

# Towards combined QCD global analysis of polarized and unpolarized PDFs and fragmentation functions

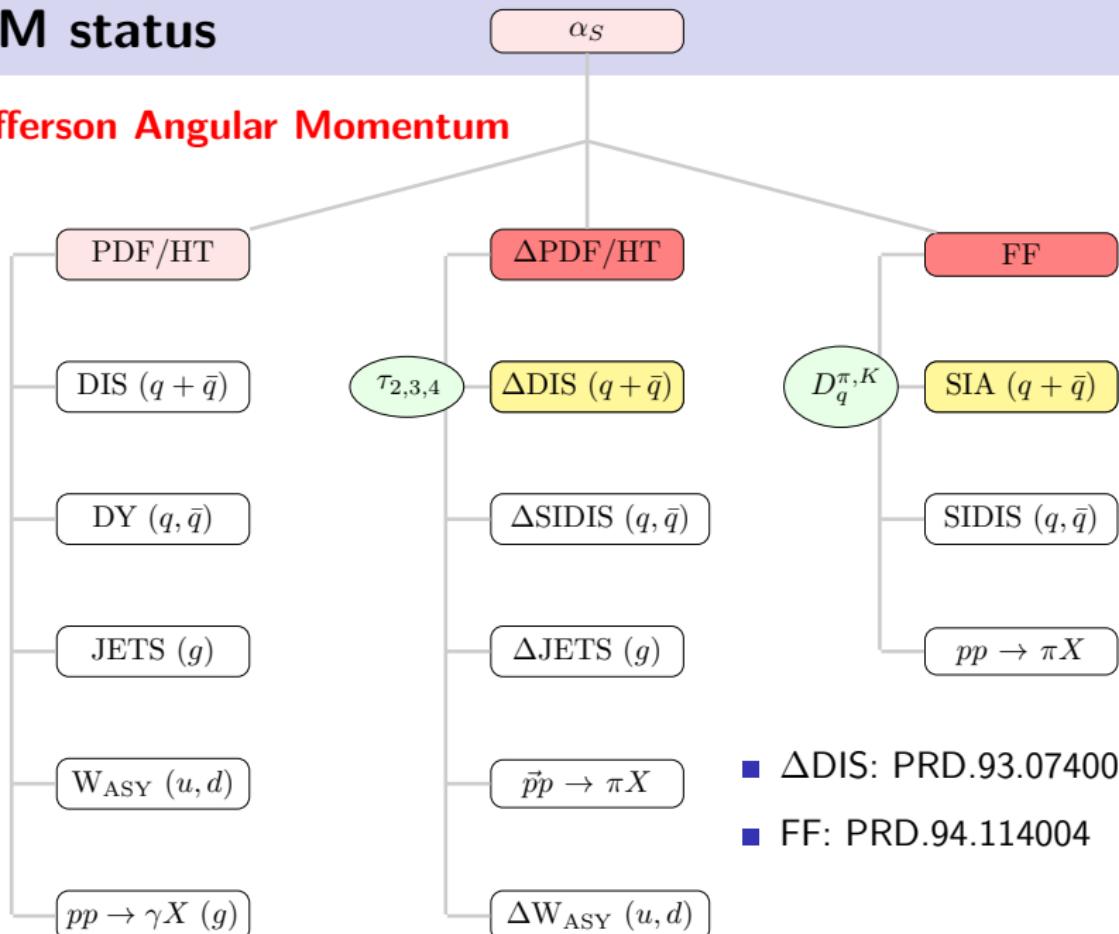
Nobuo Sato



7th Workshop of the APS Topical Group on Hadronic Physics  
Feb, 2017

# JAM status

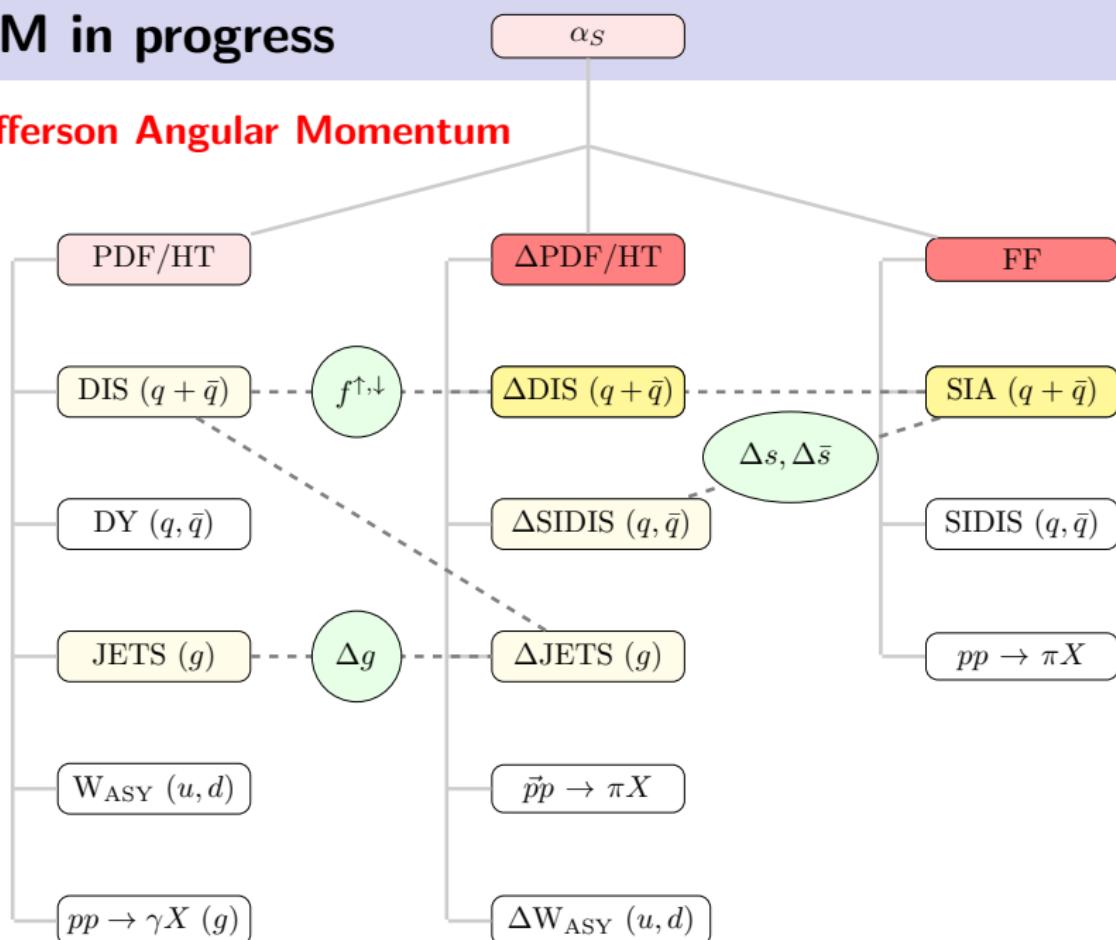
## Jefferson Angular Momentum



# JAM in progress

$\alpha_S$

## Jefferson Angular Momentum



# Theory of fitting

The goal is to estimate:

$$E[\mathcal{O}] = \int d^n a \quad \mathcal{P}(a|data) \quad \mathcal{O}(a)$$

$$V[\mathcal{O}] = \int d^n a \quad \mathcal{P}(a|data) \quad [\mathcal{O}(a) - E[\mathcal{O}]]^2$$

Monte Carlo methods

Maximum Likelihood

- $\mathcal{P}(a|data) \rightarrow \{a_k\}$
- $E[\mathcal{O}] \approx \frac{1}{N} \sum_k \mathcal{O}(a_k)$
- $V[\mathcal{O}] \approx \frac{1}{N} \sum_k [\mathcal{O}(a_k) - E[\mathcal{O}]]^2$
- Maximize  $\mathcal{P}(a|data) \rightarrow a_0$
- $E[\mathcal{O}] \approx \mathcal{O}(a_0)$
- $V[\mathcal{O}] \approx$  hessian,  $\Delta\chi^2$  envelope, ...

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- $V[\mathcal{O}] \approx \text{hessian}, \Delta\chi^2 \text{ envelope}, \dots$

# The $\Delta\chi^2$ paradox

- there is a controversy about how to treat errors in global fits
- A particular example occurred in the IC context
  - Jimenez-Delgado et al PRL114 (2015)  $\Delta\chi^2 = 1$
  - Brodsky, Gardner PRL116 (2016)  $\Delta\chi^2 \neq 1$
- similar issues in other analysis (TMD, FF)
- I will attempt clarify this!!!
- to understand this lets consider an example ...

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- consider two parameters:

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- change of variables

$$\theta_{1,2} \rightarrow t_{1,2} = \frac{\theta_{1,2} - \mu_{1,2}}{\sigma_{1,2}} \rightarrow r, \phi$$

$$\begin{aligned} d\theta_1 d\theta_2 p(\theta_1, \theta_2) &= dt_1 dt_2 \frac{1}{2\pi} \exp \left[ -\frac{1}{2}(t_1^2 + t_2^2) \right] \\ &= d\phi \, r dr \, p(t_1, t_2) = \frac{d\phi}{2\pi} \, r dr \, \exp \left[ -\frac{1}{2}r^2 \right] \end{aligned}$$

## The $\Delta\chi^2$ paradox: Example

- confidence volume:

$$CV = \int_0^{2\pi} \frac{d\phi}{2\pi} \int_0^R dr r \exp\left[-\frac{1}{2}r^2\right] = \int_0^R dr r \exp\left[-\frac{1}{2}r^2\right]$$

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- notice that

$$R^2 = t_1^2 + t_2^2 = \left(\frac{\theta_1 - \mu_1}{\sigma_1}\right)^2 + \left(\frac{\theta_2 - \mu_2}{\sigma_2}\right)^2 \equiv \chi^2$$

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- the *confidence region* for the parameters is

$$\max[t_{1,2}] = R \quad \rightarrow \quad \theta_{1,2} = \mu_{1,2} \pm \sigma_{1,2} R$$

- ... but we know that

$$\theta_{1,2} = \mu_{1,2} \pm \sigma_{1,2}$$

- this is the origin of the paradox.

# The $\Delta\chi^2$ paradox: Example

- The resolution → Use calculus!!

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- **The resolution → Use calculus!!**
- The expectation value:

$$\begin{aligned} \text{E} [\theta_{1,2}] &= \int_0^{2\pi} \frac{d\phi}{2\pi} \int_0^\infty dr p(r, \phi) \theta_{1,2} \\ &= \int_0^{2\pi} \frac{d\phi}{2\pi} \int_0^\infty dr r \exp \left[ -\frac{1}{2}r^2 \right] [\mu_{1,2} + t_{1,2}\sigma_{1,2}] \\ &= \mu_{1,2} \end{aligned}$$

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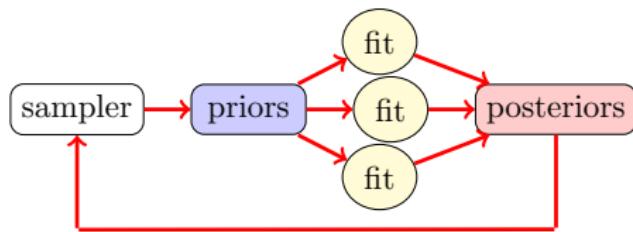
- The variance value:

$$\begin{aligned} \text{V} [\theta_{1,2}] &= \int_0^{2\pi} \frac{d\phi}{2\pi} \int_0^\infty dr p(r, \phi) [\theta_{1,2} - \mu_{1,2}]^2 \\ &= \int_0^{2\pi} \frac{d\phi}{2\pi} \int_0^\infty dr r \exp\left[-\frac{1}{2}r^2\right] [t_{1,2}\sigma_{1,2}]^2 \\ &= \int_0^{2\pi} \frac{d\phi}{2\pi} \int_0^\infty dr r \exp\left[-\frac{1}{2}r^2\right] [r\{\cos, \sin\}(\phi)\sigma_{1,2}]^2 \\ &= \sigma_{1,2}^2 \end{aligned}$$

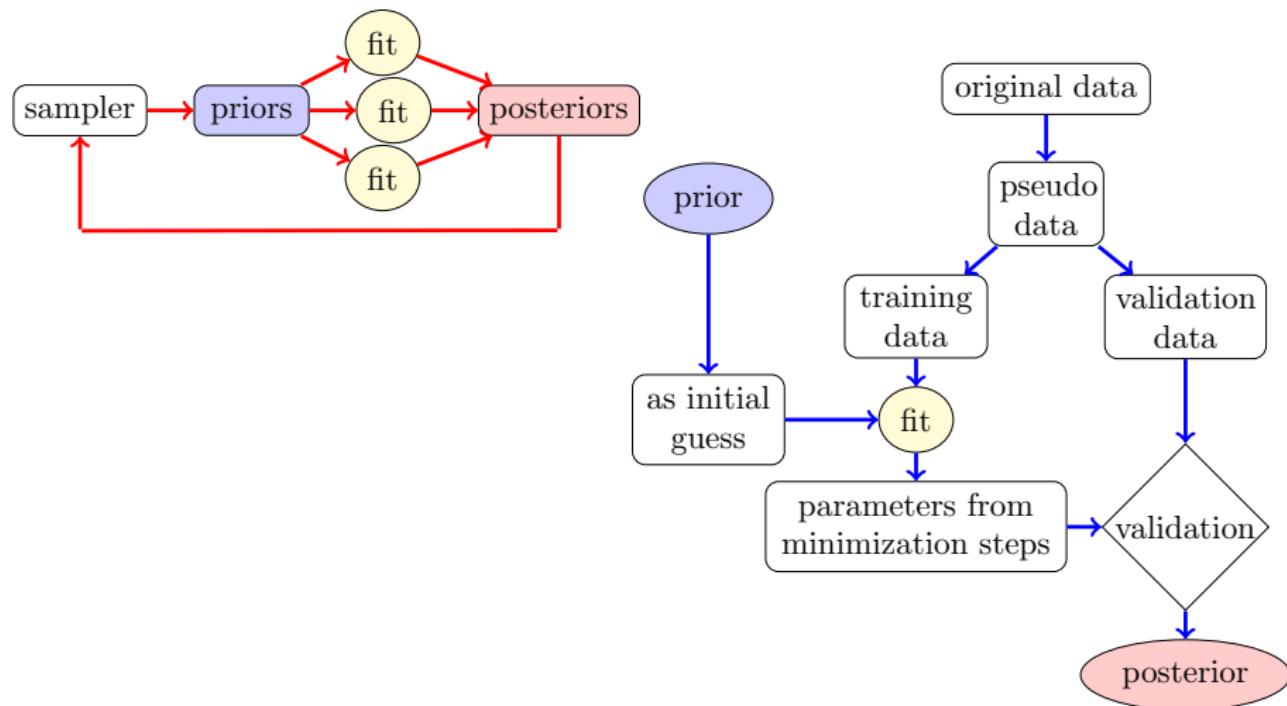
# The $\Delta\chi^2$ paradox: The resolution

- If:
  - the likelihood function is approximately gaussian in the parameters
  - If the model is well described by a linear approximation
- Then,
  - $1\sigma$  CL is characterized by the hessian criterion with  $\Delta\chi^2 = 1$  for any arbitrary number of parameters
- Else,
  - Use MC methods

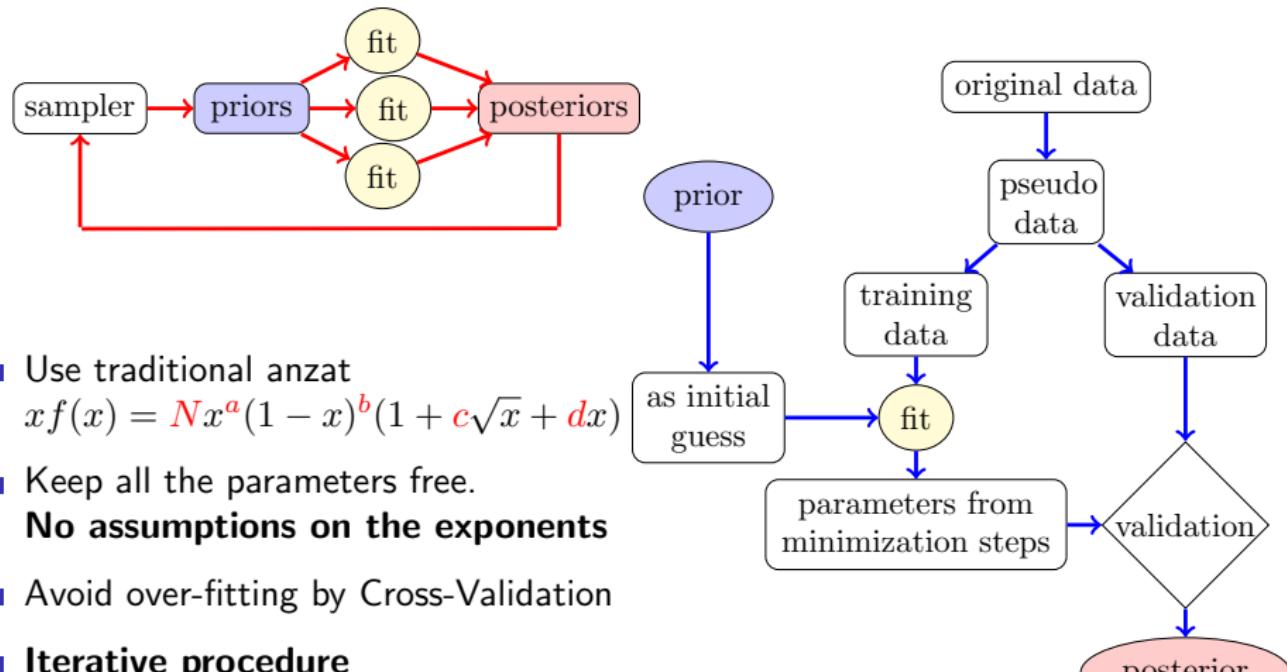
# Iterative Monte Carlo analysis (IMC)



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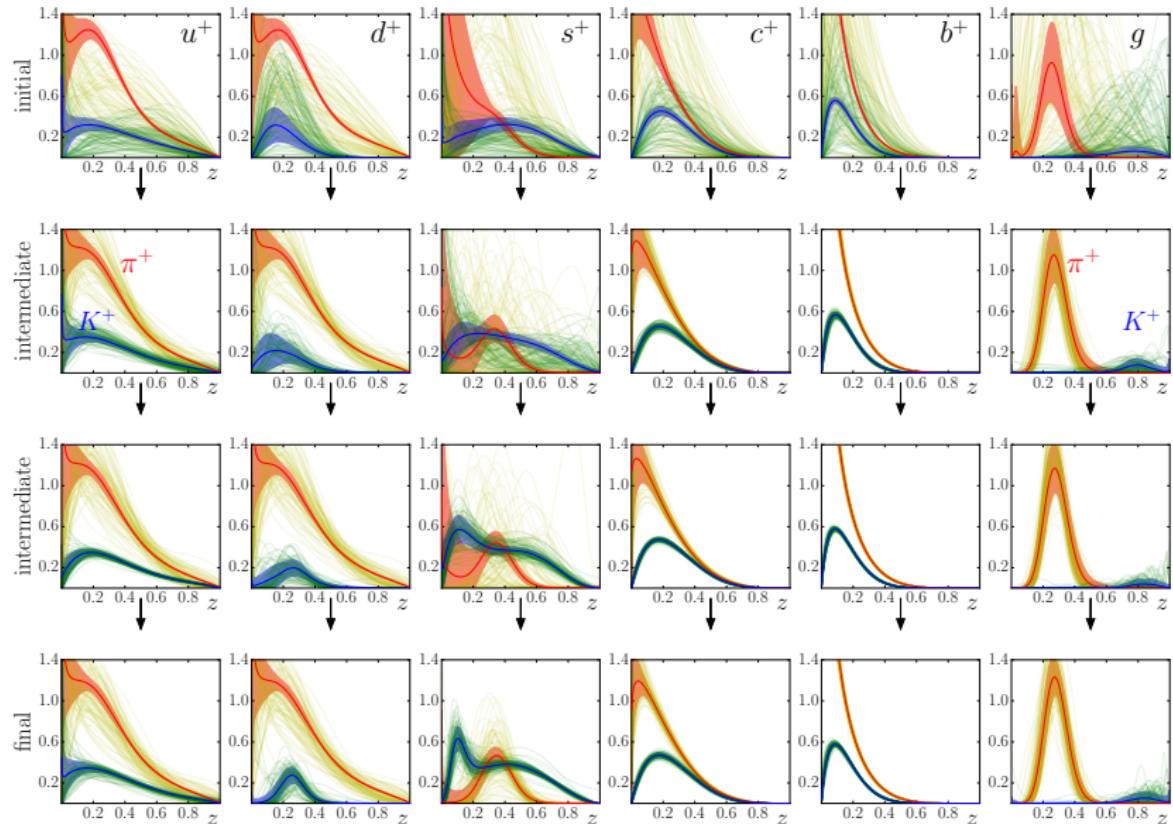


# Iterative Monte Carlo analysis (IMC)



- Use traditional anzat  
 $xf(x) = Nx^a(1-x)^b(1+c\sqrt{x}+dx)$
- Keep all the parameters free.  
**No assumptions on the exponents**
- Avoid over-fitting by Cross-Validation
- **Iterative procedure**  
→ Adaptive MC integration (like in Vegas)
- Robust estimation of uncertainties

# IMC in action (example in FF case)

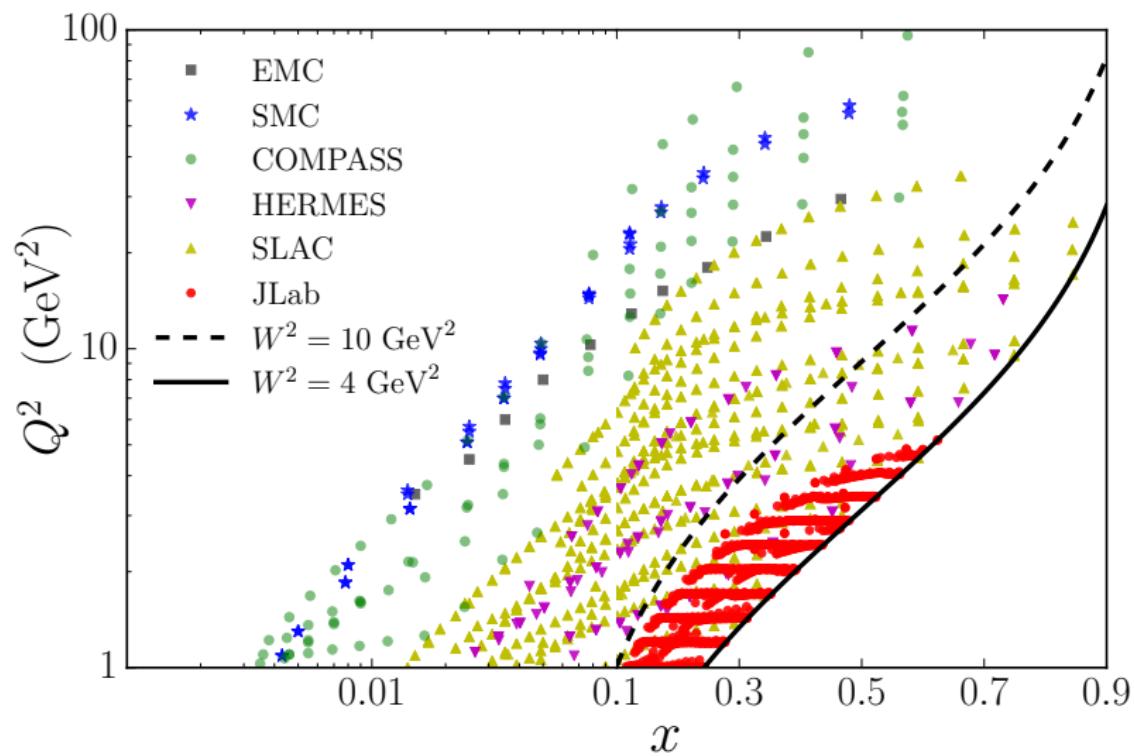


# Spin PDFs from polarized DIS

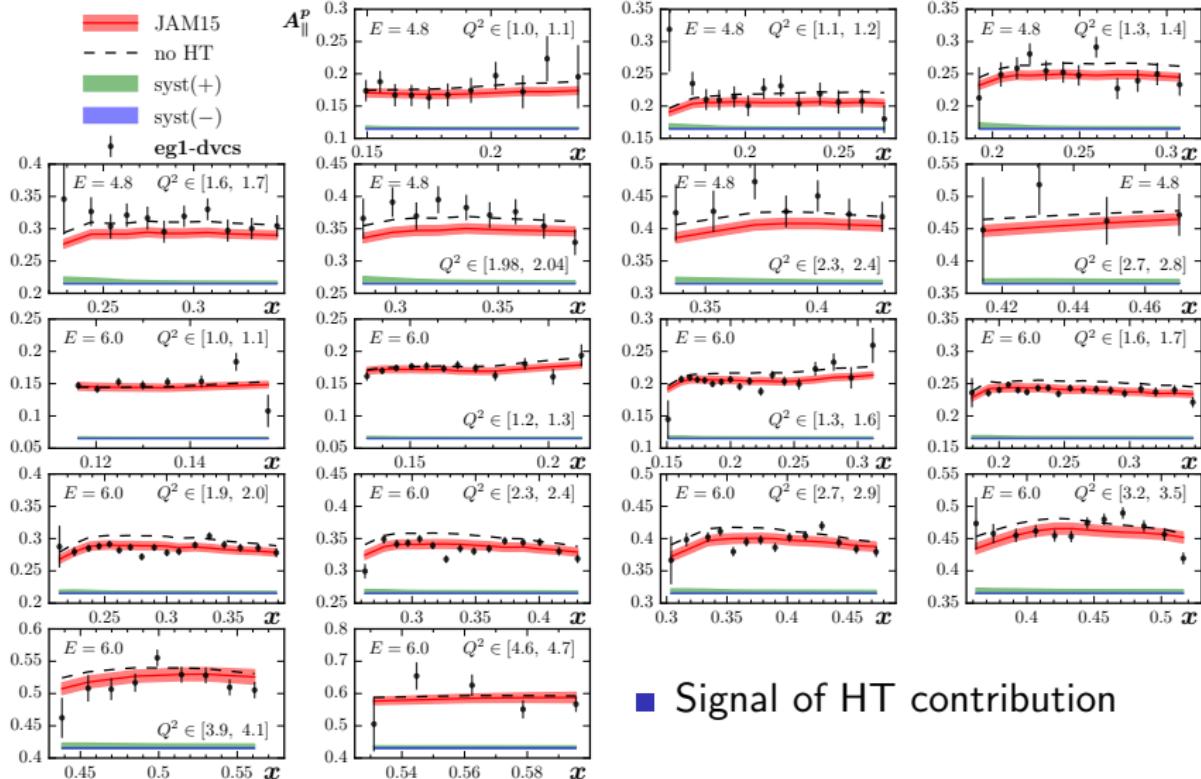
In Collaboration with:

- J. J. Ethier (College of William and Mary)
- W. Melnitchouk (Jefferson Lab)
- A. Accardi (Hampton U. and Jefferson Lab)
- S. E. Kuhn (Old Dominion U.)

# Global polarized DIS data

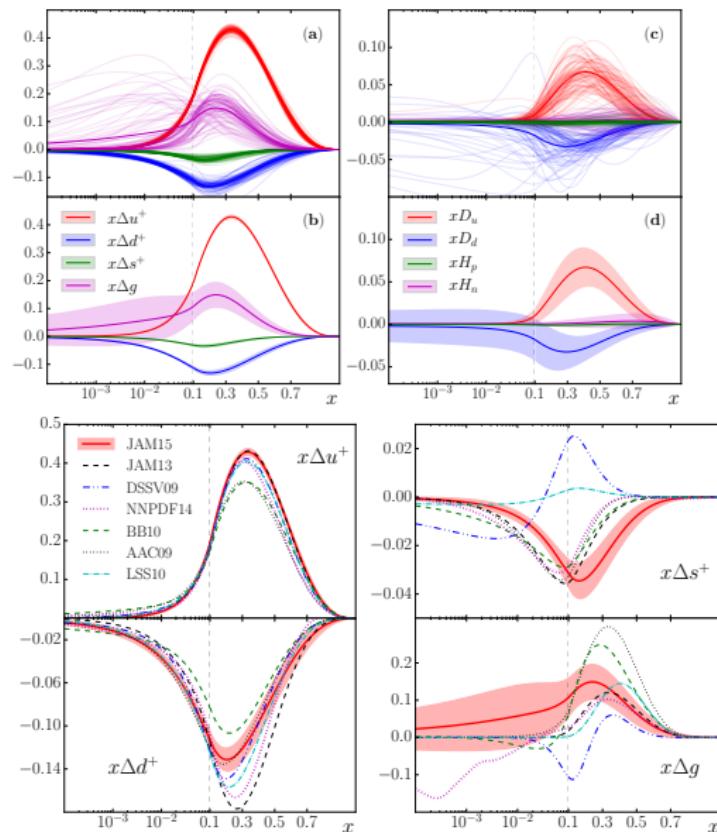


# Data vs theory: proton JLab eg1-dvcs



■ Signal of HT contribution

# Results



moment	truncated	full
$\Delta u^+$	$0.82 \pm 0.01$	$0.83 \pm 0.01$
$\Delta d^+$	$-0.42 \pm 0.01$	$-0.44 \pm 0.01$
$\Delta s^+$	$-0.10 \pm 0.01$	$-0.10 \pm 0.01$
$\Delta \Sigma$	$0.31 \pm 0.03$	$0.28 \pm 0.04$
$\Delta G$	$0.5 \pm 0.4$	$1 \pm 15$
$d_2^p$	$0.011 \pm 0.004$	$0.011 \pm 0.004$
$d_2^n$	$-0.002 \pm 0.002$	$-0.002 \pm 0.002$
$h_p$	$-0.000 \pm 0.001$	$0.000 \pm 0.001$
$h_n$	$0.001 \pm 0.002$	$0.001 \pm 0.003$

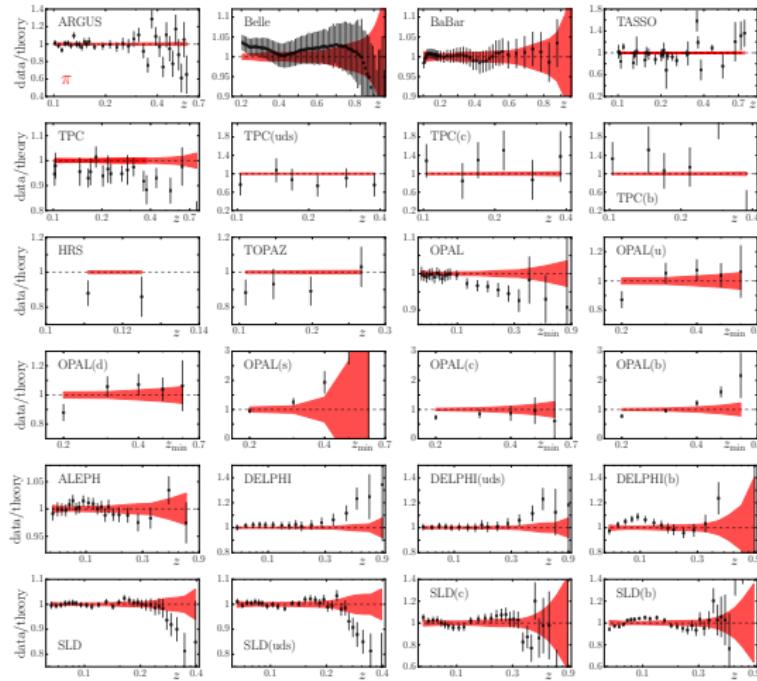
- $\chi^2/N_{npts} = 1.07$
- Sign of  $\tau_3$  distributions is the same as  $\tau_2$
- **Negative  $\Delta s^+$**
- $\Delta g$  compatible with the most recent DSSV fits
- moments of  $\Delta g$  not constrained.

# Fragmentation Functions from SIA data

In Collaboration with:

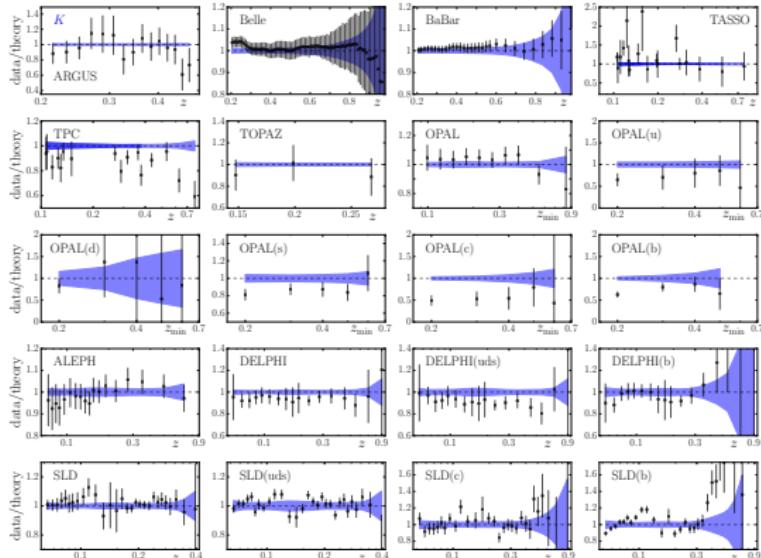
- J. J. Ethier (College of William and Mary)
- W. Melnitchouk (Jefferson Lab)
- A. Accardi (Hampton U. and Jefferson Lab)
- S.Kumano (KEK, J-PARC)
- M.Hirai (Nippon Institute of Technology)

# $\pi$ analysis ( $\chi^2/N_{\text{npts}} = 1.31$ )



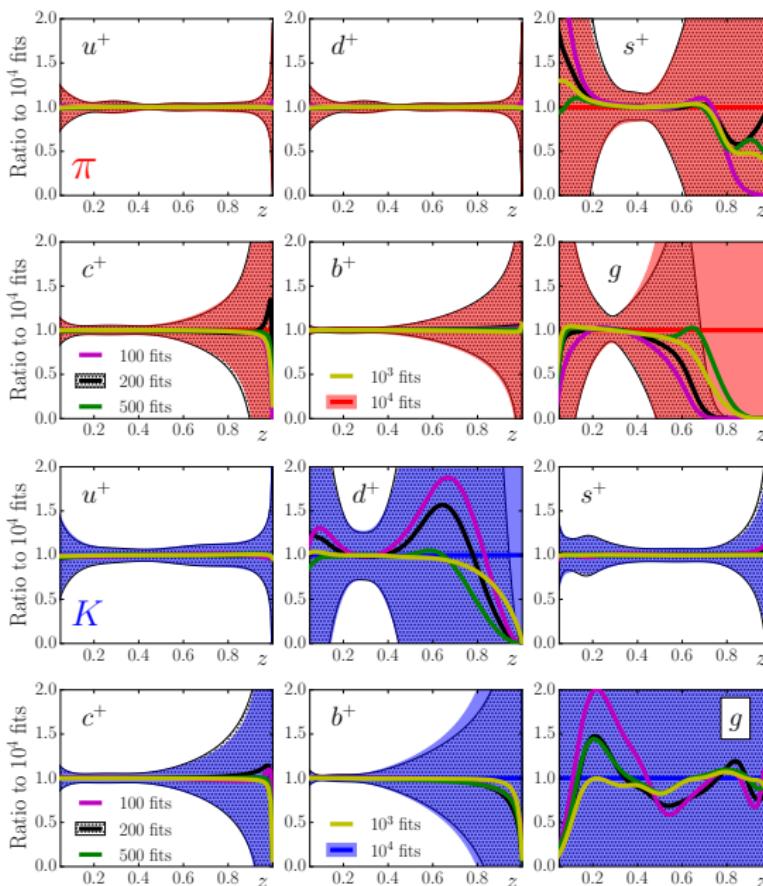
- $z_{\text{cut}} > 0.1$  for low energies
- $z_{\text{cut}} > 0.05$  for high energies
- We use BaBar prompt data set
- Belle data set needs 10% normalization
- Good agreement at low- $z$  for inclusive data sets
- Data inconsistencies at large- $z$  for  $Q^2 = M_z^2$

# $K$ analysis ( $\chi^2/N_{\text{npts}} = 1.01$ )



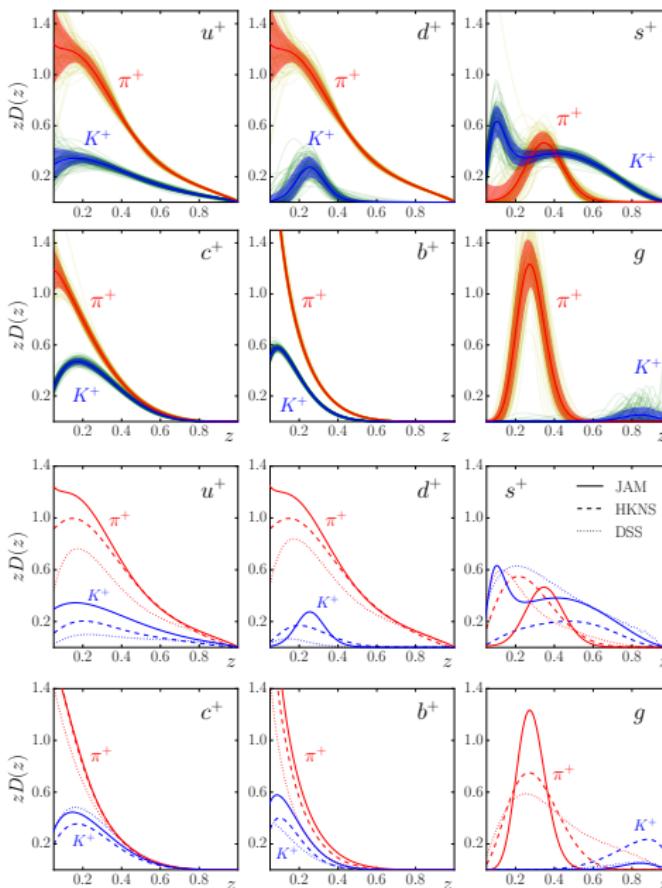
- $z_{\text{cut}} > 0.2$  for low energies to avoid hadron mass corrections
- $z_{\text{cut}} > 0.05$  for high energies
- Smaller  $\chi^2$  than  $\pi$  due to larger errors
- Consistent shapes across all  $z$
- Inconsistencies mostly due to normalizations

# Fragmentation functions



- Larger constraints on favored FFs than unfavored FFs
- More sensitivity on  $D_g^\pi$  than  $D_g^K$
- $D_g^K$  is unknown
- HQ tagged data provides similar constraints for  $\pi$  and  $K$
- Convergence attained with  $\sim 200$  posteriors

# Fragmentation functions



- Similar behavior of unfavored  $D_{s^+}^\pi$  and  $D_{d^+}^K$
- In contrast the  $D_g^\pi$  and  $D_g^K$  behave differently
- The charm and bottom FFs become compatible at large- $z$
- Favored  $D_{u^+}^\pi$  and  $D_{s^+}^K$  have similar shape at large- $z$
- JAM  $D_{s^+}^K$  is compatible with DSS. **Will it change the sign of  $\Delta s^+$ ?**

# Summary and outlook

## Polarized DIS data

- New JAM analysis to study impact of all JLab 6 GeV inclusive DIS data at low  $W$  and high  $x$
- New extraction of LT & HT distributions
- Constraints on  $d_2$

## SIA data

- New study of SIA data including recent Belle and BaBar data
- New extraction of fragmentation functions
- The setup for combined pol DIS, SIA and pol SIDIS is ready!

## JAMLIB

- JAM SPDFs and FFs are available at [github](#)
- Python, Fortran and LHAPDF interfaces provided

# JAM for public

Jefferson Lab Angular Momentum Collaboration

**LINKS**

- Home
- JAMLIB
- References
- Database
- Talks
- Collaboration
- Links

**About**

The JAM (Jefferson Lab Angular Momentum) Collaboration is an enterprise involving theorists and experimentalists from the Jefferson Lab community to study the quark and gluon spin structures of the nucleon by performing global fits of spin-dependent parton distribution functions (PDFs).

Because of the unique capabilities of Jefferson Lab's CEBAF accelerator in measuring small cross sections at extreme kinematics, the JAM spin PDFs are particularly tailored for studies of the large Bjorken- $x$  region, as well as the resonance-deep inelastic scattering transition region at low and intermediate values of  $W^2$  and  $Q^2$ .

**Library**

We provide a collection of codes/scripts (in fortran, python, mathematica), along with interpolation tables for the collinear parton distributions in the nucleon, as well as the collinear parton to hadron fragmentation functions. The library is available at [JAMLIB](#).

**Theory Framework**

Our analysis uses collinear factorization at NLO in perturbative QCD with an emphasis on the large- $x$  region which includes a treatment of higher twist as well as target mass corrections. A treatment of nuclear corrections for deuteron and  ${}^3\text{He}$  targets in DIS data sets is included.

**Results**

**Pion and Kaon fragmentation from JAM16**

Detailed description: Three side-by-side plots showing fragmentation functions. The left plot shows  $D(\pi^+)$  vs  $x$  for  $\pi^+$  and  $K^+$ . The middle plot shows  $D(d^+)$  vs  $x$  for  $\pi^+$  and  $K^+$ . The right plot shows  $D(s^+)$  vs  $x$  for  $\pi^+$  and  $K^+$ . All plots show two curves: a red one peaking at lower  $x$  and a blue one peaking at higher  $x$ .

JeffersonLab/JAMLIB

This repository Search Pull requests Issues Gist

**JAMLIB — Edit**

209 commits 1 branch 0 releases All 4 contributors

Branch master · New pull request Create new file Upload files Find file Code or download · Latest commit · 1 commit · 4 days ago

File	Description	Time
accord FT grids renaming	FF grids renaming renamed table	4 days ago
UHDF	renamed table	6 days ago
fortran	update	a month ago
gallery	ns update	7 days ago
python	update	29 days ago
gplshare	Update README.md	8 days ago
README.md		
README.rst		

**About**

The repository contains interpolation tables for the collinear parton distribution functions in the nucleon, and the collinear