

Monte Carlo Analysis of Pion Parton Distributions Using Various Threshold Resummation Methods

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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 \overline{q} constituents



Previous Pion PDFs

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



Experiments to Probe Pion Structure

• Drell-Yan (DY)



 Accelerating pion allows for time dilation and longer lifetime • Leading Neutron (LN)



Datasets -- Kinematics

- Large x_{π} -- Drell-Yan (DY)
- Small x_{π} -- 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







- For the first time, we included large $p_{\rm T}$ -dependent Drell-Yan data, which follows collinear factorization
- Large $p_{\rm T}$ does not dramatically affect the PDF

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

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



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

Add the columns to the rows



Next-to-Leading + Next-to-Leading Logarithm Order Calculation Make sure only counted once! - Subtract the matching NLL NPLL ••• LO 1 ... $\alpha_{\rm s} \log(N)^2$ $\alpha_{\rm s}\log(N)$ NLO ... $\alpha_{\rm S}^2 \log(N)^4$ $\alpha_s^2(\log(N)^2, \log(N)^3)$ NNLO $\alpha_S^k \log(N)^{2k} \quad \alpha_S^k \left(\log(N)^{2k-1} \log(N)^{2k-2} \right)$ $\dots \ \alpha_S^k \log(N)^{2k-2p} + \cdots$ N^kLO

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 $\frac{d\sigma}{dQ^2dY}$ 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 ≈ 1
 2) Keep cosine intact
- Can additionally perform a Double Mellin transform
- Explore the different methods and analyze effects

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 β_{v} parameter

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



Future Work

- Investigate high-x behavior of valence PDF through constraints from the lattice data
- Confront the small- $p_{\rm T}$ Drell-Yan data in terms of CSS formulations and extract pion TMDs
- Investigate impacts of future experiments on pion and kaon PDFs

Backup



Drell-Yan (DY) Definitions

Hadronic variable

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

Partonic variable

 \hat{S} is the center of mass momentum squared of incoming partons

$$z \equiv \frac{Q^2}{\hat{S}} = \frac{\tau}{x_1 x_2}$$





$$C_{q\bar{q}} = \delta(1-z)\frac{\delta(y) + \delta(1-y)}{2}$$

• z = 1 corresponds to partonic threshold

• All
$$\hat{S}$$
 is equal to Q^2

 All energy of hard partons turns into virtuality of photon

NLO Virtual

- Virtual corrections at NLO are proportional to $\delta(1-z)$
 - Exhibit Born kinematics



$$C_{q\bar{q}}^{\text{virtual}} = \delta(1-z)\frac{\delta(y) + \delta(1-y)}{2} \left[\frac{C_F \alpha_S}{\pi} \left(\frac{3}{2}\ln\frac{Q^2}{\mu^2} + \frac{2\pi^2}{3} - 6\right)\right]$$

NLO Real Emission

Real emissions

• Next to leading order, real gluon emissions



NLO Real Emission



Real quark emissions

•
$$C_{qg} = C_{gq}|_{y \to 1-y}$$

$$\begin{split} C_{qg}^{\text{real}} &= \frac{T_F \alpha_S}{2\pi} \Bigg[\delta(y) \Big[(z^2 + (1-z)^2) \ln \frac{Q^2 (1-z)^2}{\mu^2 z} + 2z(1-z) \Big] \\ &+ \Big[1 + \frac{(1-z)^2}{z} y(1-y) \Big] \Big[(z^2 + (1-z)^2) \Big(\frac{1}{y} \Big)_+ \\ &+ 2z(1-z) + (1-z)^2 y \Big] \Bigg] \\ & \xrightarrow{\text{barryp@jlab.org}} \end{split}$$



$$\frac{d\sigma}{dxdQ^2dy} \sim \int_{p \to \pi^+ n}^{1} (y) \times \sum_{q} \int_{x/y}^{1} \frac{d\xi}{\xi} C(\xi) q\left(\frac{x/y}{\xi}, \mu^2\right)$$

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 π^+ and neutron:

$$f_{\pi N}(\bar{x}_{L}) = \frac{g_{A}^{2}M^{2}}{(4\pi f_{\pi})^{2}} \int dk_{\perp}^{2} \frac{\bar{x}_{L} \left[k_{\perp}^{2} + \bar{x}_{L}^{2}M^{2}\right]}{x_{L}^{2} D_{\pi N}^{2}} |\mathcal{F}|^{2},$$

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

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

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

Bayesian Inference

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

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

• Perform N total χ^2 minimizations and compute statistical quantities

Expectation value
$$\mathrm{E}[\mathcal{O}] = \frac{1}{N} \sum_k \mathcal{O}(\boldsymbol{a}_k),$$
Variance $\mathrm{V}[\mathcal{O}] = \frac{1}{N} \sum_k \left[\mathcal{O}(\boldsymbol{a}_k) - \mathrm{E}[\mathcal{O}]\right]^2,$