

# Global QCD Analyses of Pion Parton Distributions and TMDPDFs

Patrick Barry

ECT\*: Mass in the Standard Model and Consequences of its Emergence

4/20/2021

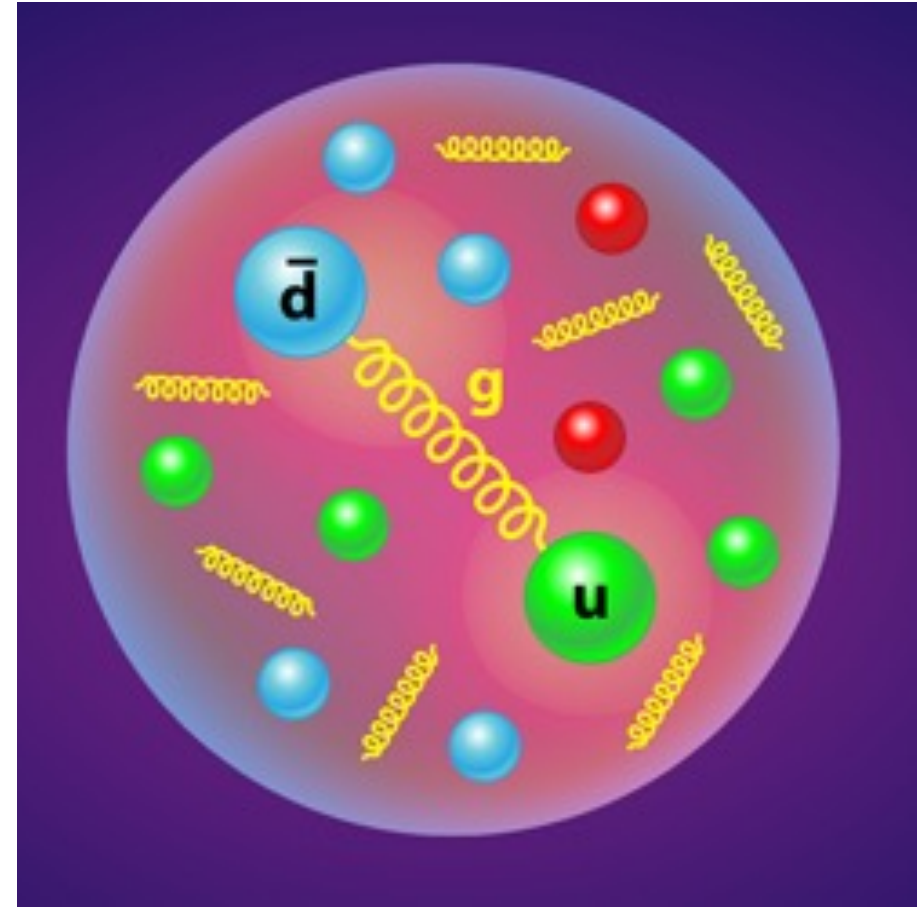
# Motivation

What to do:

- **Define** a structure of hadrons in terms of quantum field theories
- **Identify** theoretical observables that factorize into non-perturbative objects and perturbatively calculable physics
- Perform **global QCD analysis** as structures are universal and are the same in all subprocesses

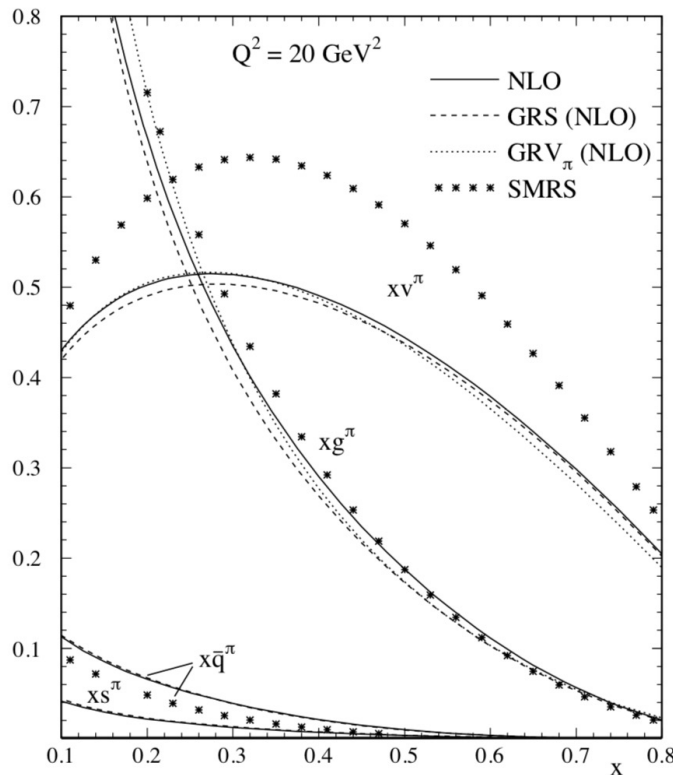
# Pions

- Pion is the **Goldstone boson** associated with spontaneous symmetry breaking of chiral  $SU(2)_L \times SU(2)_R$  symmetry
- **Lightest hadron** as  $\frac{m_\pi}{M_N} \ll 1$  and dictates the nature of hadronic interactions at low energies
- Simultaneously a pseudoscalar meson made up of  $q$  and  $\bar{q}$  constituents

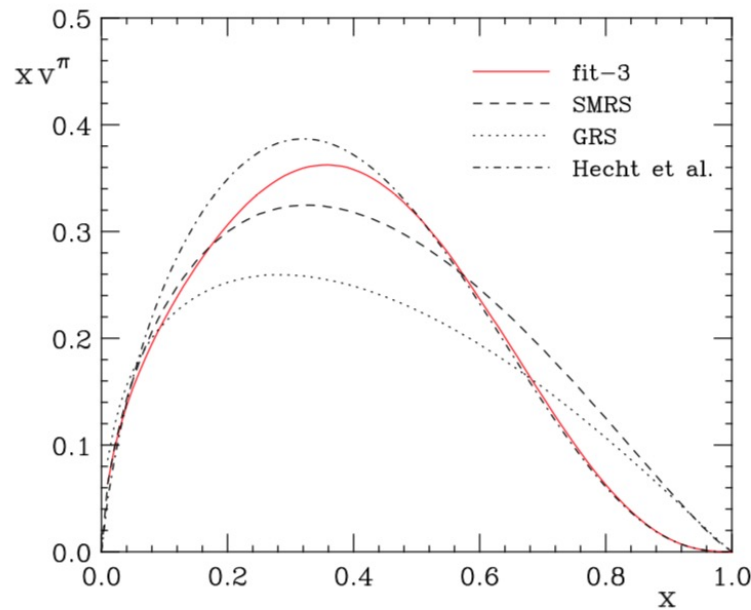


# Previous Pion PDFs

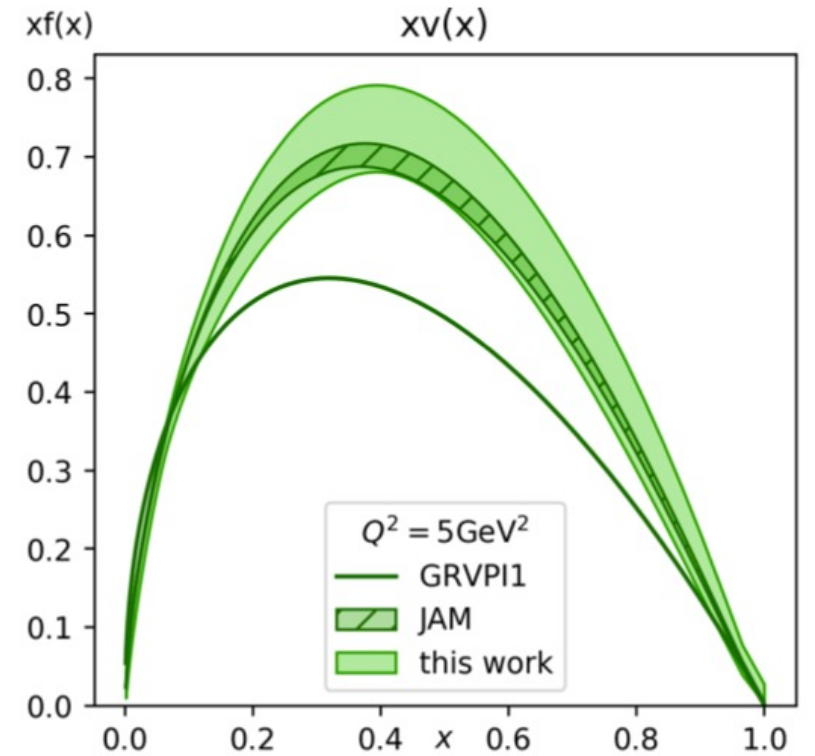
- Fits to Drell-Yan, prompt photon, or both



**GRS, GRV, and SMRS**  
 Z. Phys. C **67**, 433 (1995).  
 Eur. Phys. J. C **10** 313 (1997).  
 Phys. Rev. D **45** 2349 (1992).



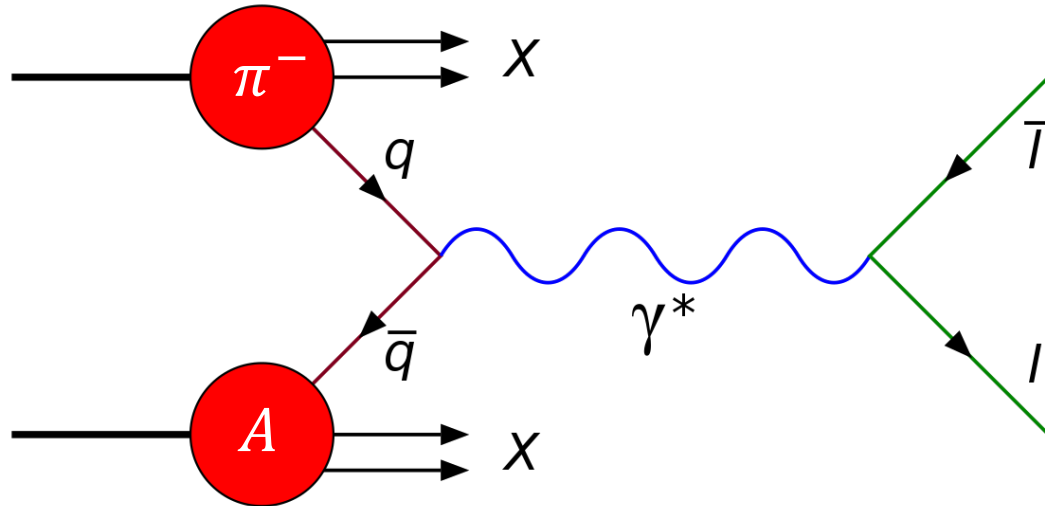
**Aicher's valence PDF**  
 Phys. Rev. Lett. **105**, 114023 (2011).



**xFitter**  
 Phys. Rev. D **102**, 014040 (2020).

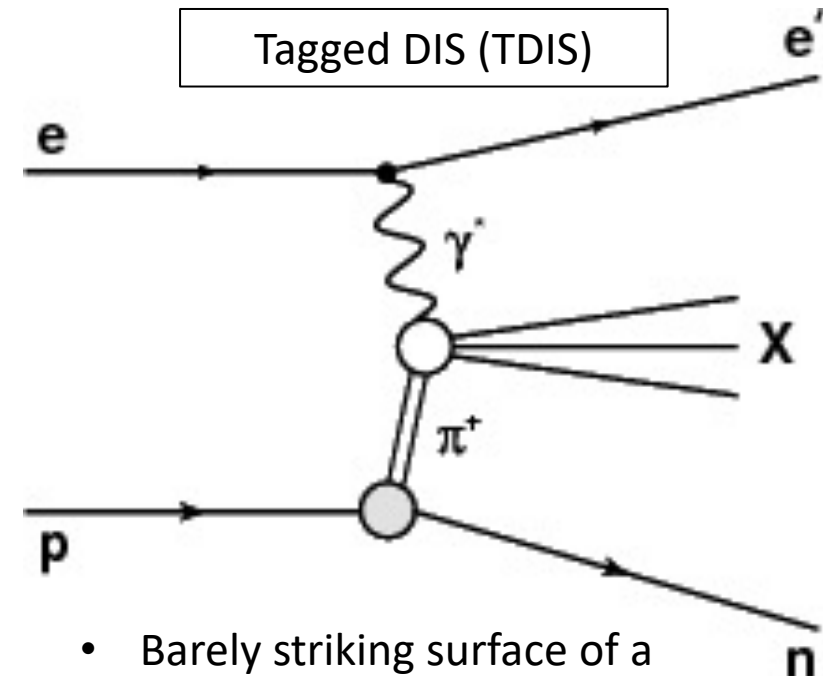
# Experiments to Probe Pion Structure

- Drell-Yan (DY)



- Accelerating pion allows for time dilation and longer lifetime

- Leading Neutron (LN)



- Barely striking surface of a target proton knocks out an almost on-shell pion to probe

# Datasets -- Kinematics

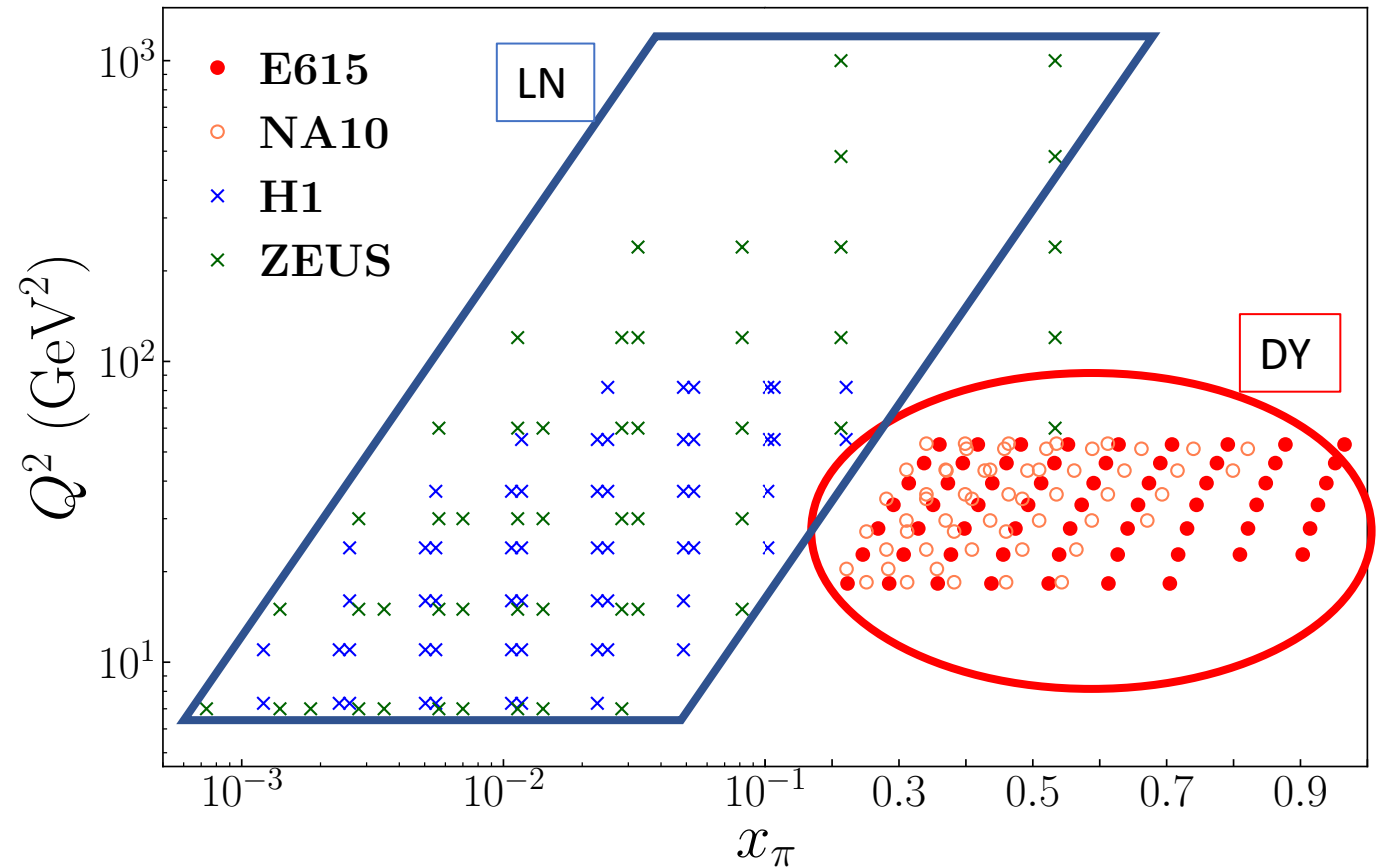
- Large  $x_\pi$  -- Drell-Yan (DY)
- Small  $x_\pi$  -- Leading Neutron (LN)
- Not much data overlap

- In DY:

$$x_\pi = \frac{1}{2} \left( x_F + \sqrt{x_F^2 + 4\tau} \right)$$

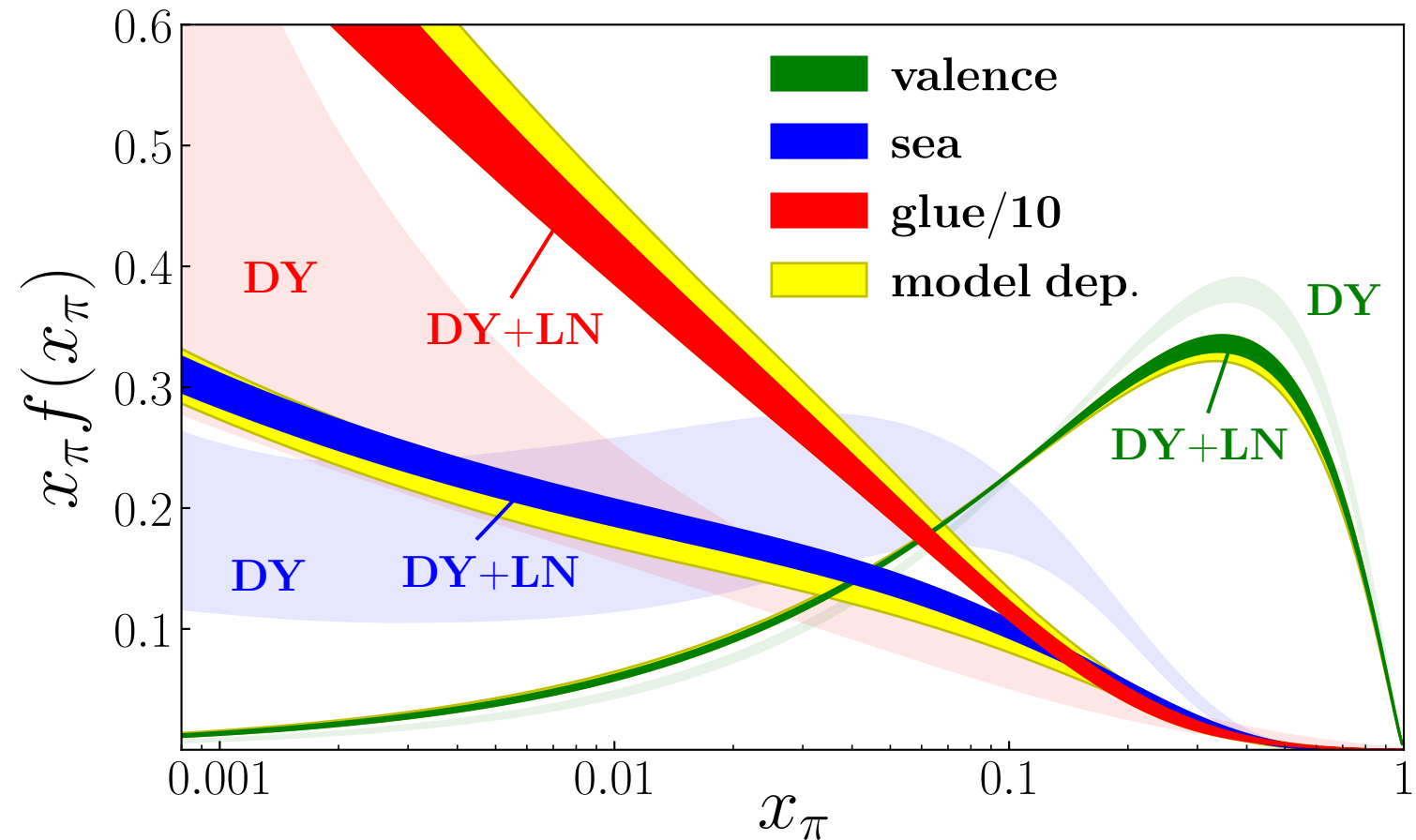
- In LN:

$$x_\pi = x_B / \bar{x}_L$$



# JAM18 Pion PDFs

- Lightly shaded bands – only Drell-Yan data
- Darkly shaded bands – fit to both Drell-Yan and LN data

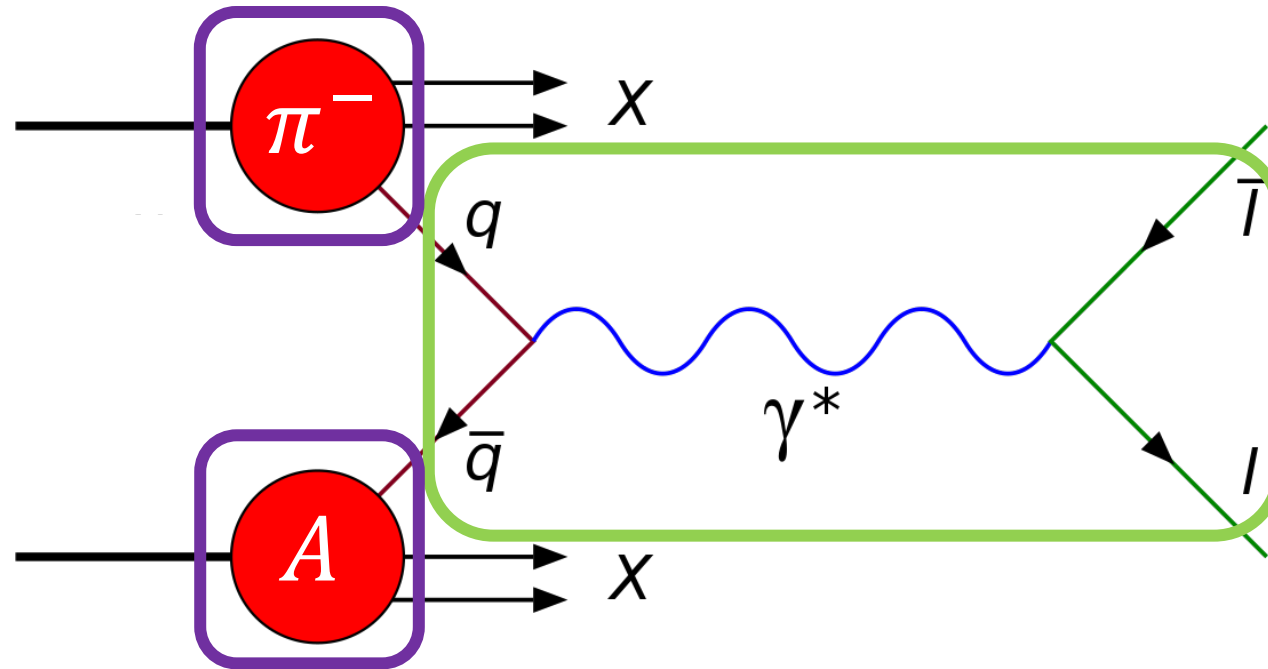


PCB, N. Sato, W. Melnitchouk and Chueng-Ryong Ji,  
Phys. Rev. Lett. **121**, 152001 (2018).

# Theoretical Input

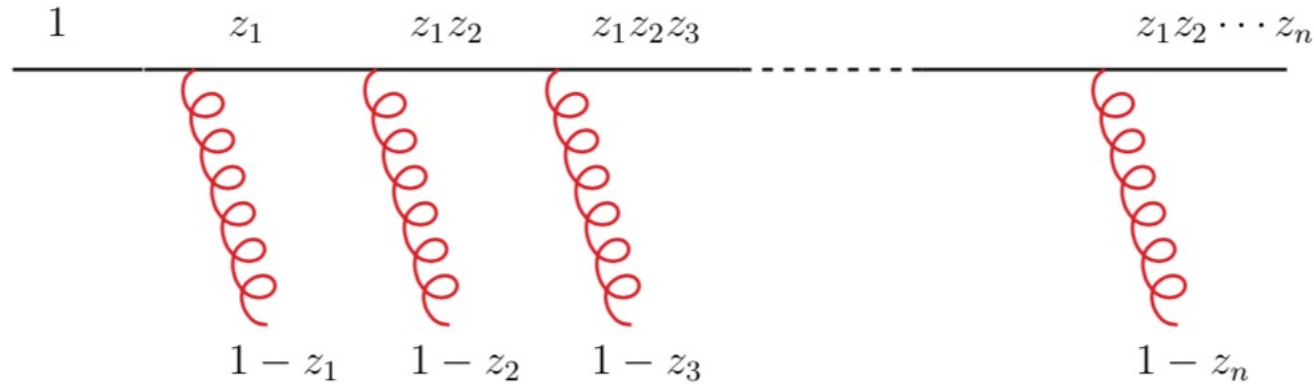


# $p_T$ -integrated Drell-Yan (DY)



$$\frac{d^2\sigma}{dx_F d\sqrt{\tau}} \propto \sum_{i,j} f_i^\pi(x_\pi, \mu) \otimes f_j^A(x_A, \mu) \otimes C_{i,j}(x_\pi, x_A, Q/\mu)$$

# Soft Gluon Resummation



- Fixed-target Drell-Yan notoriously has large- $x_F$  contamination of higher orders
- **Large logarithms** may **spoil** perturbation
- Focus on corrections to the most important  **$q\bar{q}$  channel**
- Resum contributions to all orders of  $\alpha_s$

# Issues with Perturbative Calculations

$$\hat{\sigma} \sim \delta(1-z) + \alpha_S (\log(1-z))_+ \longrightarrow \hat{\sigma} \sim \delta(1-z) [1 + \alpha_S \log(1-\tau)]$$

- If  $\tau$  is large, can potentially spoil the perturbative calculation
- Improvements can be made by resumming  $\log(1-z)_+$  terms

$$\tau = \frac{Q^2}{S}$$

# Next-to-Leading + Next-to-Leading Logarithm Order Calculation

An NLO calculation  
gathers the  $\mathcal{O}(\alpha_S)$   
terms

LL

NLL

...

N<sup>p</sup>LL

LO	1	--	...	--
NLO	$\alpha_S \log(N)^2$	$\alpha_S \log(N)$	...	--
NNLO	$\alpha_S^2 \log(N)^4$	$\alpha_S^2 (\log(N)^2, \log(N)^3)$	...	--
...	...	...	...	...
N <sup>k</sup> LO	$\alpha_S^k \log(N)^{2k}$	$\alpha_S^k (\log(N)^{2k-1}, \log(N)^{2k-2})$	...	$\alpha_S^k \log(N)^{2k-2p} + \dots$

# Next-to-Leading + Next-to-Leading Logarithm Order Calculation

Add the columns to  
the rows

	<u>LL</u>	<u>NLL</u>	...	<u>N<sup>p</sup>LL</u>
LO	1	--	...	--
NLO	$\alpha_s \log(N)^2$	$\alpha_s \log(N)$	...	--
NNLO	$\alpha_s^2 \log(N)^4$	$\alpha_s^2 (\log(N)^2, \log(N)^3)$	...	--
...	...	...	...	...
N <sup>k</sup> LO	$\alpha_s^k \log(N)^{2k}$	$\alpha_s^k (\log(N)^{2k-1}, \log(N)^{2k-2})$	...	$\alpha_s^k \log(N)^{2k-2p} + \dots$

# Next-to-Leading + Next-to-Leading Logarithm Order Calculation

Make sure only counted once!  
- Subtract the matching

	<u>LL</u>	<u>NLL</u>	...	<u>N<sup>p</sup>LL</u>
LO	1	--	...	--
NLO	$\alpha_s \log(N)^2$	$\alpha_s \log(N)$	...	--
NNLO	$\alpha_s^2 \log(N)^4$	$\alpha_s^2 (\log(N)^2, \log(N)^3)$	...	--
...	...	...	...	...
N <sup>k</sup> LO	$\alpha_s^k \log(N)^{2k}$	$\alpha_s^k (\log(N)^{2k-1}, \log(N)^{2k-2})$	...	$\alpha_s^k \log(N)^{2k-2p} + \dots$

# Origin of Landau Pole

$$\alpha_S C_{\text{soft}}^{(1)}(N) = 2 \frac{C_F}{\pi} \int_0^1 dz \frac{z^{N-1} - 1}{1-z} \int_{Q^2}^{(1-z)^2 Q^2} \frac{dk_{\perp}^2}{k_{\perp}^2} \alpha_S(k_{\perp}^2)$$

- Upper Limits imply that  $k_{\perp}^2$  will go to 0
- $\alpha_S(\mu^2 = 0)$  is **NOT** well-defined
- **Ambiguities** on how to deal with this provide needs for prescriptions

# Methods of Resummation

- Make use of the **Minimal Prescription** to avoid Landau Pole
- Rapidity distribution  $\frac{d\sigma}{dQ^2 dY}$  adds more complications
- We can perform a **Mellin-Fourier transform** to account for the rapidity
  - A cosine appears while doing Fourier transform; options:
    - 1) Take first order **expansion**, cosine  $\approx 1$
    - 2) Keep **cosine** intact
- Can additionally perform a **Double Mellin transform**
- **Explore** the different methods and **analyze** effects



# Methodology

# Bayesian Inference

- Using Bayesian statistics, we describe the posterior distribution as

$$\mathcal{P}(\mathbf{a}|\text{data}) \propto \mathcal{L}(\text{data}|\mathbf{a})\pi(\mathbf{a})$$

$$\mathcal{L}(\text{data}|\mathbf{a}) = \exp\left(-\frac{1}{2}\chi^2(\mathbf{a}, \text{data})\right)$$

$\pi$  -- Bayesian priors

# Bayesian Inference

- Minimize the  $\chi^2$  for each replica

$$\chi^2(\mathbf{a}, \text{data}) = \sum_e \left( \sum_i \left[ \frac{d_i^e - \sum_k r_k^e \beta_{k,i}^e - t_i^e(\mathbf{a}) / n_e}{\alpha_i^e} \right]^2 + \left( \frac{1 - n_e}{\delta n_e} \right)^2 + \sum_k (r_k^e)^2 \right)$$

Normalization parameter

- Perform  $N$  total  $\chi^2$  minimizations and compute statistical quantities in Monte Carlo framework

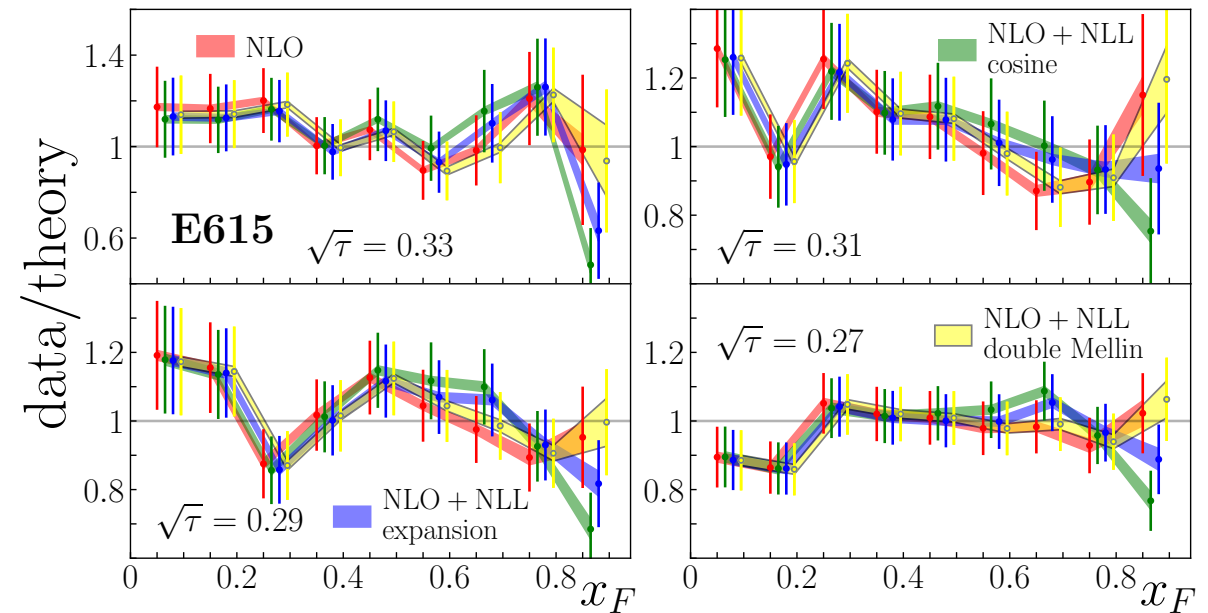
Expectation value  $E[\mathcal{O}] = \frac{1}{N} \sum_k \mathcal{O}(\mathbf{a}_k),$

Variance  $V[\mathcal{O}] = \frac{1}{N} \sum_k [\mathcal{O}(\mathbf{a}_k) - E[\mathcal{O}]]^2,$

# Threshold Resummation Results

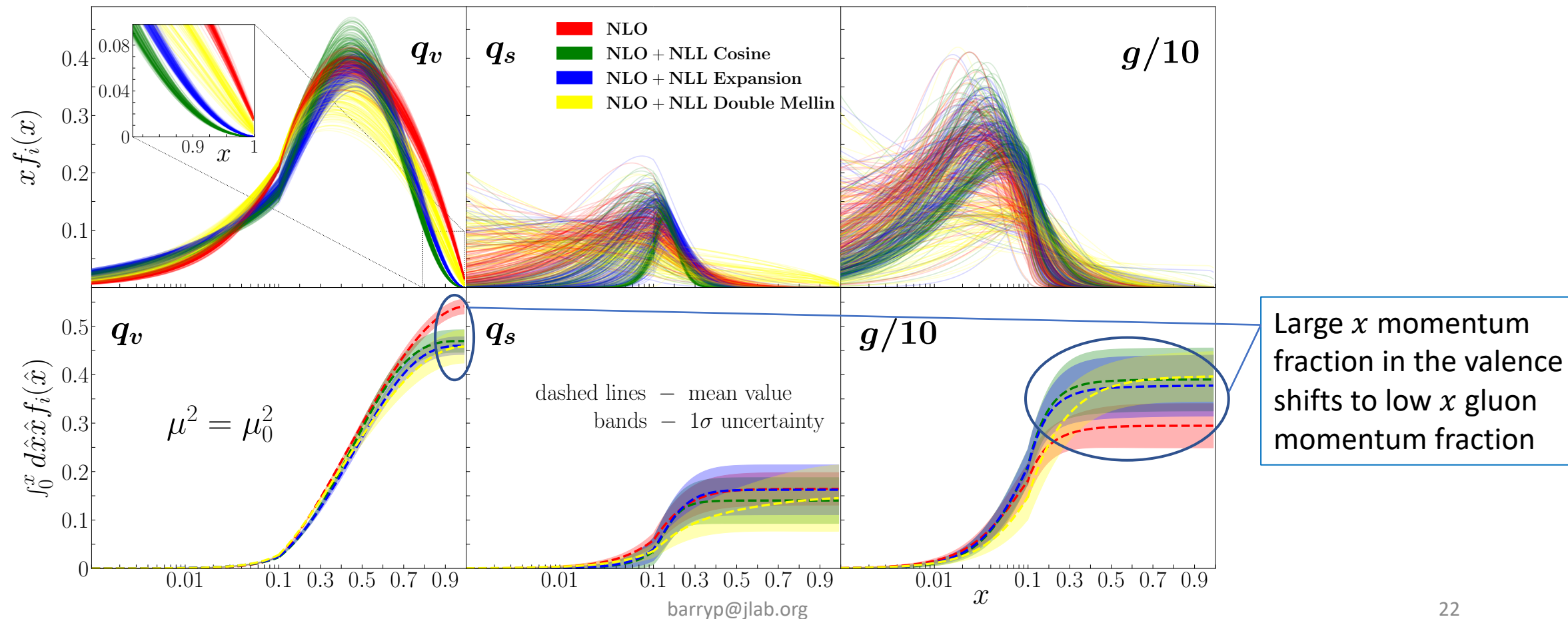
# Data and Theory Comparison – Drell-Yan

- Cosine method tends to overpredict the data at very large  $x_F$
- Double Mellin method is qualitatively very similar to NLO
- Resummation is largely a high- $x_F$  effect



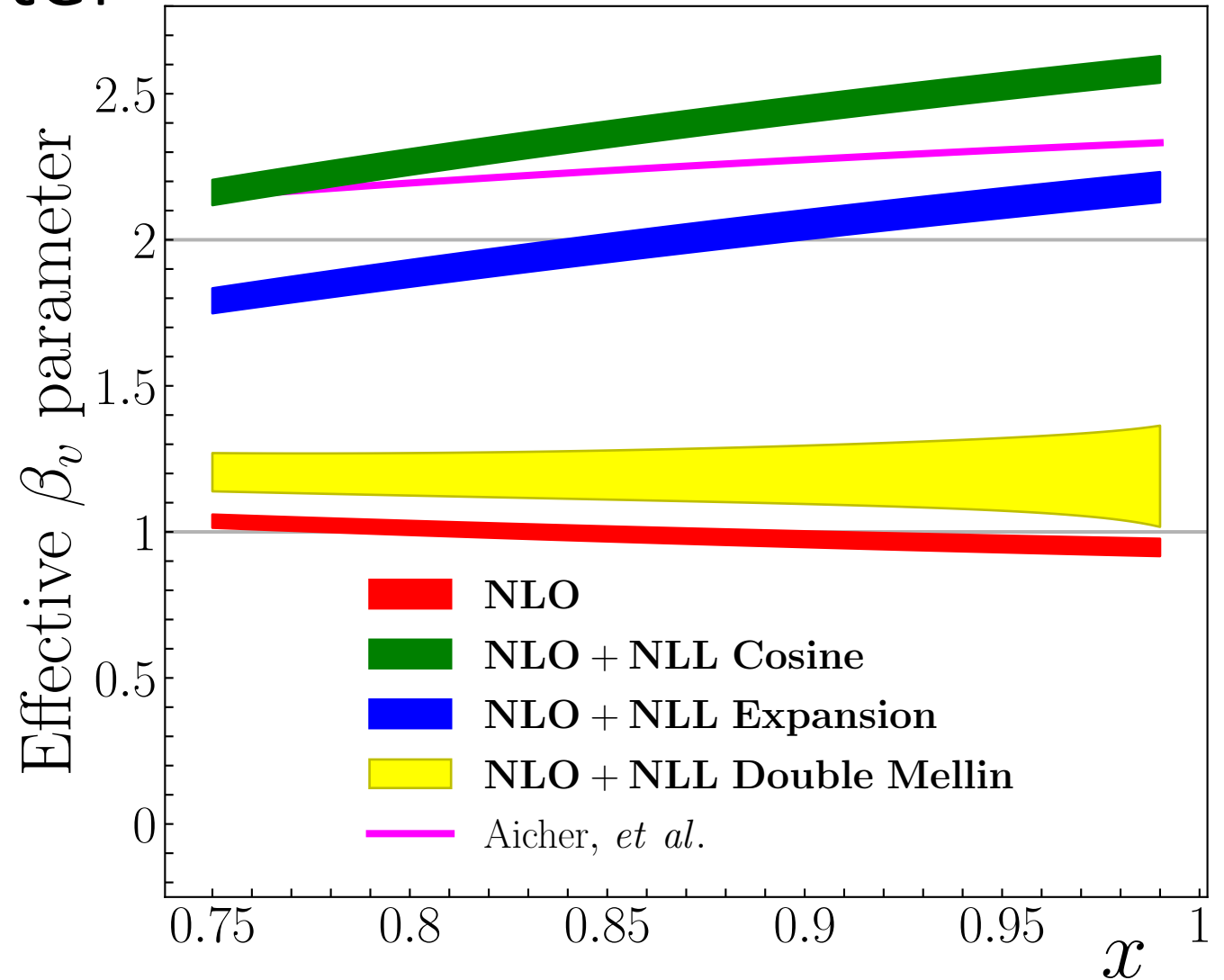
# PDF Results

- Large  $x$  behavior in valence depends on prescription



# Effective $\beta_v$ parameter

- $q_v(x) \sim (1-x)^{\beta_v}$  as  $x \rightarrow 1$
- Threshold resummation does not give universal behavior of  $\beta_v$
- NLO and double Mellin give  $\beta_v \approx 1$
- Cosine and Expansion give  $\beta_v > 2$

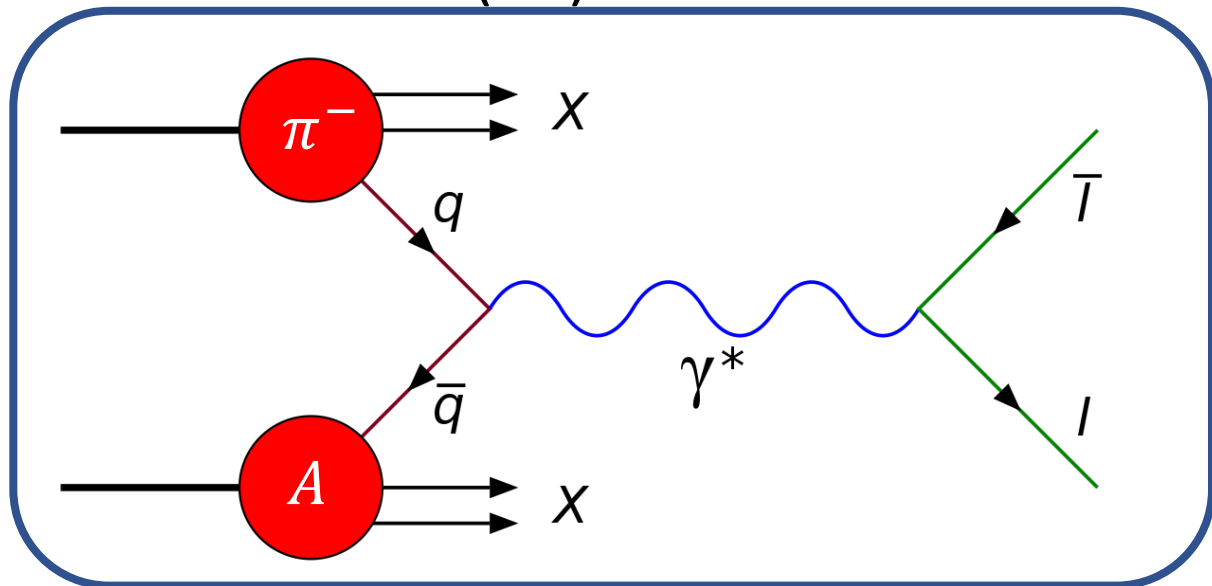


# Transverse Momentum- dependent Drell-Yan



# Experiments to Probe Pion Structure

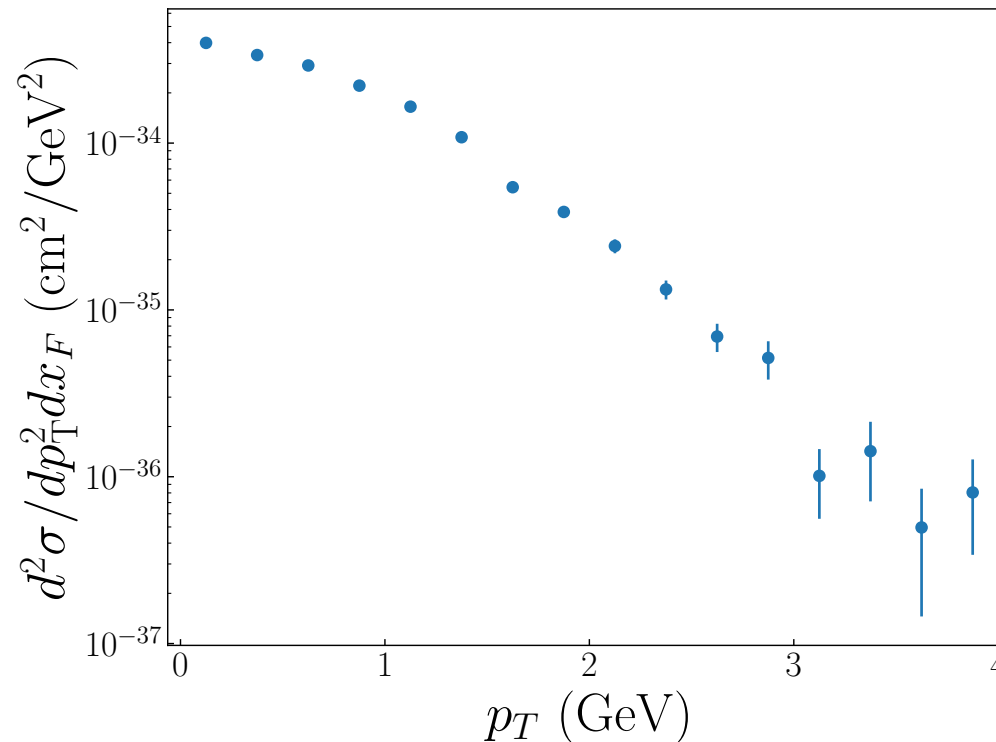
- Drell-Yan (DY)



- Traditionally, pion PDF analyses have only included the  $p_T$ -integrated cross section  $\frac{d^2\sigma}{dx_F d\sqrt{\tau}}$
- Include in this global pion PDF analysis, for the first time,  $p_T$ -dependent Drell-Yan cross sections

# $p_T$ -dependent spectrum for pion data

- Small- $p_T$  data – TMD factorization – partonic transverse momentum
- Large- $p_T$  data – collinear factorization – recoil transverse momentum

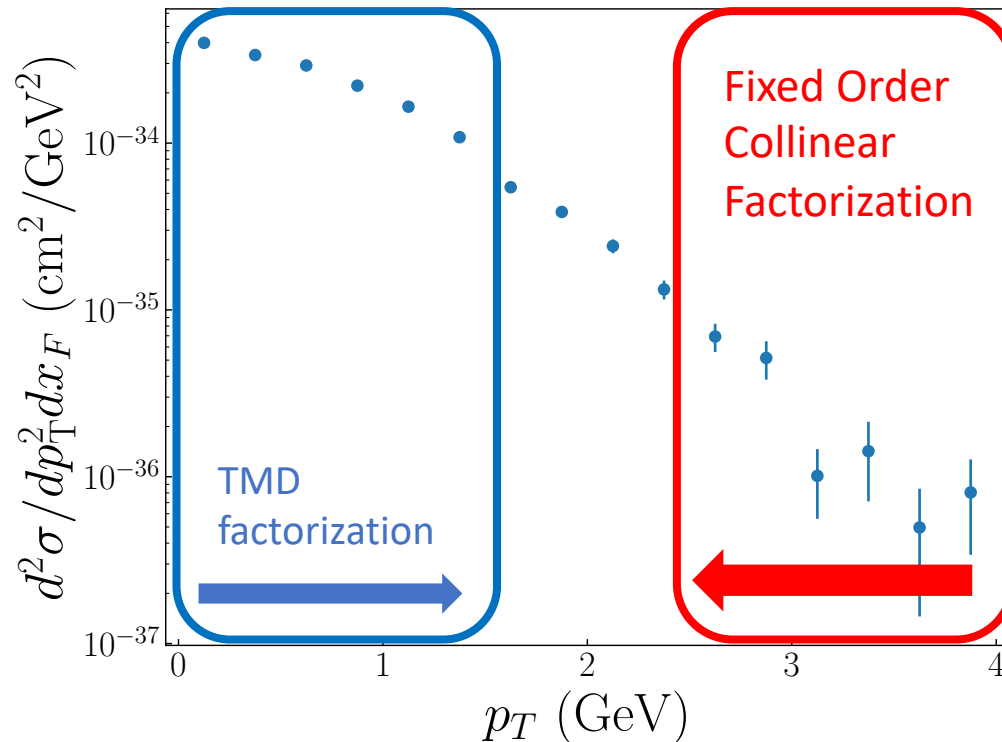


E615  $\pi W$  Drell-Yan

Phys. Rev. D **39**, 92 (1989).

# $p_T$ -dependent spectrum for pion data

- Various factorization theorems break down in certain regions of  $p_T$
- Errors are related with  $\mathcal{O}(p_T/Q)$  (low- $p_T$ ) or  $\mathcal{O}(m/p_T)$  (large- $p_T$ )

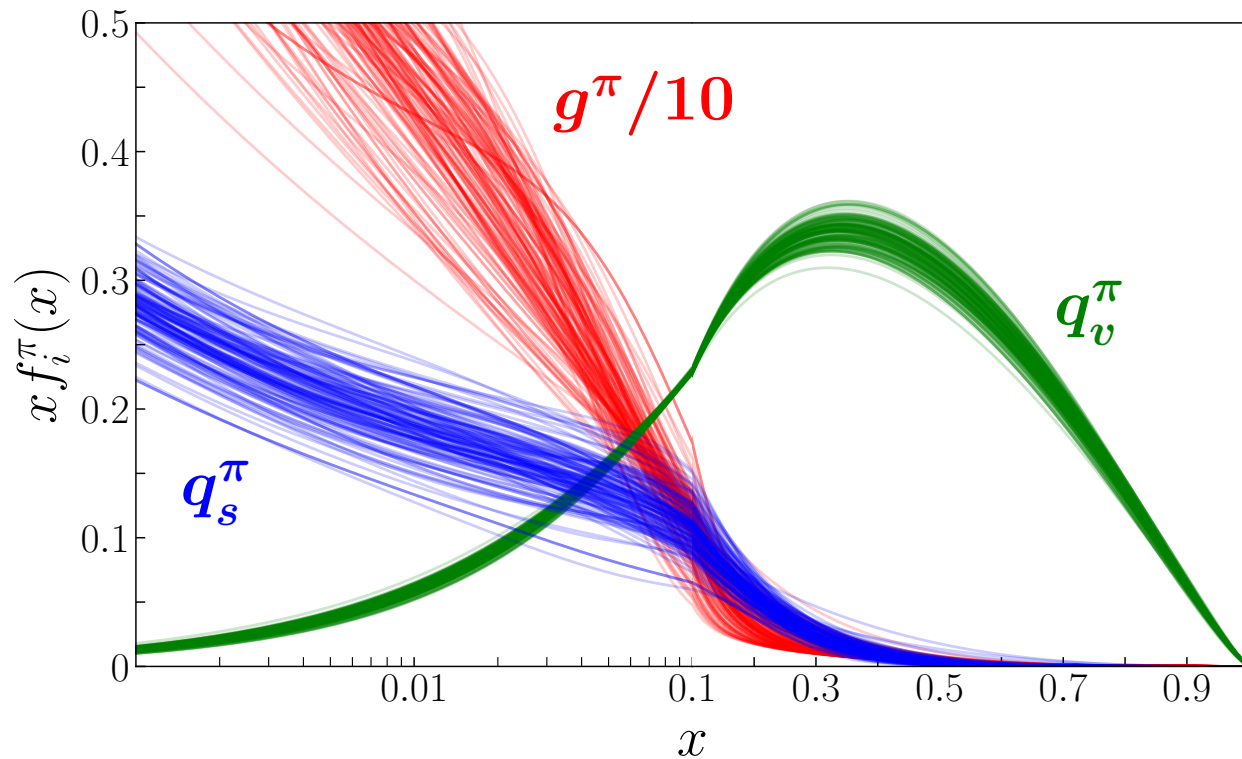


E615  $\pi W$  Drell-Yan

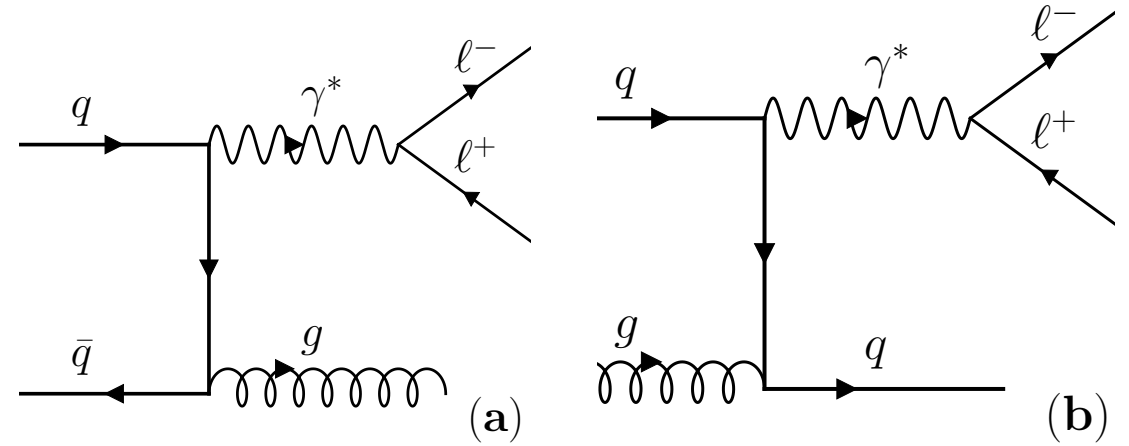
Phys. Rev. D **39**, 92 (1989).

# JAM20 Pion PDFs

Fixed Order Analysis



N.Cao, PCB, N. Sato, and W. Melnitchouk  
arXiv:2103.02159 [hep-ph]



- For the first time, we included **large  $p_T$** -dependent Drell-Yan data, which follows collinear factorization
- Large  $p_T$  does **not** dramatically affect the PDF
- Successfully describe data with a scale  $\mu = p_T/2$

# TMD factorization

- In small- $p_T$  region, Use the CSS formalism for TMD evolution

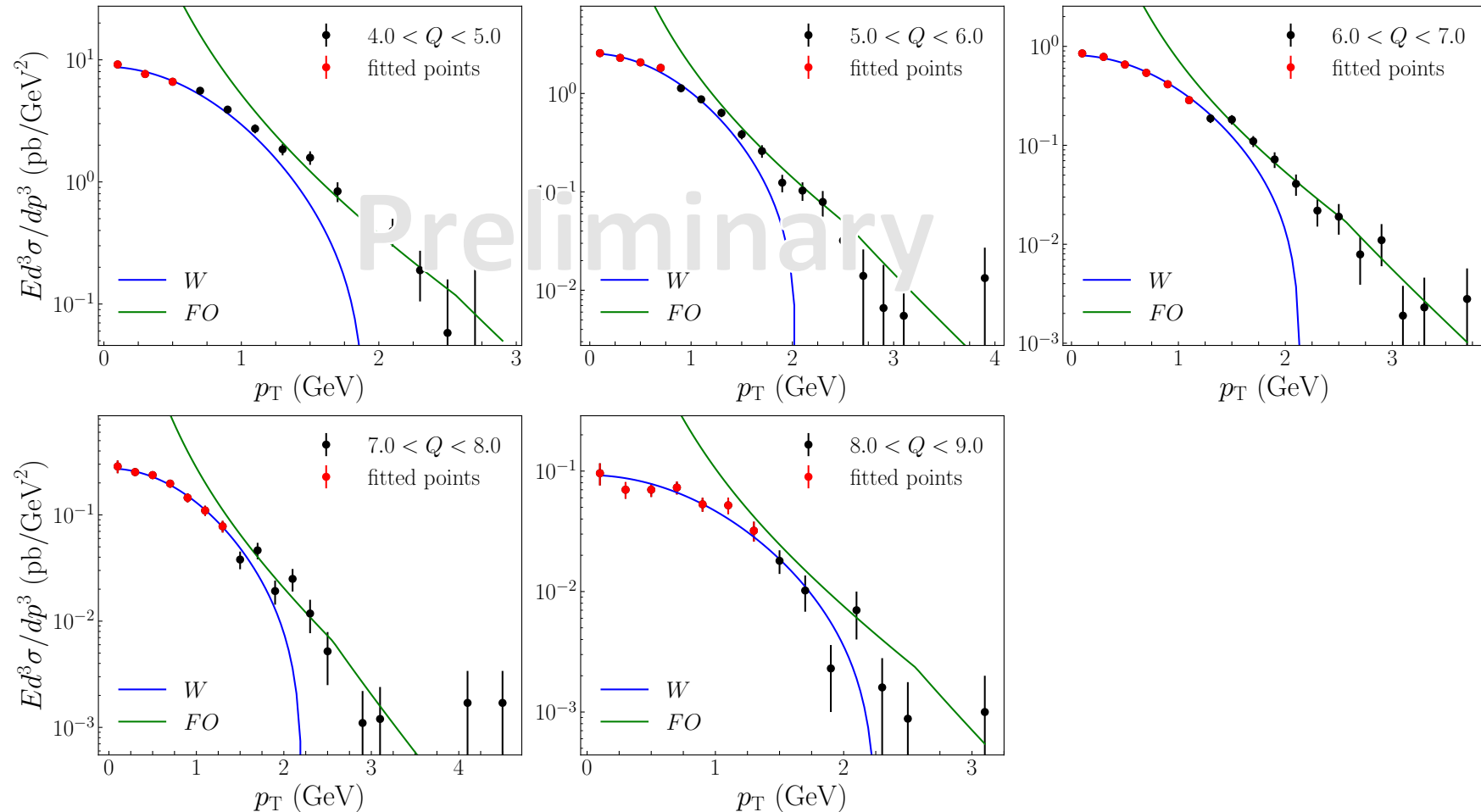
$$\begin{aligned}
 \frac{d\sigma}{dQ^2 dy dq_T^2} &= \frac{4\pi^2 \alpha^2}{9Q^2 s} \sum_{j, j_A, j_B} H_{j\bar{j}}^{\text{DY}}(Q, \mu_Q, a_s(\mu_Q)) \int \frac{d^2 \mathbf{b}_T}{(2\pi)^2} e^{i \mathbf{q}_T \cdot \mathbf{b}_T} \\
 &\times e^{-g_{j/A}(x_A, b_T; b_{\max})} \int_{x_A}^1 \frac{d\xi_A}{\xi_A} f_{j_A/A}(\xi_A; \mu_{b_*}) \tilde{C}_{j/j_A}^{\text{PDF}}\left(\frac{x_A}{\xi_A}, b_*; \mu_{b_*}^2, \mu_{b_*}, a_s(\mu_{b_*})\right) \\
 &\times e^{-g_{\bar{j}/B}(x_B, b_T; b_{\max})} \int_{x_B}^1 \frac{d\xi_B}{\xi_B} f_{j_B/B}(\xi_B; \mu_{b_*}) \tilde{C}_{\bar{j}/j_B}^{\text{PDF}}\left(\frac{x_B}{\xi_B}, b_*; \mu_{b_*}^2, \mu_{b_*}, a_s(\mu_{b_*})\right) \\
 &\times \exp \left\{ -g_K(b_T; b_{\max}) \ln \frac{Q^2}{Q_0^2} + \tilde{K}(b_*; \mu_{b_*}) \ln \frac{Q^2}{\mu_{b_*}^2} + \int_{\mu_{b_*}}^{\mu_Q} \frac{d\mu'}{\mu'} \left[ 2\gamma_j(a_s(\mu')) - \ln \frac{Q^2}{(\mu')^2} \gamma_K(a_s(\mu')) \right] \right\}
 \end{aligned}$$

Non-  
perturbative  
TMDs to extract

- Approach with **leading order** to diagnose current situation

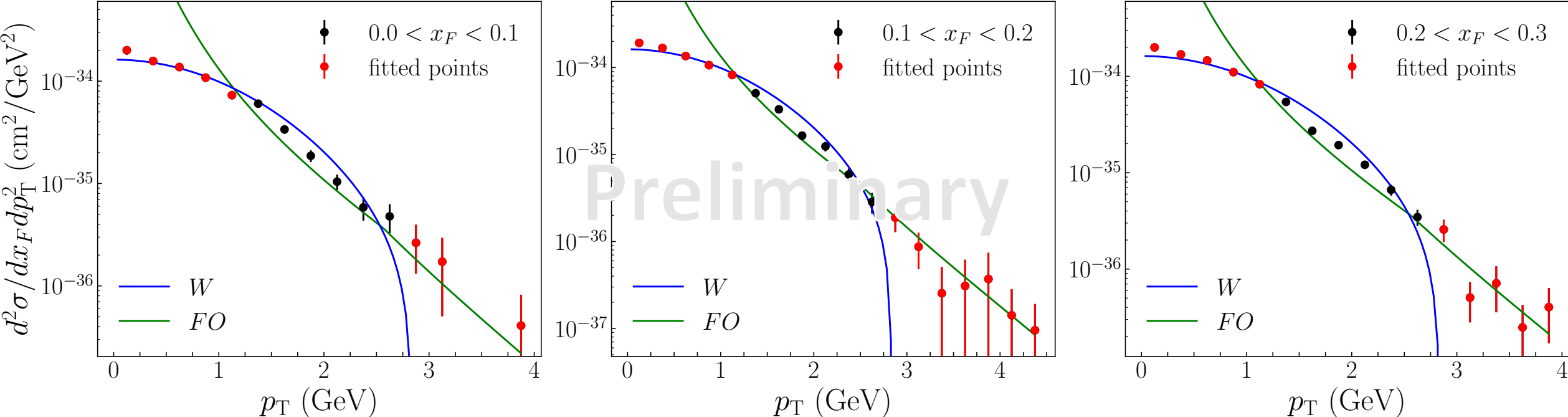
# Results – E288 $pp$ data

- Perform simultaneous fit of **small- $p_T$**  TMDs to E288 ( $pp$ ) and E615 ( $\pi A$ ) data
- FO is prediction using collinear PDFs and scale  $\mu = p_T/2$



# Results – E615 $\pi W$ data

- Proton and pion TMDs are different, but  $g_K$  are same
- Perform simultaneous fit of **small- $p_T$**  TMDs to E288 ( $pp$ ) and E615 ( $\pi A$ ) data

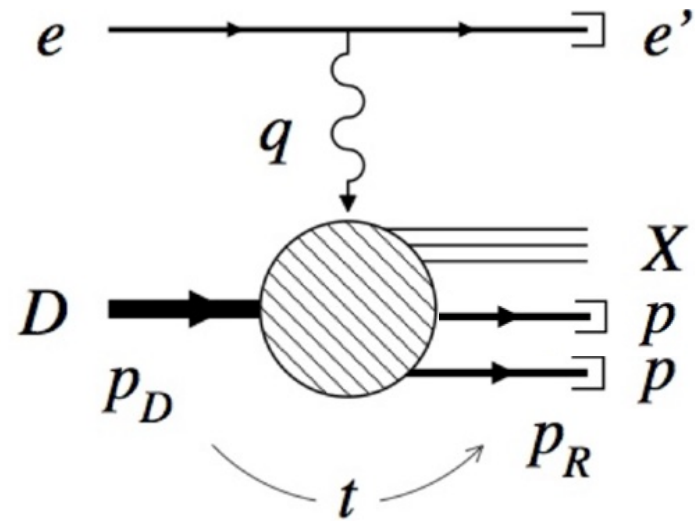


# Applications and Summary



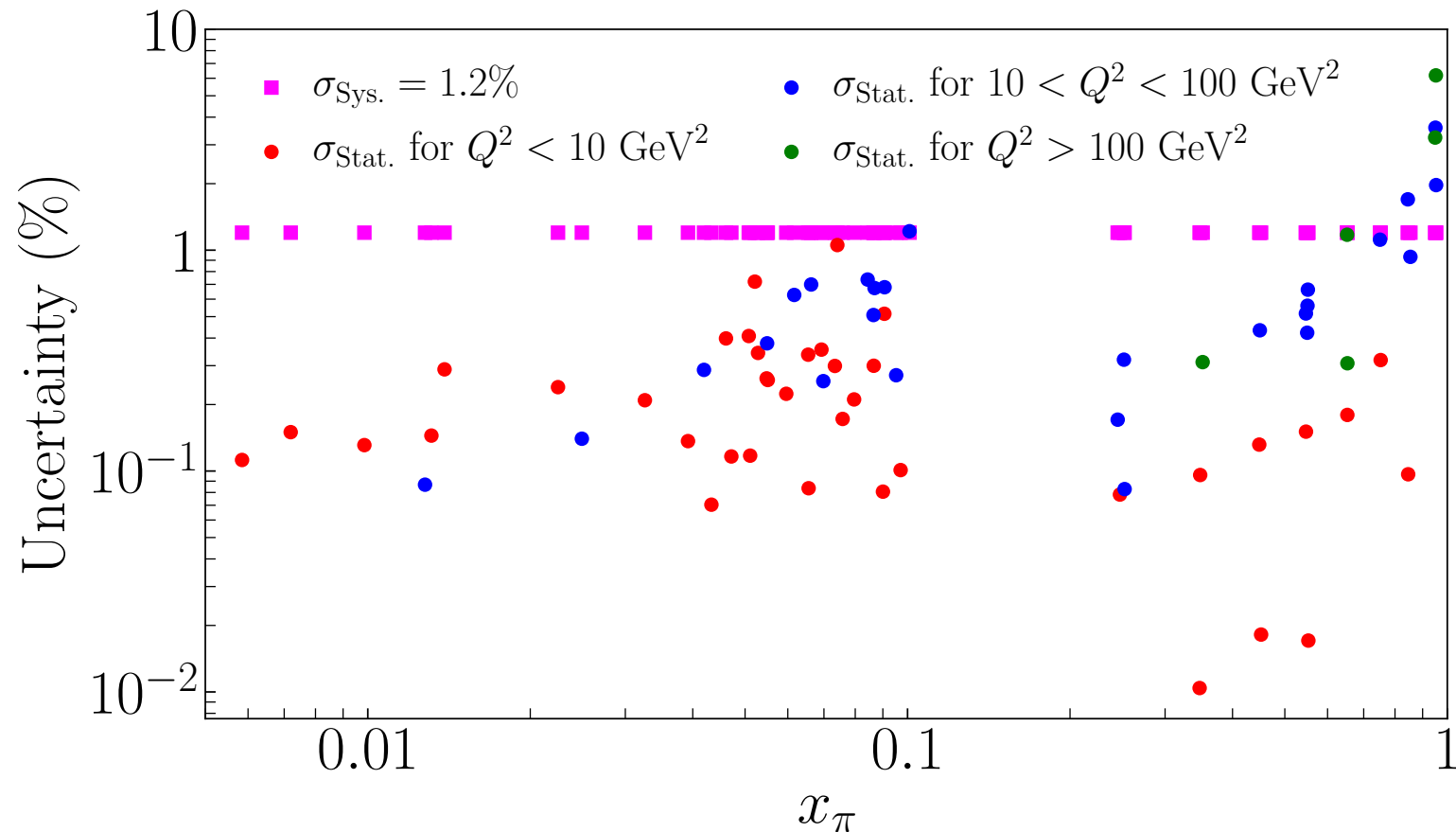
# Future Experiments

- **TDIS** experiment at 12 GeV upgrade from **JLab**, which will tag a proton in coincidence with a spectator proton
- Gives **leading proton observable**, complementary to LN, but with a fixed target experiment instead of collider (HERA)
- Proposed **COMPASS++/AMBER** also give  $\pi$ -induced **DY** data
- Both  $\pi^+$  and  $\pi^-$  beams on carbon and tungsten targets



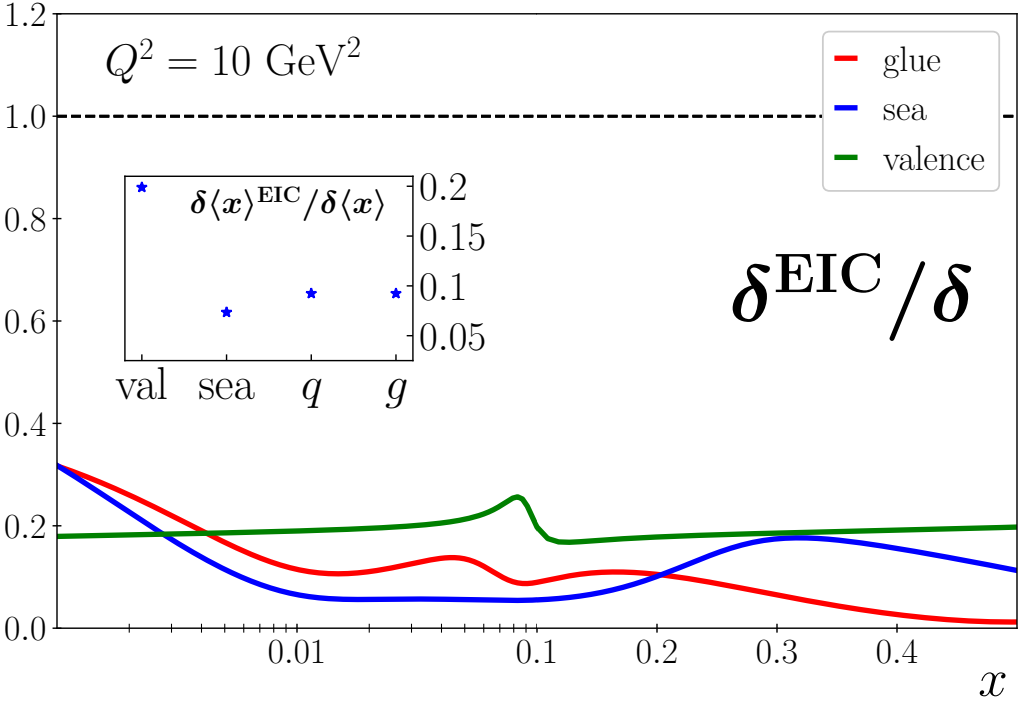
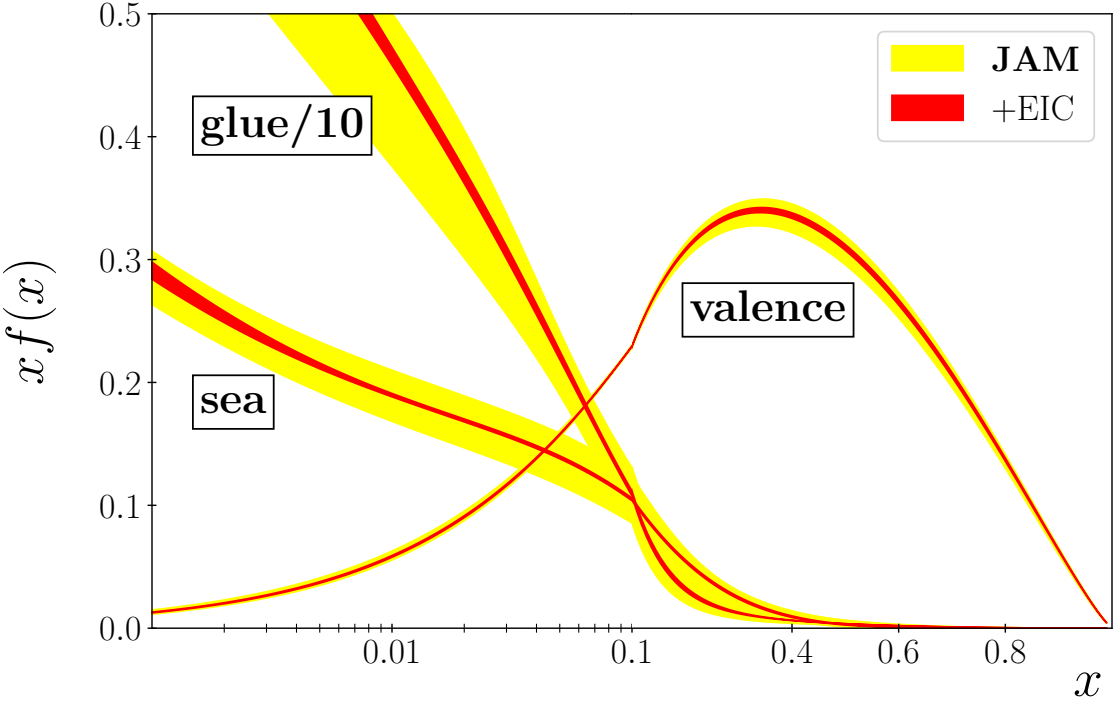
# EIC Impact on Pion PDFs

- Statistical uncertainties are small compared with HERA because of larger luminosity – systematics dominate
- $s = 5400 \text{ GeV}^2$ , 1.2% systematic uncertainty, integrated  $\mathcal{L} = 100\text{fb}^{-1}$



# EIC Impact on Pion PDFs

- Statistical uncertainties are small compared with HERA because of larger luminosity – systematics dominate
- $s = 5400 \text{ GeV}^2$ , 1.2% systematic uncertainty, integrated  $\mathcal{L} = 100\text{fb}^{-1}$



# Future Work

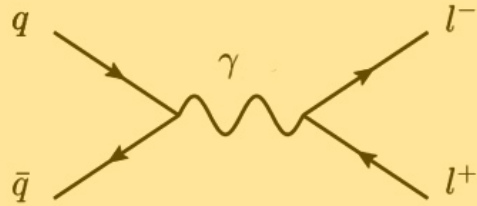
- Investigate high- $x$  behavior of valence PDF through constraints from the **lattice data**
- Perform a **simultaneous extraction** of pion PDFs and TMDs using available low- and high- $p_T$  data
- Explore matching procedure between  $W$  and  $FO$  terms in  $p_T$  spectrum
- Impacts from **future experiments** on pion and kaon PDFs

Thanks to my collaborators: Chueng-Ryong Ji (NCSU), Nobuo Sato (JLab), Wally Melnitchouk (JLab), Leonard Gamberg (Penn State Berks)

# Backup

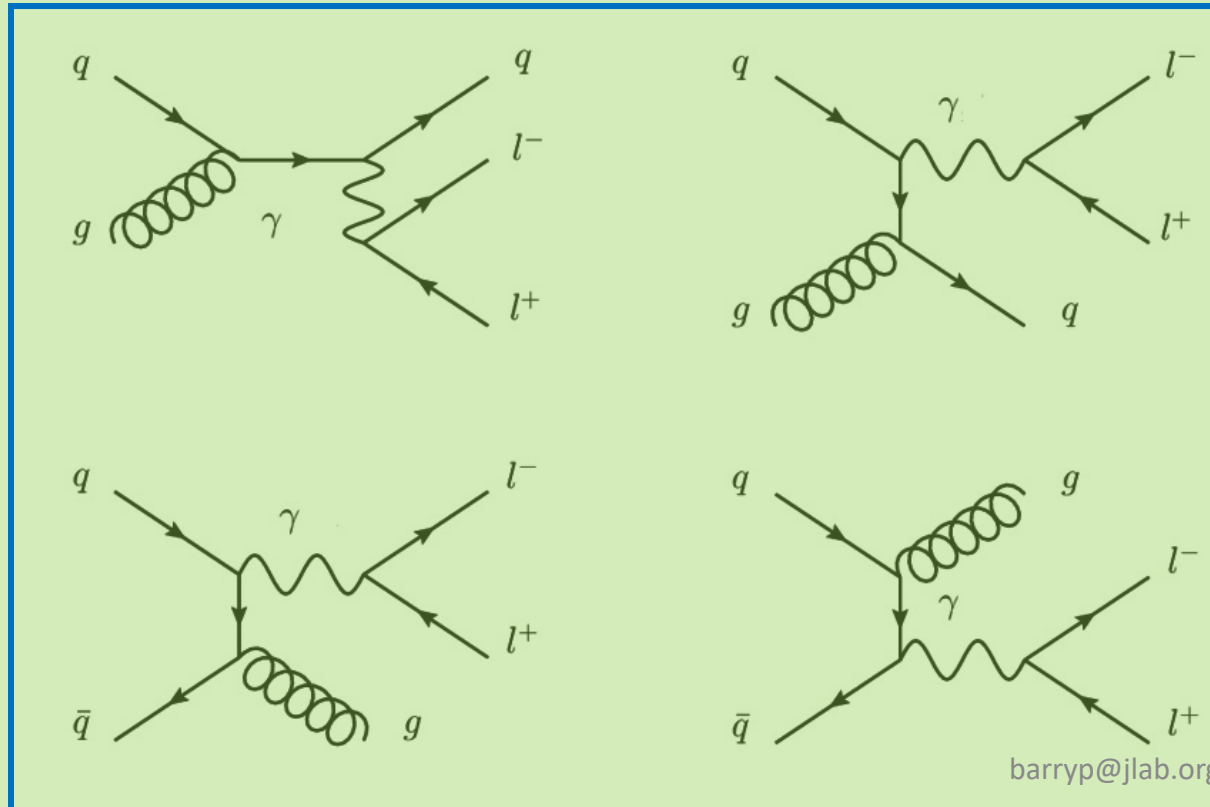
# Fixed Order Up to NLO

LO:  $\mathcal{O}(1)$



Feynman diagrams for DY amplitudes in collinear factorization

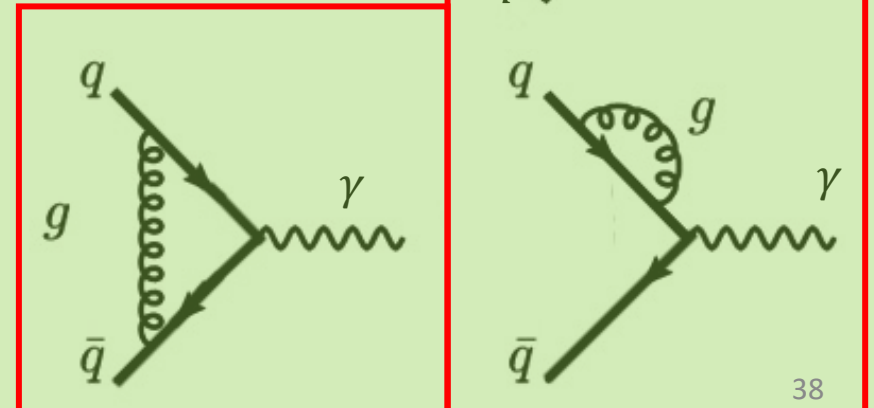
Real emissions



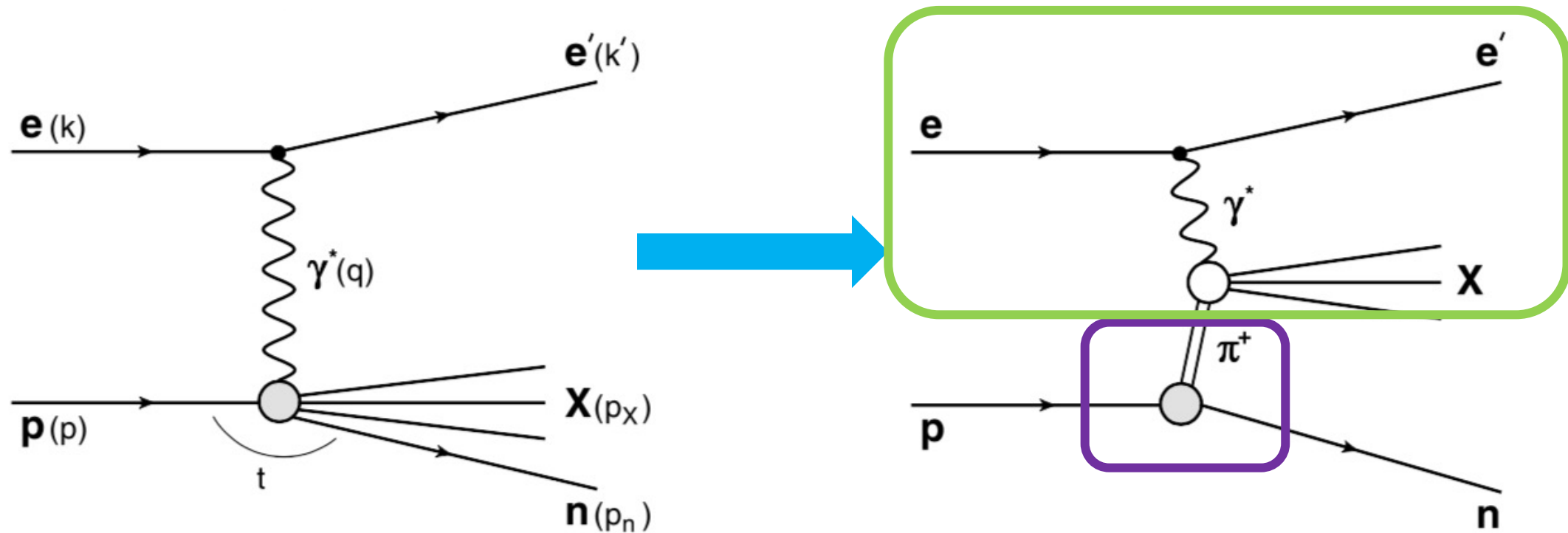
barryp@jlab.org

NLO:  $\mathcal{O}(\alpha_s)$

Virtual Corrections



# Leading Neutron (LN)

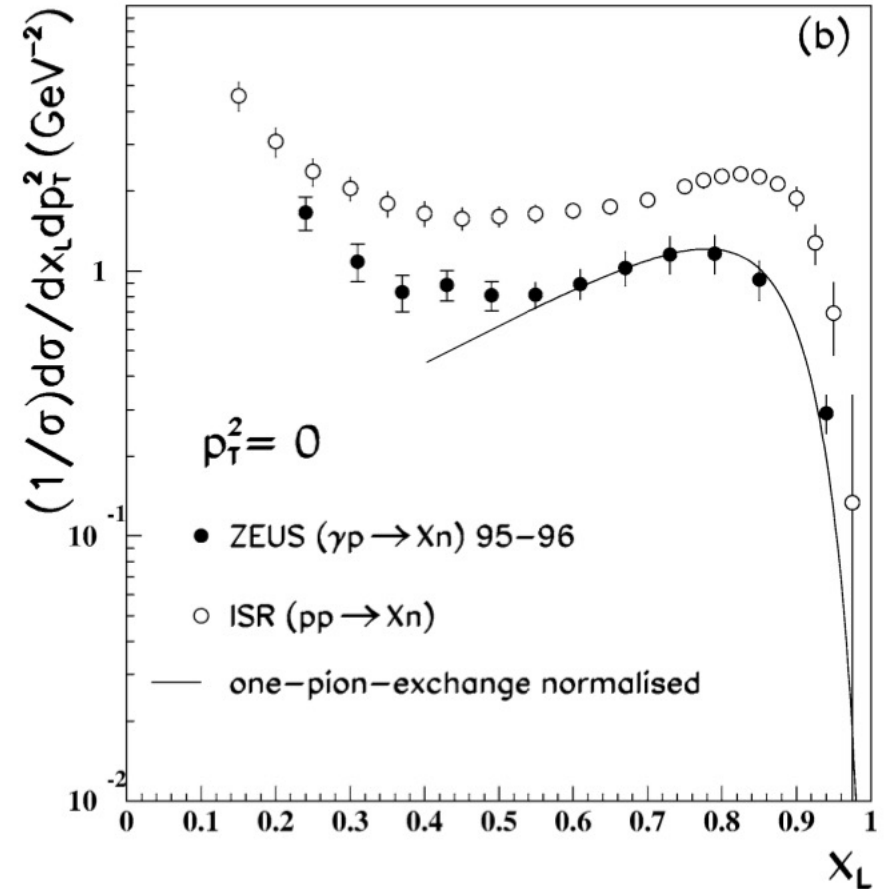
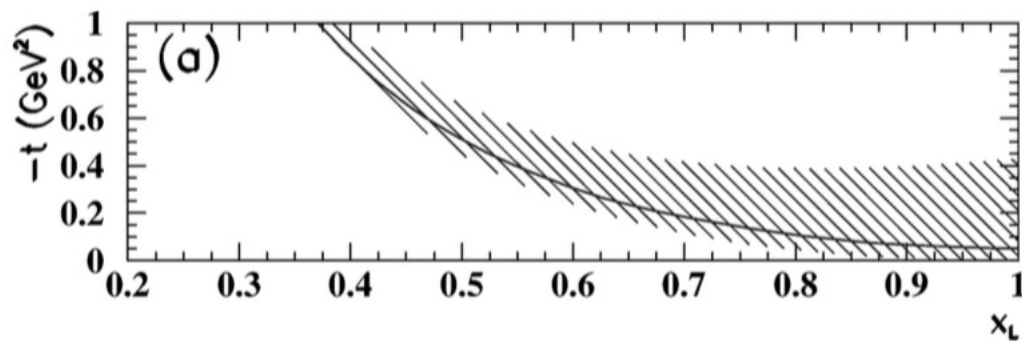


$$\frac{d\sigma}{dx dQ^2 d\bar{x}_L} \propto f_{\pi N}(\bar{x}_L) \times \sum_i \int_{x/\bar{x}_L}^1 \frac{d\xi}{\xi} C(\xi) f_i\left(\frac{x/\bar{x}_L}{\xi}, \mu^2\right)$$

barryp@jlab.org

# Large $x_L$

- $x_L$  is fraction of longitudinal momentum carried by neutron relative to initial proton
- For  $t$  to be close to pion pole, has to go near 0 – happens at large  $x_L$
- In this region, one pion exchange dominates





# Splitting Function and Regulators

Amplitude for proton to dissociate into a  $\pi^+$  and neutron:

$$f_{\pi N}(\bar{x}_L) = \frac{g_A^2 M^2}{(4\pi f_\pi)^2} \int dk_\perp^2 \frac{\bar{x}_L [k_\perp^2 + \bar{x}_L^2 M^2]}{x_L^2 D_{\pi N}^2} |\mathcal{F}|^2,$$

$$D_{\pi N} \equiv t - m_\pi^2 = -\frac{1}{1-y} [k_\perp^2 + y^2 M^2 + (1-y)m_\pi^2]$$

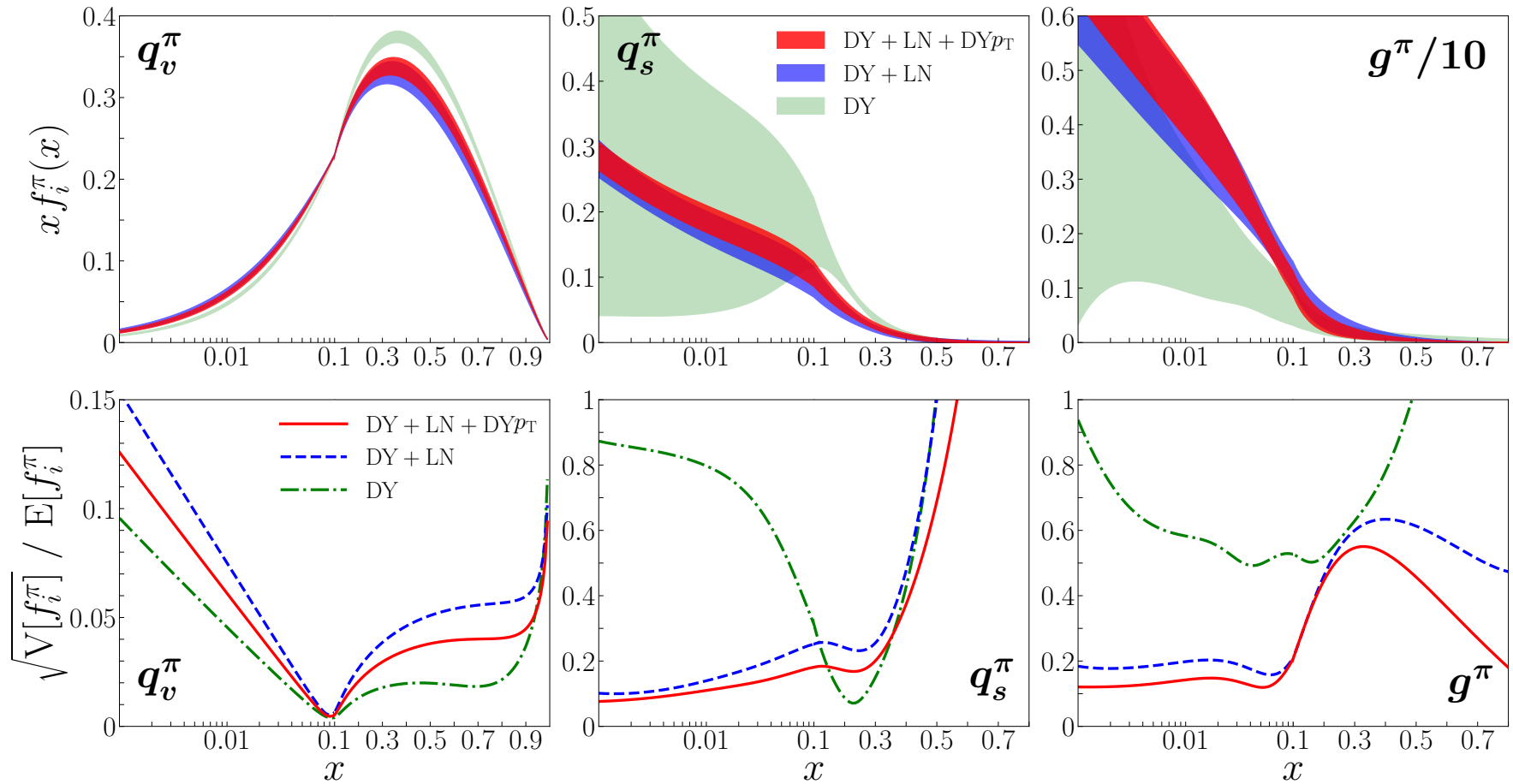
$$\mathcal{F} = \begin{cases} \text{(i)} & \exp((M^2 - s)/\Lambda^2) & \text{s-dep. exponential} \\ \text{(ii)} & \exp(D_{\pi N}/\Lambda^2) & \text{t-dep. exponential} \\ \text{(iii)} & (\Lambda^2 - m_\pi^2)/(\Lambda^2 - t) & \text{t-dep. monopole} \\ \text{(iv)} & \bar{x}_L^{-\alpha_\pi(t)} \exp(D_{\pi N}/\Lambda^2) & \text{Regge} \\ \text{(v)} & [1 - D_{\pi N}^2/(\Lambda^2 - t)^2]^{1/2} & \text{Pauli-Villars} \end{cases}$$

Best fit

- We examine five regulators, and we fit  $\Lambda$
- $\mathcal{F}$  is a UV regulator, which the data chooses

# Effects of Each Dataset

- Largest impact from LN data
- Little impact from the  $p_T$  data except for large  $x$  gluon uncertainties



# Parametrization of the PDF

- We open the shape up a little for the valence (important for resummation in DY)

$$q_v(x_\pi, Q_0^2, \mathbf{a}) = \frac{N}{N'_v} x_\pi^\alpha (1 - x_\pi)^\beta (1 + \gamma x^2)$$

where

$$N'_v = B(2 + \alpha, \beta + 1) + \gamma B(4 + \alpha, \beta + 1)$$

As was done in Aicher et al.

- And for the sea and the gluon, we parametrize by

$$f(x_\pi, Q_0^2, \mathbf{a}) = \frac{N}{N'} x_\pi^\alpha (1 - x_\pi)^\beta$$

where

$$N' = B(2 + \alpha, \beta + 1)$$

# Parameterization of the PDF (in terms of $\pi^-$ )

- We equate the valence distributions:  $\bar{u}_v^{\pi^-} = d_v^{\pi^-}$
- We equate the light sea distributions:  $u^{\pi^-} = \bar{d}^{\pi^-} = u_s^{\pi^-} = d_s^{\pi^-} = s = \bar{s}$
- Normalizations of the valence and sea PDFs are fixed by the sum rules

Quark sum rule  $\int_0^1 dx_\pi q_v^\pi = 1$

Momentum Sum Rule  $\int_0^1 dx_\pi x_\pi (2q_v^\pi + 6q_s^\pi + g^\pi) = 1$

# TMD Non-Perturbative Parametrization

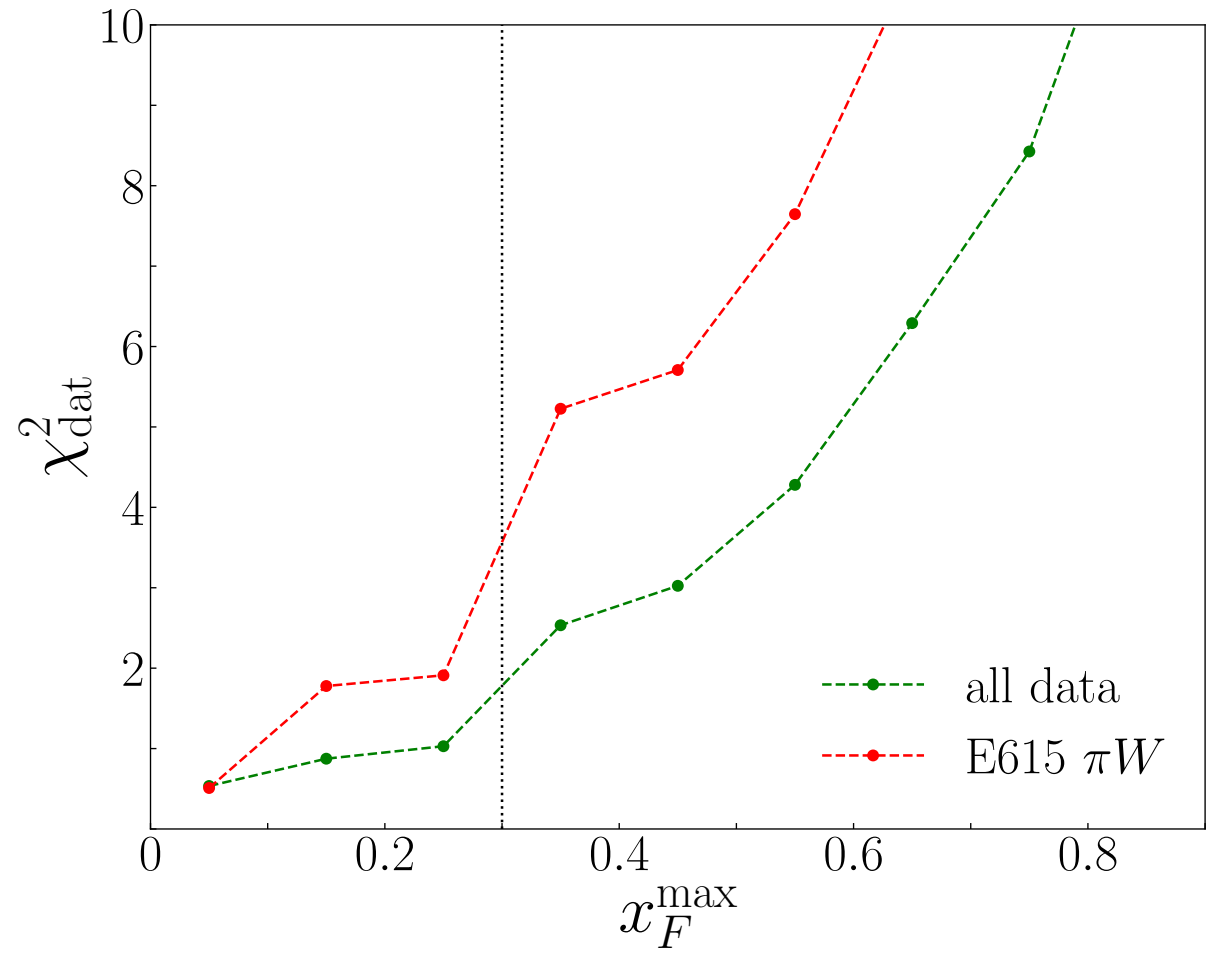
- Parametrize  $g_j$  for each hadron and  $g_K$  as in Slide 21

$$g_{j/h}(x, b_T^2) = -b_T^2 g_1 \left( \frac{1}{2} + g_3 \log(10x) \right)$$

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

# $\chi^2$ as a function of $x_F^{\max}$

- Phenomenologically determine where TMD factorization breaks down in large  $x_F$  data



# E288 $pp$

- Calculate large- $p_T$   $FO$  term with  $\mu = Q$
- **Not** a good description

