

Next generation of QCD global analysis tools

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Current paradigm

- Global analysis uses Bayesian regression
- It is done via posterior sampling

$$\rho(\boldsymbol{a}|\text{data}) = \mathcal{L}(\boldsymbol{a}, \text{data})\pi(\boldsymbol{a})$$

- \boldsymbol{a} are the “shape” parameters for QCF

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- We know how to go from $\textcolor{red}{a}$ to cross sections e.g.

$$\frac{d\sigma}{dxdQ^2} = \sum_q \int_x^1 \frac{d\xi}{\xi} H(\xi) f_q \left(\frac{x}{\xi}, \mu; \textcolor{red}{a} \right)$$

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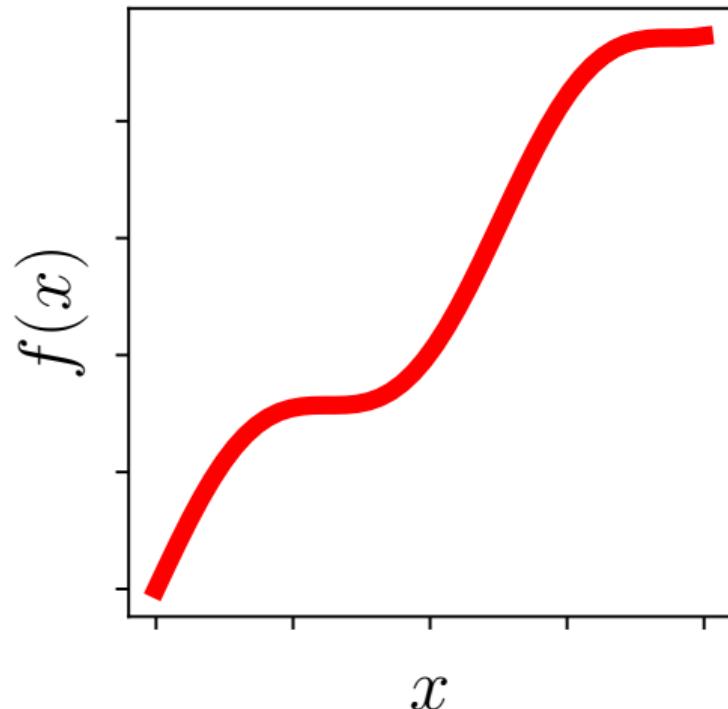
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- We DON'T have the inverse function to go from cross sections to $\textcolor{red}{a}$

Consider a simple scenario in 1D

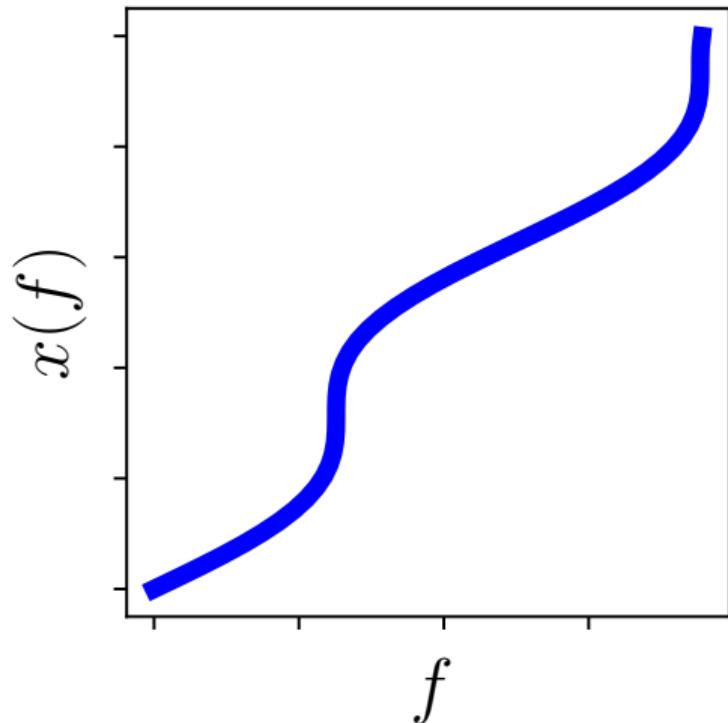
Consider a simple scenario in 1D

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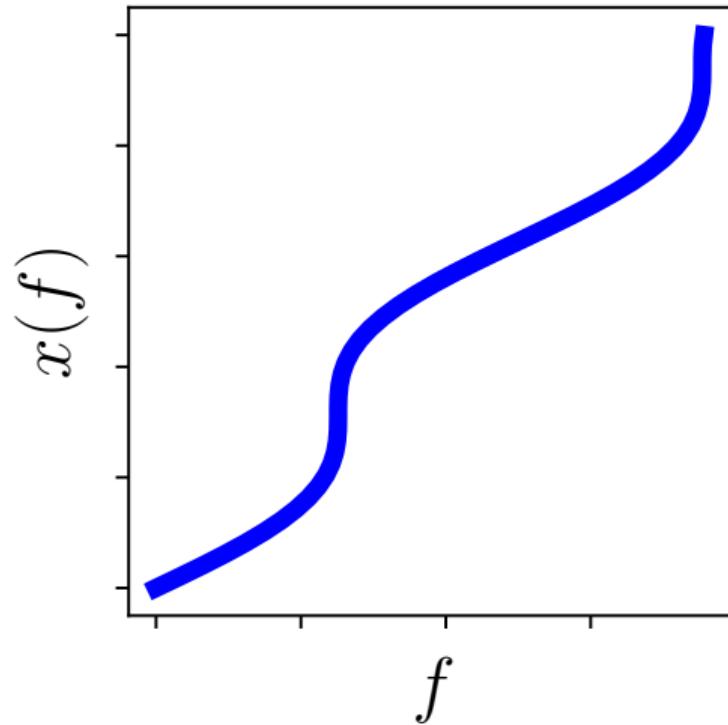


Consider a simple scenario in 1D

- Suppose we **know** $f(x)$ but **not** its inverse
- We can **flip** the graph to get the inverse

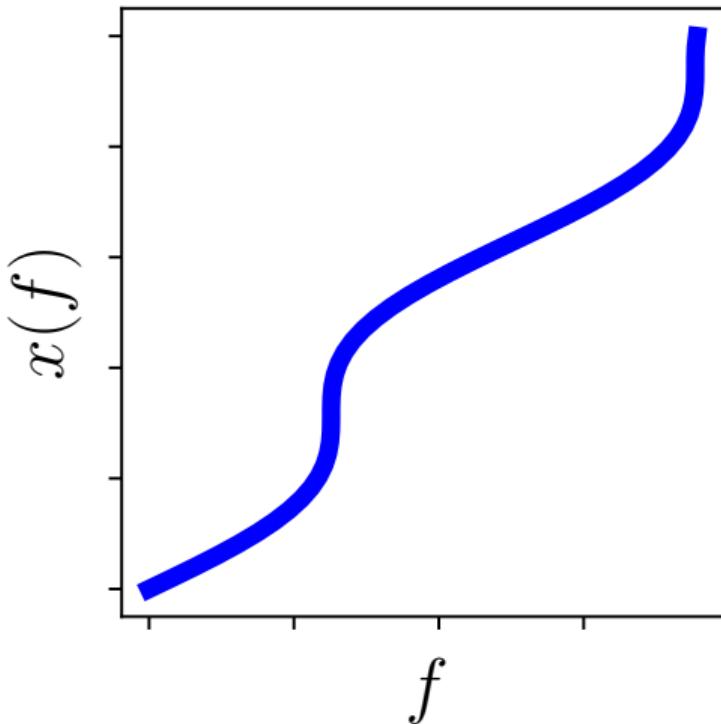


The key idea



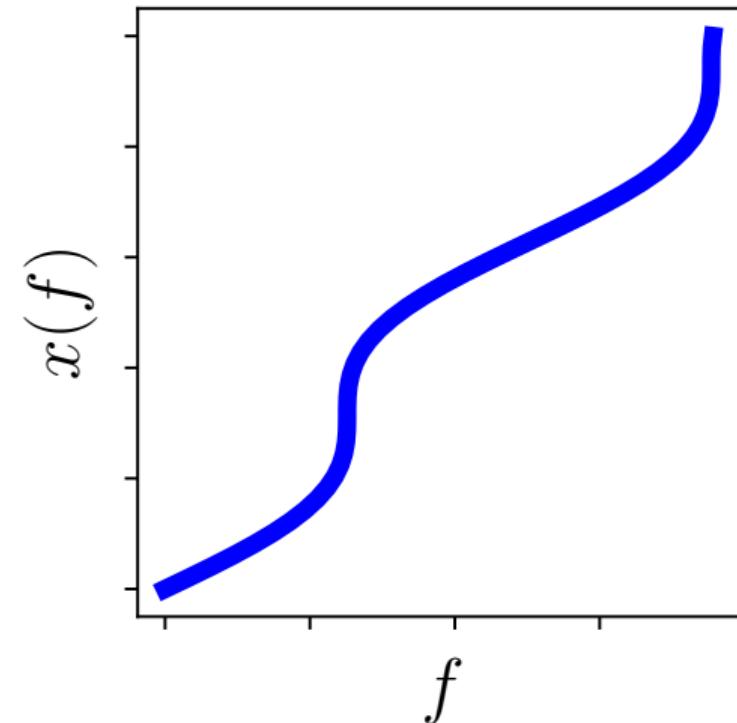
The key idea

- We can **parametrize** the inverse

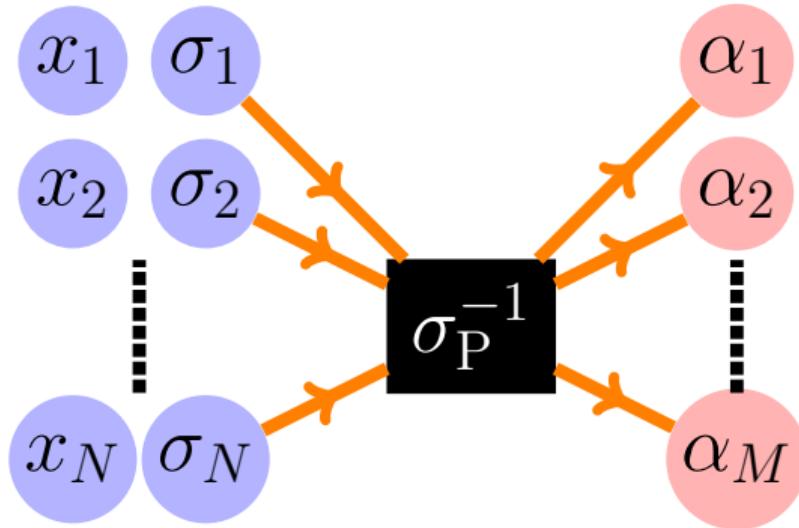


The key idea

- We can **parametrize** the inverse
- Given measurements of f we can **infer** x with the inverse

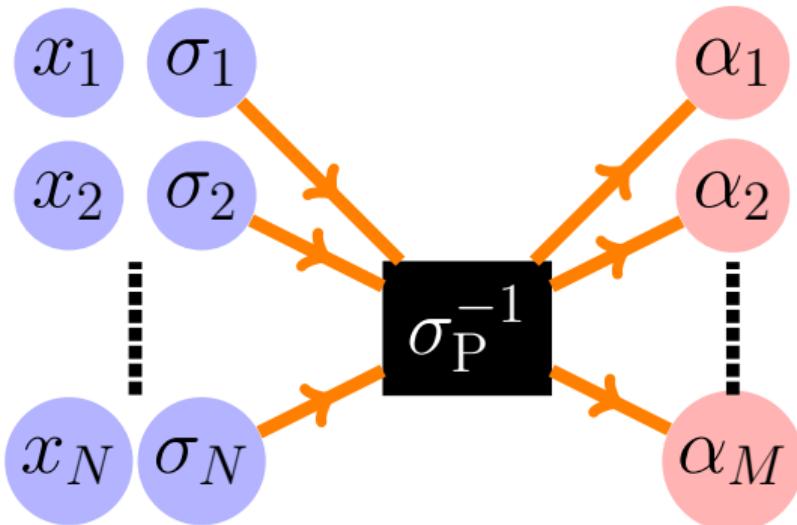


The inverse mapper for global analysis

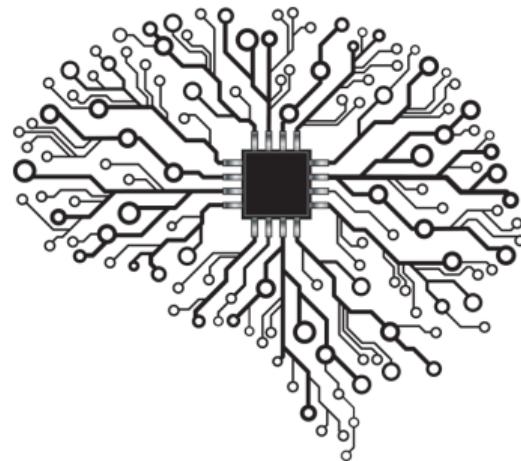


$$R^N \rightarrow R^M$$

The inverse mapper for global analysis



$$R^N \rightarrow R^M$$



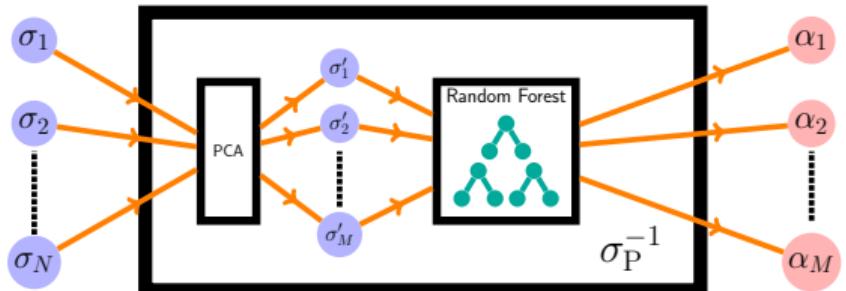
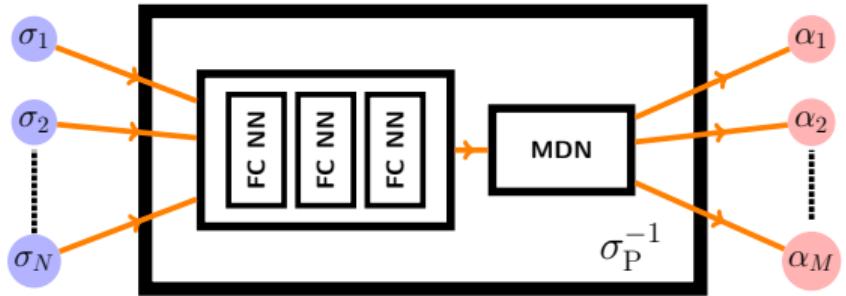
Can we use Machine Learning?

Partnership with computer scientists

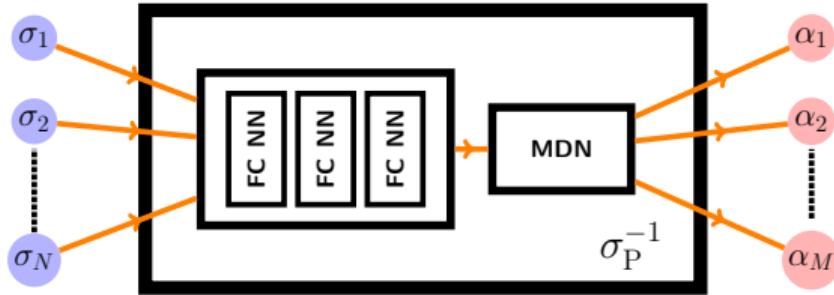
- M. Almaeen (ODU)
- Y. Awadh Alanazi (ODU)
- M. Houck (Davidson College)
- M. P. Kuchera (Davidson College)
- Y. Li (ODU)
- W. Melnitchouk (JLab)
- R. Ramanujan (Davidson College)
- NS (JLab)
- E. Tsitinidi (Davidson College)



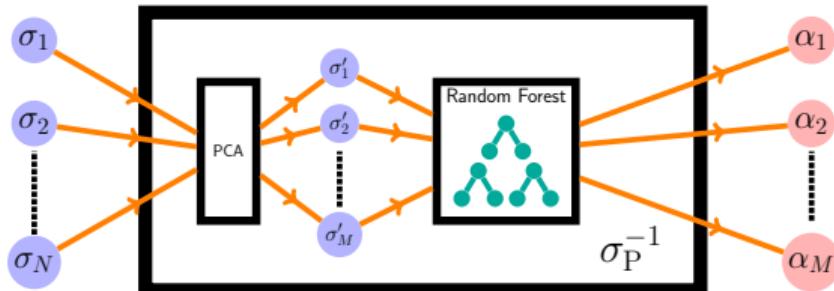
ML prototypes



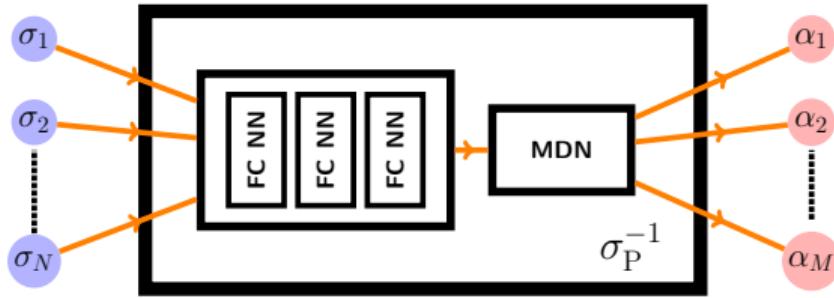
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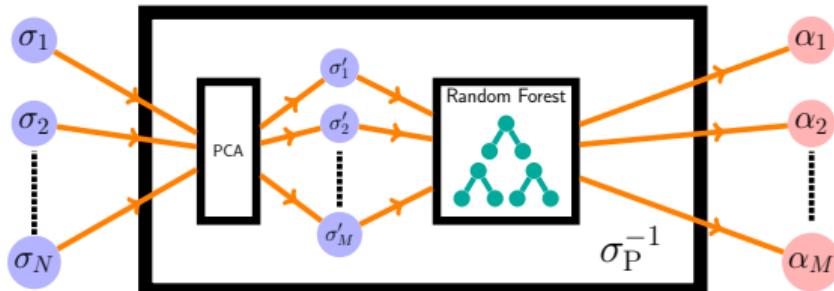
- Tested and validated in toy DIS-like examples



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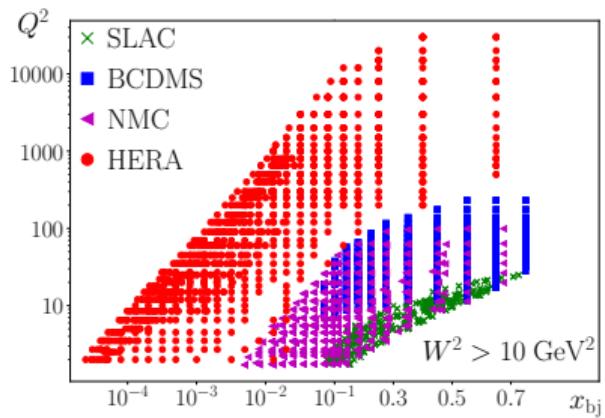


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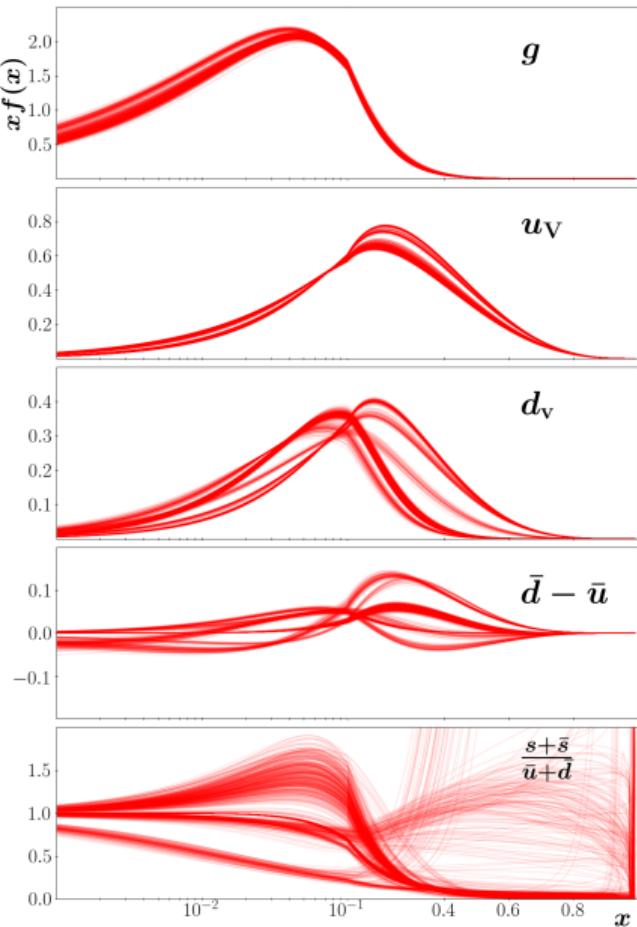
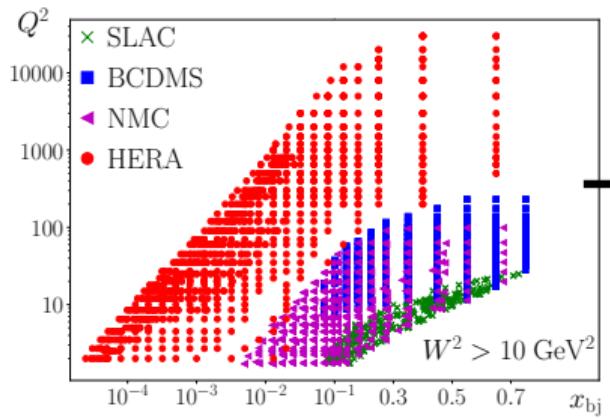


- How about real QCD analysis?

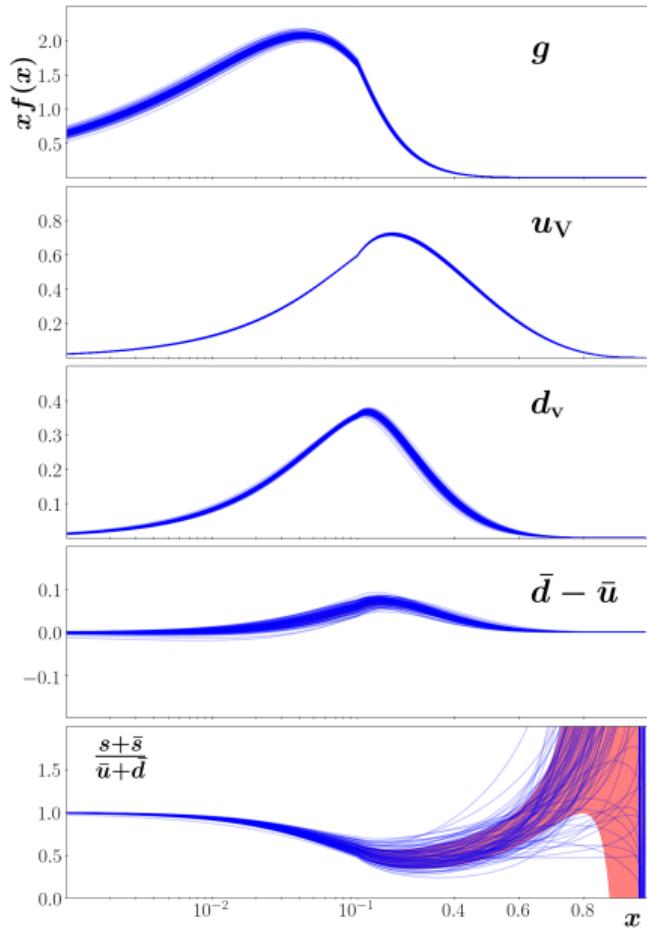
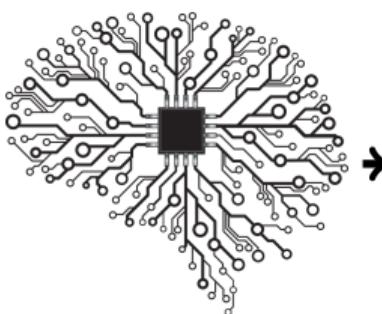
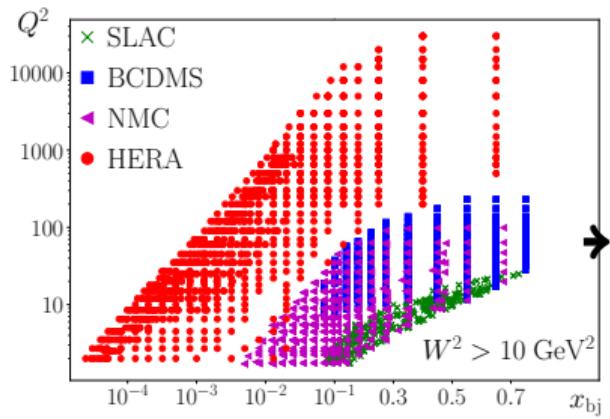
Application to unpolarized DIS



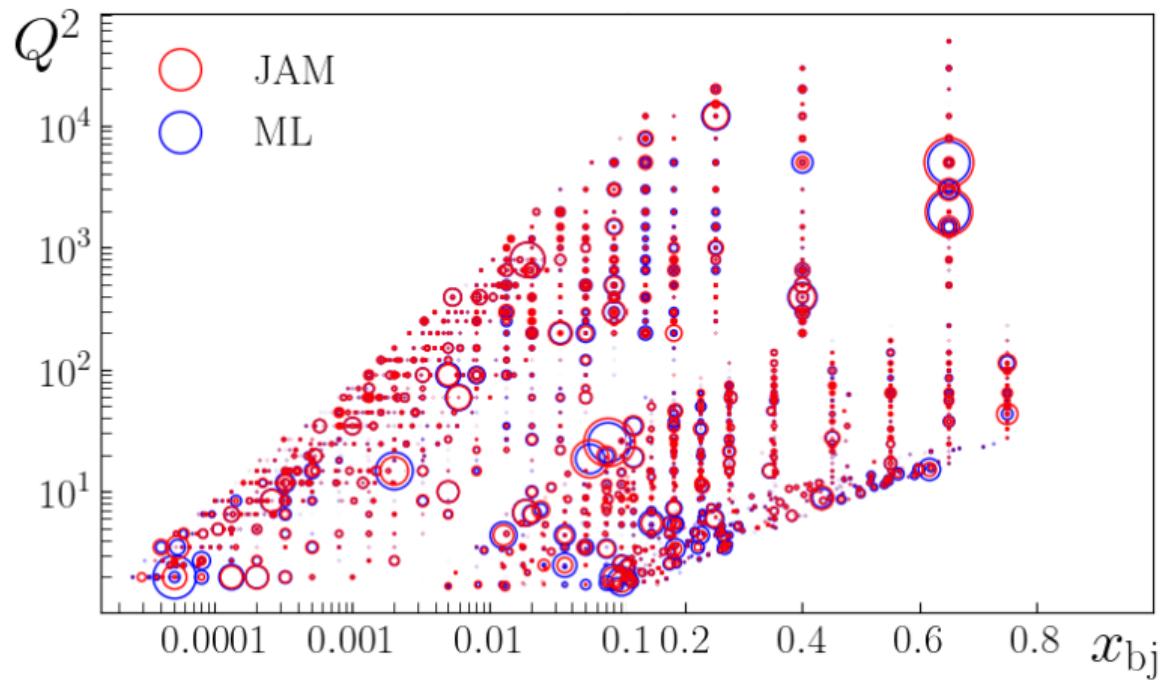
Application to unpolarized DIS



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Application to unpolarized DIS



- Proton DIS kinematics
- Blobs $\propto \chi^2$
- $\frac{\chi^2_{\text{JAM}}}{N_{\text{pts}}} = 1.25$
- $\frac{\chi^2_{\text{ML}}}{N_{\text{pts}}} = 1.36$

Summary and outlook

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 - + simultaneous extraction of PDFs and FFs
 - + boosts the SIDIS program to study hadron structure
 - + the ultimate strategy for TMD and GPD physics
 - + ripe for Machine Learning techniques