

Exploring the nature of matter

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Christopher Newport U. & Jefferson Lab

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Outline

- **Intro: what are we made of?** (And how do we know?)
 - Quarks & gluons
 - Quantum Chromodynamics (QCD)
 - Color confinement
- **Dynamical mass generation**
 - How are massive particles created out of nearly massless quarks?
- **Parton Distributions Functions (PDFs) in the proton**
 - Quark and gluons at the “edge of confinement”
 - Global QCD analysis
 - Theory + Experiment + Data/Computer Science
- **Perspectives**

Acknowledgments

- **My CTEQ-JLab (CJ) collaboration colleagues:**
 - **Matteo Cerutti (CNU), Xiaoxian Jing (Meta), Ishara Fernando (UVa), W.Melnitchouk (JLab), J.F.Owens (Florida State U.)**
 - C.E. Keppel & Sanghwa Park (JLab), **Shujie Li (LBL)**, P. Monaghan (CNU)
- **My other collaborators:**
 - **C. Costa, A. Simonelli (JLab), A. Bacchetta (Pavia U., Italy), A. Krause (ODU), A. Signori (Torino U., Italy)**
 - **[nCTEQ] F. Olness (Southern Methodist U.), R. Ruiz (Cracow, Poland), I.Schienbein (Grenoble)**
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 - DE-SC0025004 “Large-x partons...”
 - DE-AC05-06OR23177 (JLab theory)
 - APS-Simons travel grant
- **Artwork:** Bernhard Musch (2009)
Scattering discussion: R. Ruiz (2024)



General References

QCD global analysis from protons to nuclei:

- Accardi, [PoS DIS2015 \(2015\) 001](#)
- Jimenez-Delgado, Melnitchouk, Owens, [J.Phys.G40 \(2013\) 093102](#)
- Ethier, Nocera, [Ann.Rev.Nucl.Part.Sci. \(2020\) 70, 1-34](#)

QCD global analysis and statistical methods:

- Kovarik, Nadolsky, Soper, [Rev.Mod.Phys. 92 \(2020\) 4, 045003](#)
- Buckley, White and White, “[Practical Collider Physics](#)”, IOP publishing 2021

References - technical

Inclusive Jet functions & dynamical mass generation

- Accardi, Costa, Signori, [*Phys.Rev.D 108 \(2023\) 114011*](#)
- Accardi, Signori, [*Phys.Lett.B 798 \(2019\) 134993*](#) & [*Eur.Phys.J.C 80 \(2020\) 825*](#)
- Accardi, Bacchetta, [*Phys.Lett.B 773 \(2017\) 632*](#)

Large-x fits with nuclear corrections & applications

- **CJ15:** Accardi et al., [*PRD 93 \(2016\) 114017*](#)
 - Accardi, DNP 2020 / Fernando, GHP 2021 / Accardi, APS 2022
- **$F_2(n)$:** Li, Accardi et al., [*Phys.Rev.D 109 \(2024\) 074036*](#)

Light quark asymmetry fits

- **CJ15a:** Accardi, Owens, Melnitchouk, [*Phys.Lett.B 801 \(2020\) 135143*](#)
- **CJ22:** Accardi, Jing, Owens, Park, [*Phys.Rev.D 107 \(2023\) 113005*](#)

PDF uncertainties

- Hunt-Smith, Accardi, Melnitchouk, Sato, Thomas, White, [*Phys.Rev.D 106 \(2022\) 036003*](#)

References - recent and in prep.

Inclusive Jet functions & dynamical mass generation

- Simonelli, Accardi, Cerutti, Costa, Signori
 - “Unveiling the Collins-Soper kernel in inclusive DIS at threshold”
 - Rigorous QCD factorization with jet functions

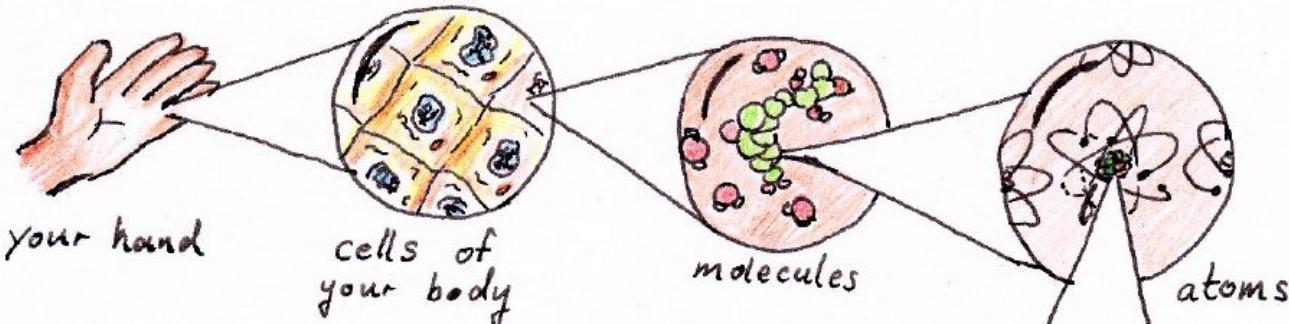
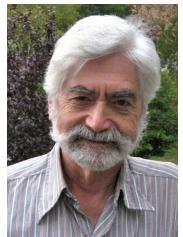
Large-x fits with power and nuclear corrections & applications

- Accardi, Cerutti, Fernando, Li, Owens, Park
 - “Interplay of off-shell and higher-twist corrections in DIS at large x”
 - Non negligible QCD analysis systematic effects
 - Short version @ DIS 2024: [arXiv:2407.03589](https://arxiv.org/abs/2407.03589)
- Accardi, Krause
 - “Testing large-x QCD analysis in a spectator model”

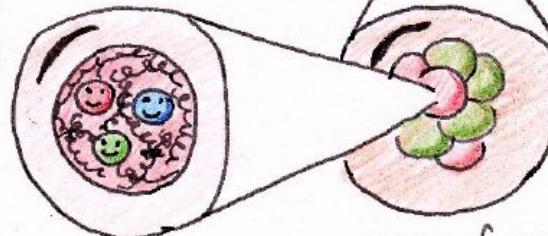
Introduction: What are we made of?

What are we made of?

- Quarks and gluons!



Here they
are! 3 quarks ☺
bound by
gluons ☺



single
nucleon
 10^{-15} m

core of an atom:
protons + neutrons
("nucleons")



E. Rutherford,
1897, 1917

Quarks:

proposed: Gell-Mann & Zweig, 1964

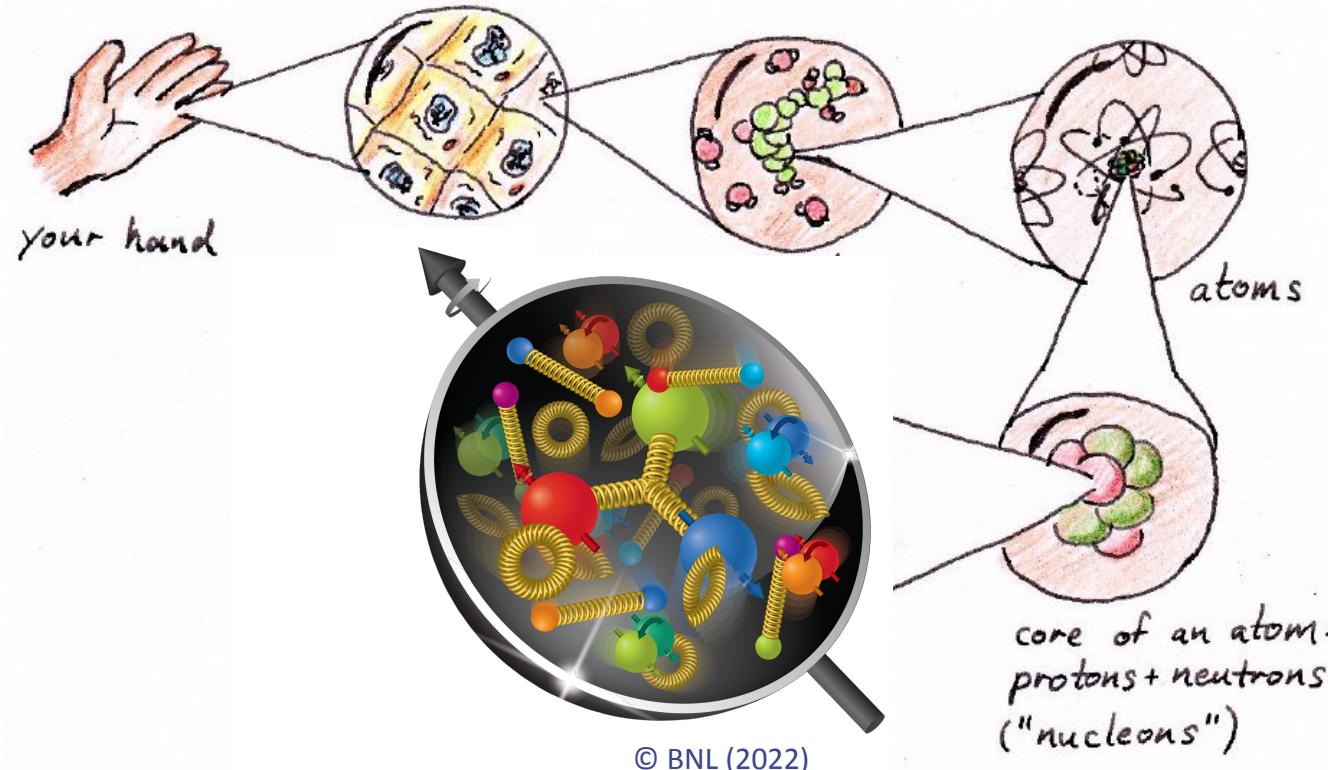
discovered: SLAC (USA), 1968

Gluons:

discovered: Desy (GER) 1979

What are we made of?

- Quarks and gluons!



How do we know?

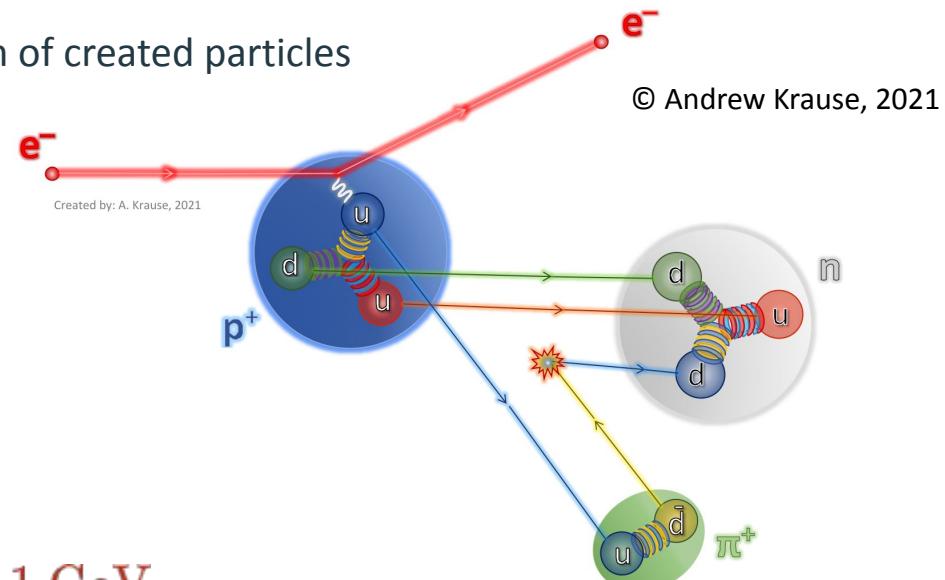
- By using high-energy scattering as a microscope!
(actually, a femtoscope)
 - And also looking at the spectrum of created particles

- Need a wavelength smaller than the proton's size:

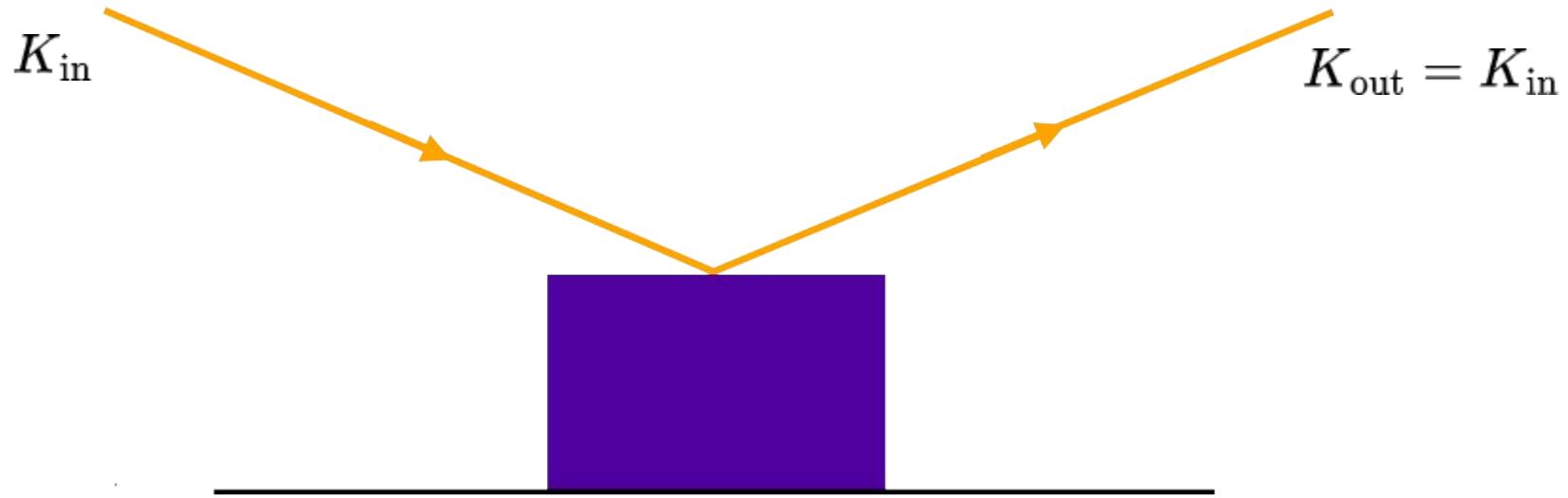
$$\lambda \approx \frac{hc}{E} < 10^{-15} \text{ m}$$

$$\implies E > 1 \text{ GeV}$$

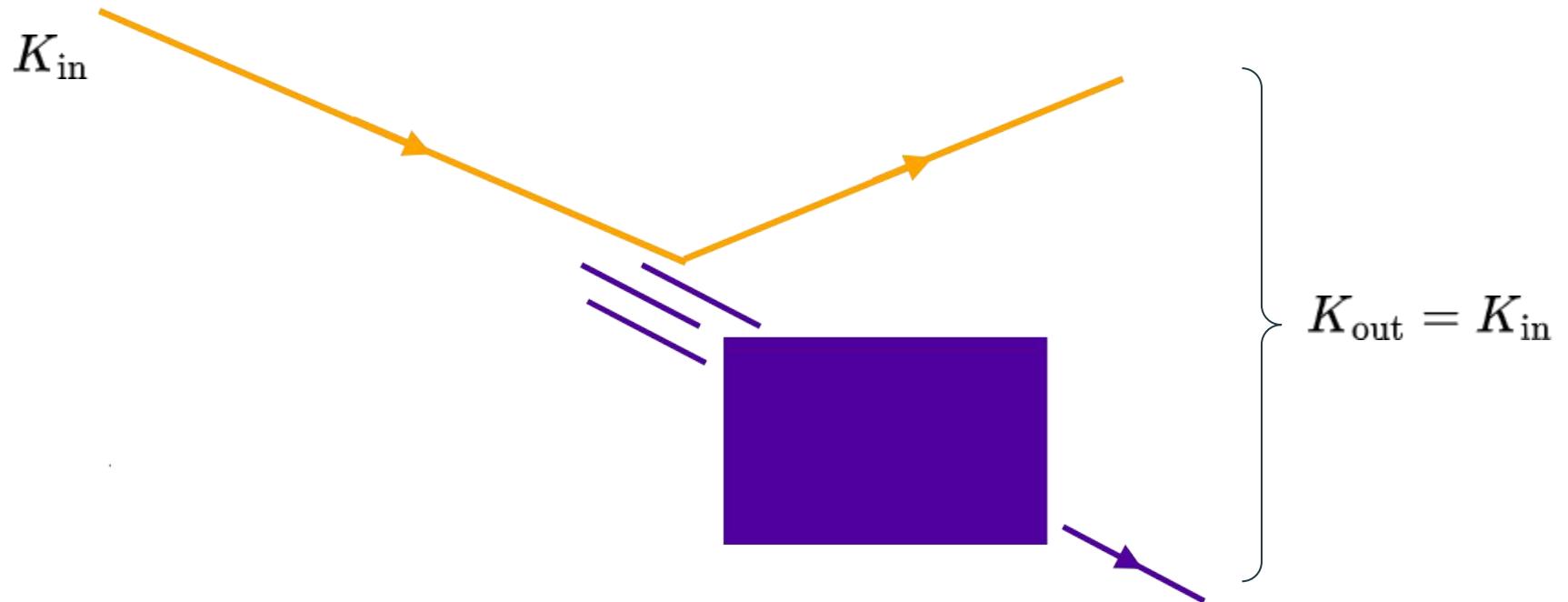
$1 \text{ eV} = 10^{-19} \text{ J}$: energy needed to move an electric charge $e = 10^{-19} \text{ C}$ through a potential difference of 1 V



Scattering 101: elastic

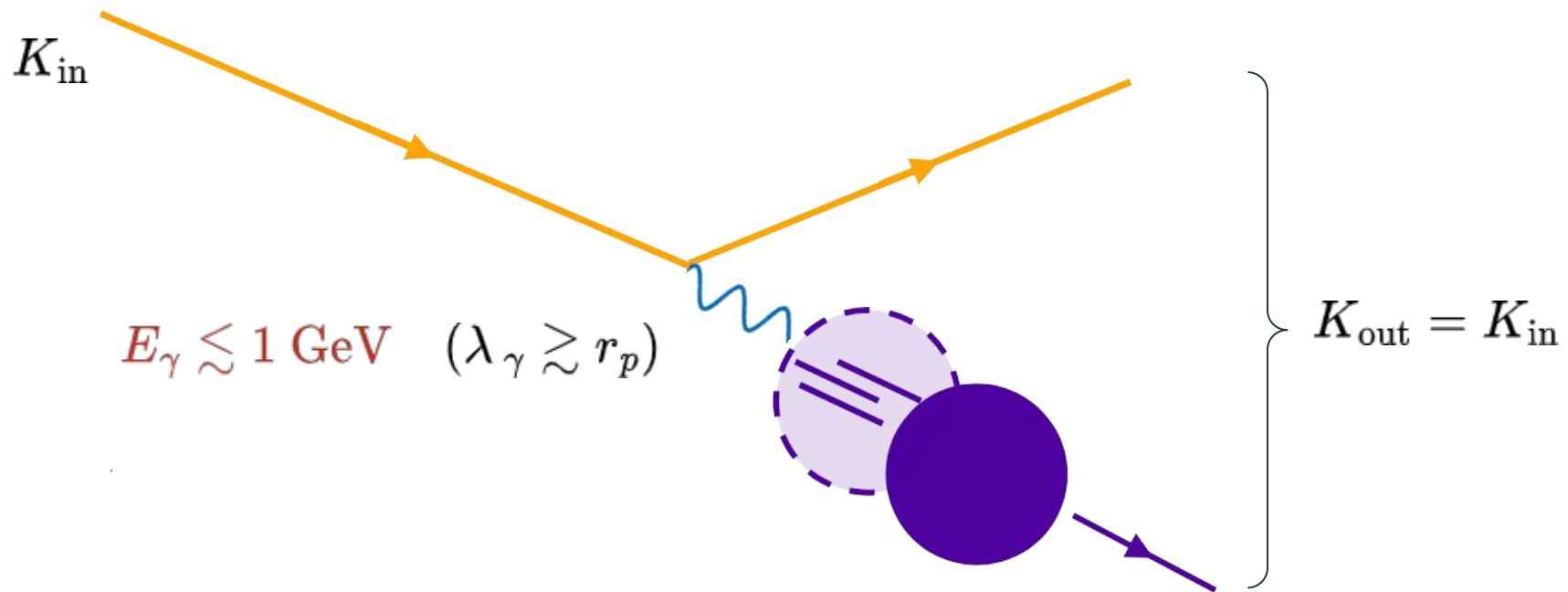


Scattering 101: elastic



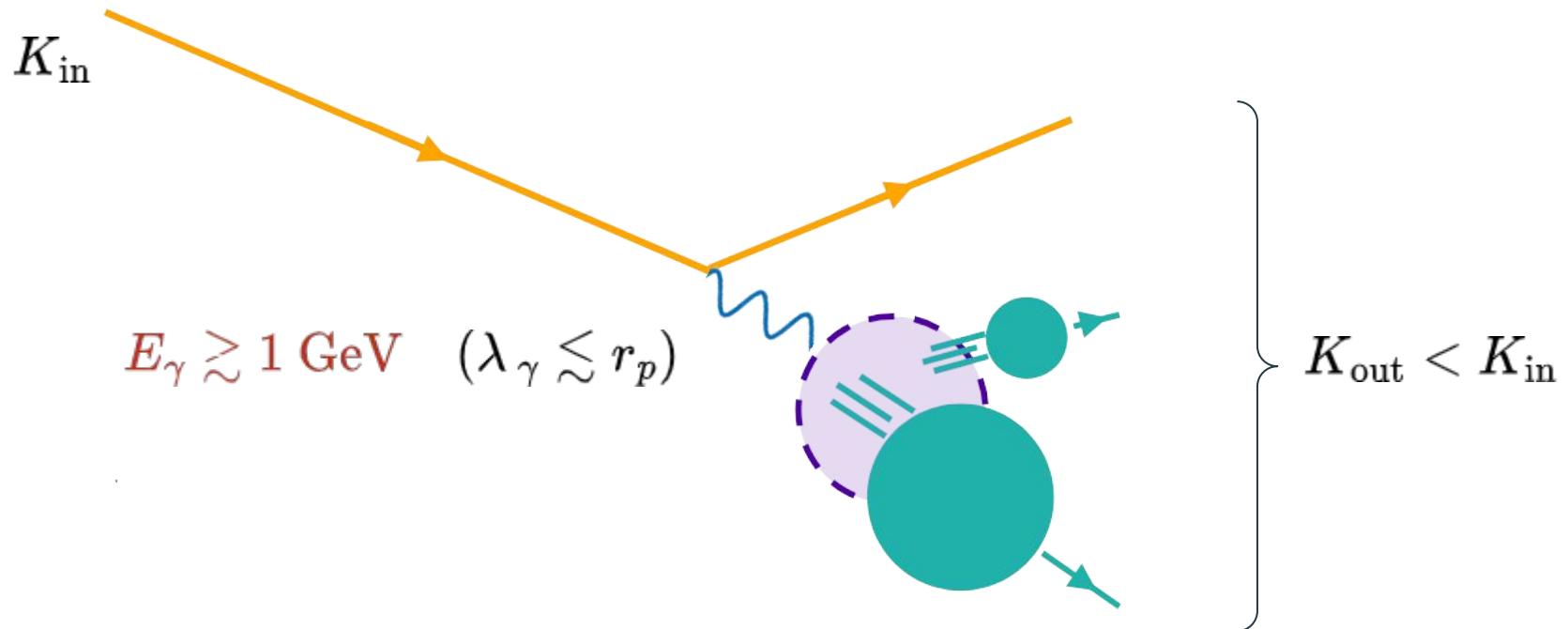
Electron-proton: elastic

- Exchanged photon wavelength: $\lambda_\gamma = h/p_\gamma \approx hc/E_\gamma$



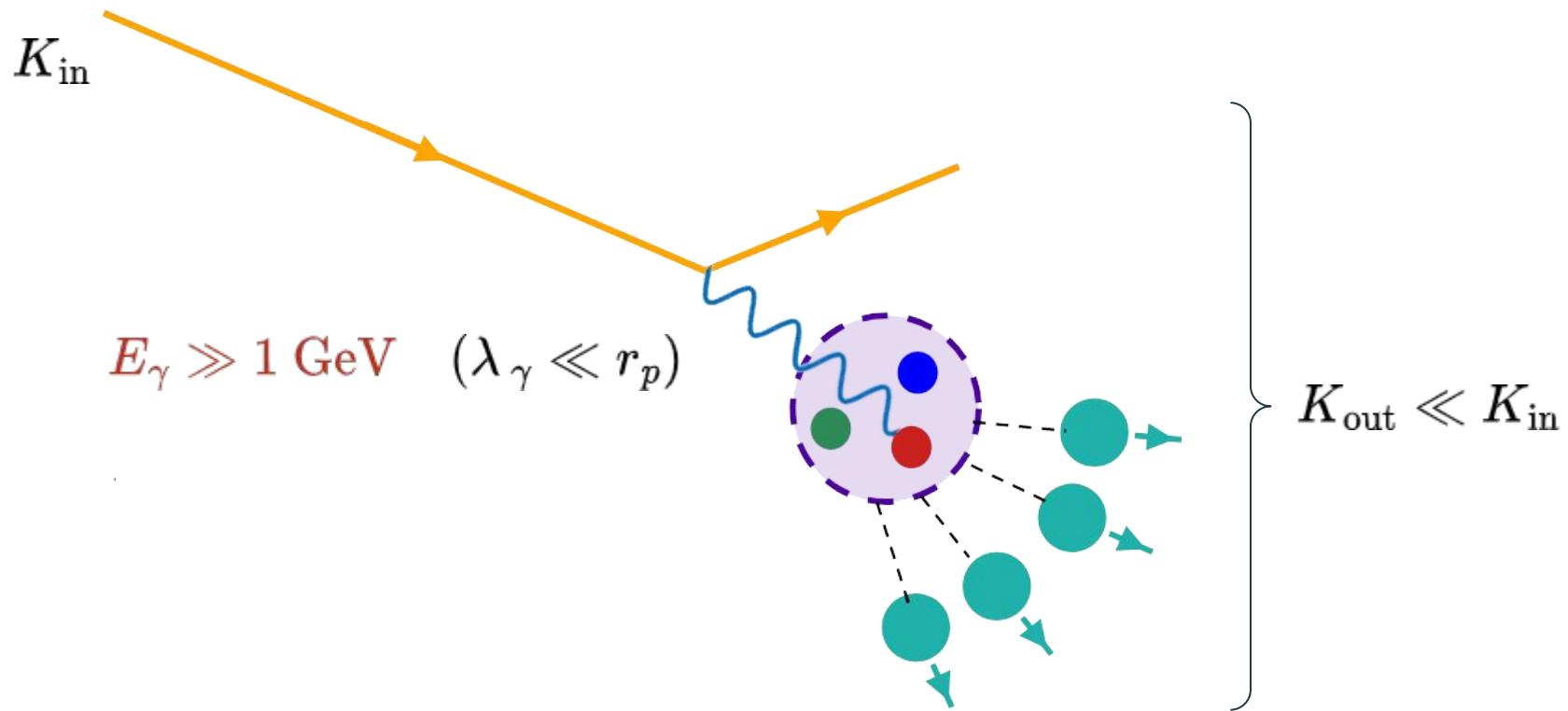
Electron-proton: inelastic

- Exchanged photon wavelength: $\lambda_\gamma = h/p_\gamma \approx hc/E_\gamma$



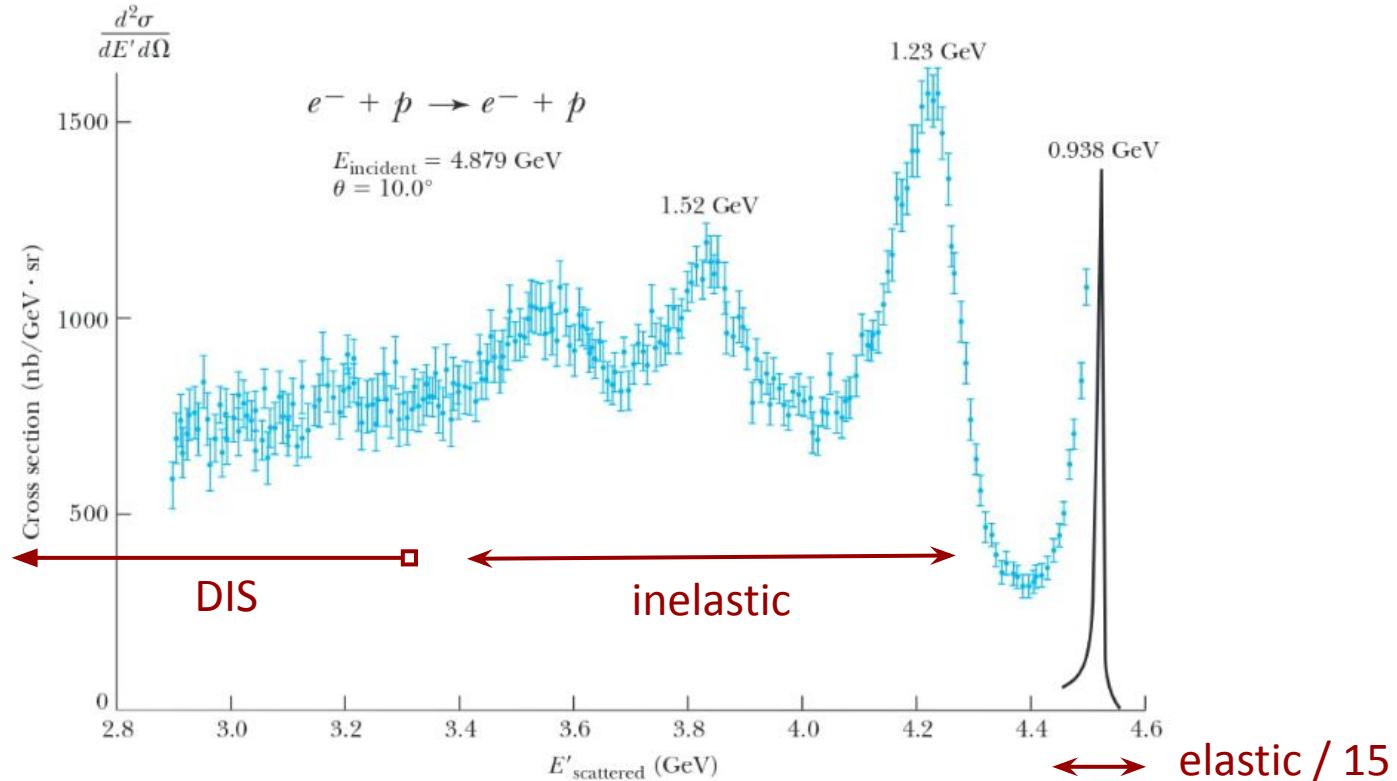
Electron-proton: deeply inelastic scattering (DIS)

- Exchanged photon wavelength: $\lambda_\gamma = h/p_\gamma \approx hc/E_\gamma$



Cross section

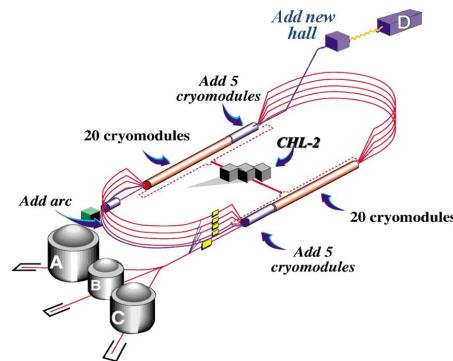
- Roughly, the probability that the electron hits the target



Bartel et al., Physics Letters B 28, 148 (1968)

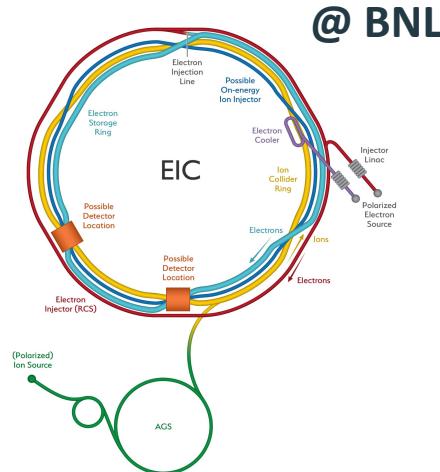
Modern particle accelerators

JLab 12 GeV



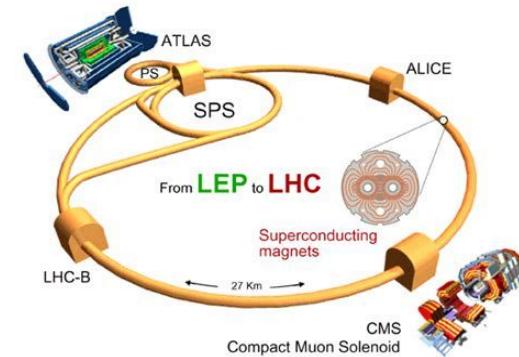
$e^- (12 \text{ GeV}) \rightarrow \text{proton}$

Electron-Ion Collider



$e^- (10 \text{ GeV}) \rightarrow \leftarrow (275 \text{ GeV}) p$

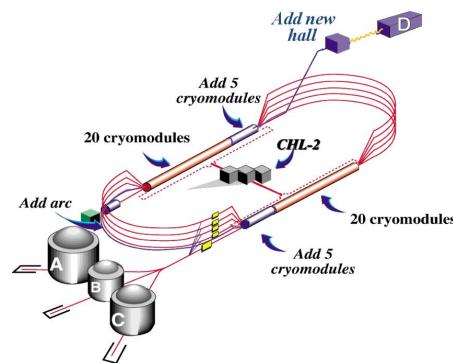
Large Hadron Collider @ CERN



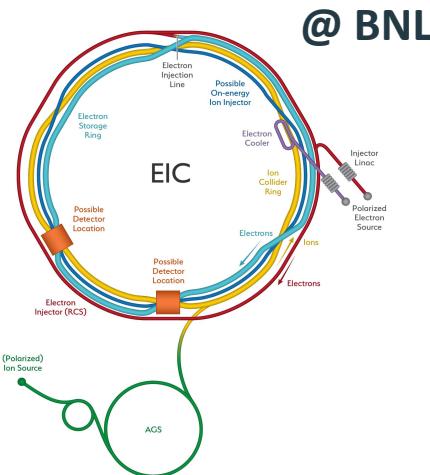
$p (7000 \text{ GeV}) \rightarrow \leftarrow (7000 \text{ GeV}) p$

Modern particle accelerators

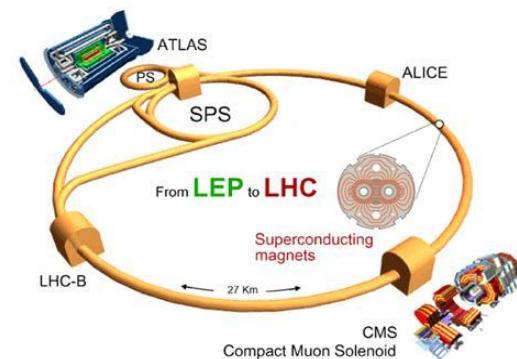
JLab 12 GeV



Electron-Ion Collider



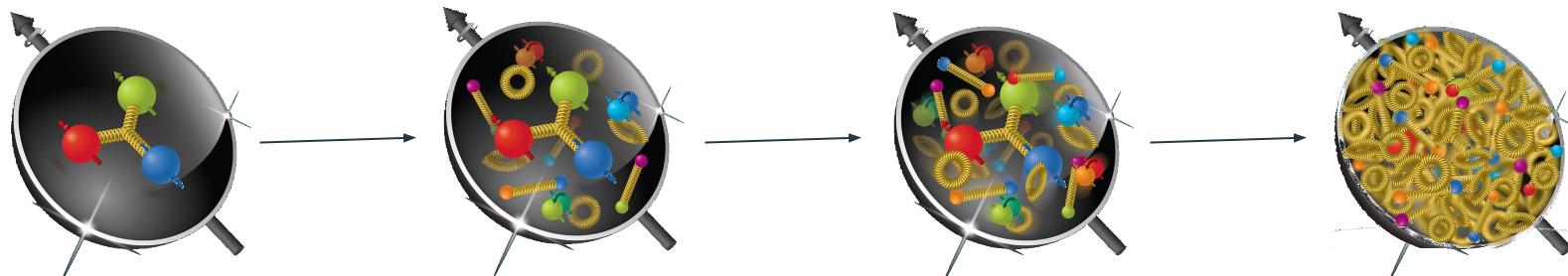
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Modern particle accelerators

JLab 12 GeV

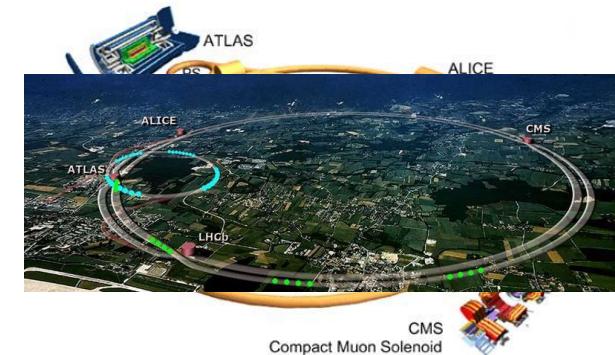


Electron-Ion Collider



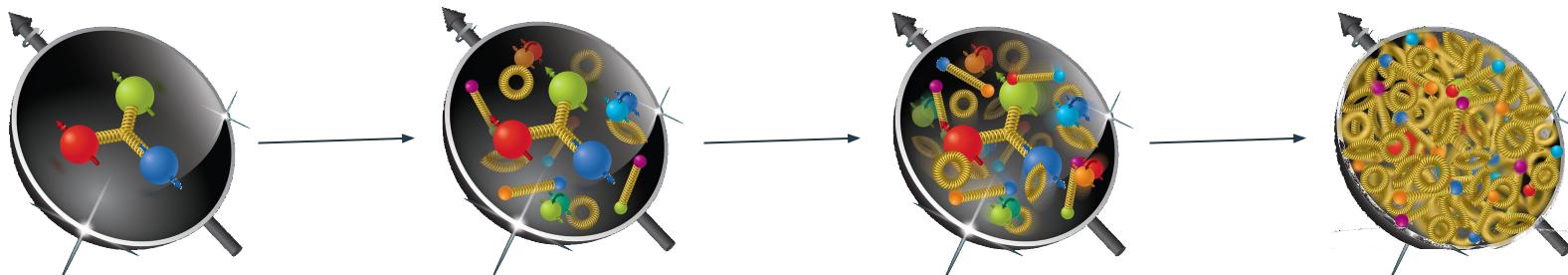
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Large Hadron Collider @ CERN



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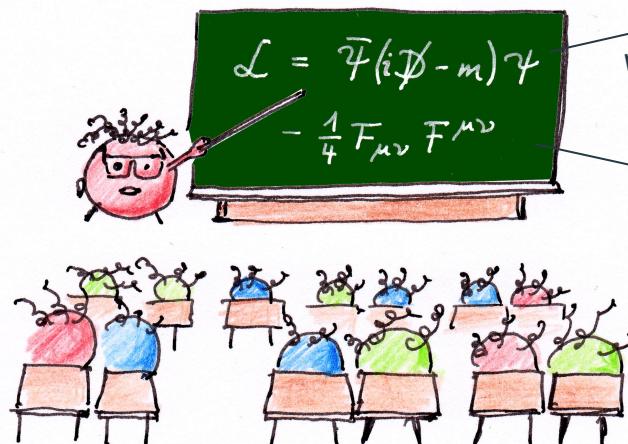


Quantum Chromodynamics (QCD)

- Simple rules

- Matter = quarks + antiquarks:
→ 3 colors, 6 flavors, spin = $\pm \frac{1}{2}$
- Interaction = exchange of gluons

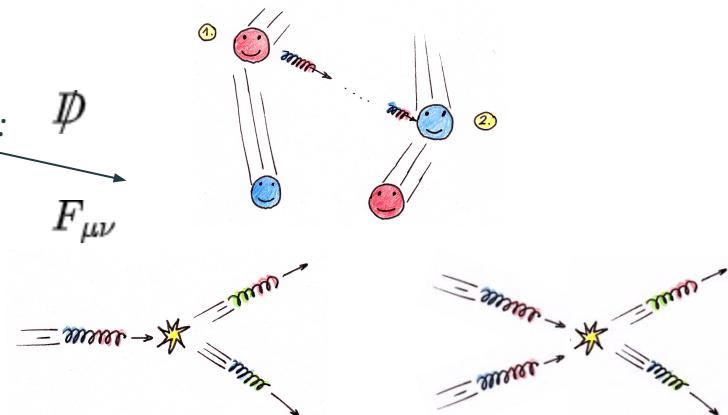
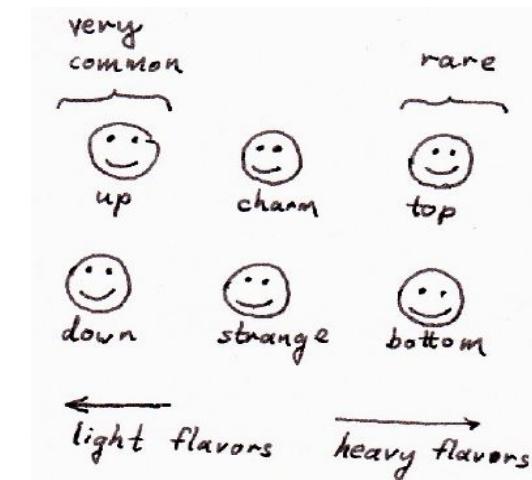
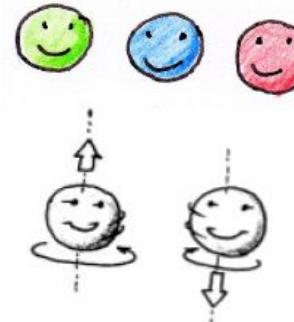
- “Lagrangian”: contains the theory



matter: $\psi, \bar{\psi}$

interactions: \cancel{D}

$F_{\mu\nu}$

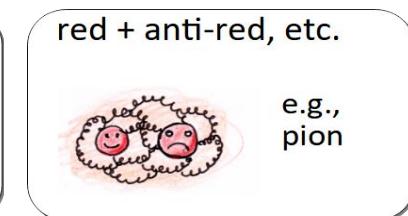
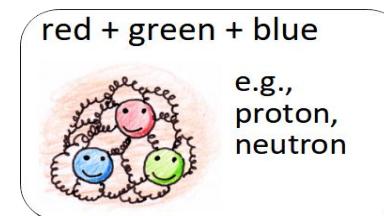
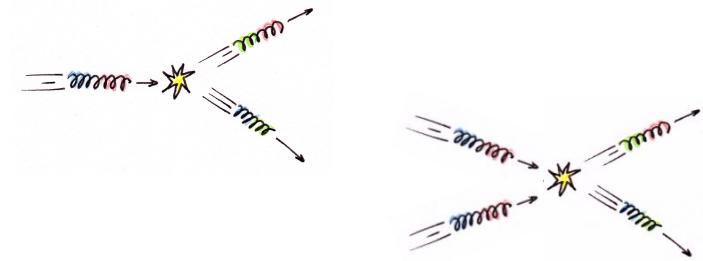


Unlike any other fundamental theory!

- QCD is analogous to electromagnetism, but:
 - **gluons can split...**
 - **...and bang on each other**
(no linear superposition principle!!)
- This happens so often that quarks are always surrounded by a dense network of gluons that holds together a group of several quarks
- Quarks form particles that are **always “color-neutral”** as a whole:

Color is *confined* inside hadrons!

- Not your everyday charge...



Mysteries

- Gluon binding energy provides most of the proton mass:

$$m_{\text{proton}} = 1 \text{ GeV}/c^2$$

vs.

$$m_u + m_u + m_d \approx 0.015 \text{ GeV}/c^2$$

- Are we actually made of a force field??



- Why is the spin of proton = $\frac{1}{2}$?

An unlikely way of combining:

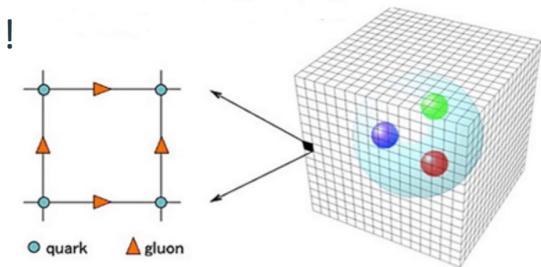
- the spin of (many) quark and gluons
- and their orbital angular momentum!



- In fact, why isn't there any free quark??

Color confinement

- Only “color neutral” particles have ever been (can ever be?) observed
- Why??
 - Mathematical explanation from QCD Lagrangian
 - I attempted one a long time ago (1996-1998) – my first scientific love!
 - \$1M “millennium prize” from the Clay Institute!!
 - Lattice QCD – computational explanation
 - But “black box”...
 - Look at the effects of confinement
 - Dynamical mass generation
 - Parton Distribution Functions at the “edge of confinement”



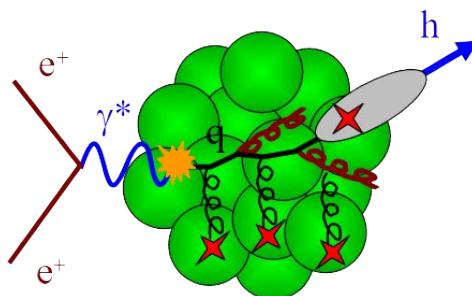
my research!

Dynamical mass generation:
how do massive hadrons emerge
from nearly massless quarks?

Parton propagation and fragmentation

Review: Accardi et al., [Riv. Nuovo Cim. 032 \(2010\)](#)

- Nuclei as femtometer-scale detectors



Transverse momentum broadening

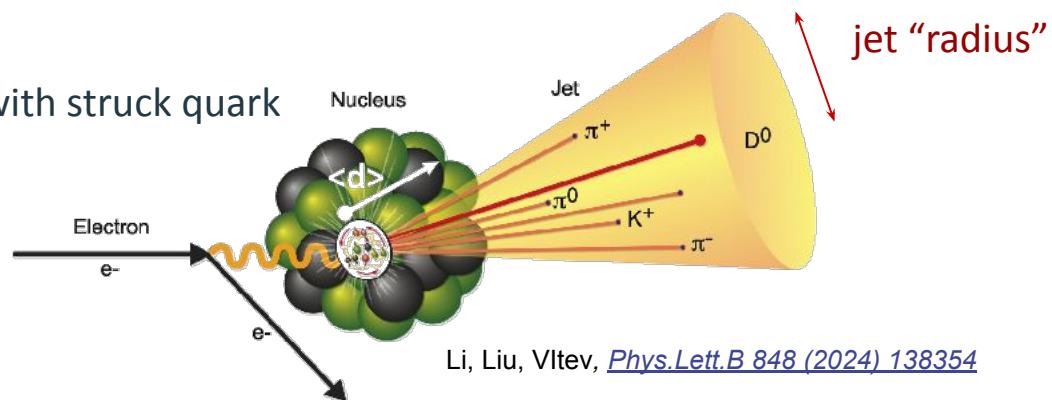
$$\Delta p_T^2 = \langle p_T^2 \rangle_A - \langle p_T^2 \rangle_D$$

Hadron attenuation

$$R_M = (N^h/N^e)_A / (N^h/N^e)_D$$

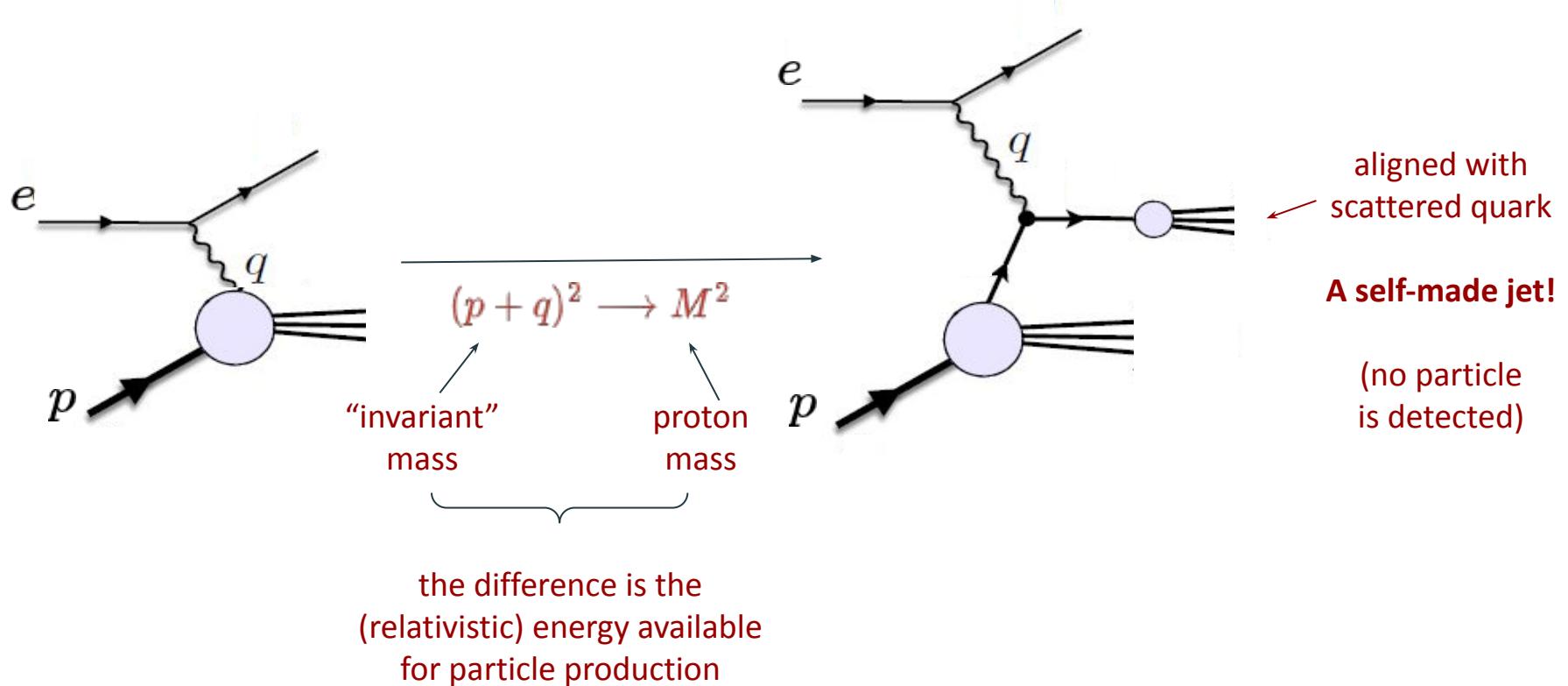
- At the Electron-ion collider

- Higher energy
- "Jets" of particles aligned with struck quark



Li, Liu, Vitev, [Phys.Lett.B 848 \(2024\) 138354](#)

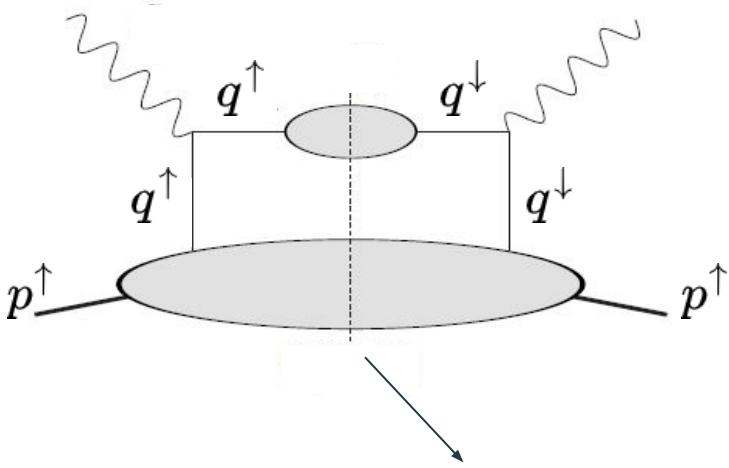
“Inclusive jets” in vacuum



Quantum mechanically

- Polarized electron+proton scattering:

$$\sigma(\vec{e} + p^\uparrow \rightarrow X) \sim \mathcal{M}\mathcal{M}^* \sim$$



$$q^\uparrow \quad q^\downarrow = \int \mu \rho_1(\mu^2) d\mu^2 \sim \langle \mu \rangle$$

“Spectral function”:
prob. distrib. of
produced mass

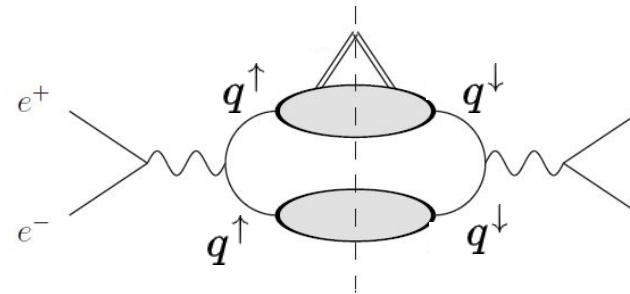
measurable!
In “inclusive LT
asymmetries”

calculable:
lattice QCD / Dyson-Schwinger

Quantum mechanically

- Polarized electron+positron scattering:

$$\sigma(\vec{e}^- + \vec{e}^+ \rightarrow \vec{\Lambda}) \sim$$



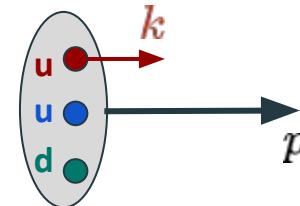
- We are looking also for observables in proton-proton collisions
- **Many observables** → can think of a “global analysis” (see later)

Parton Distributions at the “edge of confinement”

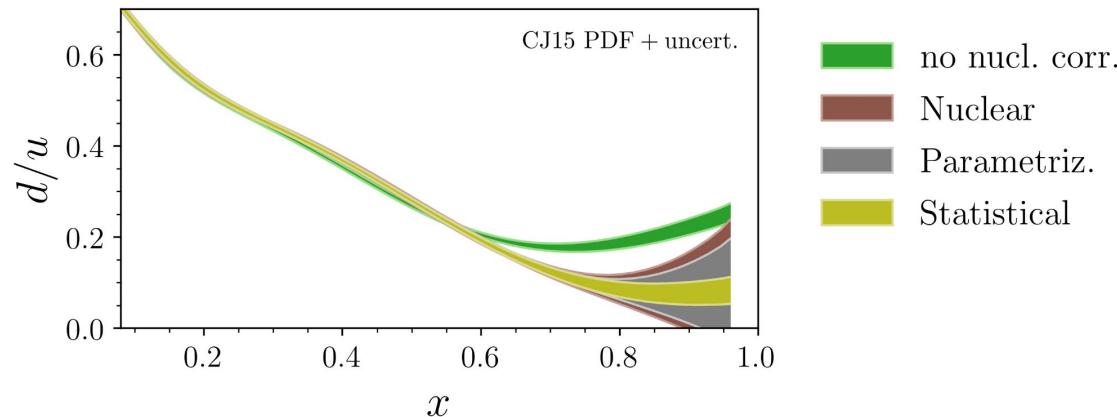
What happens when a single quark
contributes most of the proton’s
momentum?

Parton Distribution Functions

- PDF = probability distribution for a parton (quark or gluon) inside a proton to have momentum fraction $x = \frac{k}{p}$

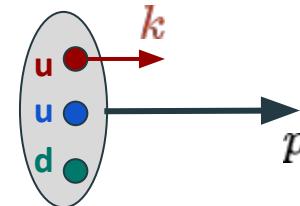


- Large $x \rightarrow 1$
 - The proton “is” a single quark – up or down
 - Confinement most directly determines the behavior of the d/u ratio

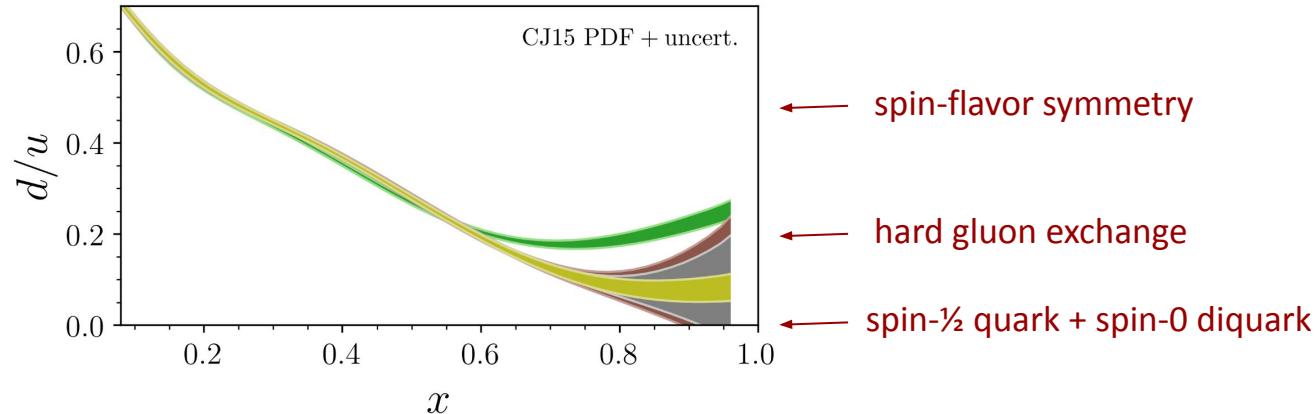


Parton Distribution Functions

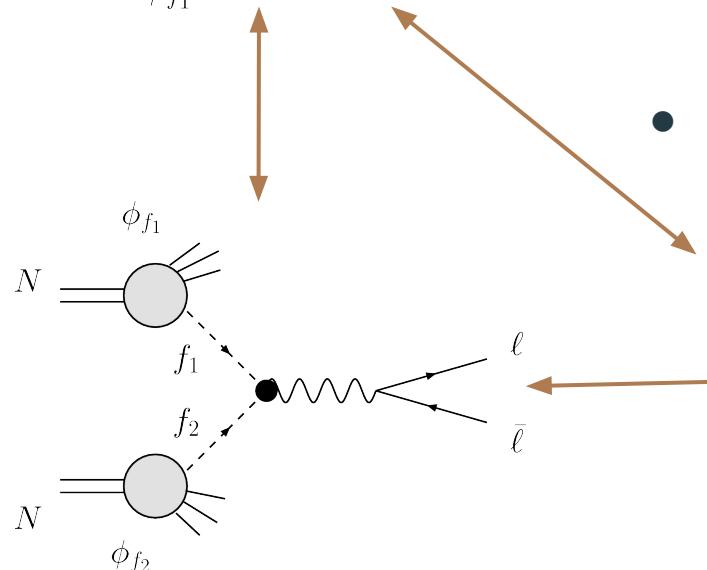
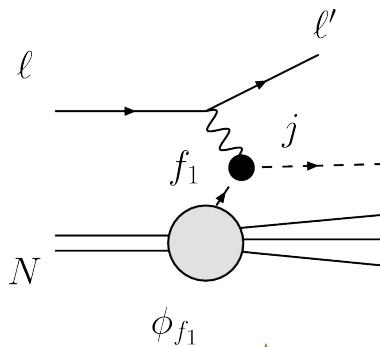
- PDF = probability distribution for a parton (quark or gluon) inside a proton to have momentum fraction $x = \frac{k}{p}$



- Large $x \rightarrow 1$
 - The proton “is” a single quark – up or down
 - Confinement most directly determines the behavior of the d/u ratio



How to measure the PDFs?



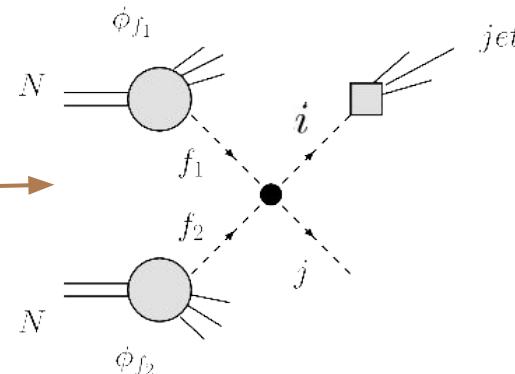
- **QCD factorization**

of short and long distance interactions

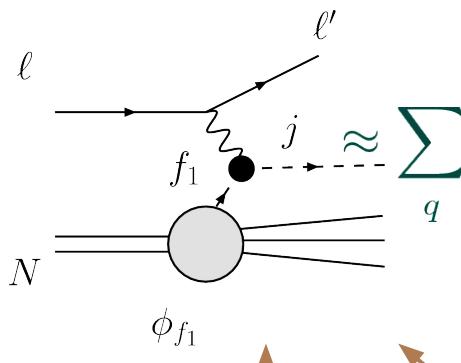
$$d\sigma_{\text{hadron}} = \sum_{f_1, f_2, i, j} \phi_{f_1} \otimes \hat{\sigma}_{\text{parton}}^{f_1 f_2 \rightarrow ij} \otimes \phi_{f_2}$$

perturbative. QCD calc.
PDFs

- **Universality:** same PDFs for many processes



How to measure the PDFs?



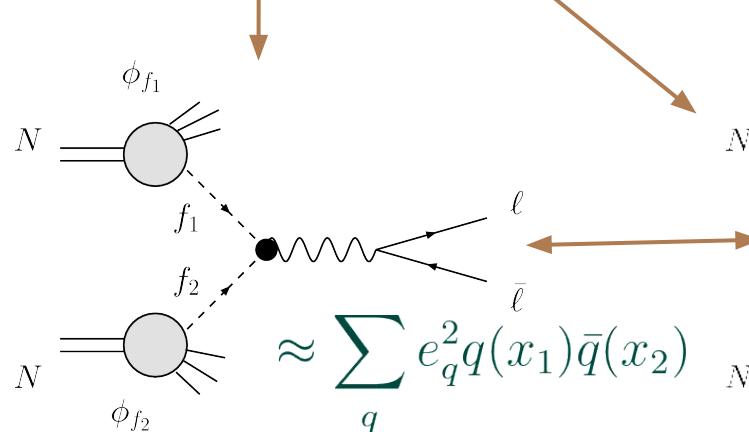
- **QCD factorization**

of short and long distance interactions

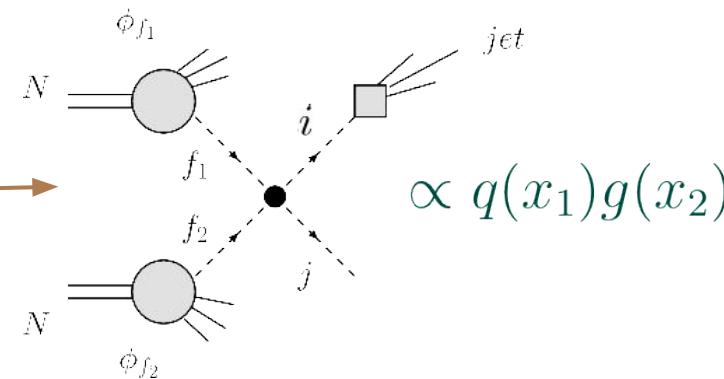
$$d\sigma_{\text{hadron}} = \sum_{f_1, f_2, i, j} \phi_{f_1} \otimes \hat{\sigma}_{\text{parton}}^{f_1 f_2 \rightarrow ij} \otimes \phi_{f_2}$$

perturbative. QCD calc.

PDFs



- **Universality:** same PDFs for many processes



Global QCD fits

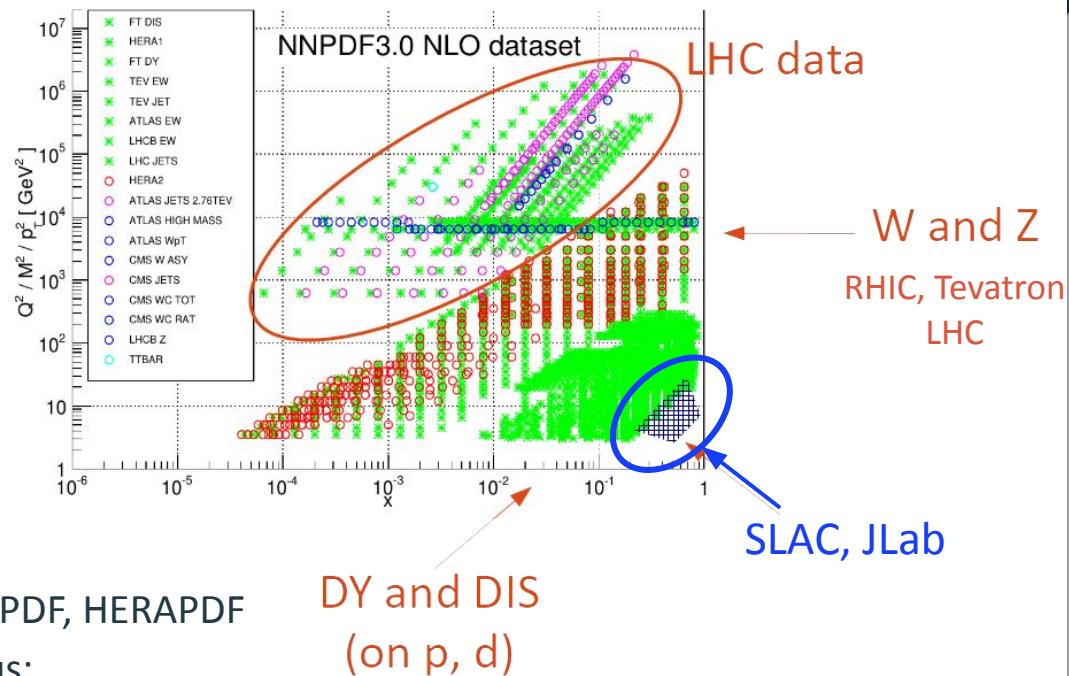
- QCD factorization & universality:
can fit PDFs to a variety of hard scattering data

- Hadron-hadron collisions
 - Jets
 - Electro-weak boson production
- Electron-proton DIS
- Electron-Deuteron DIS

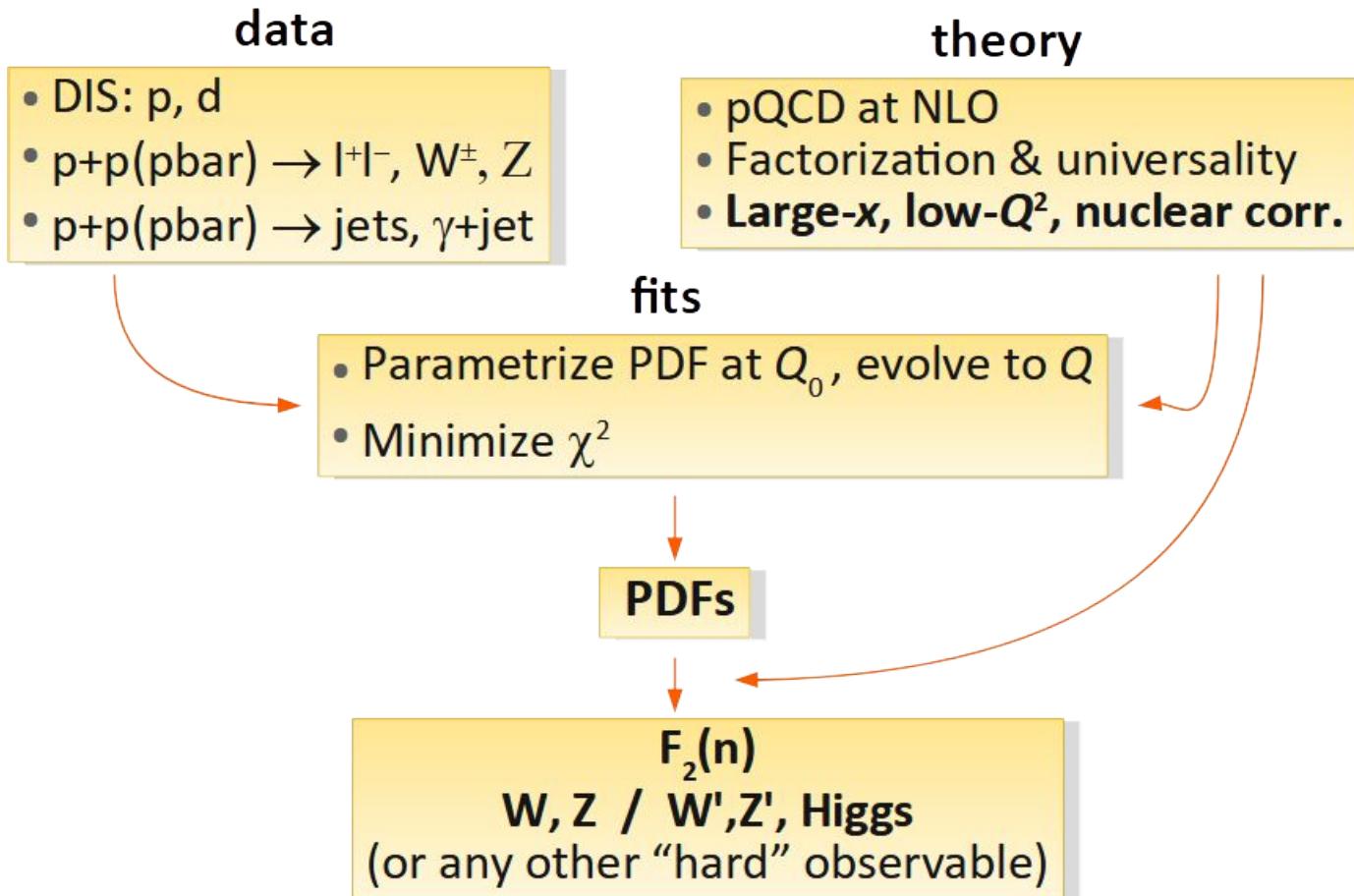
- >1000's data points
- 40+ years of experience,
 - "High-energy" fitters:
 - CTEQ-TEA, MMHT, NNPDF, HERAPDF
 - Lower-energy / nuclear focus:
 - **CTEQ-JLab**, AKP, JAM

$$d\sigma_{\text{hadron}} = \sum_{f_1, f_2, i, j} \phi_{f_1} \otimes \hat{\sigma}_{\text{parton}}^{f_1 f_2 \rightarrow ij} \otimes \phi_{f_2}$$

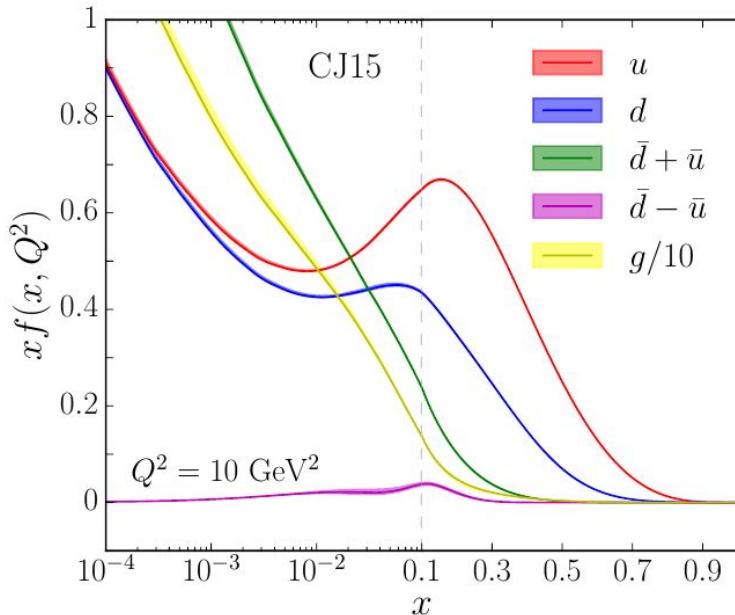
pert. QCD calc.
PDFs



Global QCD fits



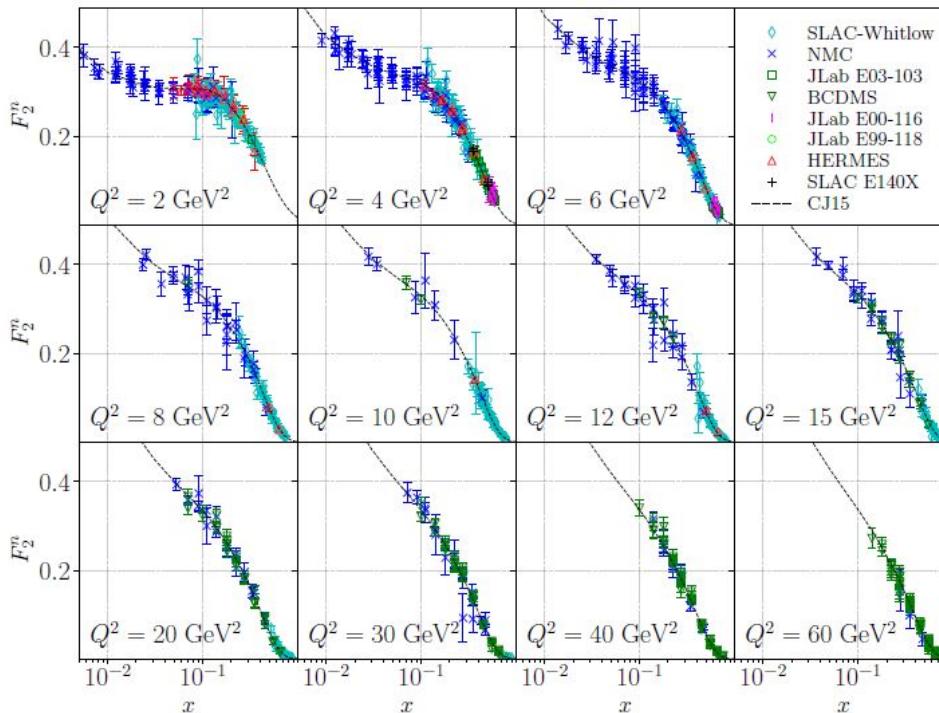
The CJ15 PDFs



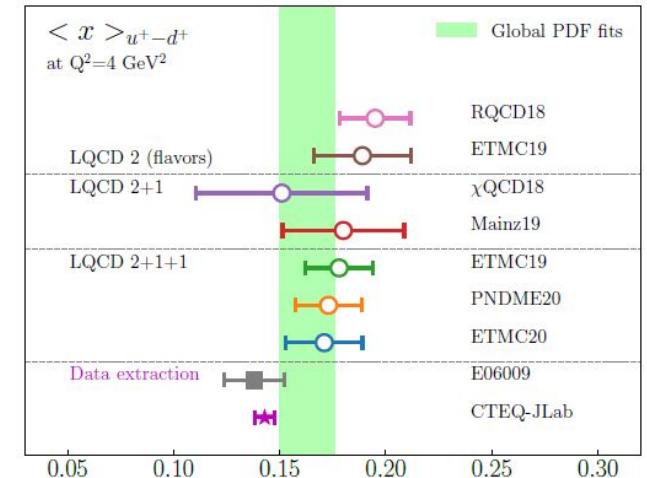
- Fitted with $\chi^2 = 1.04 / \text{datum}$
- Propagation of exp. errors
 - Hessian analysis
 - Correlated errors used if available
- “PDF error band” for $\Delta\chi^2 = 2.71$
 - $\rightarrow 90\% \text{ c.l. in a perfect world}$
 - Many alternative methods
 - See review by Kovarik et al.
- Theoretical systematics more difficult
 - Recent effort by fitting community

Neutron structure function extraction

- There is no free neutron target
 - Use (fitted) nuclear correction model
 - Confine model dependence in deuteron / free p + n correction ratio

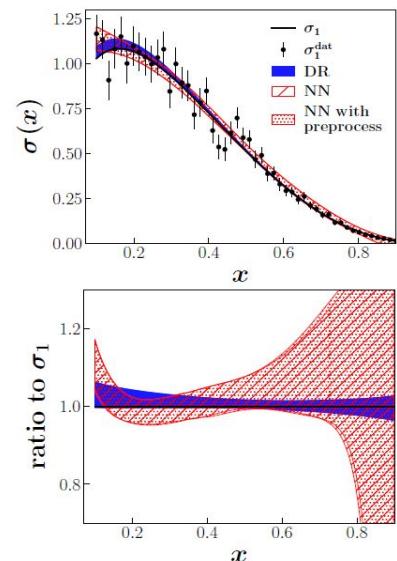


Apply data:
→ e.g. to benchmark lattice QCD



(Some of the) challenges

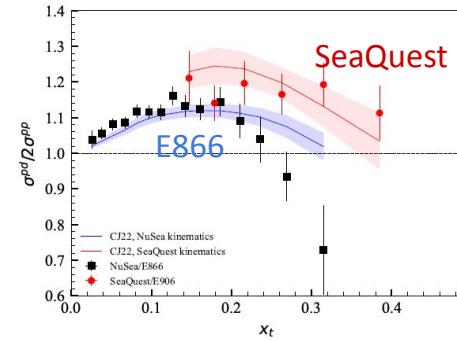
- **Very large data set:**
 - Need to: modernize code, speed up calcs!
- **Need fit flexibility → many parameters, highly correlated**
→ numerical instability in Hessian matrix
 - Increase numerical robustness?
 - Use neural networks?
 - But: Cross Valid. leads to likelihood deformation!
 - Better parametric methods
 - (iterative) bootstrap?
 - Fully Bayesian methods (MCMC, ...)?
- **Visualization tools / information compression**



CJ22

(Some of the) challenges

- **Incompatible data sets**
 - Increase “tolerance”: blow up errors
 - More solid (Bayesian) treatment
 - E.g., multi-Gaussian mixtures
- K. Mohan @ DIS 2023
- **How can AI/ML help?**



CJ22

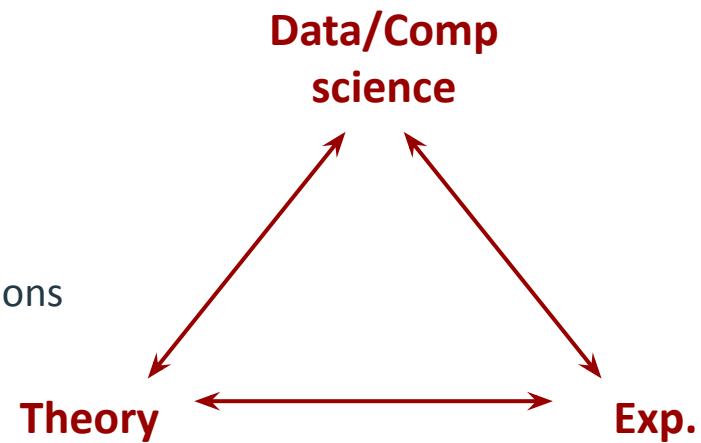
To conclude...

Summary and Perspective

- **What are we made of?**
 - Study the effects of confinement
 - 2 intertwined research lines
- **Global QCD analysis**
 - Many new data & better precision in near future
 - Needs matching uncertainty determination
 - **Interdisciplinary challenge**
 - **Expanded CTEQ-JLab collaboration**
- **Dynamical mass generation**
 - Theory: factorization with inclusive jet functions
 - New observables
 - **Novel global analysis**
 - PDFs + i-JFN → mass generation

Summary and Perspective

- **What are we made of?**
 - Study the effects of confinement
 - 2 intertwined research lines
- **Global QCD analysis**
 - Many new data & better precision in near future
 - Needs matching uncertainty determination
 - **Interdisciplinary challenge**
 - **Expanded CTEQ-JLab collaboration**
- **Dynamical mass generation**
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Thank you!

Backup #1:

Evidence for quarks and gluons

Evidence for quarks and gluons - a whirlwind tour

□ Baryon spectroscopy – light sector (u, d, s), ground state

- $J=3/2^+$: $|q_1\uparrow, q_2\uparrow, q_3\uparrow\rangle$ totally symmetric w.fn.
- $J=1/2^+$: $|q_1\uparrow, q_2\uparrow, q_3\downarrow\rangle$

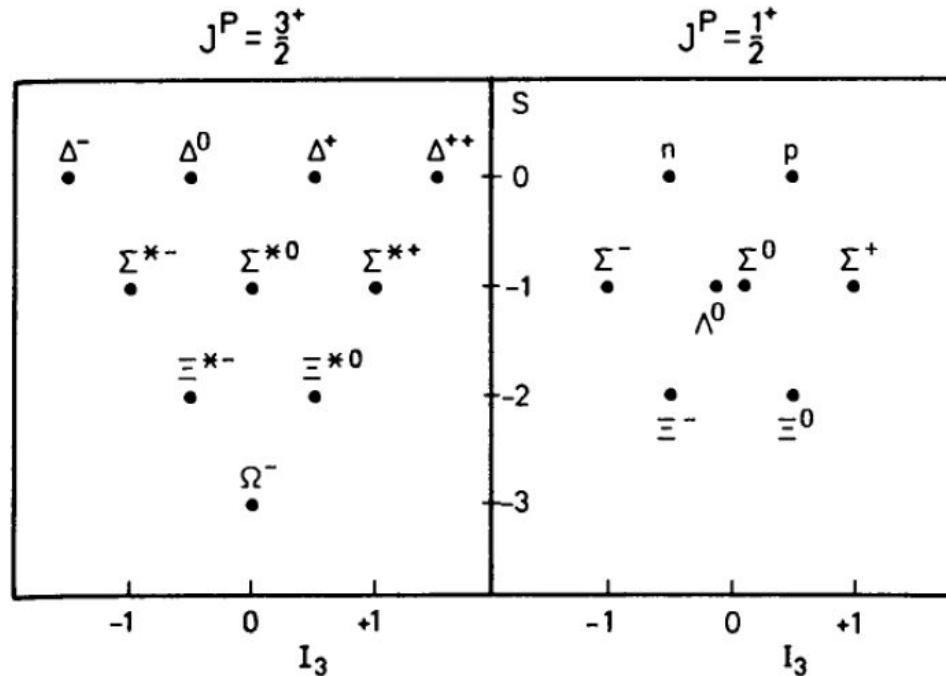


Fig:
[from Povh et al.]

Evidence for quarks and gluons - a whirlwind tour

□ Baryon spectroscopy – light sector (u, d, s), ground state

- $J=3/2^+$: $|q_1\uparrow, q_2\uparrow, q_3\uparrow\rangle$ spin symmetric, color antisymmetric
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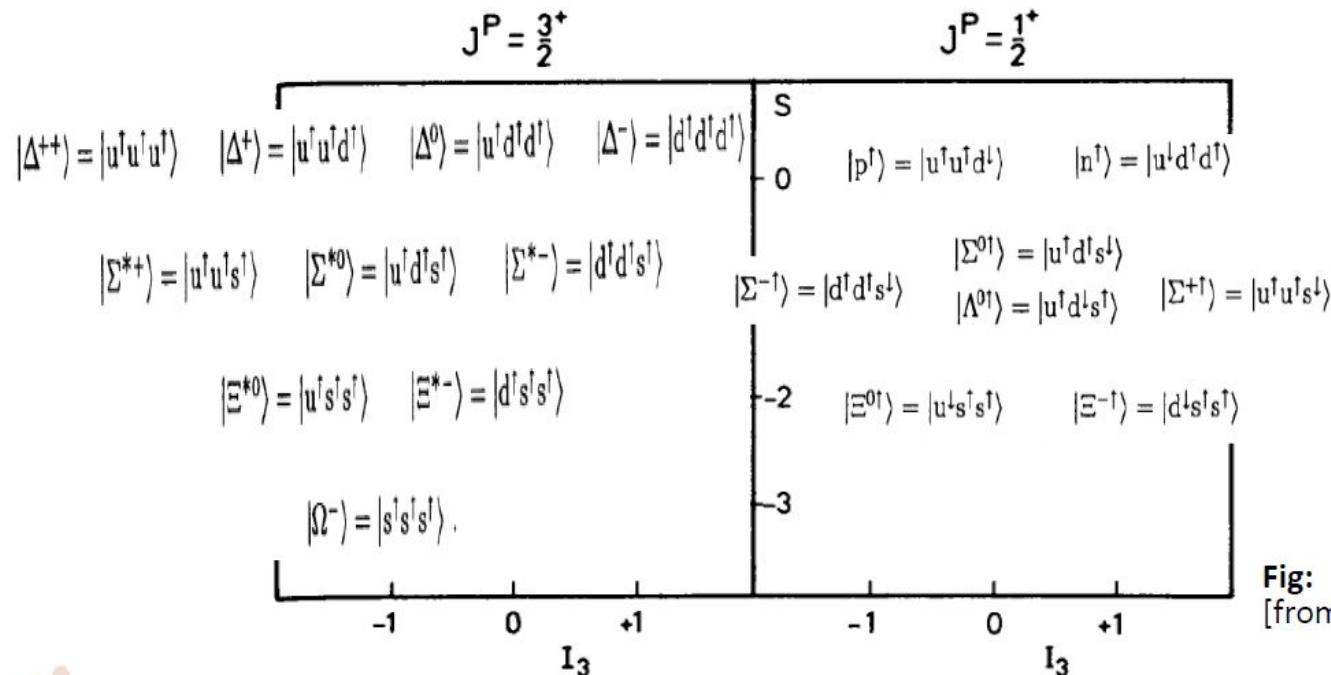


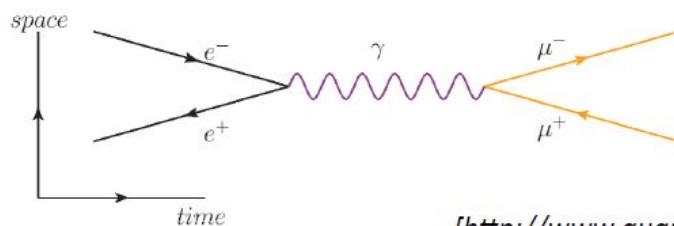
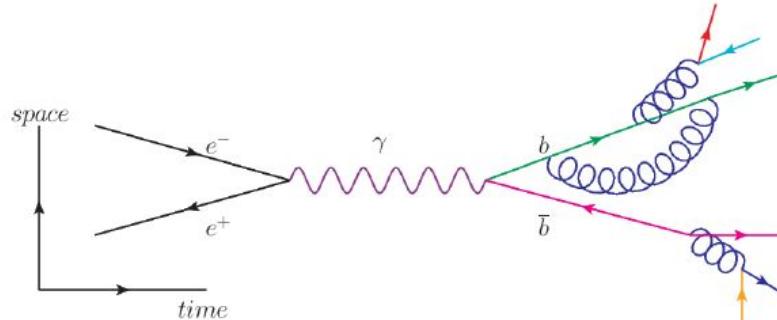
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Evidence for quarks and gluons - a whirlwind tour

☐ $e^+ + e^-$ annihilation into hadrons

- quark-mediated process $e^+ + e^- \rightarrow q + \bar{q} \rightarrow \text{hadrons}$

$$R = \frac{\sigma(e^+e^- \text{ hadrons})}{\sigma(e^+e^- \mu^+\mu^-)} = N_{\text{colors}} \sum_q e_q^2$$



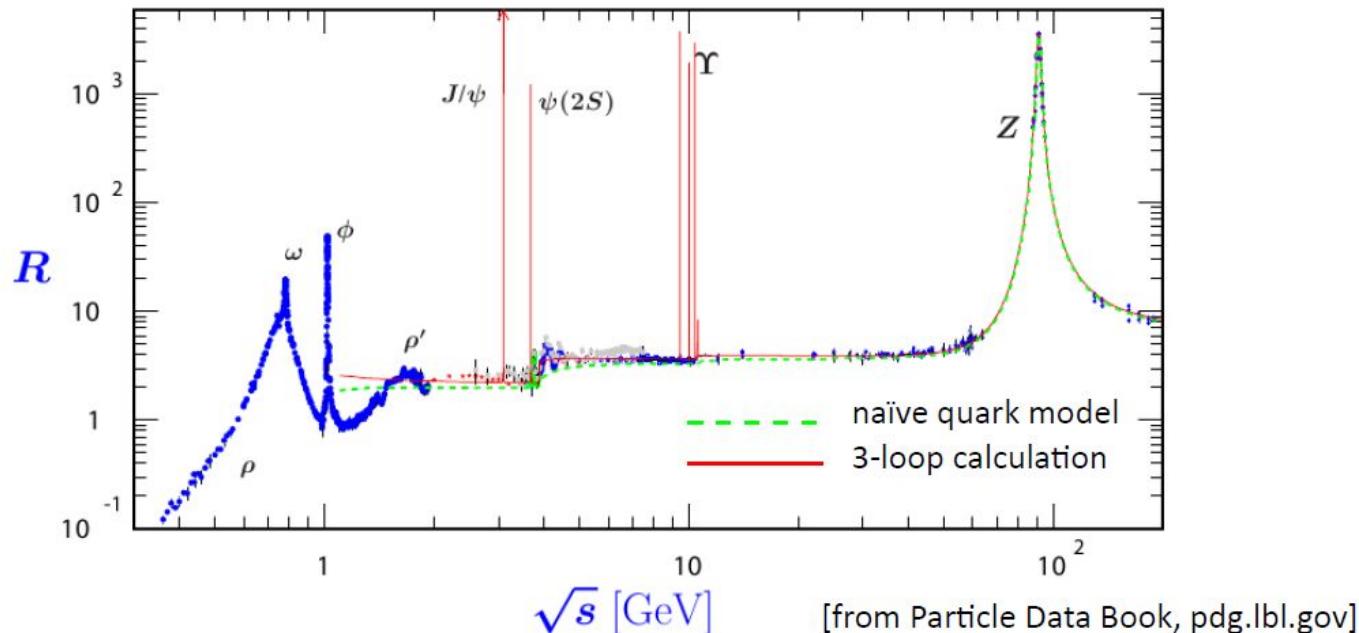
[<http://www.quantumdiaries.org/author/richard-ruiz/>]

Evidence for quarks and gluons - a whirlwind tour

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Evidence for quarks and gluons - a whirlwind tour

□ Jets in high-energy $e^+ + e^-$ collisions

- Hadron produced in 2, 3, ... N, high-energy collimated “jets”
- Evidence of common origin from a parton

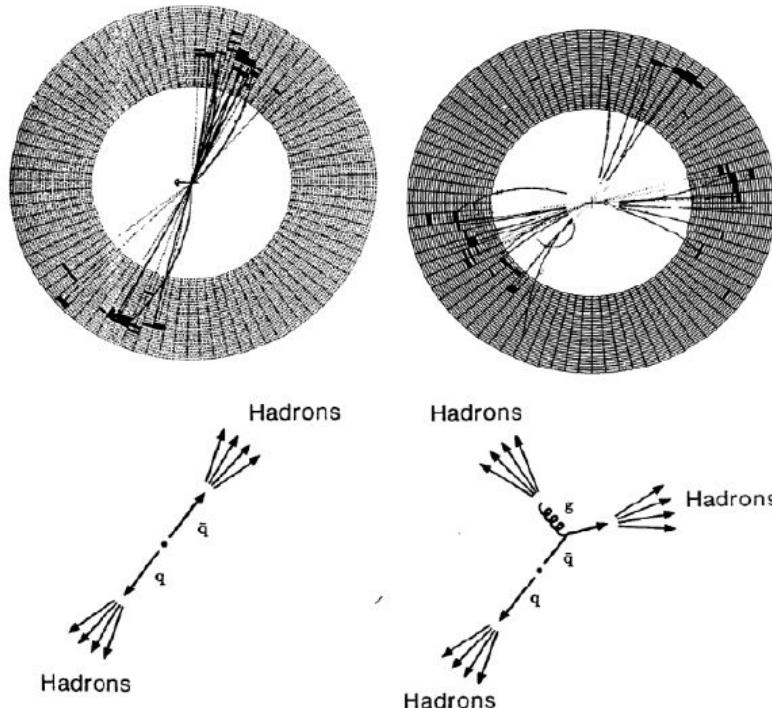


Fig.: 2- and 3-jet events observed by the JADE detector at PETRA [from Povh et al.]

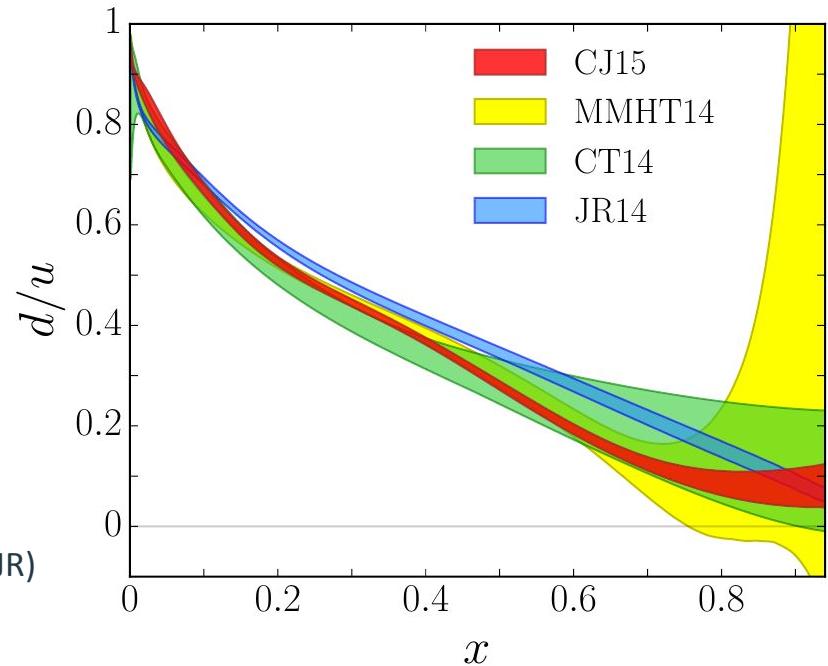
Backup #2:

Experimental uncertainties
or

“ Why do different global fits give
different PDF uncertainties? ”

Global fits are not created equal...

- **Uncertainty determination**
 - Hessian (JR)
 - Hessian + “Tolerance” (*)
 - $T \sim 10$ (CT14)
 - $T \sim 5-7$ (MMHT14)
 - $T = 1.646$ (CJ15)
 - Data Resampling (JAM)
 - DR + Cross Validation (NNPDF)
- **Parametrization**
 - $x^a (1-x)^b P(x)$ – most groups
 - d-quark:
 - extended (CJ, CT) or std (MMHT, JR)
 - Neural Nets – NNPDF
- **Data choice and coverage, ...**
 - Can use SLAC, JLab only if considering TMC, HT corrections
 - Highest x reach for d/u on proton if using reconstructed W asymmetries (vs. decay lepton asymmetries)
 - ...



(*) CJ vs. CT comparison on “equal” footing:
[Accardi, Hobbs, Jing, Nadolsky, EPJC 81 \(2021\) 7](#)

Bayesian estimators

- **Bayes theorem** $p(\mathbf{a}|\mathbf{m}) = \frac{1}{\mathcal{Z}} p(\mathbf{m}|\mathbf{a}) p(\mathbf{a})$

with “evidence”

$$\mathcal{Z} = \int d\mathbf{a} p(\mathbf{m}|\mathbf{a}) p(\mathbf{a})$$

and “likelihood”

$$p(\mathbf{m}|\mathbf{a}) = \mathcal{N} \exp \left[-\frac{1}{2} \chi^2(\mathbf{a}, \mathbf{m}) \right]$$

Typical choice
in PDF analyses

- Algorithms for sampling of likelihood → probability density in $\{\mathbf{a}_k\}$
 - **HMC:** Hamiltonian Monte Carlo (an example of Markov-Chain MC methods)
 - **NS:** Nested Sampling, primarily aimed at estimating the evidence
→ Samples the likelihood as a byproduct

- **Expectation values**

$$E_{\text{Bayes}}\{\mathcal{O}(\mathbf{a})\} = \frac{1}{n} \sum_{k=1}^n \mathcal{O}(\mathbf{a}_k),$$

and variance

$$V_{\text{Bayes}}\{\mathcal{O}(\mathbf{a})\} = \frac{1}{n} \sum_{k=1}^n [\mathcal{O}(\mathbf{a}_k) - E_{\text{Bayes}}\{\mathcal{O}(\mathbf{a})\}]^2$$

Generalized Hessian Approximation

Hunt-Smith et al., PRD 106 (2022) 036003

- **Start as usual:**
 - Find minimum of likelihood
 - Diagonalize Hessian → e_k eigenvectors, w_k eigenvalues
- **Change variables:** $\mathbf{a}(\mathbf{t}) = \mathbf{a}_0 + \sum_{k=1}^{n_{\text{par}}} t_k \frac{\mathbf{e}_k}{\sqrt{w_k}}$, then $p(\mathbf{a}|m) \rightarrow p(\mathbf{t}|m)$
- **Assume likelihood factorized along Hessian eigendirection, then**

$$E_{\text{Hess}}\{\mathcal{O}(\mathbf{a})\} = \int d^n t \, p(\mathbf{t}|m) \mathcal{O}(\mathbf{a}(\mathbf{t})) \approx \mathcal{O}(\mathbf{a}_0)$$

$$V_{\text{Hess}}\{\mathcal{O}(\mathbf{a})\} \approx \sum_k T_k^2 \left(\left. \frac{\partial \mathcal{O}(\mathbf{a}(\mathbf{t}))}{\partial t_k} \right|_{\mathbf{a}_0} \right)^2$$

- **Here** $T_k^2 = \int dt_k \, p_k(t_k|m) t_k^2$ **is the “tolerance” :**
 - $T_k = 1$ where likelihood is Gaussian;
 - Approximates well the likelihood in non-Gaussian directions
 - Maintains a “68%” or “ 1σ ” kind of meaning also when $\neq 1$

CT, MSTW → T=5-10

- **Often T_k determined “ad hoc” to account for statistical inconsistency of data**

Data resampling

- Data Resampling (DR) approximates Bayes' posterior using frequentist logic
 - Assume some prior (typically “flat”)
 - Reshuffle data within data uncertainty (Gaussian distribution)
 - Maximize likelihood
 - Repeat n_{rep} times → $\{\mathbf{a}_k\}$
- Estimate

$$E_{\text{freq}}\{\mathcal{O}(\mathbf{a})\} = \frac{1}{n_{\text{rep}}} \sum^{n_{\text{rep}}} \mathcal{O}(\mathbf{a}_{\text{rep}}),$$

$$V_{\text{freq}}\{\mathcal{O}(\mathbf{a})\} = \frac{1}{n_{\text{rep}}} \sum^{n_{\text{rep}}} [\mathcal{O}(\mathbf{a}_{\text{rep}}) - E_{\text{freq}}\{\mathcal{O}(\mathbf{a})\}]^2$$

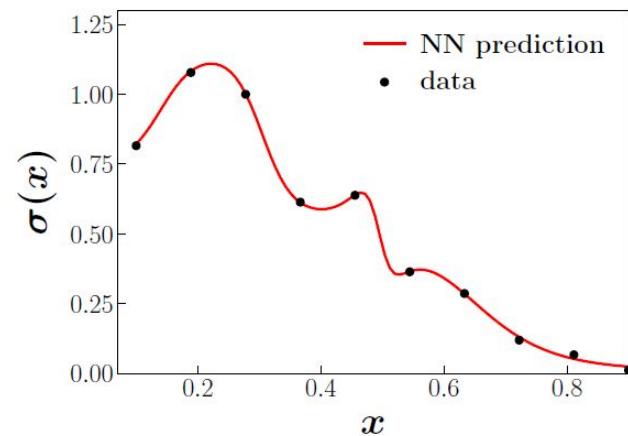
- Good in parameter space region well constrained by data

Neural Networks and overfitting

- Neural networks provide:
 - Efficient, very flexible parametrizations
 - Hundreds of parameters
 - Essentially a parameter free functional form (“nonparametric method”)
- Use Data Resampling and aims at maximizing the same likelihood function

$$p(\mathbf{m}|\mathbf{a}) = \mathcal{N} \exp \left[-\frac{1}{2} \chi^2(\mathbf{a}, \mathbf{m}) \right]$$

- Without intervention, will overfit the data
 - The plot shows an extreme example

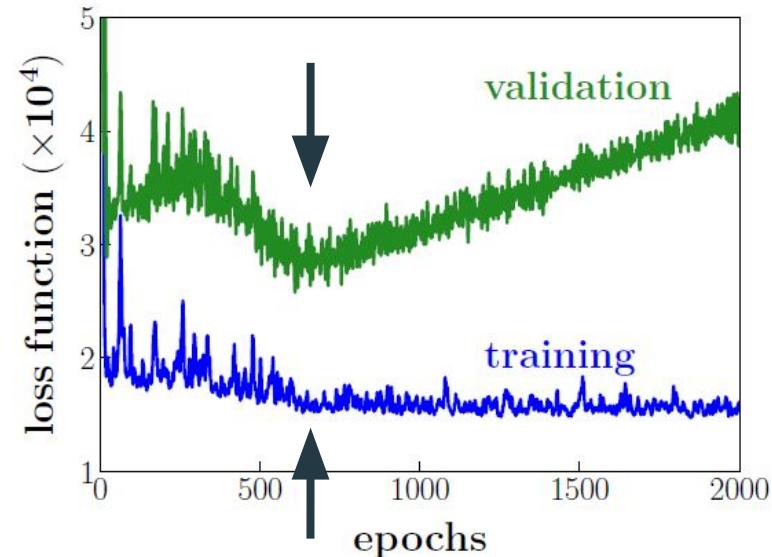


Cross-validation (CV) and stopping

- Needs a “stopping criterion”
 - to avoid fitting statistical noise instead of physics
- Randomly separate the data into 2 groups, say
 - 70% → training (T)
 - 30 % → validation (V)
- Fit the training, calculate $\chi^2(T)$ and $\chi^2(V)$
- Resample data, repeat
- “Stop” training when $\chi^2(V)$ is minimum:

$$\sigma = E[\sigma_{\text{fit}}]$$

$$\delta\sigma = V[\sigma_{\text{fit}}]$$



Statistical uncertainties

- The method can effectively modify the likelihood!

- Even with perfectly compatible (toy) data!

[N. Hunt-Smith et al., PRD 106 \(2022\) 036003](#)

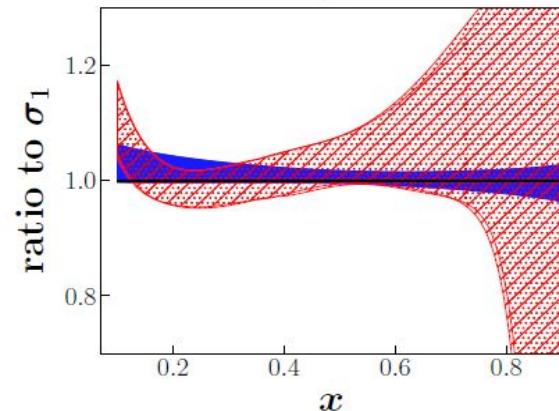
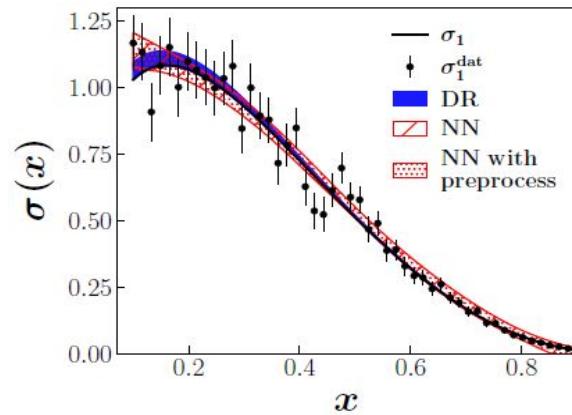
- Bayesian Methods

(Markov Chain MC, Nested Sampling)

- Explore the likelihood function
- Well approximated by
→ Hessian, Data Resampling (DR)

- Cross Validation, NN-based fits

- Inflate the uncertainties
- Deform the likelihood



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