

Global fits for deep inelastic scattering and related processes

Nobuo Sato

ODU/JLab

2019 Fall Meeting of the APS
Division of Nuclear Physics



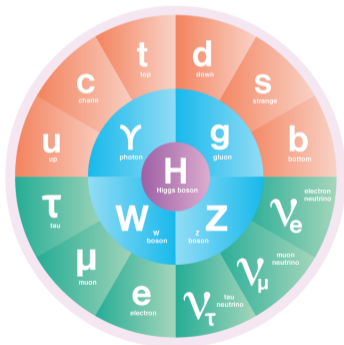
Outline

- Motivations
- Global QCD analysis in a nutshell
- Regression strategies
- JAM 19

Motivations

Motivations

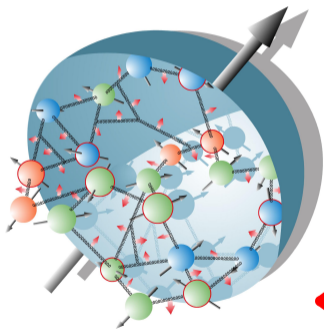
hadrons as **emergent phenomena** of QCD



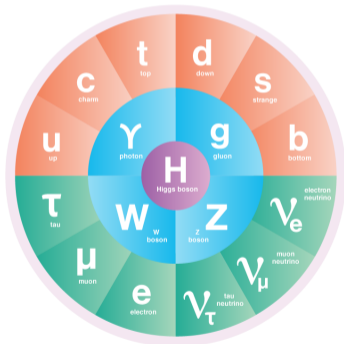
quarks and gluons

Motivations

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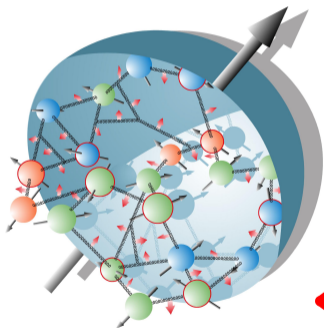
nucleon structure



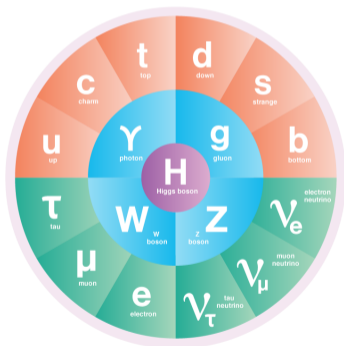
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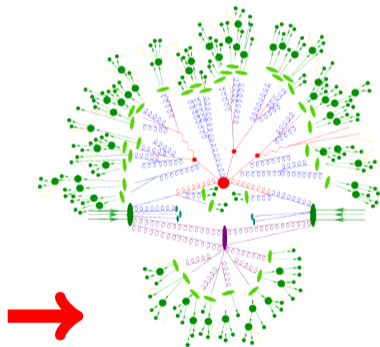
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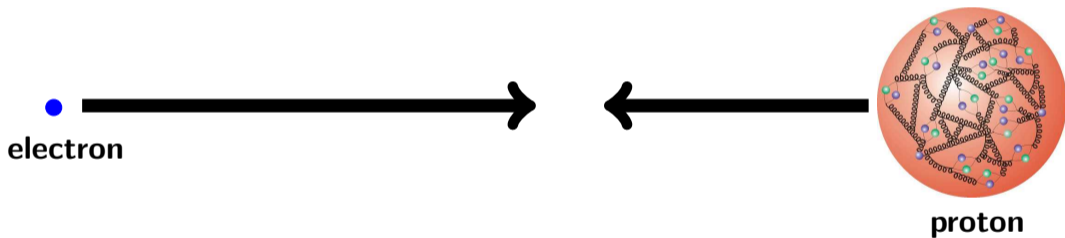


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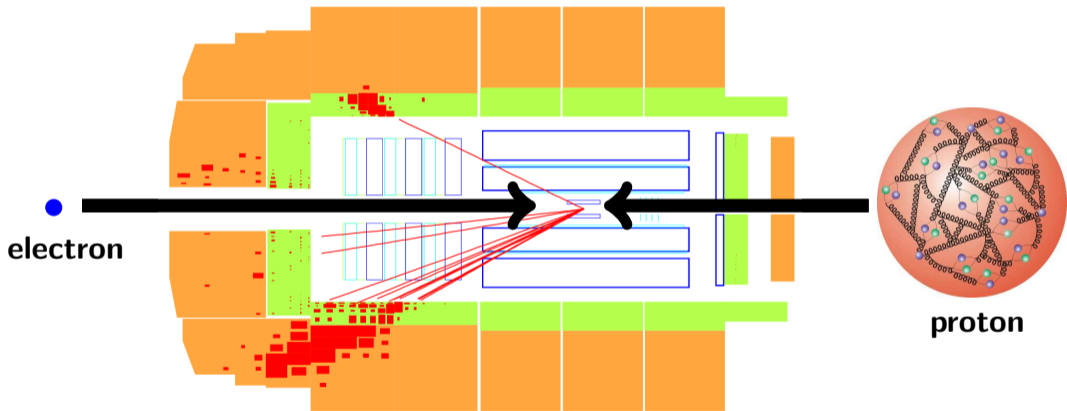


hadronization

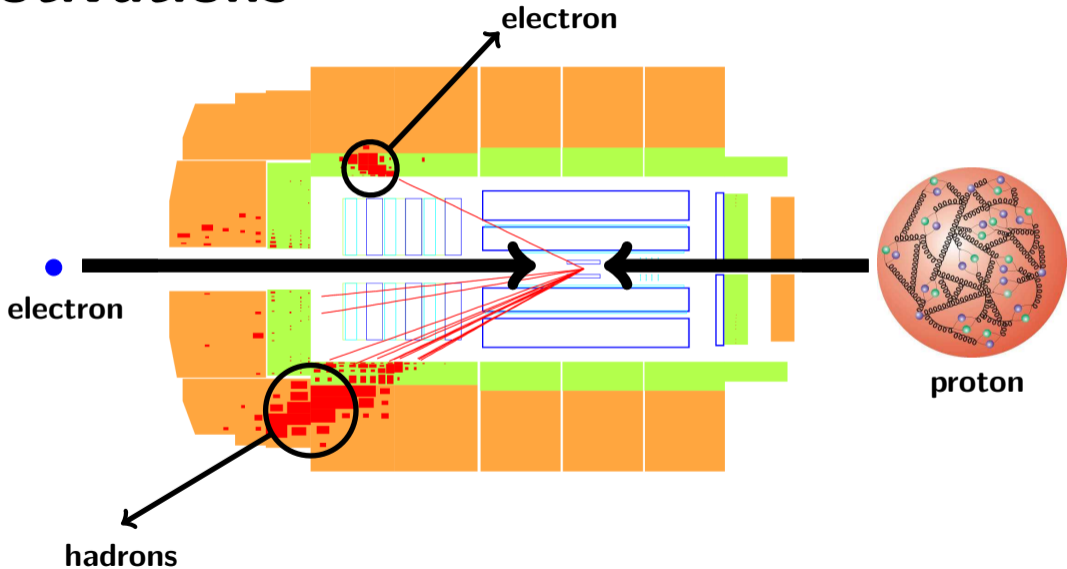
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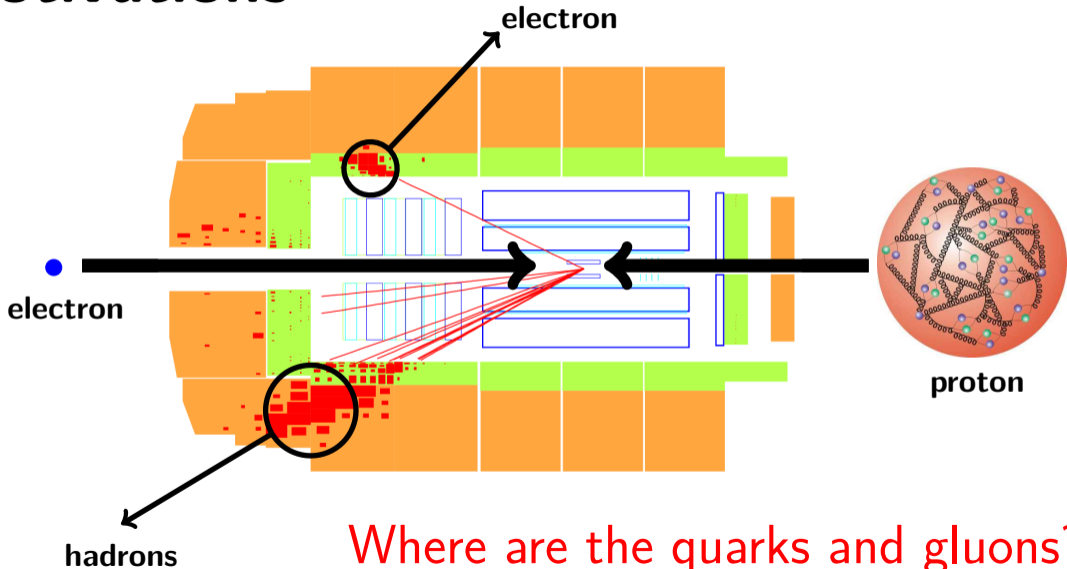
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Motivations

- Quark and gluon d.o.f. **cannot be measured** directly



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- Quark and gluon d.o.f. **cannot be measured** directly
- Experimental measurements **can be interpreted** in terms of quark and gluon d.o.f.



Motivations

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- QCD **factorization** theorems (theory)
- Experimental **cross section** measurements
- Global QCD analysis (Bayesian **regression**)

Motivations

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1. Define **nucleon structure/hadronization** objects in QFT
2. Identify cross sections that **factorize** in terms of such QFT objects
3. Perform a **global QCD analysis**

Motivations

What do we mean by “**structure of nucleon**”? e.g.

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Acronyms for 1D distributions

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- $f_{j/h}(\xi)$: “Parton Distribution Functions”
PDFs

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- $f_{j/h}(\xi)$: “Parton Distribution Functions”

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- $d_{h/j}(\zeta)$: “Fragmentation Functions”

FFs

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- f_j cannot be solved in closed form
→ **inverse problem**

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Another example: SIDIS

$$F_1^h(x, z, Q) = x \sum_j e_j^2 \int_x^1 \frac{d\xi}{\xi} \int_z^1 \frac{d\zeta}{\zeta} C_1(\xi, \zeta, \mu) f_j\left(\frac{x}{\xi}, \mu\right) d_j\left(\frac{z}{\zeta}, \mu\right)$$

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Global QCD analysis in a nutshell

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$$f_j(\xi) = N_j \xi^{a_j} (1 - \xi)^{b_j} P(\xi; \mathbf{w}_j)$$

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$$\mathbf{p} = (\dots, N_j, a_j, b_j, \mathbf{w}_j, \dots, \tilde{N}_j, \tilde{a}_j, \tilde{b}_j, \tilde{\mathbf{w}}_j, \dots)$$

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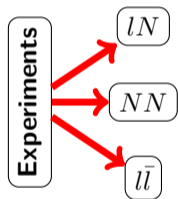
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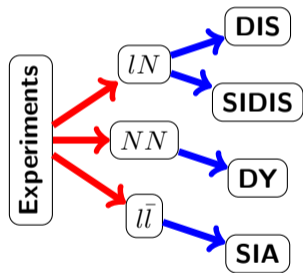
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$$\mathcal{O} = f, d, \sigma, \dots$$

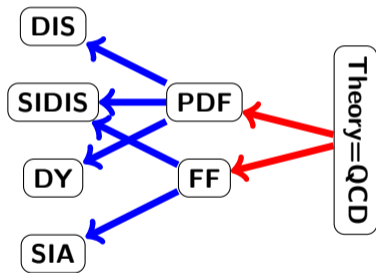
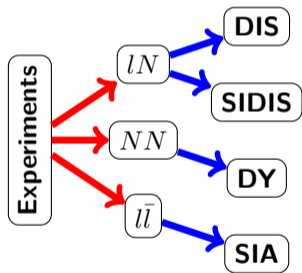
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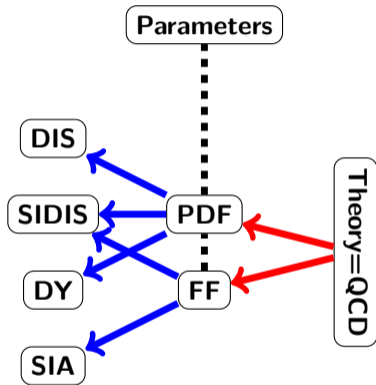
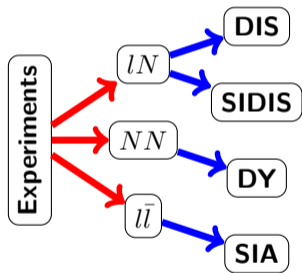
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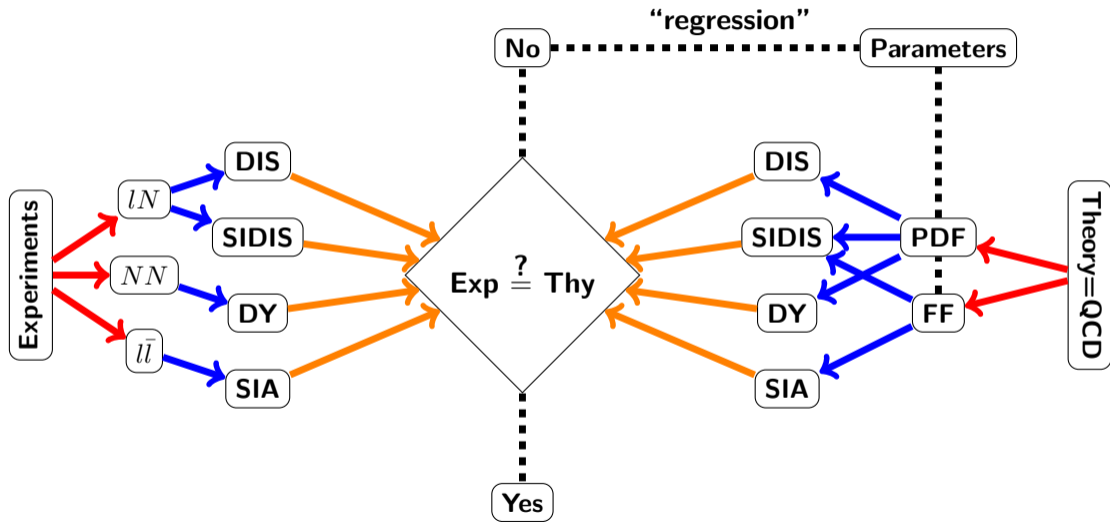
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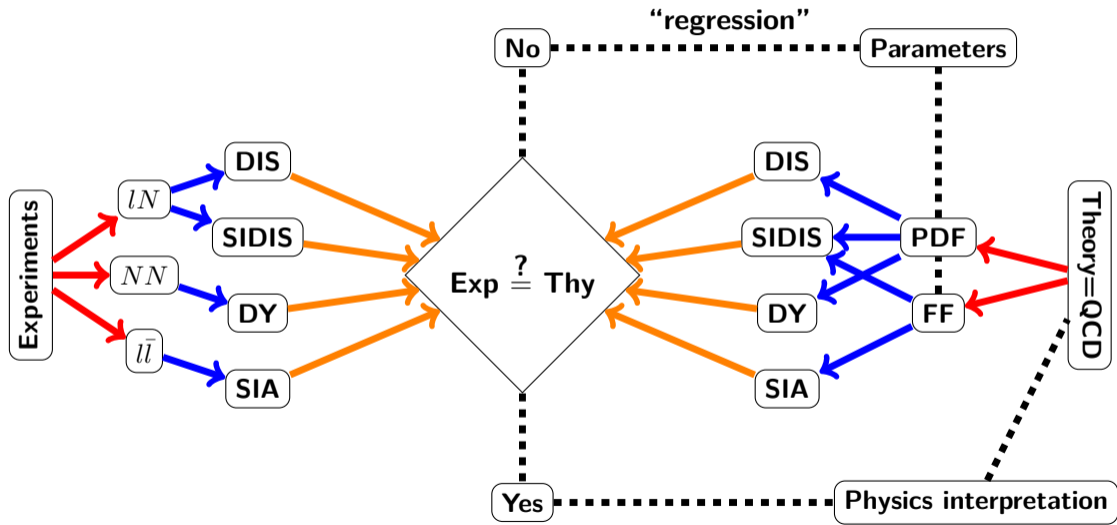
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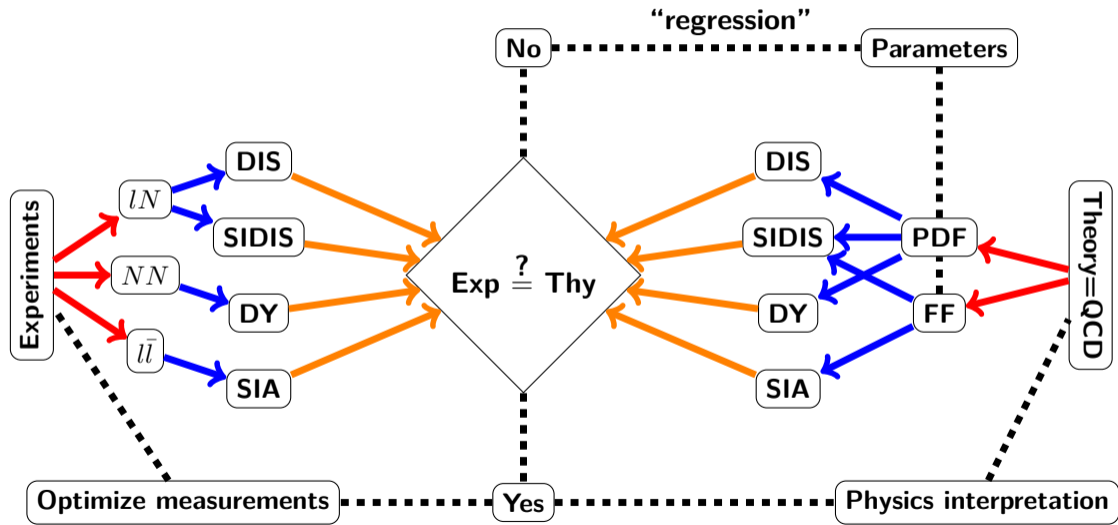
Global QCD analysis in a nutshell



Global QCD analysis in a nutshell



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Regression strategies

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

= hessian, lagrange

Regression strategies

- Data resampling (JAM, NNPDF)

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+ Generate N resampled data

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$$\sigma_{i,k} = \sigma_i + R_{i,k} \delta \sigma_i$$

- + $\{\mathbf{p}_k : 1 \dots N\}$ from N fits to resampled data

- + Use flat priors as guess for the N fits

Regression strategies

- Other approaches

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- + Hybrid Markov Chain (Gbedo, Mangin-Brinet)

Regression strategies

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- + Hybrid Markov Chain (Gbedo, Mangin-Brinet)

- + Nested sampling (JAM)

- challenging for higher dimensions $O(100)$

JAM19: “A less strange proton”

arXiv:1905.03788

NS, Andres, Ethier, Melnitchouk

Session KH: Nucleon Structure I

9:50AM , Wednesday, October 16, 2019

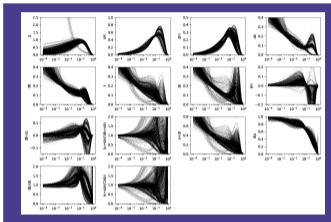
Room: Salon B

The JAM 19 challenge

- Simultaneous extraction of f s and d s
- Dimension of parameter space is $\mathcal{O}(100)$
- NLL evaluation ~ 1 min per point in parameter space

JAM19 multi-step strategy

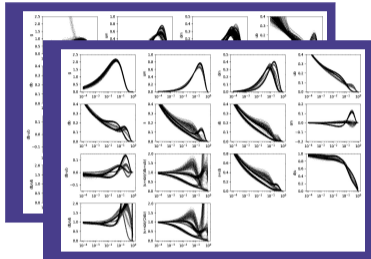
PDFs



+DIS (No HERA)

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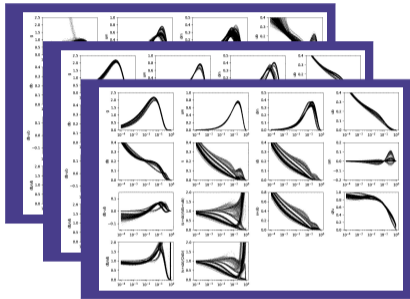


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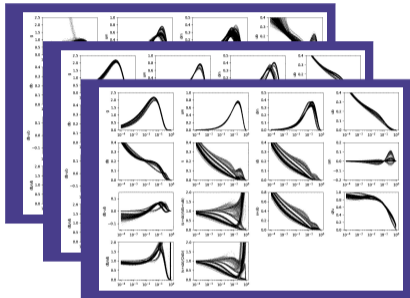
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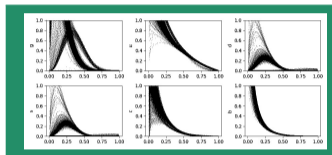
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pion FFs



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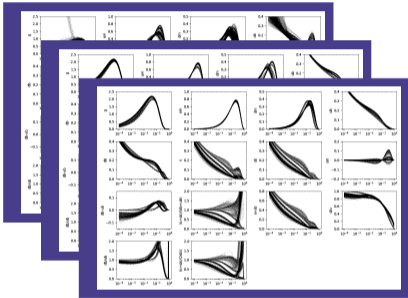
+SIA pions

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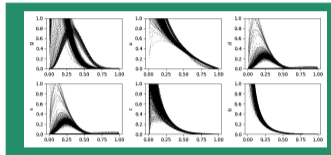
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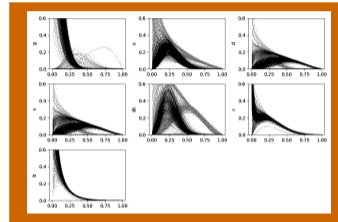
PDFs



pion FFs



kaon FFs



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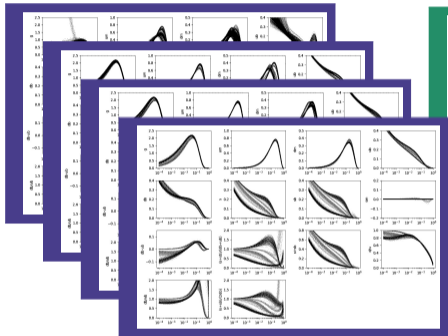
+SIA kaons

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JAM19 multi-step strategy

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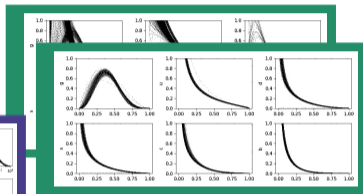


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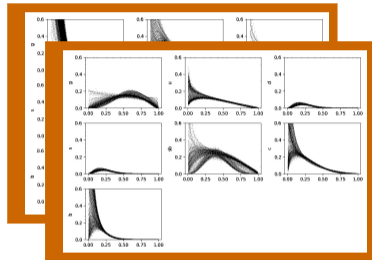
pion FFs



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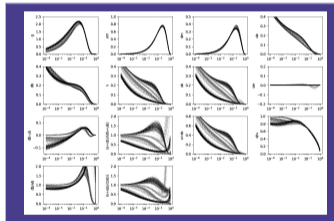
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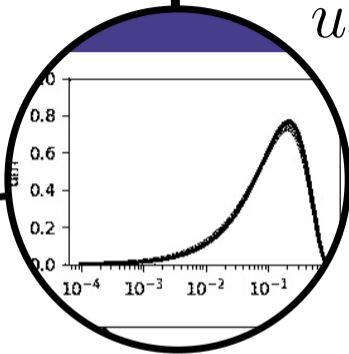
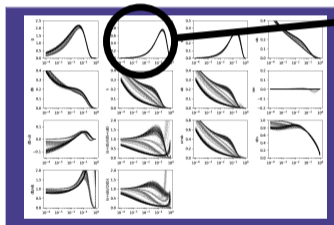
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Discriminating multiple solutions

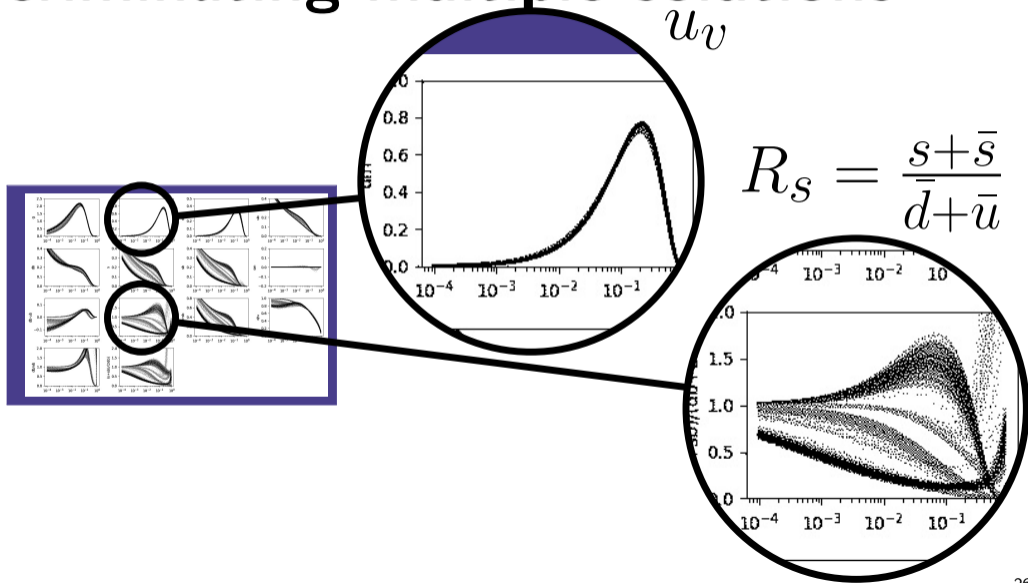


Discriminating multiple solutions

uv

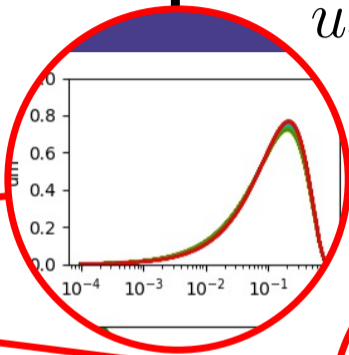
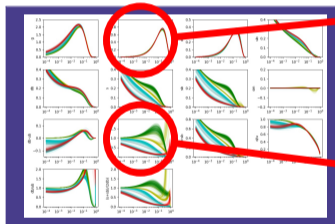


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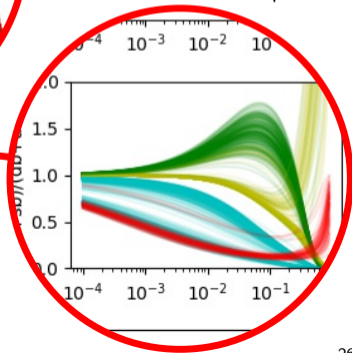


Discriminating multiple solutions

u_v



$$R_s = \frac{s + \bar{s}}{d + \bar{u}}$$



k-means clustering

k -means clustering: 2D example

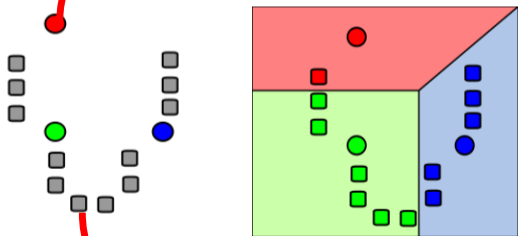
k -means clustering: 2D example

e.g. $f(x) = x^\alpha (1 - x)^\beta$

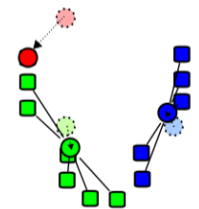
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(α^*, β^*) : centroid



define clusters



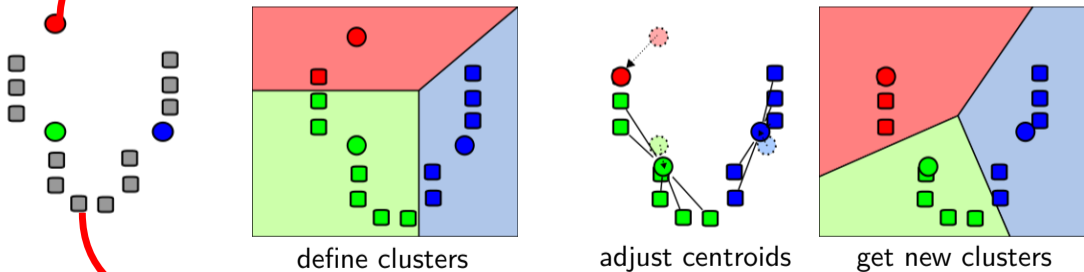
adjust centroids

(α_i, β_i) : replica

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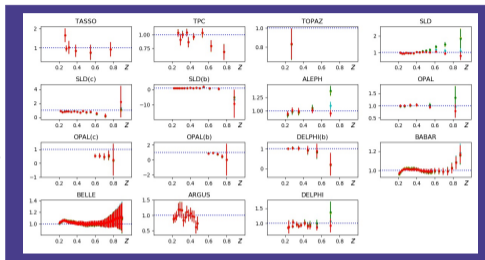
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(α_i, β_i) : replica

data over theory

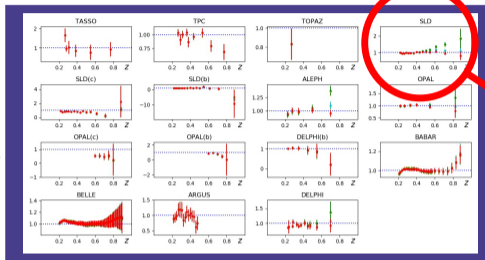
Data/theory



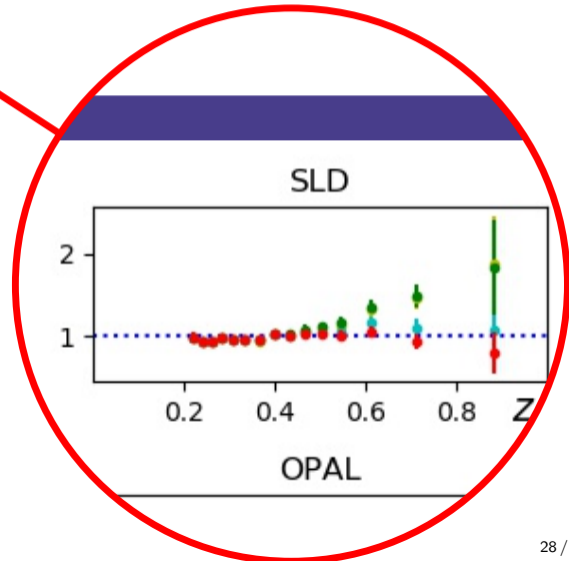
Z

data over theory

Data/theory

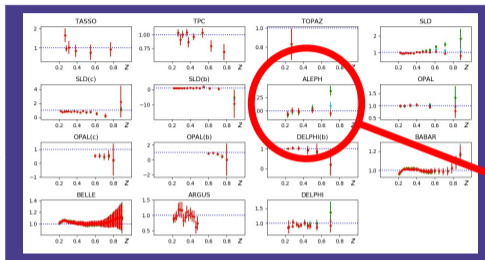


Z

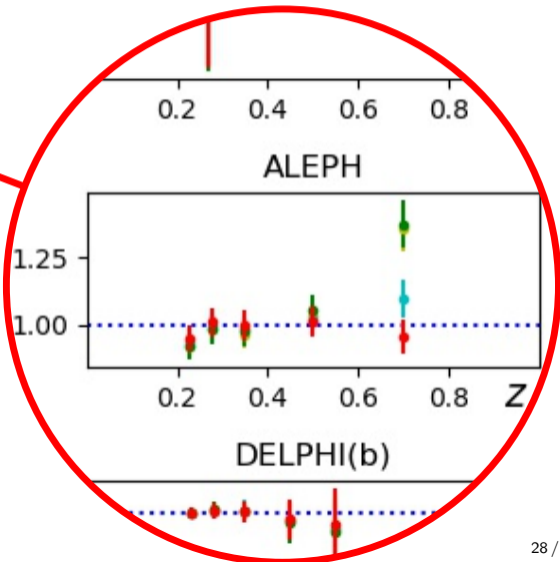


data over theory

Data/theory

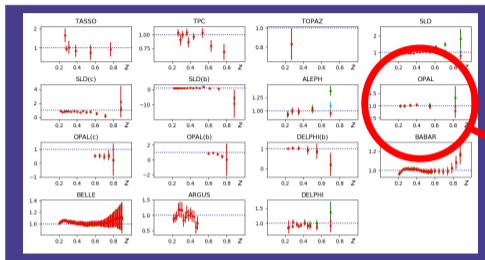


Z

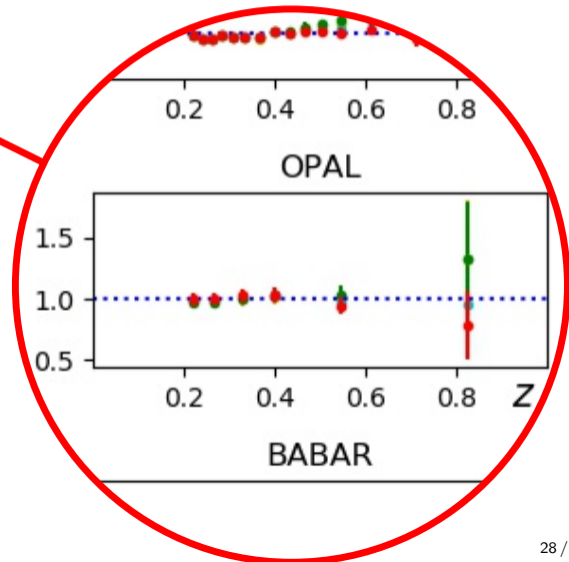


data over theory

Data/theory

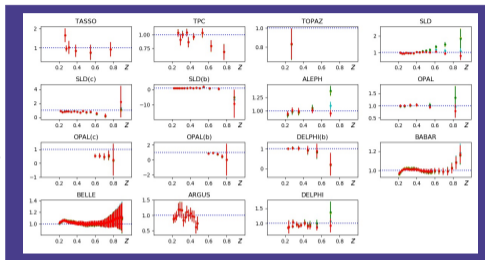


Z



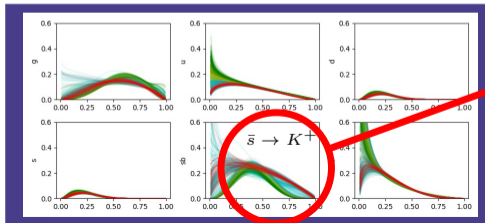
data over theory

Data/theory

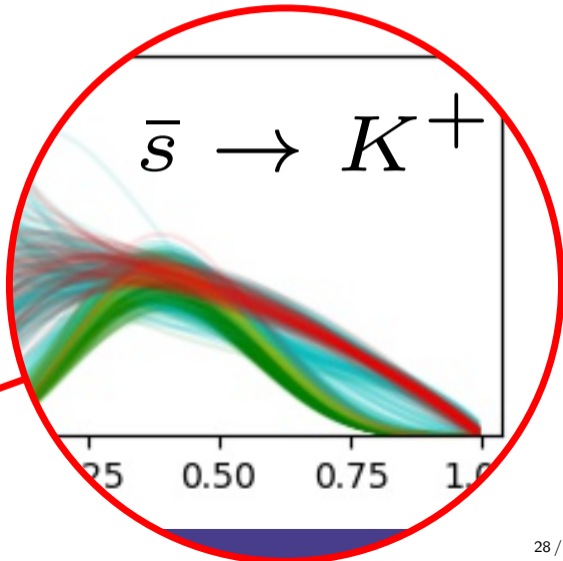


Z

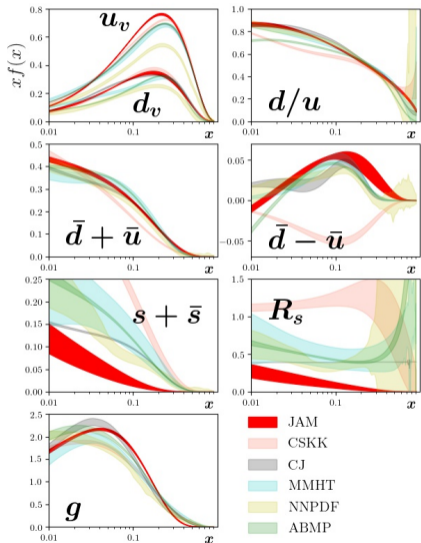
$zD_q^{K^+}$



Z

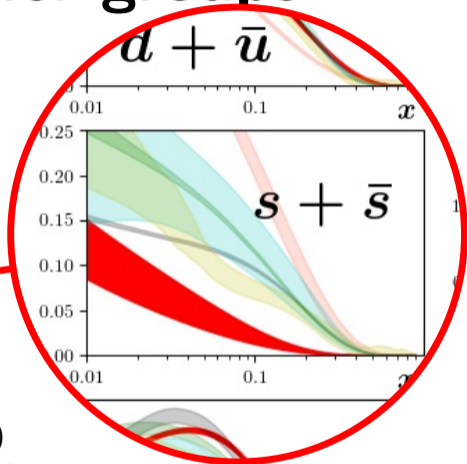
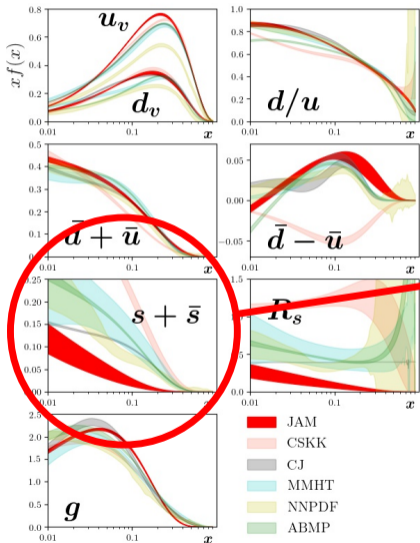


Comparison with other groups



- ✓ DIS (p, d)
- ✓ DY (pp, pd)
- ✓ SIA (π^\pm, K^\pm)
- ✓ SIDIS (π^\pm, K^\pm)

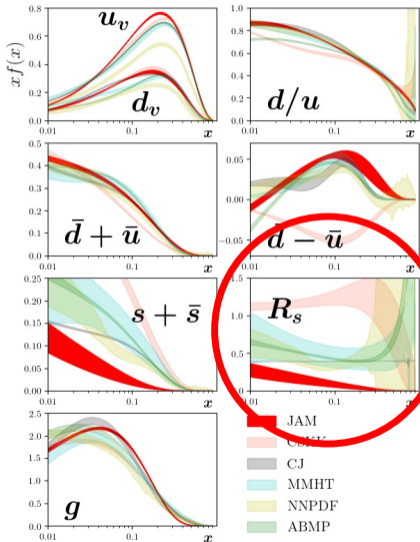
Comparison with other groups



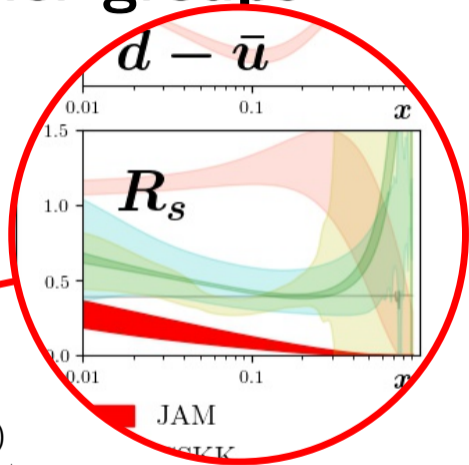
- ✓ DIS (p, d)
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Strong strange suppression

Comparison with other groups

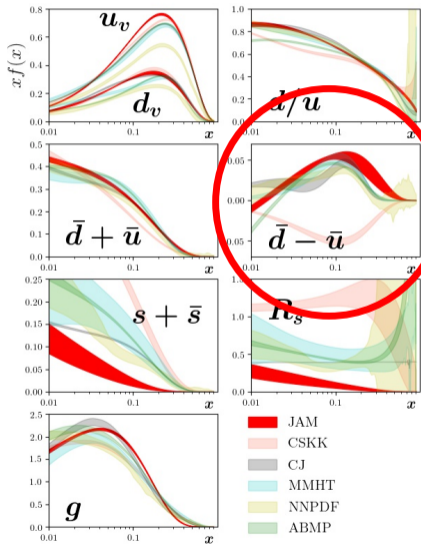


- ✓ DIS (p, d)
- ✓ DY (pp, pd)
- ✓ SIA (π^\pm, K^\pm)
- ✓ SIDIS (π^\pm, K^\pm)

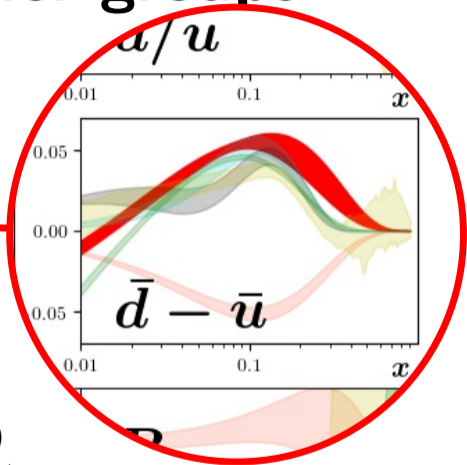


Strong strange suppression

Comparison with other groups



- ✓ DIS (p, d)
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- ✓ SIA (π^\pm, K^\pm)
- ✓ SIDIS (π^\pm, K^\pm)



Large $\bar{d} - \bar{u}$

Summary and outlook

- Understanding hadrons as emergent phenomena of QCD

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+ Factorization theorems

Summary and outlook

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 - + Factorization theorems
 - + Experimental cross sections

Summary and outlook

- Understanding hadrons as emergent phenomena of QCD
 - + Factorization theorems
 - + Experimental cross sections
 - + Global analysis of nucleon structures and hadronization

Summary and outlook

- Challenges of the **inverse problem**

Summary and outlook

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Summary and outlook

- Challenges of the **inverse problem**
 - + Efficient sampling of the posterior distribution
 - + Identification of the **best solution**
 - + Treatment of **non compatible** data sets
(not discussed in this talk)

Summary and outlook

- Next generation of global analysis tools using
Machine Learning

Summary and outlook

- Next generation of global analysis tools using **Machine Learning**

- + **M. Kuchera** Session FE: Mini-Symposium: Towards a US Electron Ion Collider: Physics, Accelerator, and Detectors II
11:00 AM, Tuesday, October 15, 2019
Room: Salon 5

Summary and outlook

- Next generation of global analysis tools using **Machine Learning**

- + **M. Kuchera** Session FE: Mini-Symposium: Towards a US Electron Ion Collider: Physics, Accelerator, and Detectors II
11:00 AM, Tuesday, October 15, 2019
Room: Salon 5
- + **M. Houk & E. Tsitinidi** Session HA: Conference Experience for Undergraduates Poster Session
4:00 PM, Tuesday, October 15, 2019
Room: Salon 1