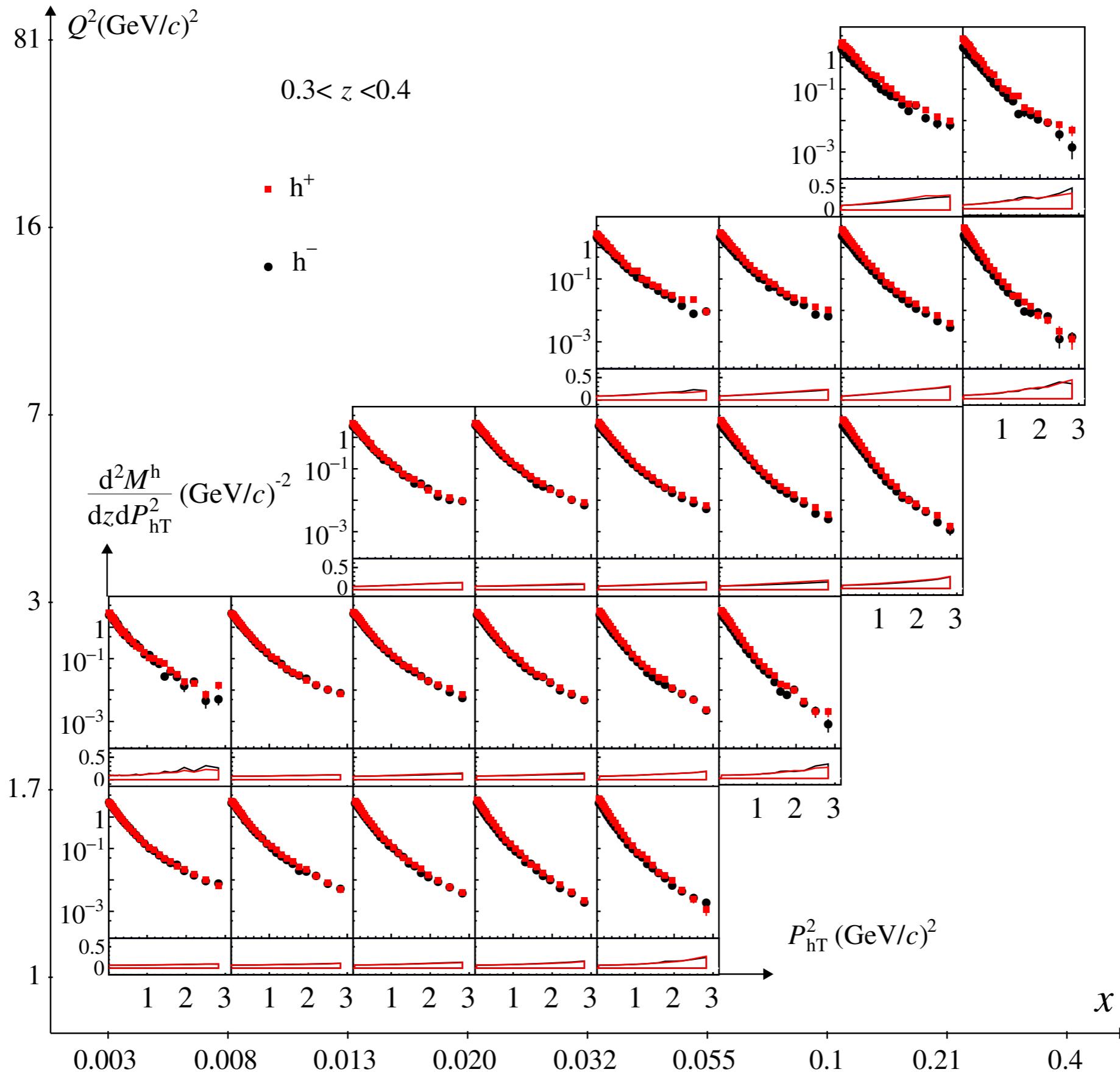
Ex: $Q^2 > 1.5 \text{ GeV}^2$; $0.25 < z < 0.6$; $\text{PhT} < 0.2Qz \Rightarrow \chi^2/\text{dof} = 1.02$ (477 bins)

Analysis of revised SIDIS data from COMPASS

[Phys.Rev. D97 (2018) no.3, 032006]



Revised Compass Data



Revised Compass Data: binning

[Eur. Phys. J. C (2013) 73:2531]

Bin	x_{bj}^{min}	x_{bj}^{max}	$\langle x_{bj} \rangle$	Q^2_{min}	Q^2_{max}	$\langle Q^2 \rangle$					
1	0.0045	0.0060	0.0052	1.0	1.25	1.11	13	0.0250	0.0350	0.0295	1.0
2	0.0060	0.0080	0.0070	1.0	1.30	1.14	14	0.0250	0.0400	0.0316	1.2
3	0.0060	0.0080	0.0070	1.3	1.70	1.48	15	0.0250	0.0400	0.0318	1.5
4	0.0080	0.0120	0.0099	1.0	1.50	1.22	16	0.0250	0.0400	0.0319	2.5
5	0.0080	0.0120	0.0099	1.5	2.10	1.76	17	0.0250	0.0400	0.0323	3.5
6	0.0120	0.0180	0.0148	1.0	1.50	1.22	18	0.0400	0.0500	0.0447	1.5
7	0.0120	0.0180	0.0148	1.5	2.50	1.92	19	0.0400	0.0700	0.0533	2.5
8	0.0120	0.0180	0.0150	2.5	3.50	2.90	20	0.0400	0.0700	0.0536	3.5
9	0.0180	0.0250	0.0213	1.0	1.50	1.23	21	0.0400	0.0700	0.0550	6.0
10	0.0180	0.0250	0.0213	1.5	2.50	1.92	22	0.0700	0.1200	0.0921	3.5
11	0.0180	0.0250	0.0213	2.5	3.50	2.94	23	0.0700	0.1200	0.0932	6.0
12	0.0180	0.0250	0.0216	3.5	5.00	4.07					10.0
											7.57

z bins	\rightarrow								
0.2÷0.25	0.25÷0.3	0.3÷0.35	0.35÷0.4	0.4÷0.5	0.5÷0.6	0.6÷0.7	0.7÷0.8		

[Phys.Rev. D97 (2018) no.3, 032006]

TABLE I. Bin limits for the four-dimensional binning in x , Q^2 , z and P_{hT}^2 .

Bin limits									
x	0.003	0.008	0.013	0.02	0.032	0.055	0.1	0.21	0.4
Q^2 (GeV/c^2)	1.0	1.7	3.0	7.0	16	81			
z	0.2	0.3	0.4	0.6	0.8				
P_{hT}^2 (GeV/c^2)	0.02	0.04	0.06	0.08	0.10	0.12	0.14	0.17	0.196
	0.23	0.27	0.30	0.35	0.40	0.46	0.52	0.60	0.68
	0.76	0.87	1.00	1.12	1.24	1.38	1.52	1.68	1.85
	2.05	2.35	2.65	3.00					

Number of experimental data

Same kinematical cuts in x, Q^2, z, Ph_T

Same data for DY 203

Z 90

SIDIS eN 1514

Total: 3931 data



SIDIS μN
2124
data points

Preliminary results

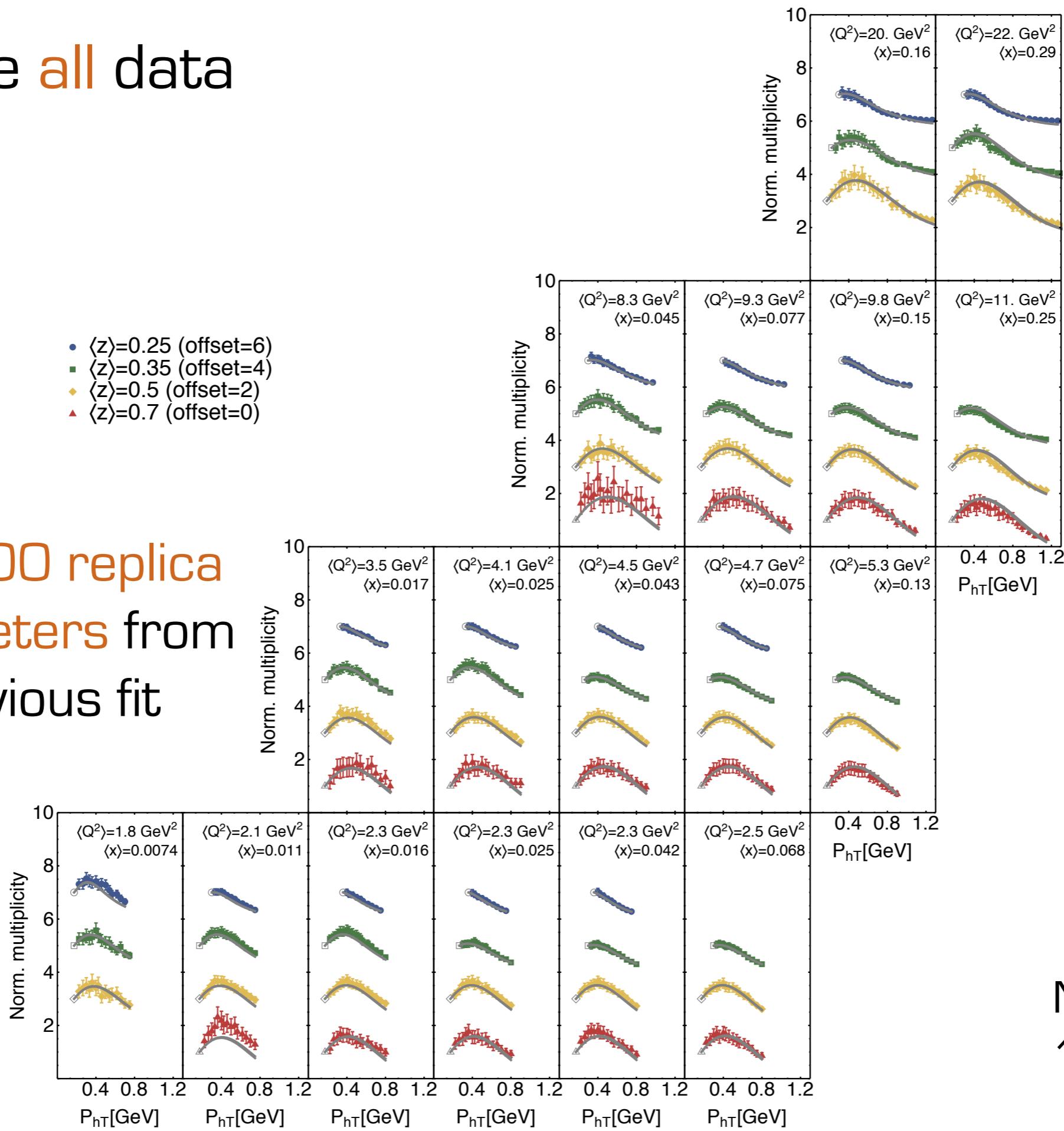


Include all data

SIDIS h^+

- $\langle z \rangle = 0.25$ (offset=6)
- $\langle z \rangle = 0.35$ (offset=4)
- $\langle z \rangle = 0.5$ (offset=2)
- $\langle z \rangle = 0.7$ (offset=0)

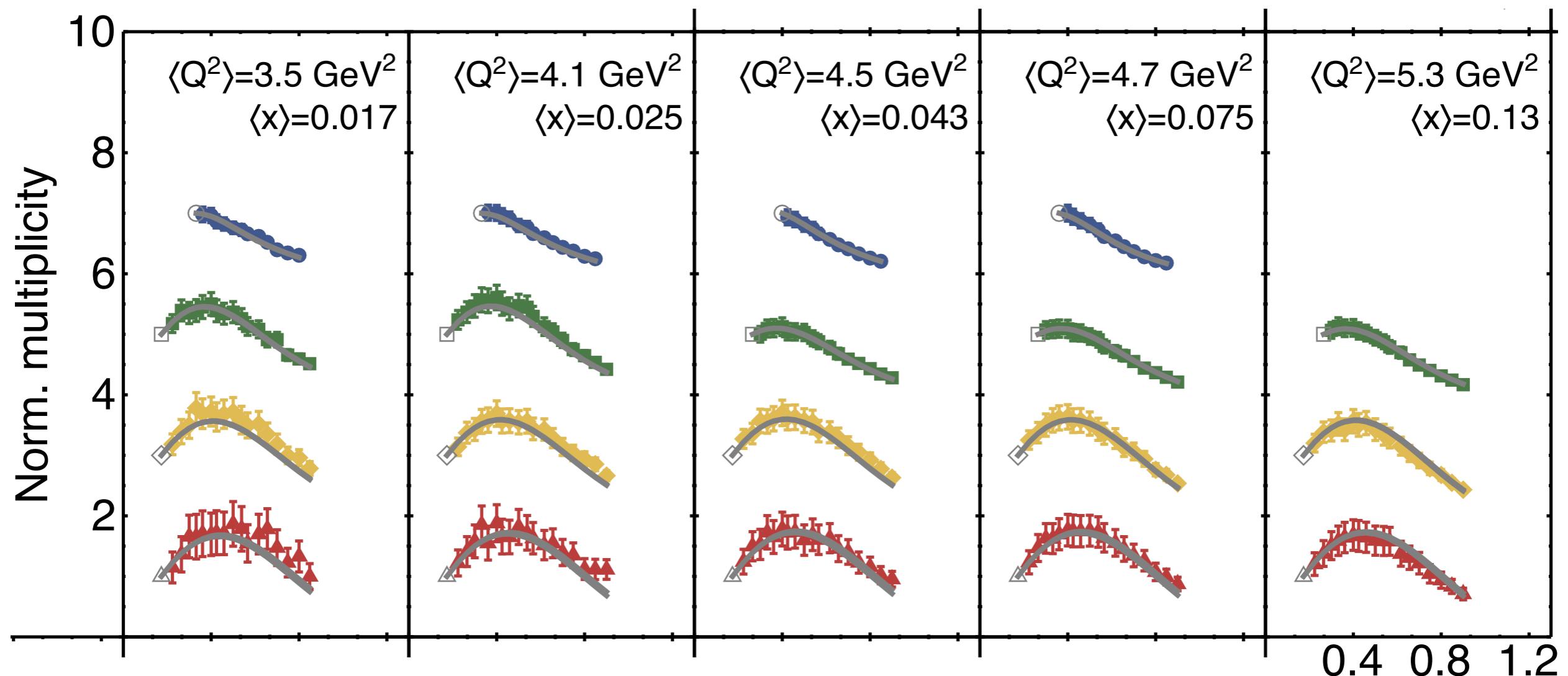
Use 200 replica
parameters from
previous fit



Normalized at
1st data point
of bin

Include all data

SIDIS h^+

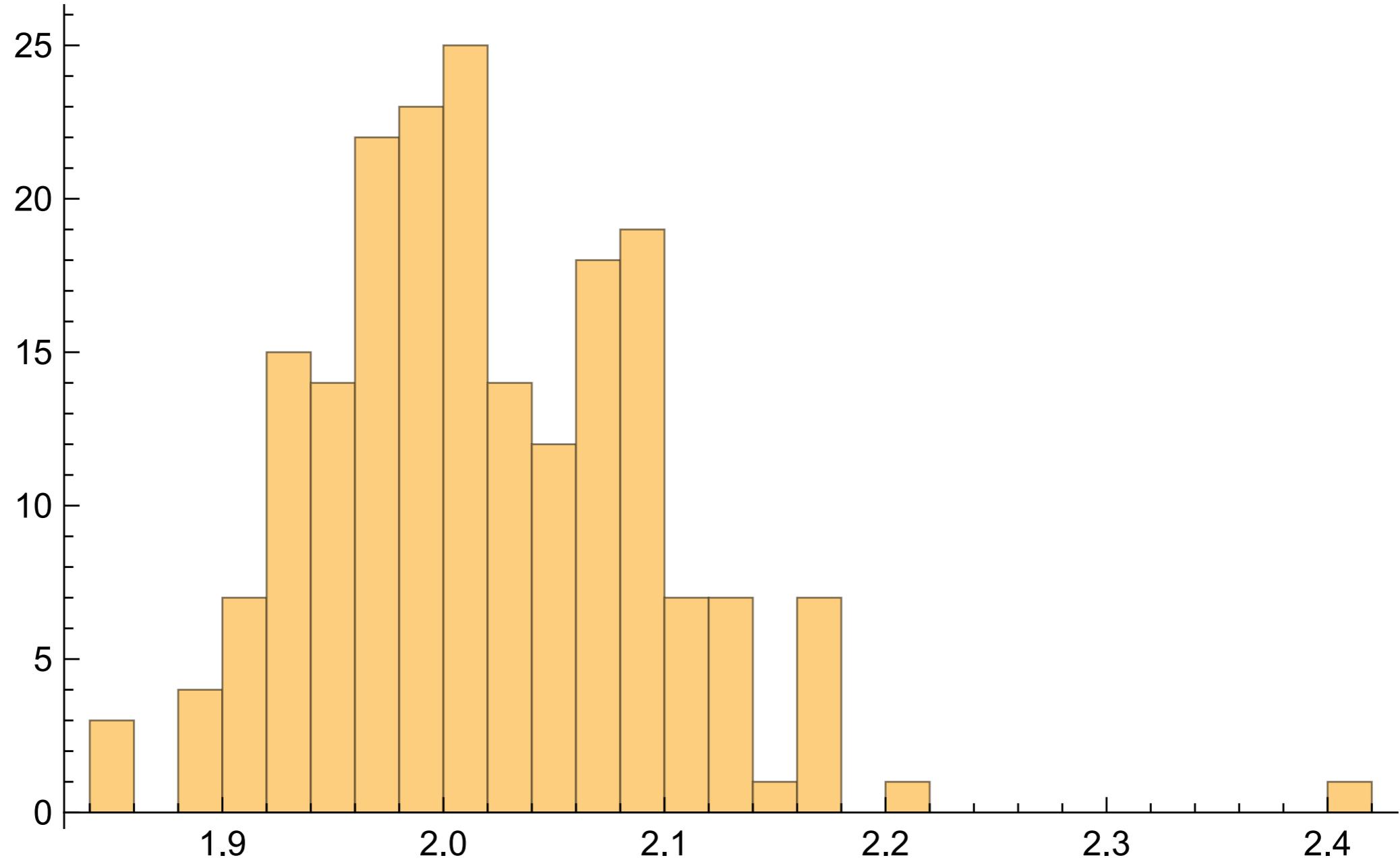


Use 200 replica
parameters from
previous fit

Normalized at
1st data point
of bin

Include all data

SIDIS h⁺



Use 200 replica
parameters from
previous fit

$$\chi^2/\text{dof} = 2.01$$

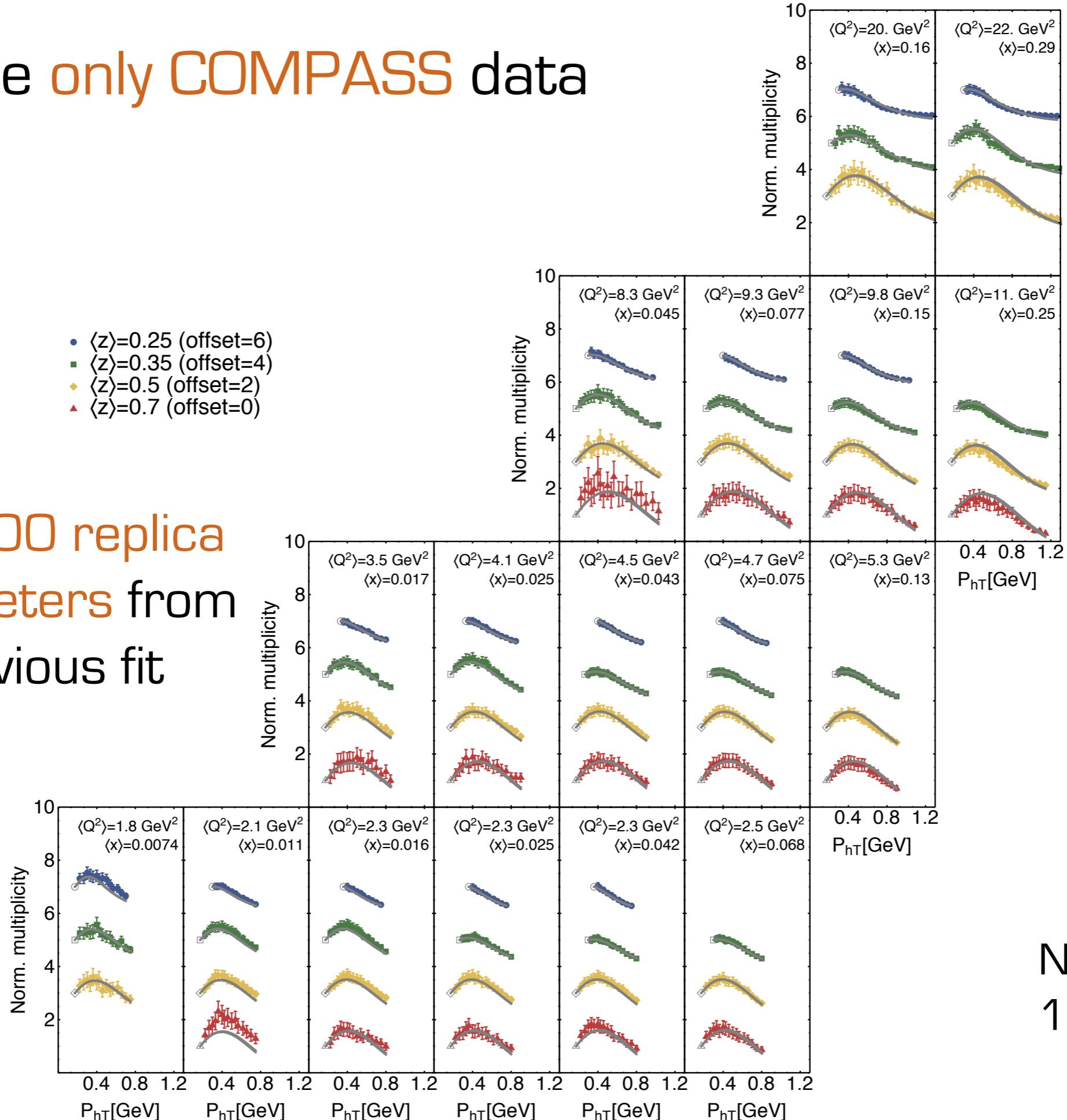
Normalized at
1st data point
of bin

Include only COMPASS data

SIDIS h^+

- $\langle z \rangle = 0.25$ (offset=6)
- $\langle z \rangle = 0.35$ (offset=4)
- $\langle z \rangle = 0.5$ (offset=2)
- $\langle z \rangle = 0.7$ (offset=0)

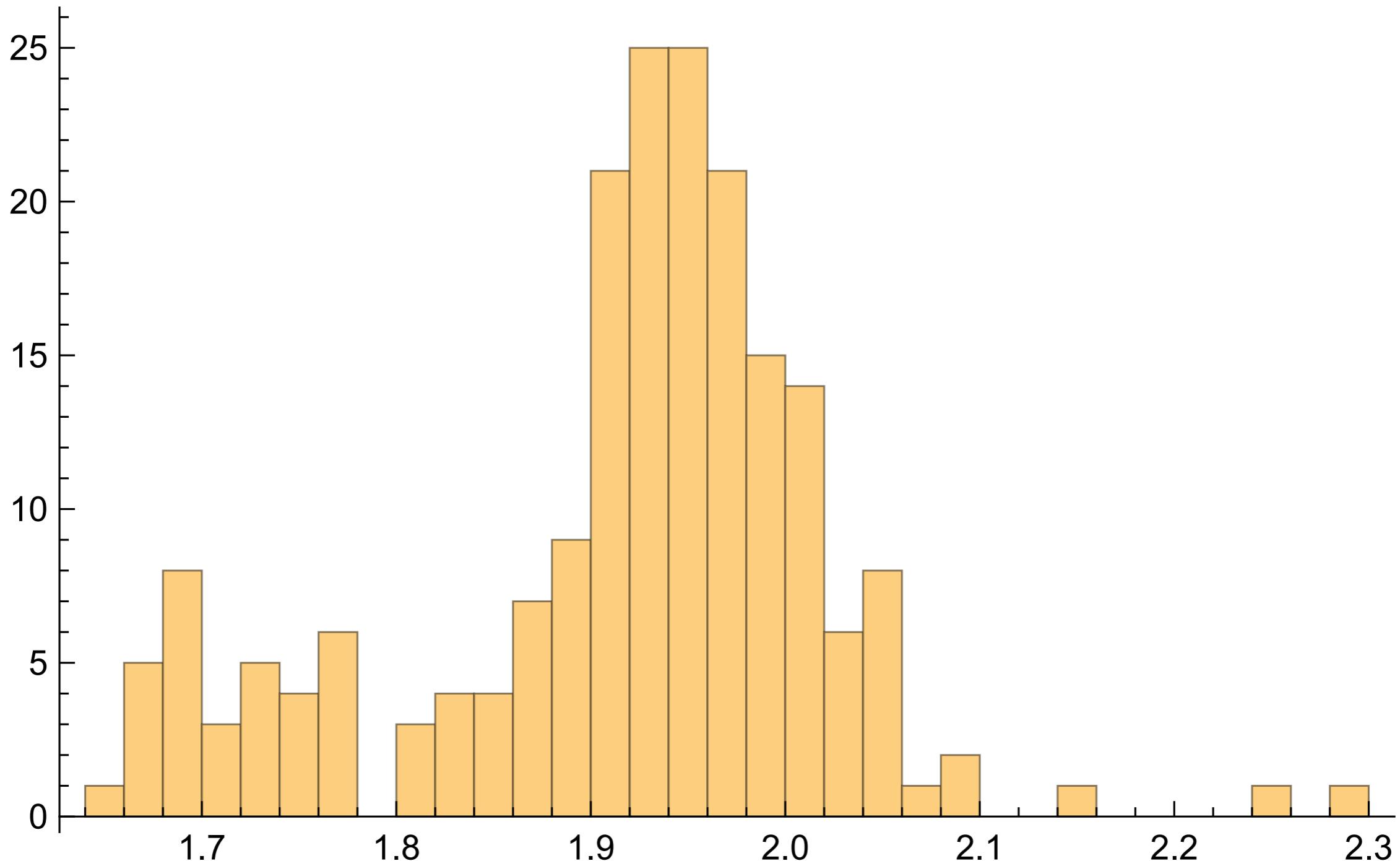
Use 200 replica
parameters from
previous fit



Normalized at
1st data point
of bin

Include only COMPASS data

SIDIS h^+



Use 200 replica
parameters from
previous fit

$\chi^2/\text{dof} = 1.91$

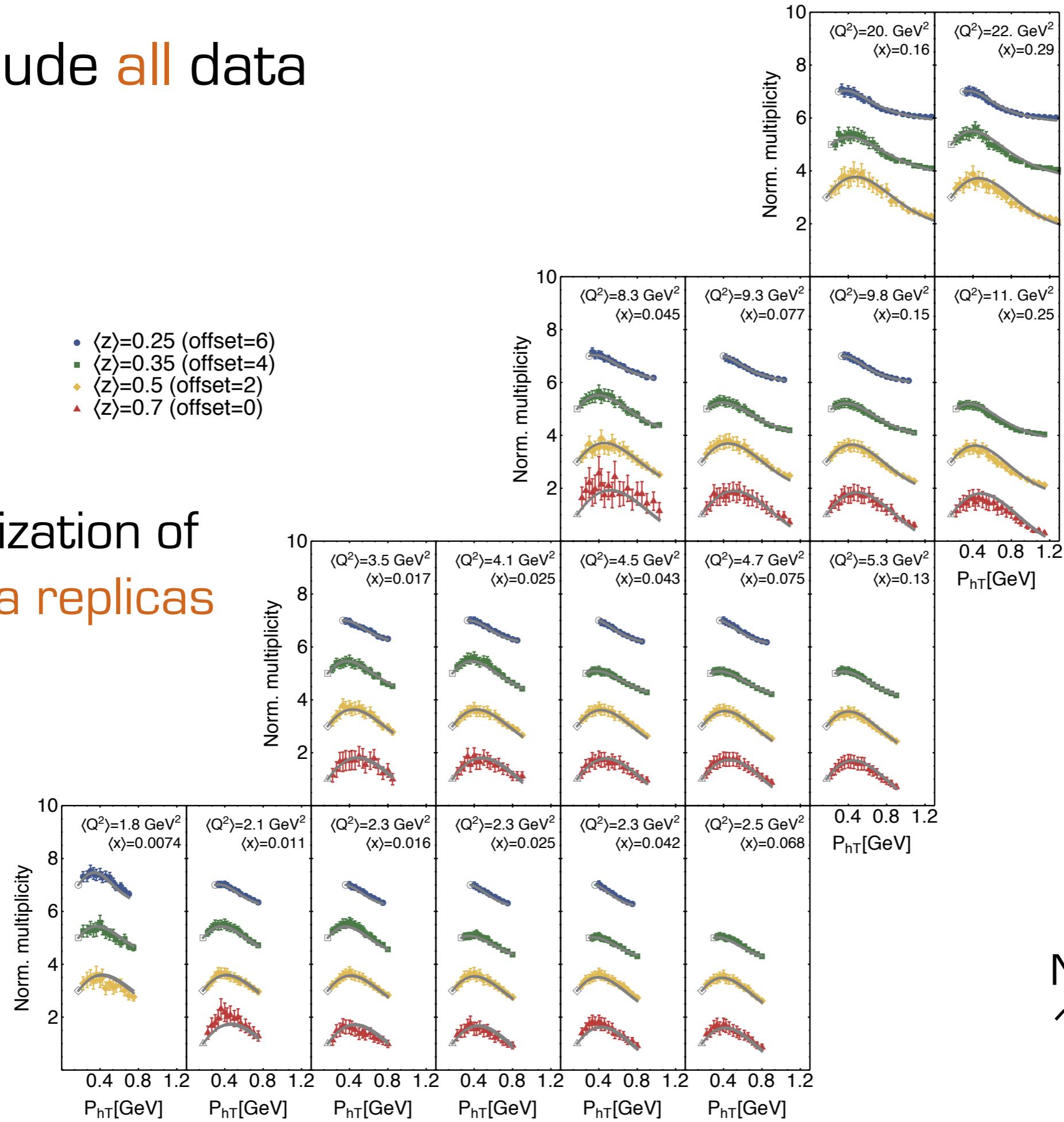
Normalized at
1st data point
of bin

Include all data

SIDIS h^+

- $\langle z \rangle = 0.25$ (offset=6)
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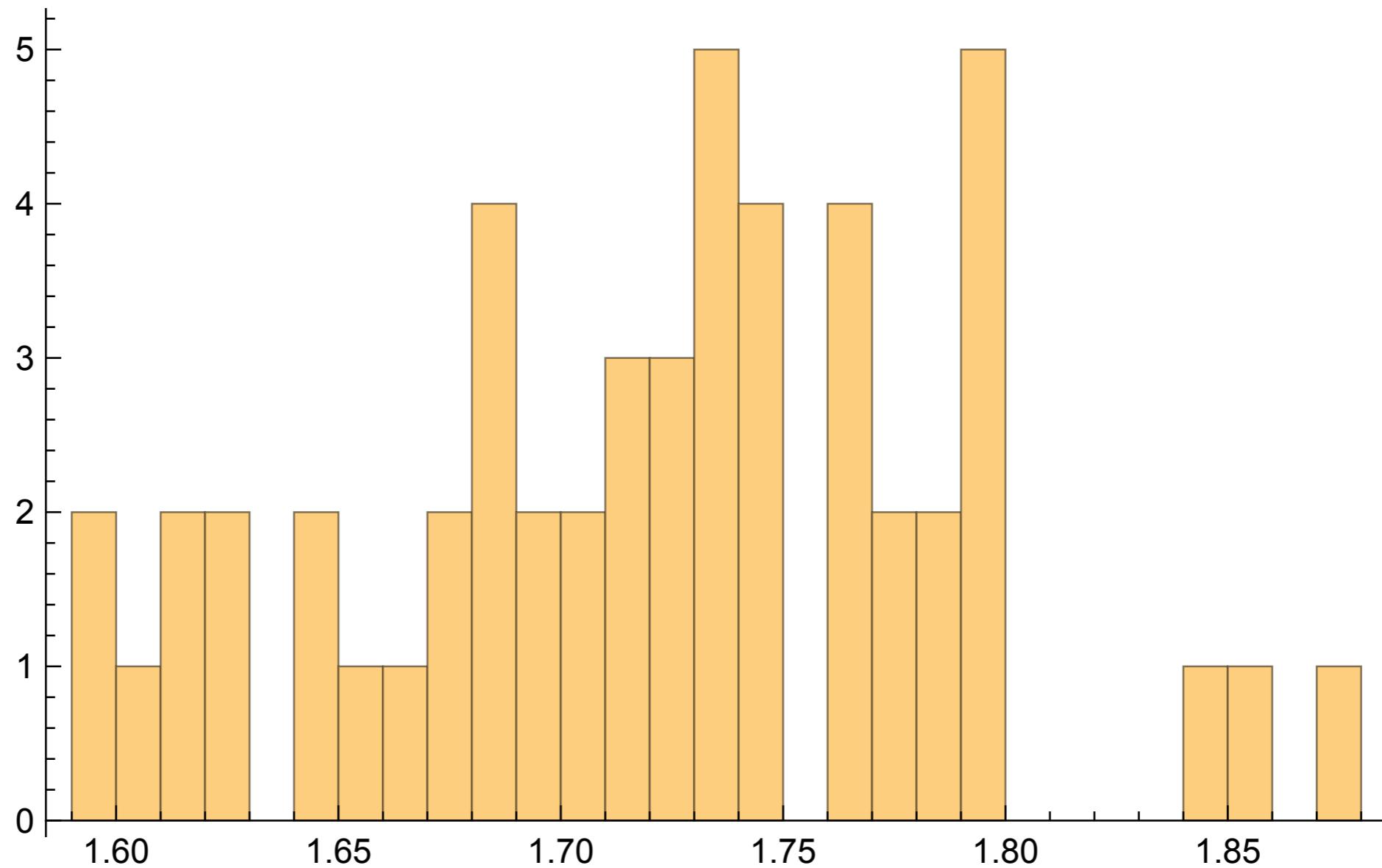
Minimization of 50 data replicas



Normalized at
1st data point
of bin

Include all data

SIDIS h⁺



Minimization of
50 data replicas

$\chi^2/\text{dof} = 1.71$

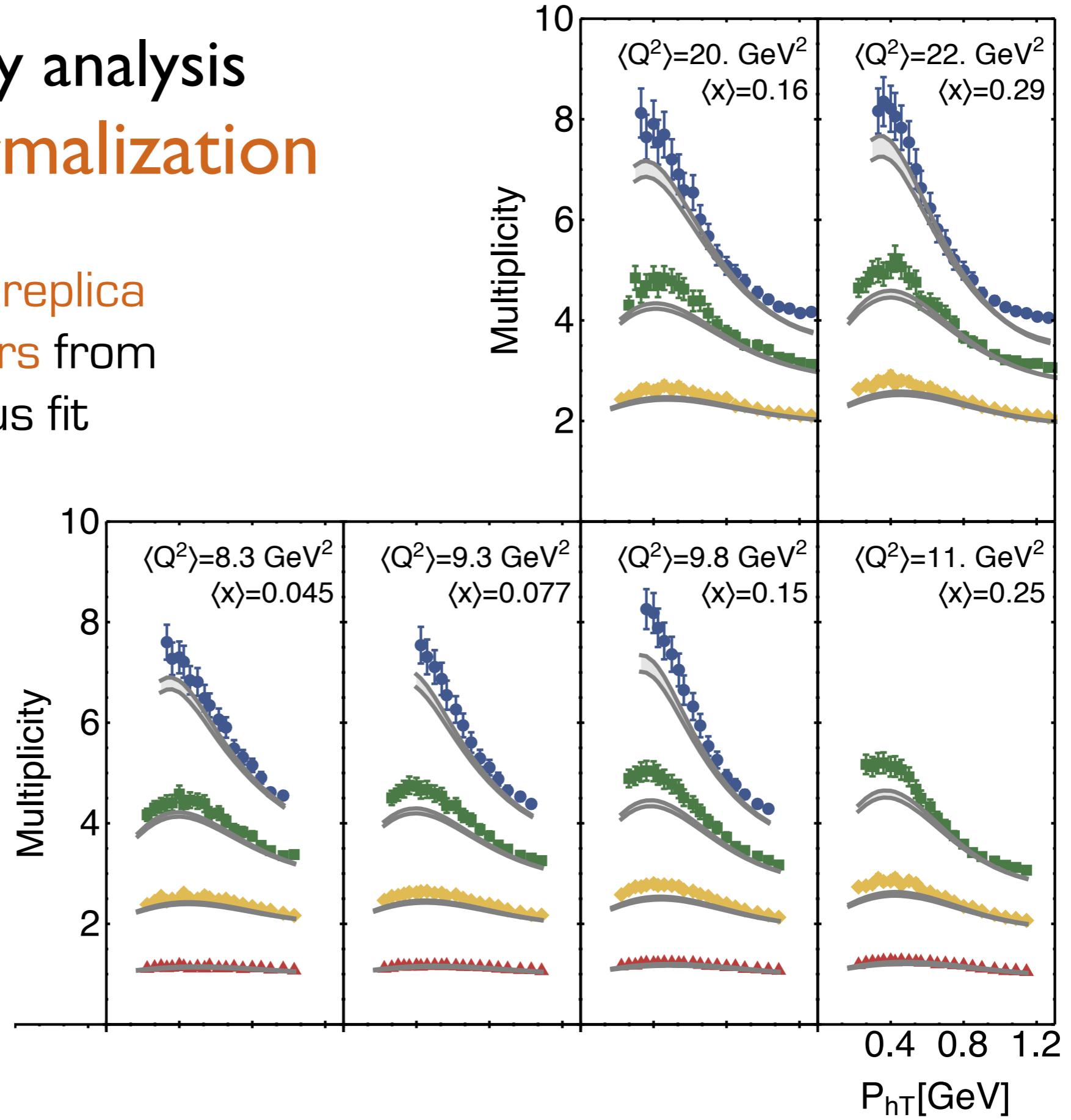
Normalized at
1st data point
of bin

Exploratory analysis without normalization

Use 200 replica
parameters from
previous fit

- $\langle z \rangle = 0.25$ (offset=4)
- $\langle z \rangle = 0.35$ (offset=3)
- ◊ $\langle z \rangle = 0.5$ (offset=2)
- ▲ $\langle z \rangle = 0.7$ (offset=1)

SIDIS h^+



Exploratory analysis without normalization

Use 200 replica
parameters from → $\chi^2/\text{dof} > 4$
previous fit

SIDIS h⁺

Exploratory analysis without normalization

Use 200 replica
parameters from
previous fit → $\chi^2/\text{dof} > 4$

Sensitive to z value

Less stable with regards to
kinematical cuts

...

SIDIS h⁺

Conclusions

For the first time we demonstrated that it is possible to fit simultaneously SIDIS, DY and Z boson

We extracted a reasonable functional form for TMD from more than 8000 data points

We tested the universality and applicability of the TMD framework and it works quite well
(most of the discrepancies come from normalization)

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We extracted a reasonable functional form for TMD from more than 8000 data points

We tested the universality and applicability of the TMD framework and it works quite well
(most of the discrepancies come from normalization)

Conclusions and open issues

For the first time we demonstrated that it is possible to fit simultaneously SIDIS, DY and Z boson

We extracted TMDs from more than 8000 data points

We tested the universality and applicability of the TMD framework and it works quite well

Revised Compass Data

- Reduced number of data points
- compatible with parameters obtained from previous analysis
- removing normalization requires further considerations

BACKUP

Best fit values

TMD PDFs	N_1 [GeV ²]	α	σ		λ [GeV ⁻²]	
All replicas	0.28 ± 0.06	2.95 ± 0.05	0.17 ± 0.02		0.86 ± 0.78	
Replica 105	0.285	2.98	0.173		0.39	
TMD FFs	N_3 [GeV ²]	β	γ	δ	λ_F [GeV ⁻²]	N_4 [GeV ²]
All replicas	0.21 ± 0.02	1.65 ± 0.49	2.28 ± 0.46	0.14 ± 0.07	5.50 ± 1.23	0.04 ± 0.01
Replica 105	0.212	2.10	2.52	0.094	5.29	0.04

TABLE XI: 68% confidence intervals of best-fit values for parametrizations of TMDs at $Q = 1$ GeV.

Flavor independent scenario:

$$N_1 = 0.28 \pm 0.06 \text{ GeV}^2$$

$$N_3 = 0.21 \pm 0.02 \text{ GeV}^2$$

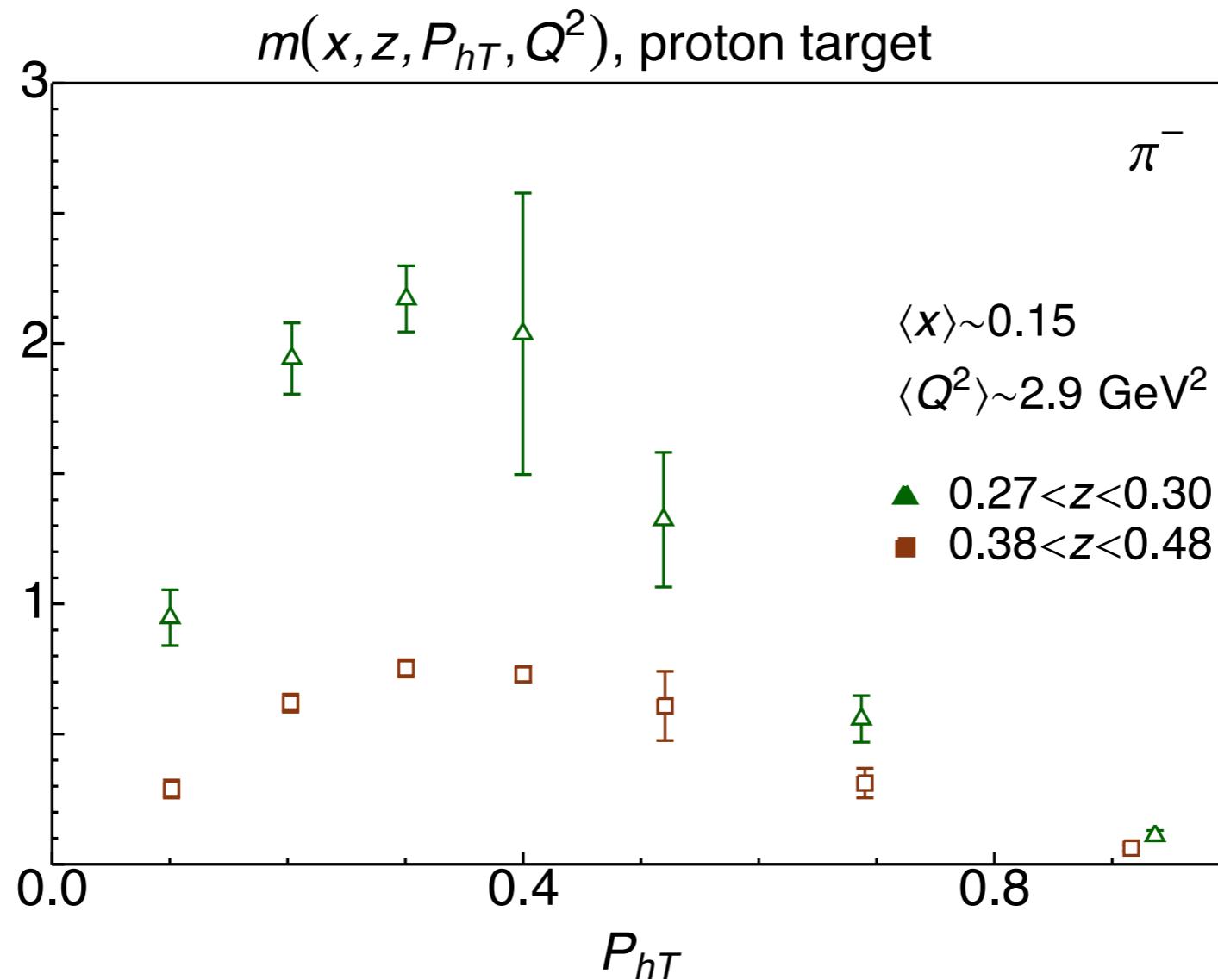
$$N_4 = 0.04 \pm 0.01 \text{ GeV}^2$$

$$g_2 = 0.13 \pm 0.01 \text{ GeV}^2$$

best value from 200 replicas

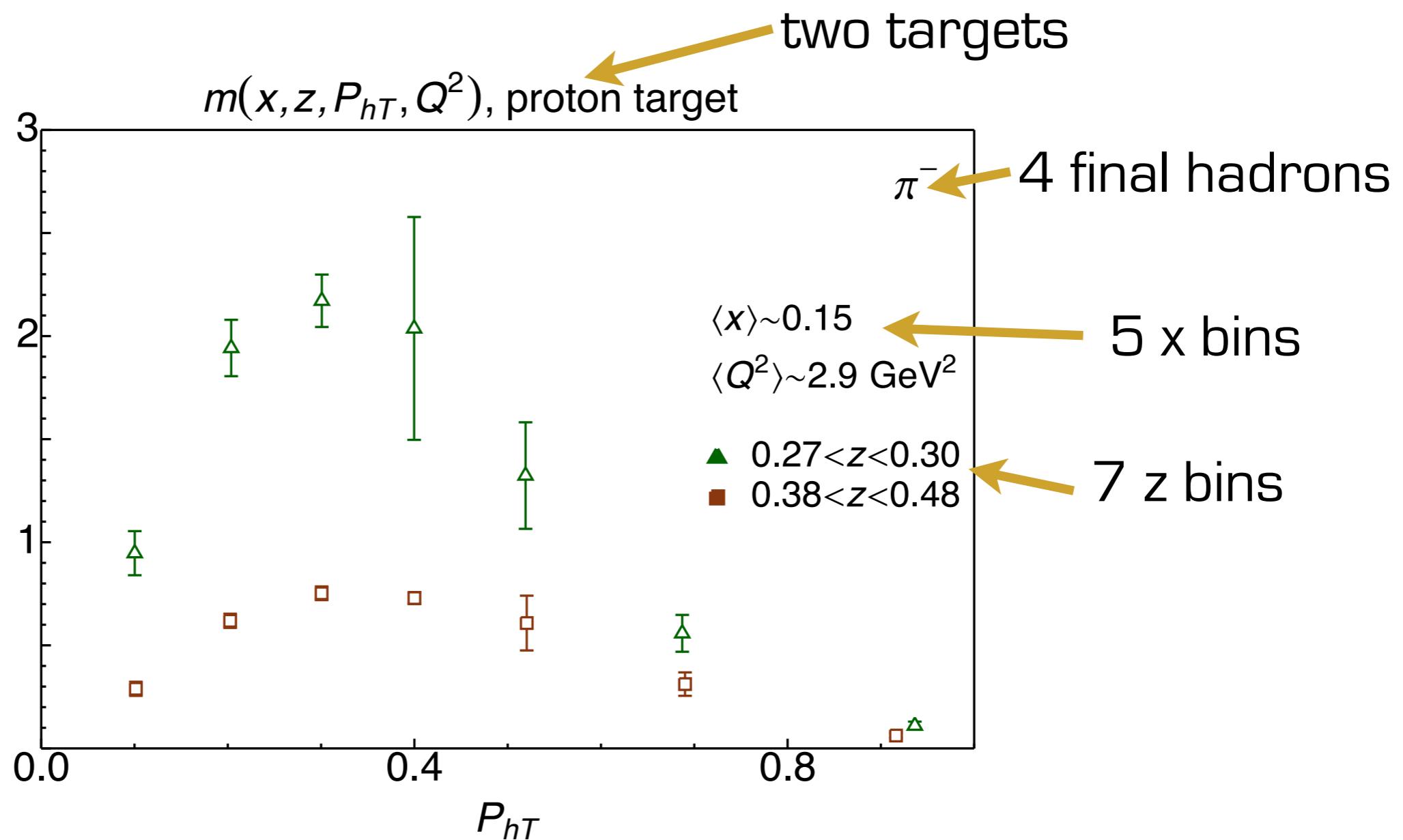
compatible with other extractions

The replica method



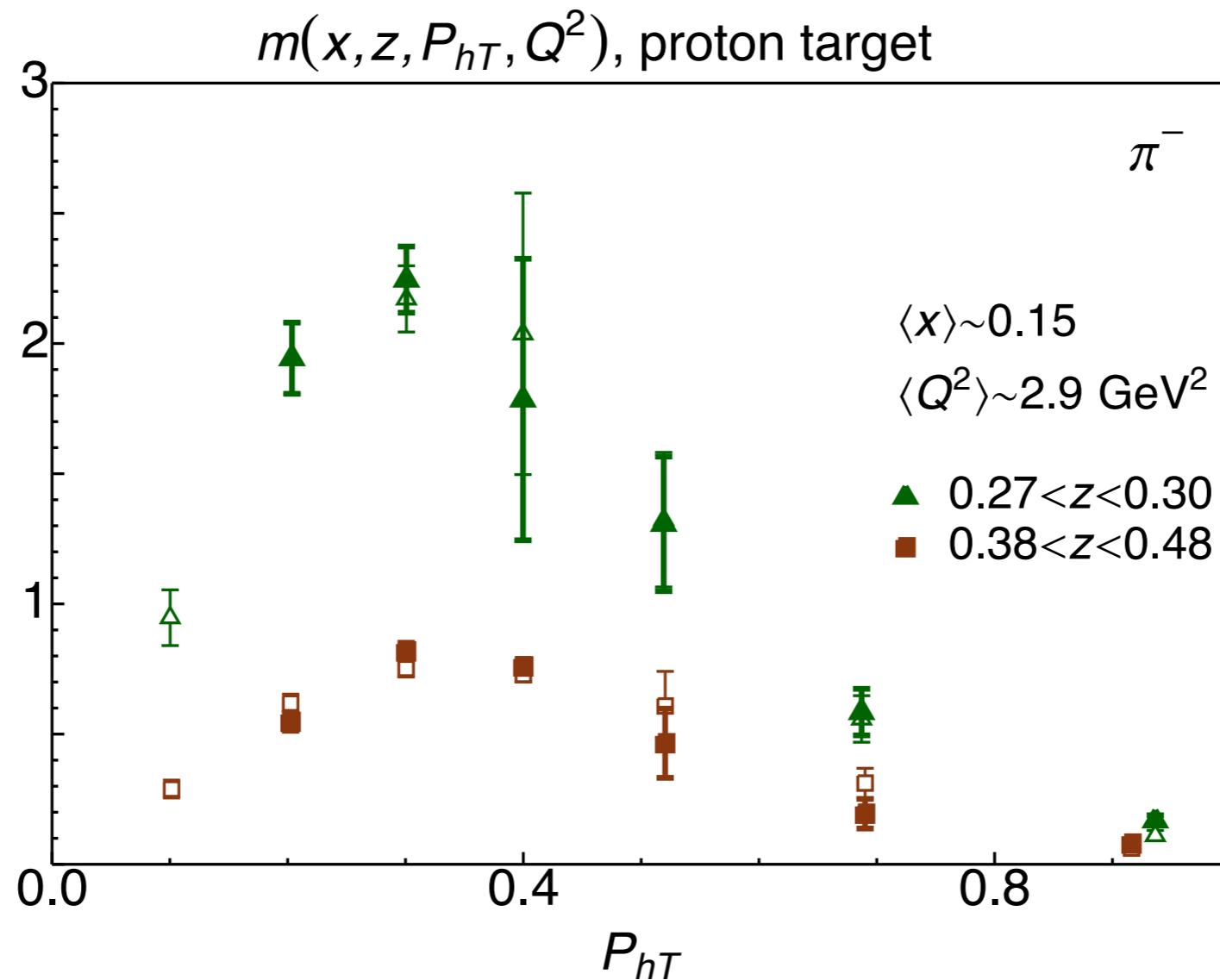
Example of original data

The replica method



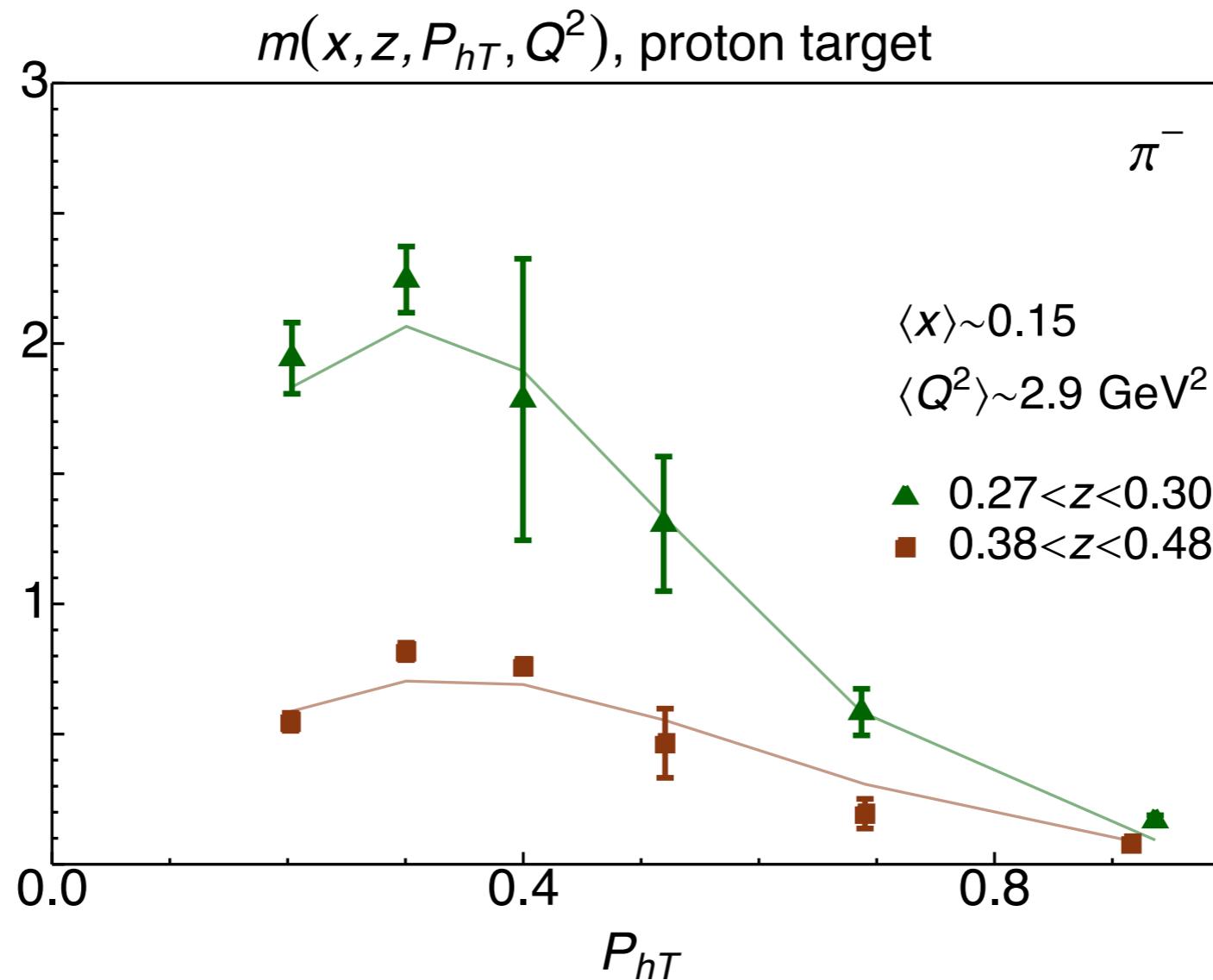
Example of original data

The replica method



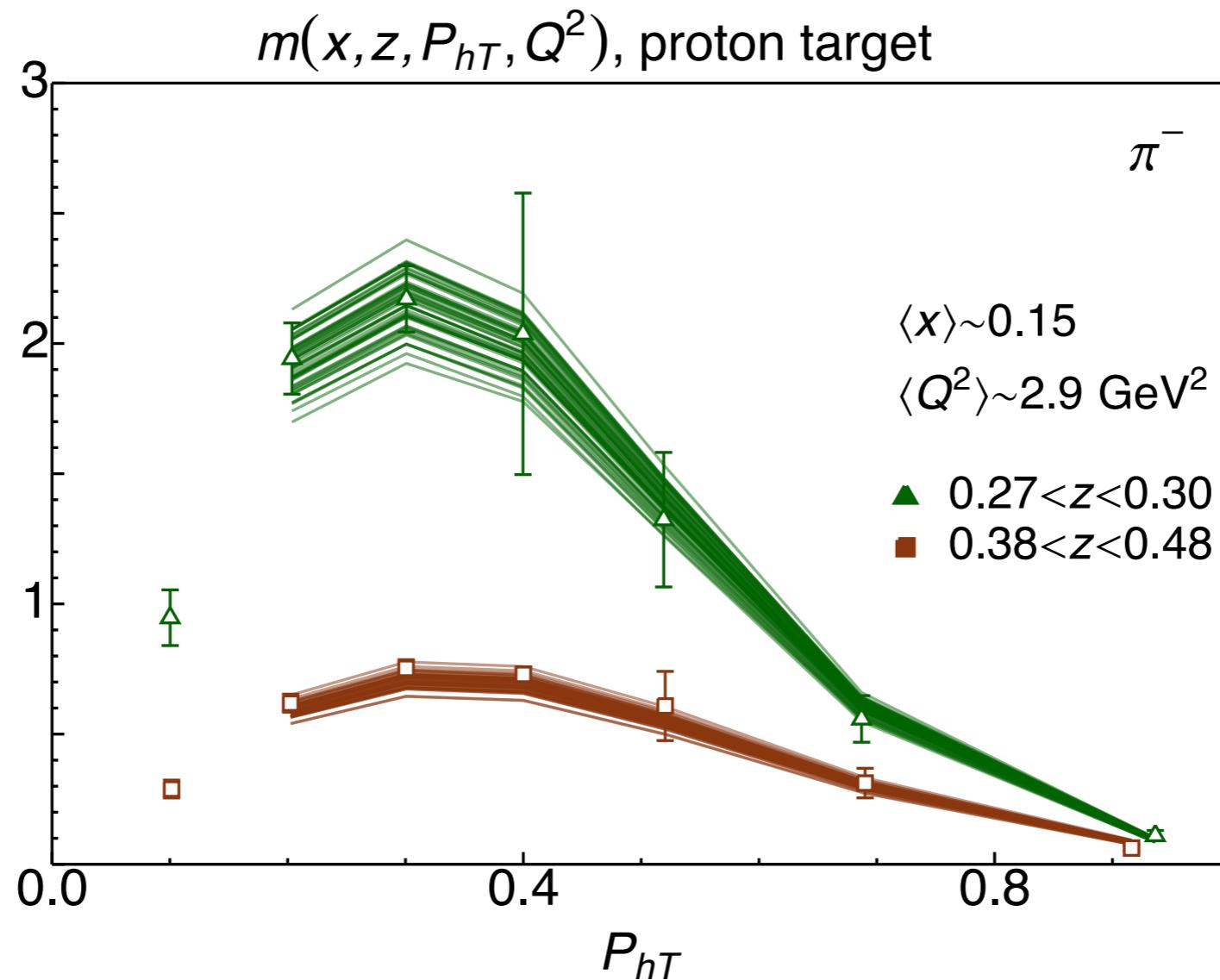
Data are replicated (with Gaussian distribution)

The replica method



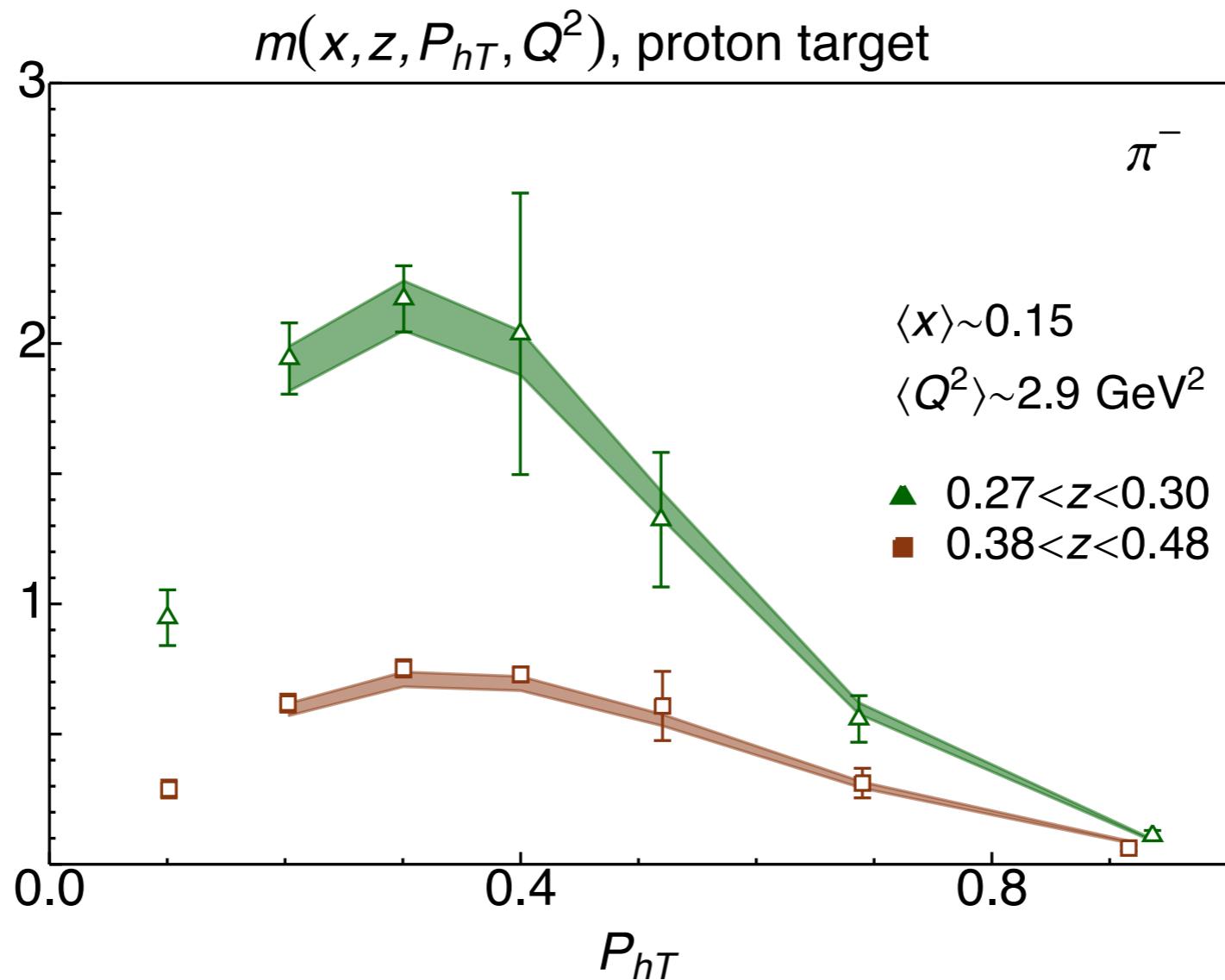
The fit is performed on the replicated data

The replica method



The procedure is repeated 200 times

The replica method



For each point, a central 68% confidence interval is identified

Previous fit studies

	Framework	HERMES	COMPASS	DY	Z production	N of points
KN 2006 hep-ph/0506225	LO-NLL	✗	✗	✓	✓	98
Pavia 2013 (+Amsterdam, Bilbao) arXiv:1309.3507	No evo (QPM)	✓	✗	✗	✗	1538
Torino 2014 (+JLab) arXiv:1312.6261	No evo (QPM)	✓ (separately)	✓ (separately)	✗	✗	576 (H) 6284 (C)
DEMS 2014 arXiv:1407.3311	NLO-NNLL	✗	✗	✓	✓	223
EIKV 2014 arXiv:1401.5078	LO-NLL	1 (x, Q^2) bin	1 (x, Q^2) bin	✓	✓	500 (?)
Pavia 2017 (+ JLab)	LO-NLL	✓	✓	✓	✓	8059

Data selection

SIDIS proton-target data		HERMES $p \rightarrow \pi^+$	HERMES $p \rightarrow \pi^-$	HERMES $p \rightarrow K^+$	HERMES $p \rightarrow K^-$
Reference					
Cuts		$Q^2 > 1.4 \text{ GeV}^2$	$0.2 < z < 0.7$	$P_{hT} < \text{Min}[0.2 Q, 0.6 Qz] + 0.5 \text{ GeV}$	
Points		188	186	187	185
Max. Q^2			9.2 GeV^2		
x range			$0.06 < x < 0.4$		
Notes					

Motivations behind kinematical cuts

TMD factorization ($P_{hT}/z \ll Q^2$)

Avoid target fragmentation (low z)
and exclusive contributions (high z)

Data selection

SIDIS
deuteron-target
data

	HERMES $D \rightarrow \pi^+$	HERMES $D \rightarrow \pi^-$	HERMES $D \rightarrow K^+$	HERMES $D \rightarrow K^-$	COMPASS $D \rightarrow h^+$	COMPASS $D \rightarrow h^-$
Cuts	$Q^2 > 1.4 \text{ GeV}^2$ $0.2 < z < 0.7$ $P_{hT} < \text{Min}[0.2 Q, 0.6 Qz] + 0.5 \text{ GeV}$					
Points	188	188	186	187	3024	3021
Max. Q^2		9.2 GeV^2			10 GeV^2	
x range		$0.06 < x < 0.4$			$0.006 < x < 0.12$	
Notes					Observable: $\frac{m_N^h(x, z, P_{hT}^2, Q^2)}{m_N^h(x, z, \text{Min}[P_{hT}^2], Q^2)}$	

to avoid problems
with Compass data normalization

Data selection

	E288 200	E288 300	E288 400	E605
Cuts	$q_T < 0.2 Q + 0.5 \text{ GeV}$			
Points	45	45	78	35
\sqrt{s}	19.4 GeV	23.8 GeV	27.4 GeV	38.8 GeV
Q range	4-9 GeV	4-9 GeV	5-9, 11-14 GeV	7-9, 10.5-18 GeV
Kin. var.	$y=0.4$	$y=0.21$	$y=0.03$	$-0.1 < x_F < 0.2$

Drell-Yan
data

	CDF Run I	D0 Run I	CDF Run II	D0 Run II
Cuts	$q_T < 0.2 Q + 0.5 \text{ GeV} = 18.7 \text{ GeV}$			
Points	31	14	37	8
\sqrt{s}	1.8 TeV	1.8 TeV	1.96 TeV	1.96 TeV
Normalization	1.114	0.992	1.049	1.048

fixed from DEMS fit,
different from exp.

(not really relevant for TMD
parametrizations)

u and b_{*} prescriptions

$$\tilde{f}_1^a(x, b_T; \mu^2) = \sum_i (\tilde{C}_{a/i} \otimes f_1^i)(x, b_*; \mu_b) e^{\tilde{S}(b_*; \mu_b, \mu)} e^{g_K(b_T) \ln \frac{\mu}{\mu_0}} \hat{f}_{\text{NP}}^a(x, b_T)$$

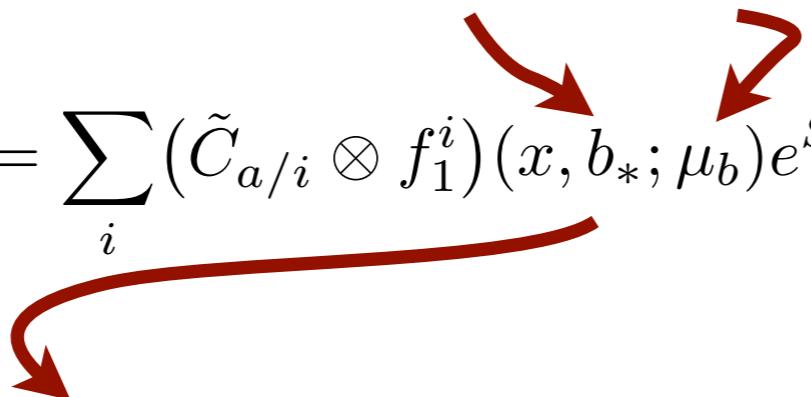
u and b_{*} prescriptions

Choice Choice

$$\tilde{f}_1^a(x, b_T; \mu^2) = \sum_i (\tilde{C}_{a/i} \otimes f_1^i)(x, b_*; \mu_b) e^{\tilde{S}(b_*; \mu_b, \mu)} e^{g_K(b_T) \ln \frac{\mu}{\mu_0}} \hat{f}_{\text{NP}}^a(x, b_T)$$

μ and b_* prescriptions

Choice Choice

$$\tilde{f}_1^a(x, b_T; \mu^2) = \sum_i (\tilde{C}_{a/i} \otimes f_1^i)(x, b_*; \mu_b) e^{\tilde{S}(b_*; \mu_b, \mu)} e^{g_K(b_T) \ln \frac{\mu}{\mu_0}} \hat{f}_{\text{NP}}^a(x, b_T)$$


$$\mu_b = 2e^{-\gamma_E}/b_* \quad b_* \equiv \frac{b_T}{\sqrt{1 + b_T^2/b_{\max}^2}}$$

Collins, Soper, Sterman, NPB250 (85)

$$\mu_b = 2e^{-\gamma_E}/b_* \quad b_* \equiv b_{\max} \left(1 - e^{-\frac{b_T^4}{b_{\max}^4}} \right)^{1/4}$$

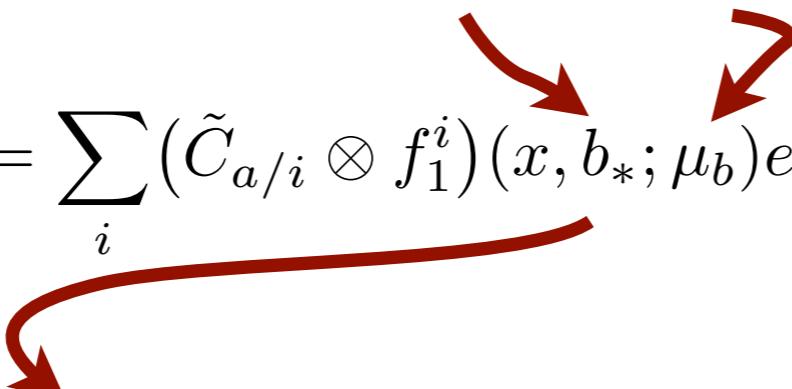
Bacchetta, Echevarria, Mulders, Radici, Signori
[arXiv:1508.00402](https://arxiv.org/abs/1508.00402)

$$\mu_b = Q_0 + q_T \quad b_* = b_T$$

DEMS 2014

μ and b_* prescriptions

Choice Choice

$$\tilde{f}_1^a(x, b_T; \mu^2) = \sum_i (\tilde{C}_{a/i} \otimes f_1^i)(x, b_*; \mu_b) e^{\tilde{S}(b_*; \mu_b, \mu)} e^{g_K(b_T) \ln \frac{\mu}{\mu_0}} \hat{f}_{\text{NP}}^a(x, b_T)$$


$$\mu_b = 2e^{-\gamma_E}/b_* \quad b_* \equiv \frac{b_T}{\sqrt{1 + b_T^2/b_{\max}^2}}$$

Collins, Soper, Sterman, NPB250 (85)

$$\mu_b = 2e^{-\gamma_E}/b_* \quad b_* \equiv b_{\max} \left(1 - e^{-\frac{b_T^4}{b_{\max}^4}} \right)^{1/4}$$

Bacchetta, Echevarria, Mulders, Radici, Signori
[arXiv:1508.00402](https://arxiv.org/abs/1508.00402)

$$\mu_b = Q_0 + q_T \quad b_* = b_T \quad \text{DEMS 2014}$$

Complex-b prescription

Laenen, Sterman, Vogelsang, PRL 84 (00)

Nonperturbative ingredients 1

$$\tilde{f}_1^a(x, b_T; \mu^2) = \sum_i (\tilde{C}_{a/i} \otimes f_1^i)(x, b_*; \mu_b) e^{\tilde{S}(b_*; \mu_b, \mu)} e^{g_K(b_T) \ln \frac{\mu}{\mu_0}} \hat{f}_{\text{NP}}^a(x, b_T)$$

Nonperturbative ingredients 1

$$\tilde{f}_1^a(x, b_T; \mu^2) = \sum_i (\tilde{C}_{a/i} \otimes f_1^i)(x, b_*; \mu_b) e^{\tilde{S}(b_*; \mu_b, \mu)} e^{g_K(b_T) \ln \frac{\mu}{\mu_0}} \hat{f}_{\text{NP}}^a(x, b_T)$$

Choice
↓

Nonperturbative ingredients 1

$$\tilde{f}_1^a(x, b_T; \mu^2) = \sum_i (\tilde{C}_{a/i} \otimes f_1^i)(x, b_*; \mu_b) e^{\tilde{S}(b_*; \mu_b, \mu)} e^{g_K(b_T) \ln \frac{\mu}{\mu_0}} \hat{f}_{\text{NP}}^a(x, b_T)$$



Choice

$$e^{-\frac{b_T^2}{\langle b_T^2 \rangle}}$$

almost everybody

$$e^{-\frac{b_T^2}{\langle b_T^2(x) \rangle_a}}$$

Pavia 2013, KN 2006

$$e^{-\lambda_1 b_T} (1 + \lambda_2 b_T^2)$$

DEMS 2014

Low- b_T modifications

$$\log(Q^2 b_T^2) \rightarrow \log(Q^2 b_T^2 + 1)$$

see, e.g., Bozzi, Catani, De Florian, Grazzini
[hep-ph/0302104](#)

see talks by Collins, Boglione, (Rogers?)

Low- b_T modifications

$$\log(Q^2 b_T^2) \rightarrow \log(Q^2 b_T^2 + 1)$$

see, e.g., Bozzi, Catani, De Florian, Grazzini
[hep-ph/0302104](#)

$$b_*(b_c(b_T)) = \sqrt{\frac{b_T^2 + b_0^2/(C_5^2 Q^2)}{1 + b_T^2/b_{\max}^2 + b_0^2/(C_5^2 Q^2 b_{\max}^2)}}$$

$$b_{\min} \equiv b_*(b_c(0)) = \frac{b_0}{C_5 Q} \sqrt{\frac{1}{1 + b_0^2/(C_5^2 Q^2 b_{\max}^2)}}$$

Collins et al.
[arXiv:1605.00671](#)

see talks by Collins, Boglione, (Rogers?)

Data selection

$$Q^2 > 1.4 \text{ GeV}^2$$

$$0.2 < z < 0.7$$

$$P_{hT}, q_T < 0.2 Q + 0.5 \text{ GeV}$$

$$P_{hT} < 0.8 \text{ GeV} \text{ (if } z < 0.3\text{)}$$

Data selection

$$Q^2 > 1.4 \text{ GeV}^2$$

$$0.2 < z < 0.7$$

$$P_{hT}, q_T < 0.2 Q + 0.5 \text{ GeV}$$

$$P_{hT} < 0.8 \text{ GeV} \text{ (if } z < 0.3\text{)}$$

Total number of data points: 8156

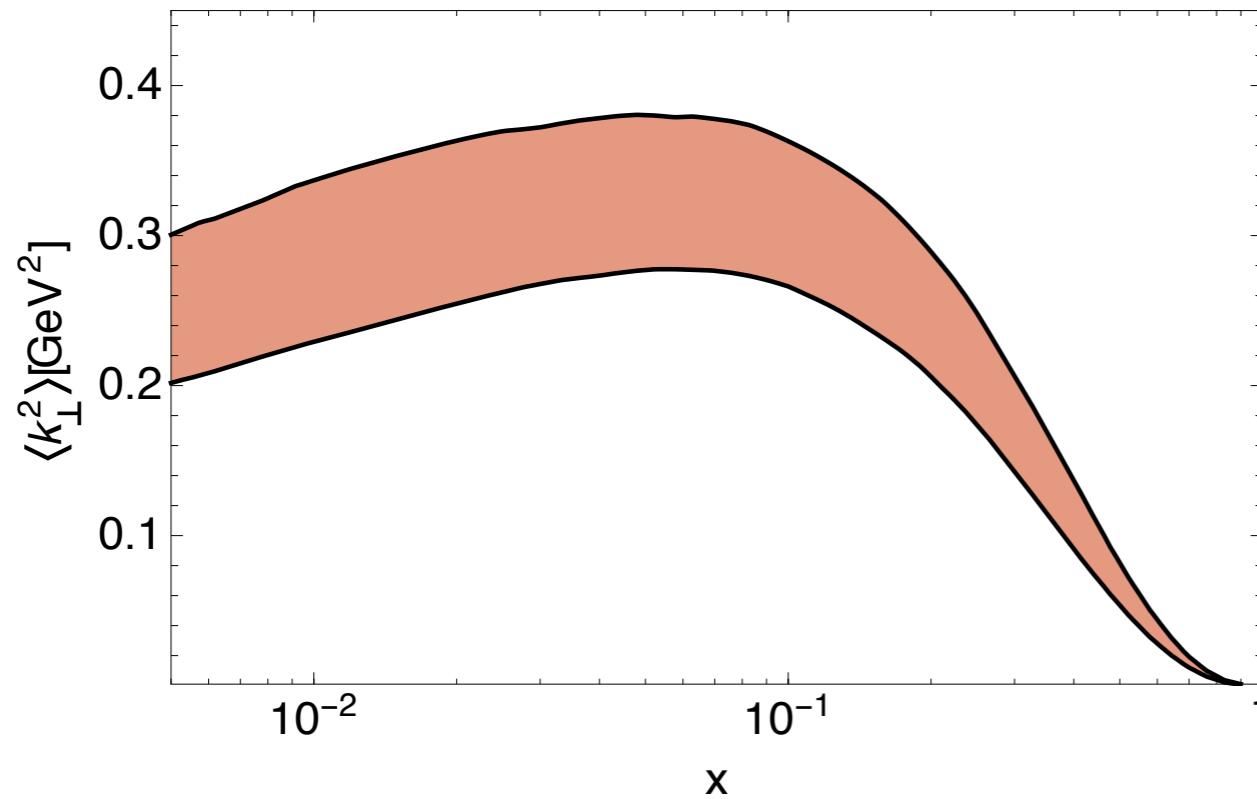
Total $\chi^2/\text{dof} = 1.45$

Preliminary

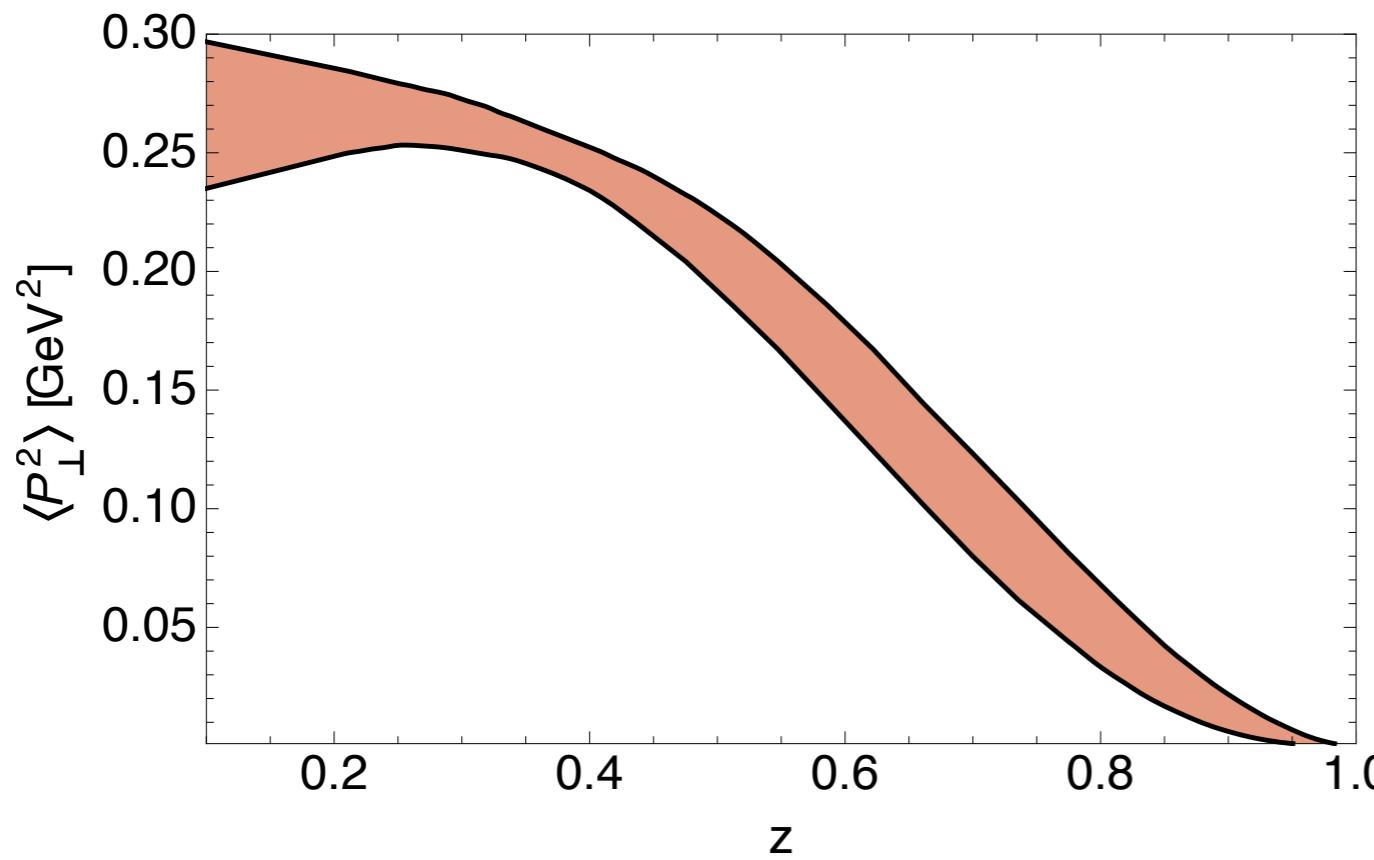
Pavia 2016 perturbative ingredients

$\tilde{f}_1^a(x, b_T; \mu^2) = \sum_i (\tilde{C}_{a/i} \otimes f_1^i)(x, b_*; \mu_b) e^{\tilde{S}(b_*; \mu_b, \mu)} e^{g_K(b_T) \ln \frac{\mu}{\mu_0}} \hat{f}_{\text{NP}}^a(x, b_T)$			
$A_1(\mathcal{O}(\alpha_S^1))$	✓	$A_2(\mathcal{O}(\alpha_S^2))$	$A_3(\mathcal{O}(\alpha_S^3))$
$B_1(\mathcal{O}(\alpha_S^1))$	✓	$B_2(\mathcal{O}(\alpha_S^2))$...
$C_0(\mathcal{O}(\alpha_S^0))$	✓	$C_1(\mathcal{O}(\alpha_S^1))$	$C_2(\mathcal{O}(\alpha_S^2))$
<hr/>			
$H_0(\mathcal{O}(\alpha_S^0))$	✓	$H_1(\mathcal{O}(\alpha_S^1))$	$H_2(\mathcal{O}(\alpha_S^2))$
			...
		$Y_1(\mathcal{O}(\alpha_S^1))$	$Y_2(\mathcal{O}(\alpha_S^2))$
			...

Mean transverse momentum



In TMD PDF



In TMD FF

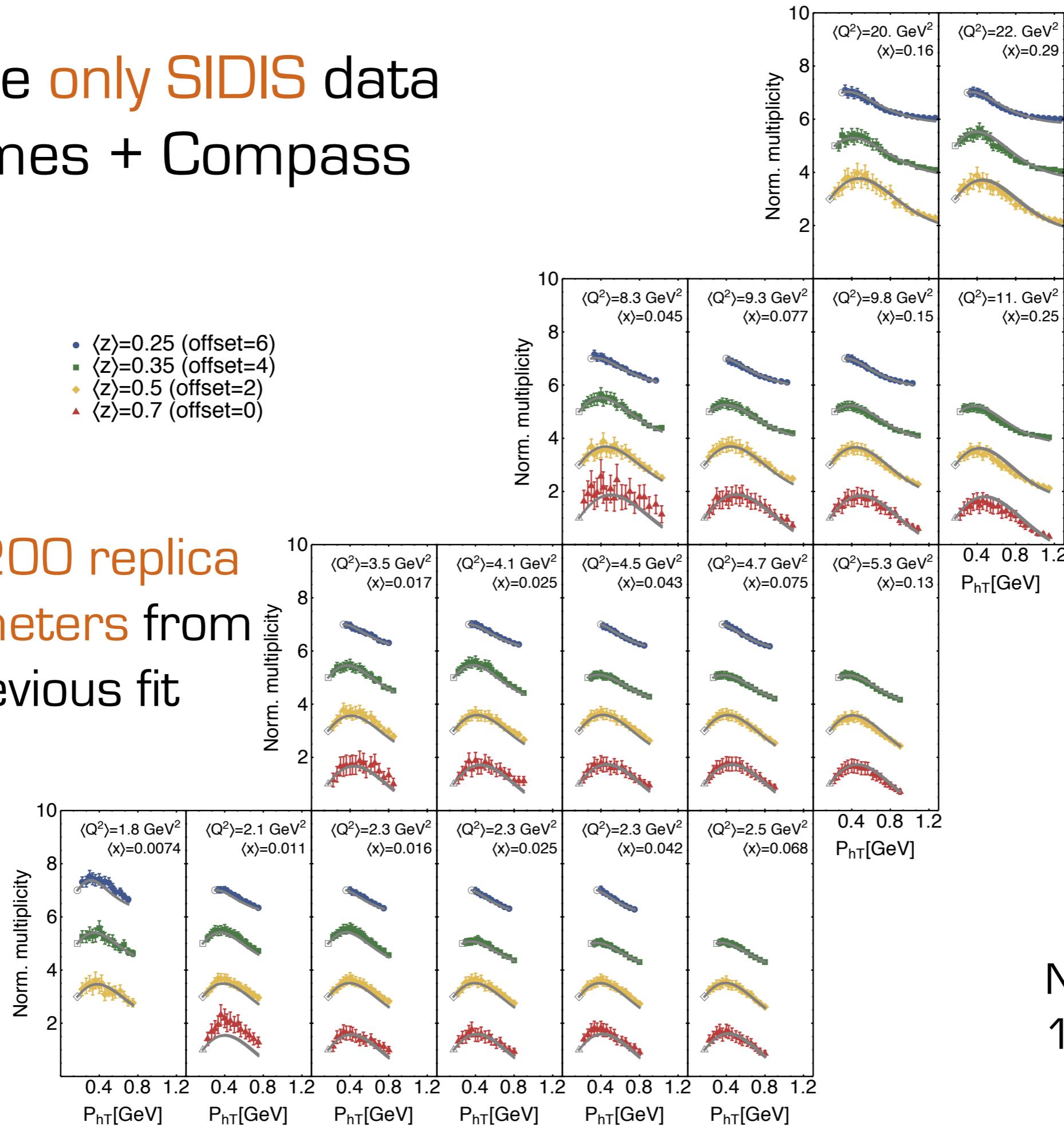
$Q^2 = 1 \text{ GeV}_{62}^2$

Include only SIDIS data Hermes + Compass

SIDIS h^+

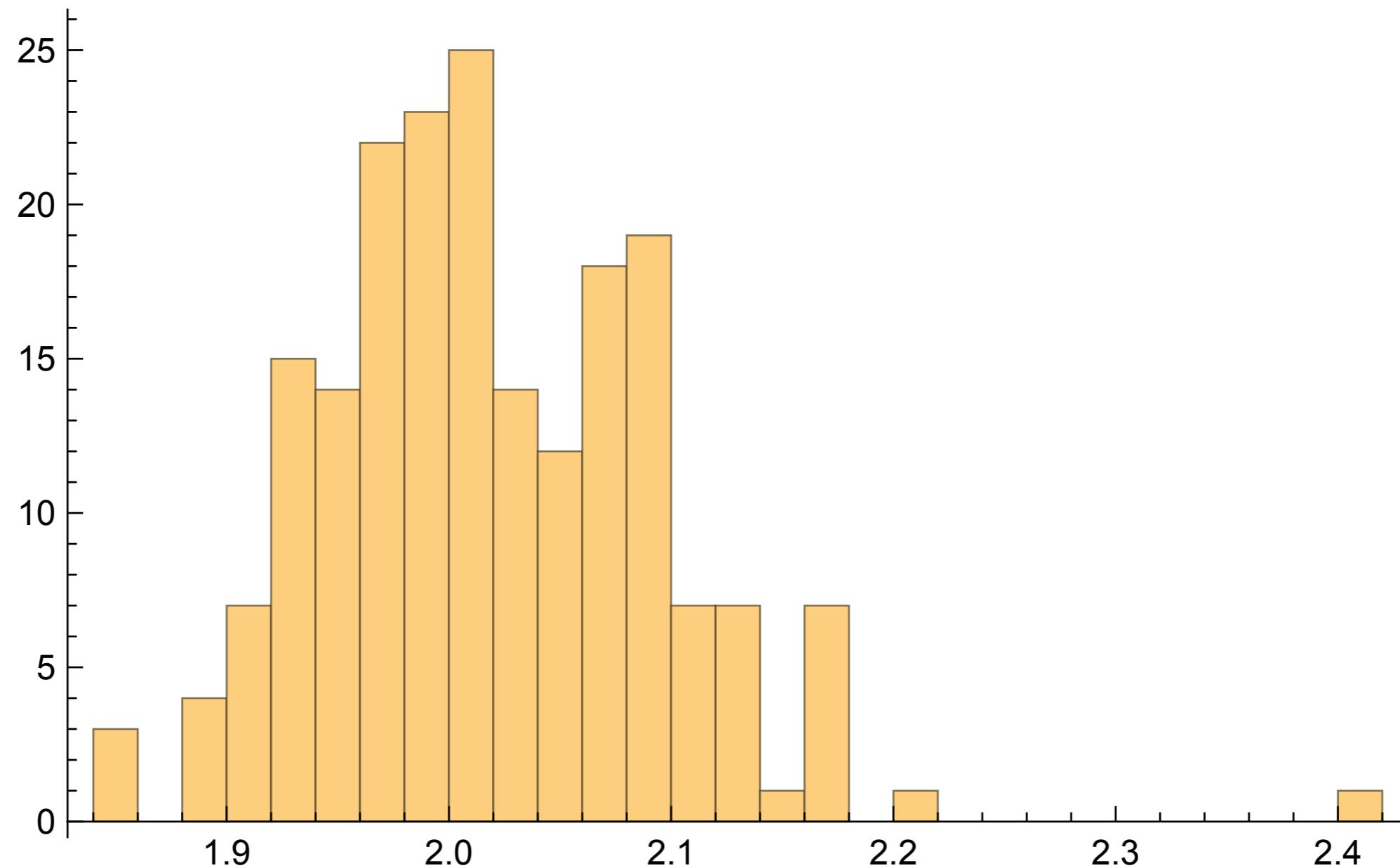
- $\langle z \rangle = 0.25$ (offset=6)
- $\langle z \rangle = 0.35$ (offset=4)
- ◊ $\langle z \rangle = 0.5$ (offset=2)
- ▲ $\langle z \rangle = 0.7$ (offset=0)

Use 200 replica
parameters from
previous fit



Include only SIDIS data

SIDIS h⁺



Use 200 replica
parameters from
previous fit

$$\chi^2/\text{dof} = 2.07$$

Normalized at
1st data point
of bin