

5th Joint DNP/JPS Meeting Hawaiʻi, October 24, 2018

First MC global analysis of nucleon transversity — with lattice QCD constraints —

Wally Melnitchouk

with Huey-Wen Lin (MSU) Alexei Prokudin (PSU Berks) Nobuo Sato (Connecticut) Harvey Shows (UCLA)



JLab Angular Momentum collaboration

PRL 120, 152502 (2018)

Motivation

- Along with unpolarized distribution $f_1^q = q^{\uparrow} + q^{\downarrow}$ and helicity distribution $g_1^q = q^{\uparrow} - q^{\downarrow}$, the transversity distribution $h_1^q = q^{\rightarrow} - q^{\leftarrow}$ completes the full set of twist-2 PDFs in a spin-1/2 hadron, such as a nucleon
- Since h_1 is chiral-odd, it is difficult to measure experimentally \rightarrow need at least two hadrons in reaction, *e.g.* SIDIS
 - \rightarrow phenomenology ~3 decades behind f_1 , ~2 decades behind g_1
- Some SIDIS data available from HERMES, COMPASS, ...
 dedicated program under way at Jefferson Lab
- Several global QCD analyses, based on maximum likelihood method (Anselmino *et al.*, Radici *et al.*, Kang *et al.*) have been performed
 Suggest discrepancy with lattice QCD calculations...

Motivation

■ Most analyses find tensor charge $g_T = \int dx \left(h_1^u(x) - h_1^d(x)\right)$ with central values ~ 0.6 (with large uncertainty) → Bhattacharya et al (2015) → Gockeler et al (2005) → Bhattacharya et al (2013)

• Lattice calculations give $g_T \approx 1!$

 \rightarrow problem with data? lattice? analysis? transverse spin crisis?

Perform robust MC analysis of data, with *simultaneous* determination of PDFs & fragmentation functions, to test consistency of experiment and lattice QCD



- JAM Collaboration utilizes Bayesian approach to global analysis, with robust fitting methodology based on MC sampling methods (e.g., iterative Monte Carlo, nested sampling, ...)
 - → allows more flexible parameterizations of input PDFs and fragmentation functions
 - → systematic exploration of the parameter space by computing likelihood function directly
 - → map likelihood function into an MC-weighted parameter sample, allowing rigorous determination of PDF uncertainties
- Methodology used previously to analyze

 - --- fragmentation functions Sato et al. PRD 94, 114004 (2016)
 - pion PDFs Barry, Sato, WM, Ji PRL 121, 152001 (2018)

Analysis of data requires estimating expectation values E and variances V of "observables" O (functions of PDFs) which are functions of parameters

$$E[\mathcal{O}] = \int d^{n} a \,\mathcal{P}(\vec{a}|\text{data}) \,\mathcal{O}(\vec{a})$$
$$V[\mathcal{O}] = \int d^{n} a \,\mathcal{P}(\vec{a}|\text{data}) \left[\mathcal{O}(\vec{a}) - E[\mathcal{O}]\right]^{2}$$

"Bayesian master formulas"

■ Using Bayes' theorem, probability distribution \mathcal{P} given by $\mathcal{P}(\vec{a}|\text{data}) = \frac{1}{Z} \mathcal{L}(\text{data}|\vec{a}) \pi(\vec{a})$

in terms of the likelihood function $\mathcal L$

Likelihood function

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

is a Gaussian form in the data, with χ^2 function

$$\chi^{2}(\vec{a}) = \sum_{i} \left(\frac{\text{data}_{i} - \text{theory}_{i}(\vec{a})}{\delta(\text{data})} \right)^{2}$$

with priors $\pi(\vec{a})$ and "evidence" Z

$$Z = \int d^n a \, \mathcal{L}(\text{data}|\vec{a}) \, \pi(\vec{a})$$

 \rightarrow Z tests if *e.g.* an *n*-parameter fit is statistically different from (*n*+1)-parameter fit

- \Box Standard method for evaluating E, V via maximum likelihood
 - \rightarrow maximize probability distribution

 $\mathcal{P}(\vec{a}|\text{data}) \rightarrow \vec{a}_0$

 \rightarrow if \mathcal{O} is linear in parameters, and if probability is symmetric in all parameters

 $E[\mathcal{O}(\vec{a})] = \mathcal{O}(\vec{a}_0), \quad V[\mathcal{O}(\vec{a})] \to \text{Hessian} \quad H_{ij} = \frac{1}{2}$

$$I_{ij} = \frac{1}{2} \frac{\partial \chi^2(\vec{a})}{\partial a_i \partial a_j} \Big|_{\vec{a} = \vec{a}_0}$$

- In practice, since in general $E[f(\vec{a})] \neq f(E[\vec{a}])$, maximum likelihood method often fails
 - \rightarrow must resort to more robust MC approach

$$E[\mathcal{O}] \approx \frac{1}{N} \sum_{k} \mathcal{O}(\vec{a}_{k}), \quad V[\mathcal{O}] \approx \frac{1}{N} \sum_{k} \left[\mathcal{O}(\vec{a}_{k}) - E[\mathcal{O}] \right]^{2}$$

Data Analysis

$\square Collins a symmetry in e p \rightarrow e h X$

$$A_{UT}^{\sin(\phi_h + \phi_s)} = \frac{2(1-y)}{1+(1-y)^2} \frac{F_{UT}^{\sin(\phi_h + \phi_s)}}{F_{UU}}$$

- ϕ_h azimuthal angle of h
- ϕ_s azimuthal angle of target spin

semi-inclusive structure functions



Factorized parametrization for TMD PDFs

$$f^q(x,k_{\perp}^2) = f^q(x) \, \mathcal{G}_f^q(k_{\perp}^2)$$
 f
collinear PDF

$$\mathcal{G}_{f}^{q}(k_{\perp}^{2}) \propto \exp\left[-\frac{k_{\perp}^{2}}{\langle k_{\perp}^{2} \rangle_{f}^{q}}\right]$$
assume Gaussian shape

Data Analysis

- SIDIS charged-pion data on *p* & *d* targets
 total of 106 data points from HERMES and COMPASS
 - → 4 linear combinations of transversity TMD PDFs and Collins TMD fragmentation function:
 - u, d, \bar{q} transversity PDFs $h_1^q(x) = N_q x^{a_q} (1-x)^{b_q}$
 - favored + unfavored Collins FFs
 - respective widths
- World averaged value of lattice tensor charge $g_T^{\text{latt}} = 1.01(6)$
- Total of 23 parameters:
 - 19 for F_{UT}
 - 4 for F_{UU} (fitted to 978 points from HERMES $\pi \& K$ production)

Data Analysis



 $\rightarrow \chi^2 / N_{\text{dat}} = 68.9 / 106 \approx 0.65$

 \rightarrow for SIDIS-only fit, $\chi^2/N_{\rm dat}$ almost indistinguishable

Transversity distributions



significantly reduced uncertainties with lattice constraint

Transversity distributions



- → distributions do not look Gaussian
- \rightarrow full MC analysis gives $g_T^{\text{full}} = 1.0(1)$
- → SIDIS-only fit gives $g_T^{\text{SIDIS}} = 0.9(8)$, while maximum likelihood analysis would have given $g_T \approx 0.5$
- \rightarrow no tension with lattice data

Future

New data expected from SoLID experiment at Jefferson Lab



Z. Ye, N. Sato et al. PLB 767, 91 (2017)