

DATA DRIVEN EXTRACTION OF PROTON'S

VALENCE QUARK RATIO

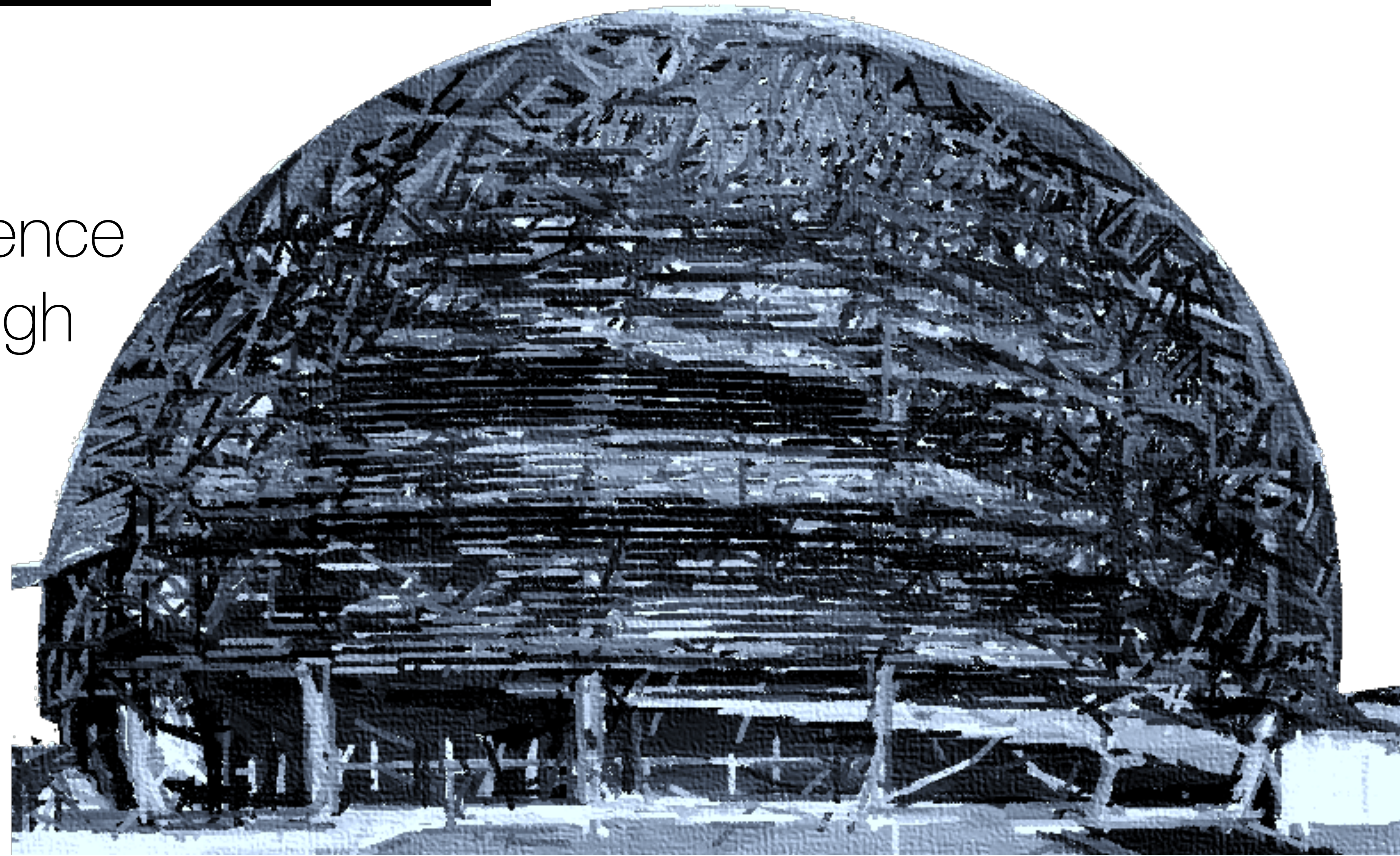
DANIELE BINOSI

ECT* - FONDAZIONE BRUNO KESSLER

Perceiving the Emergence
of Hadron Mass through

AMBER@CERN **6**

SEPTEMBER 27 - SEPTEMBER 29 2021

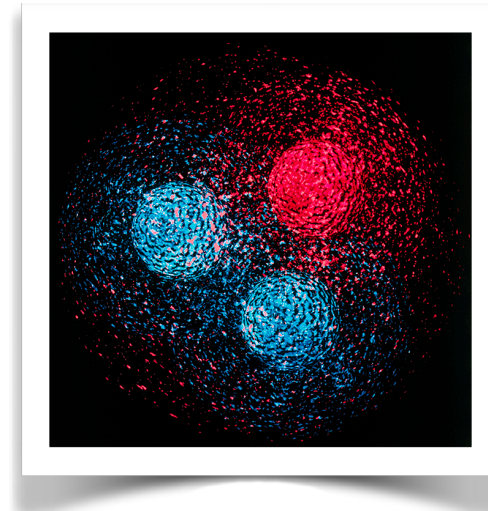


ECT*

EUROPEAN CENTRE
FOR THEORETICAL STUDIES
IN NUCLEAR PHYSICS AND RELATED AREAS

FONDAZIONE BRUNO KESSLER

Proton STRUCTURE



Nature's lowest-mass **bound-state** in the scattering of two u -quarks and one d -quark

Quantum ChromoDynamics describes the proton's wave function Ψ

Knowledge of Ψ yields proton's gluons and quarks *number density distributions*

DISTRIBUTION FUNCTIONS

Powerful discriminator between competing descriptions of proton structure:

ratio of u and d quark densities in the far-valence domain

1 THEORY

Consider DIS reactions from neutron and proton targets
ratio of Poincaré-invariant structure functions

$$\frac{F_2^n(x)}{F_2^p(x)} = \frac{\underbrace{u_v(x) + 4d_v(x)}_{\text{valence}} + \underbrace{6d_s(x) + \Sigma(x)}_{\text{sea}}}{4u_v(x) + d_v(x) + 6u_s(x) + \Sigma(x)}$$

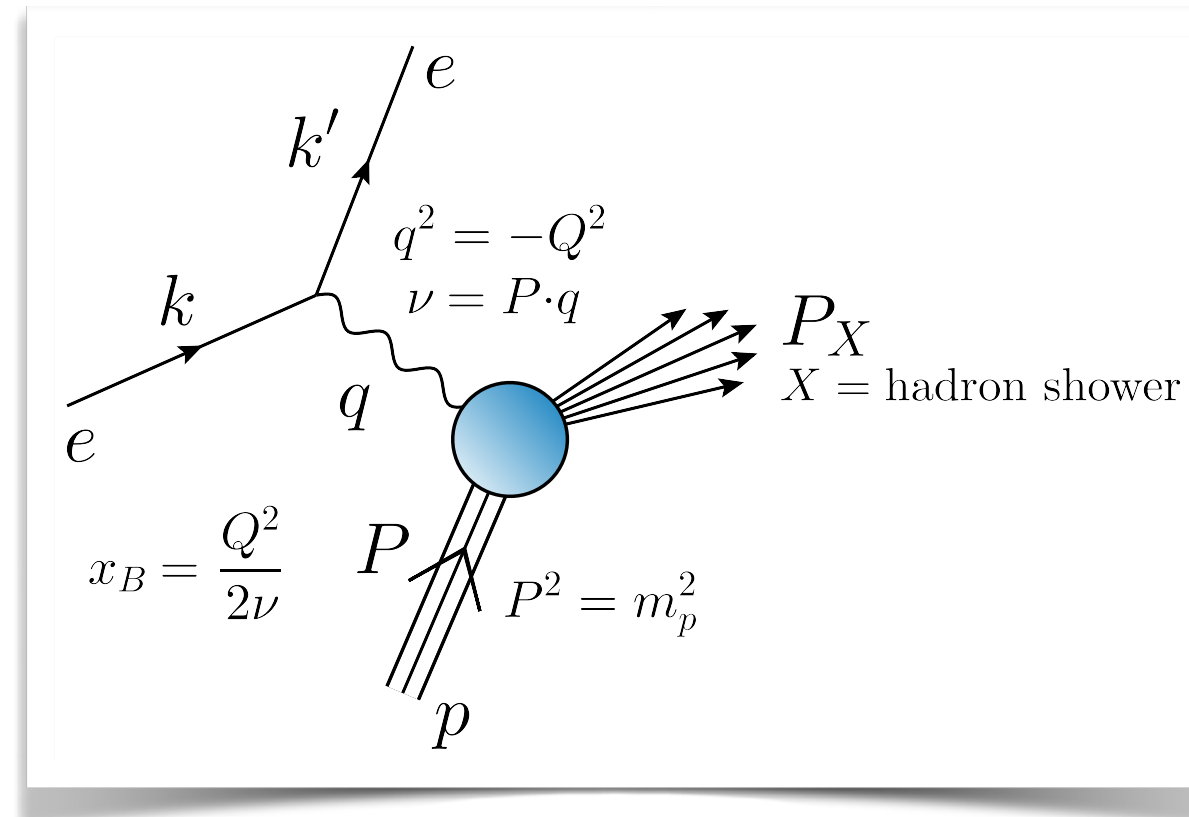
Sea distributions negligible on $x \gtrsim 0.2$

$$\frac{F_2^n(x)}{F_2^p(x)} \underset{x \gtrsim 0.2}{\approx} \frac{1 + 4d_v(x)/u_v(x)}{4 + d_v(x)/u_v(x)}$$

d/u ratio fixed point under QCD evolution at $x = 1$

Nachtmann bounds:

$$1/4 \leq F_2^n(x)/F_2^p(x) \leq 4$$



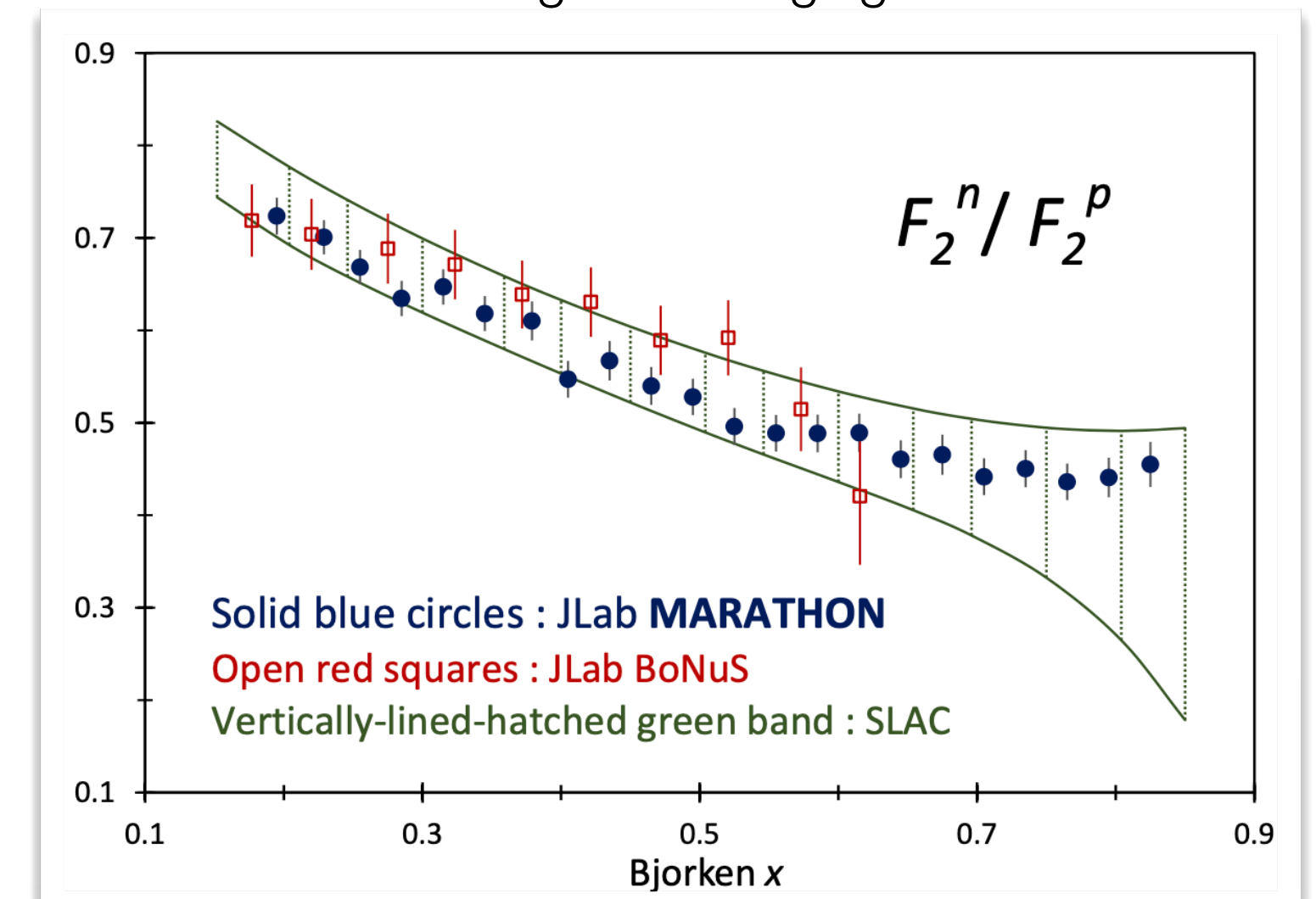
Jefferson Lab EXPERIMENT 2

Main problem: construction of a **free neutron target**

Use the *deuteron* (BONuS); large systematic uncertainties beyond $x \gtrsim 0.7$ due to characterisation of proton-neutron interactions

Perform DIS measurement on ^3He and ^3H and take ratio of scattering rates (MARATHON)

Nuclear interaction effects largely cancel; handling of radioactive ^3He target challenging



WHAT DO DATA ENCODE ABOUT PROTON'S STRUCTURE?

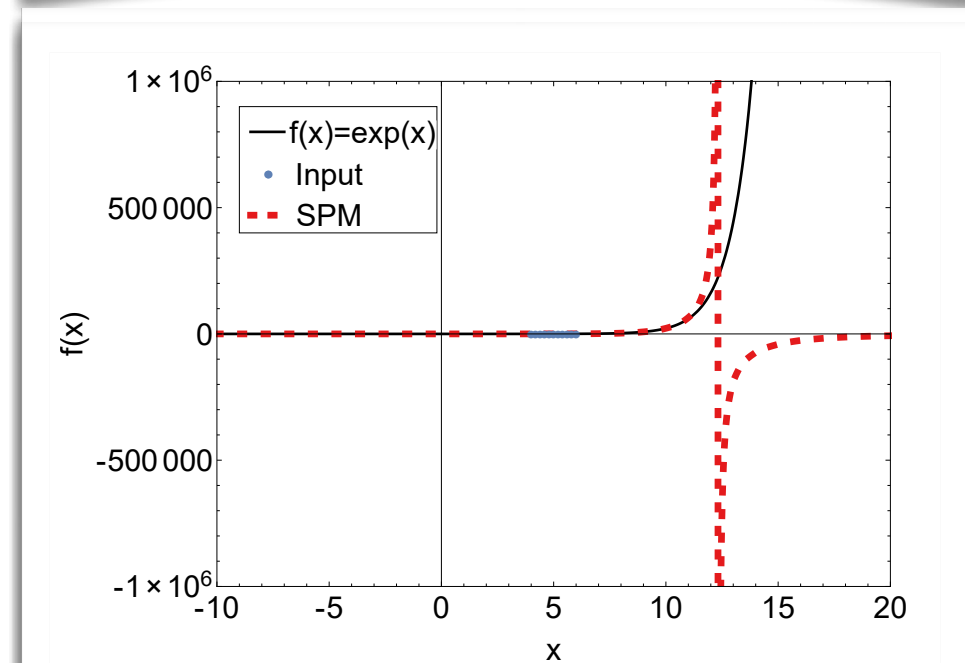
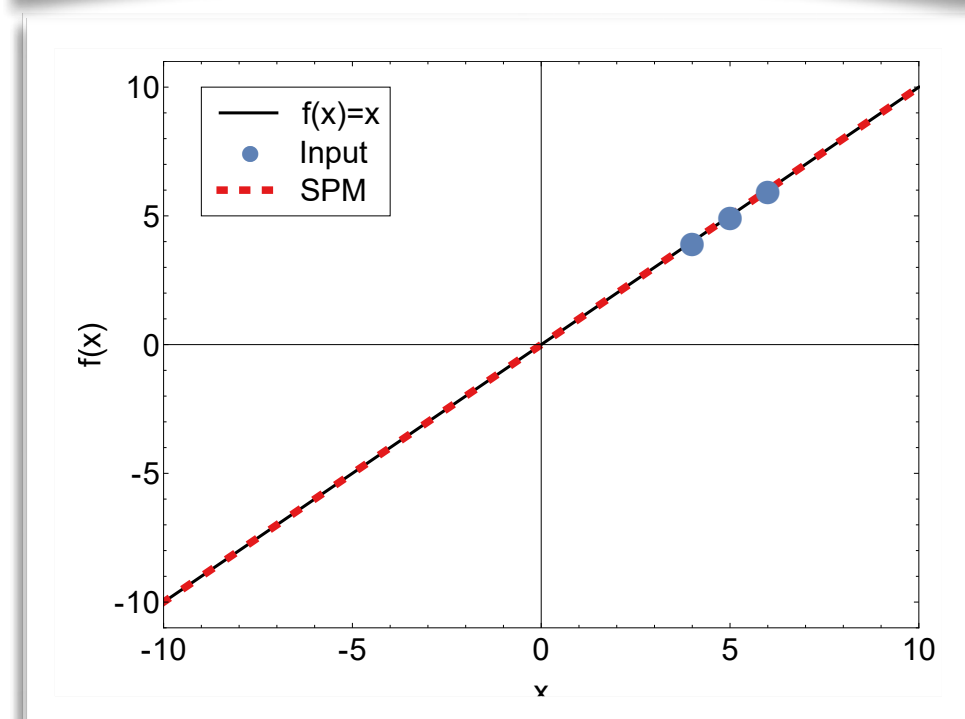
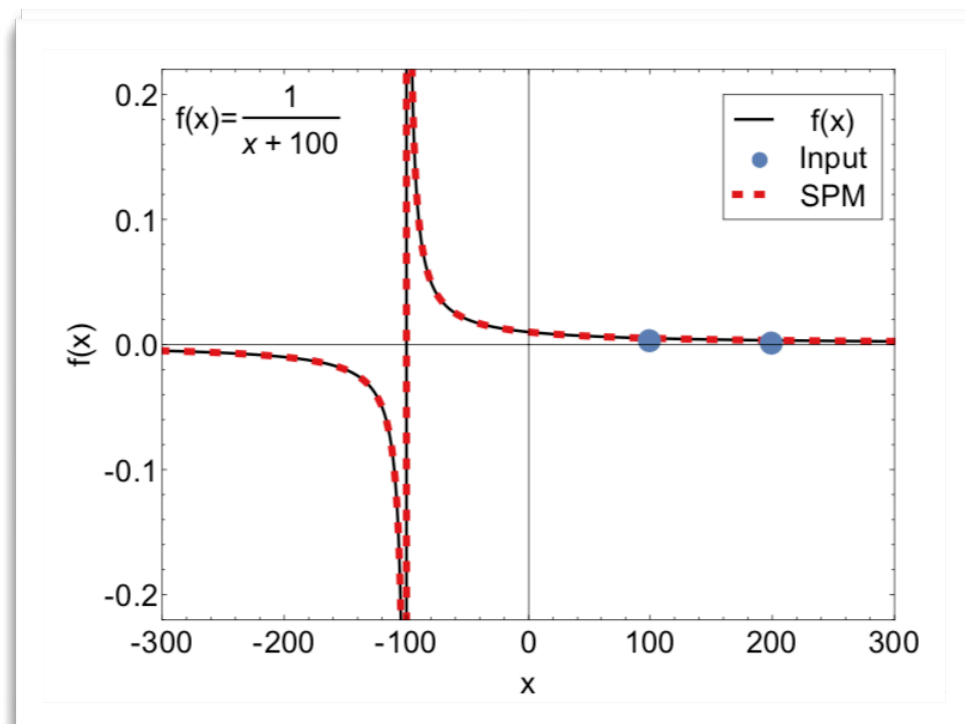
SPM AND SMOOTHING

$$D = \{(x_i, y_i = f(x_i)), i = 1, \dots, N\}$$

$$C_N(x) = \frac{y_1}{1+} \frac{a_1(x-x_1)}{1+} \frac{a_2(x-x_2)}{1+} \dots \frac{a_{N-1}(x-x_{N-1})}{1}$$

Schlessinger, PR 167 (1968)

elementary (functions) examples



LARGE DATASETS

randomly choose $4 < M \lesssim N/2$ points
 reduce (binomial) number of interpolators
 introducing **physical constraints**
 (absence of poles)

Chen et al., PRD 99 (2019)

IN THE PRESENCE OF ERRORS?

direct interpolation does not work
 requires **smoothing** with **roughness penalty**:

seek $g \in \mathcal{S}$ minimising

$$P(g, \lambda) = \lambda \sum_{i=1}^{\ell} [y_i - g(x_i)]^2 + (1 - \lambda) \int_a^b dx [g''(x)]^2$$

smoothing par. data fidelity roughness penalty

THEOREM: g is the *natural spline* interpolant of nodes $\{x_i\}$

Reinsch, NM 10 (1967)

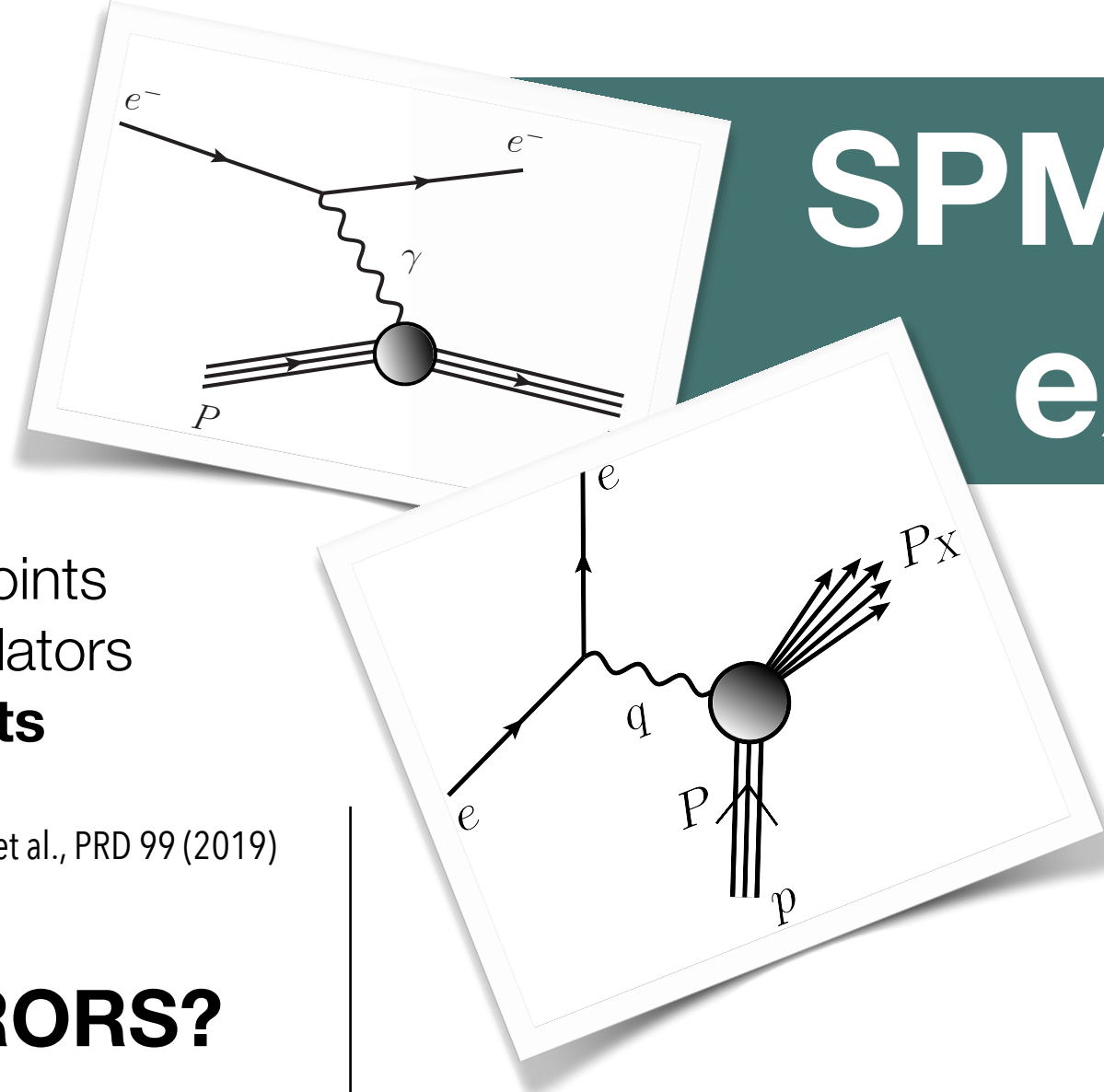
optimal smoothing parameter determined via generalised cross validation

Craven and Wahba, NM 31 (1978)

use **bootstrap** procedure to generate replicas accounting for statistical errors in data when extrapolating

$$(x_i, y_i, \sigma_i) \rightarrow (x_i, \mathcal{N}(y_i, \sigma_i))$$

SPM parameter extraction



General algorithm to extrapolate a certain parameter from given noisy experimental datasets.

generate (10^3) replicas for the given experimental central values and error
 smooth each replica with associated optimal λ

set $\{M_j = 5 + j \mid j = 1, \dots, n_M\}$ for a suitable n_M

fix M_j and get a number of monotonic SPM interpolators for each replica
 determine the replicas' parameter value (averaging over the obtained curves)
 construct the (normal) distribution of the replicas' (10^3) extracted parameter;

extract the mean p^{M_j} and standard deviation $\sigma_p^{M_j}$

final result

$$p \pm \sigma; \quad p = \sum_{j=1}^{n_M} \frac{p^{M_j}}{n_M}; \quad \sigma_p = \left[\sum_{j=1}^{n_M} \frac{(\sigma_p^{M_j})^2}{n_M^2} + \sigma_{SM}^2 \right]^{\frac{1}{2}}$$

standard deviation of p^{M_j} distribution

DOES IT WORK?

SPM AND SMOOTHING VALIDATION

does it really work? ✓
is it robust? ✓
If you want to disprove a result, show first you can replicate it

build **replicas** of the observable of **known parameter values** p^*

DATA GENERATORS

Generate data from a variety of models: functional forms; parametrisations of experimental data; “real-world” calculations

CHECKS

$\forall M$ /generators/kinematics:

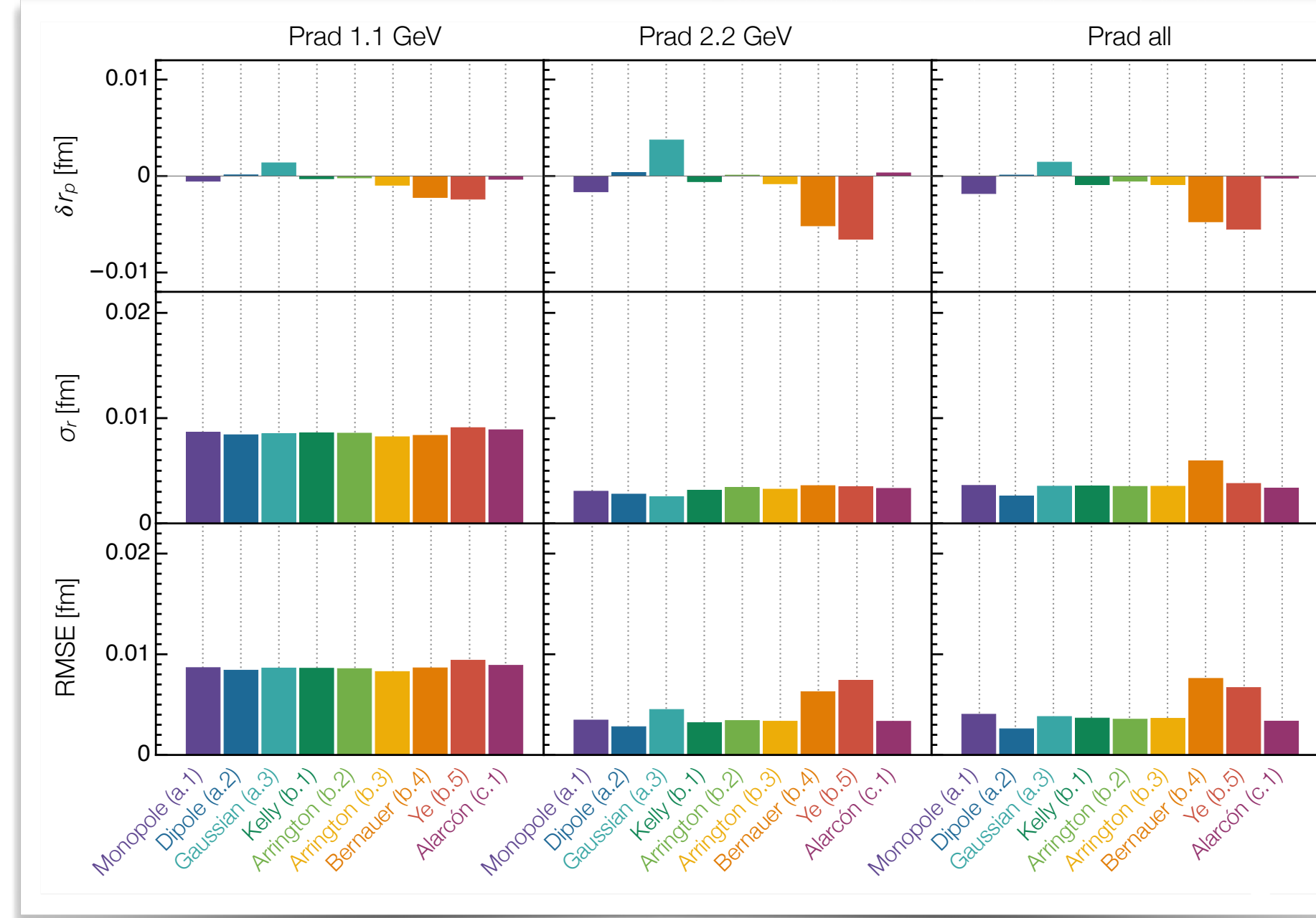
Gaussianity of p distribution ✓
robustness of p extraction ✓

bias $\delta p = p - p^*$ standard deviation σ_p Root Mean Square Error $RMSE = \sqrt{\delta p^2 + \sigma_p^2}$

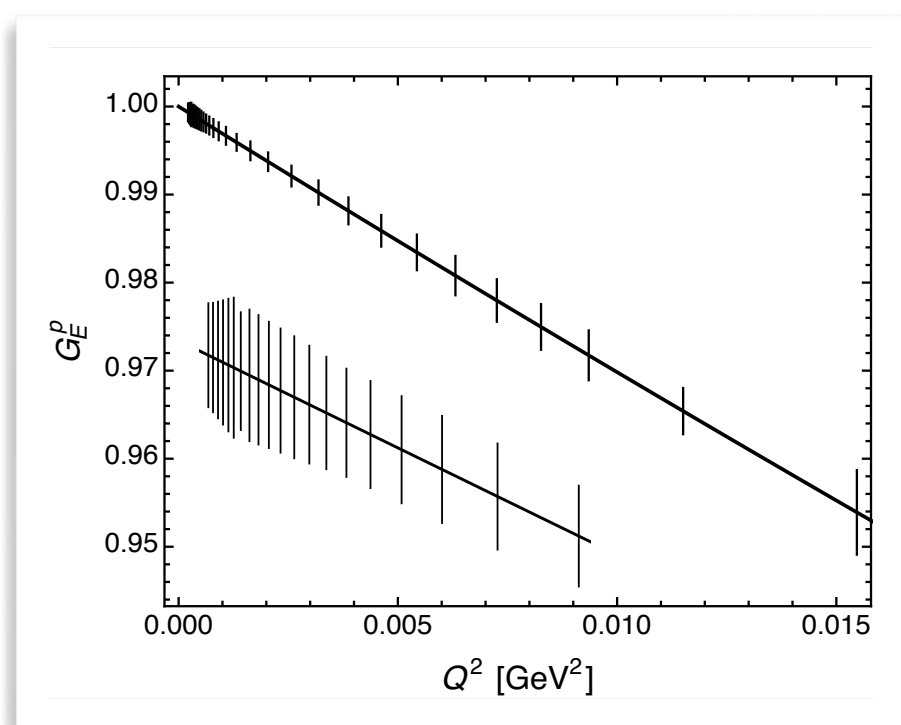
p extraction robust if

$$|\delta p| < \sigma_p \quad \checkmark$$

RMSE independent from generator ✓

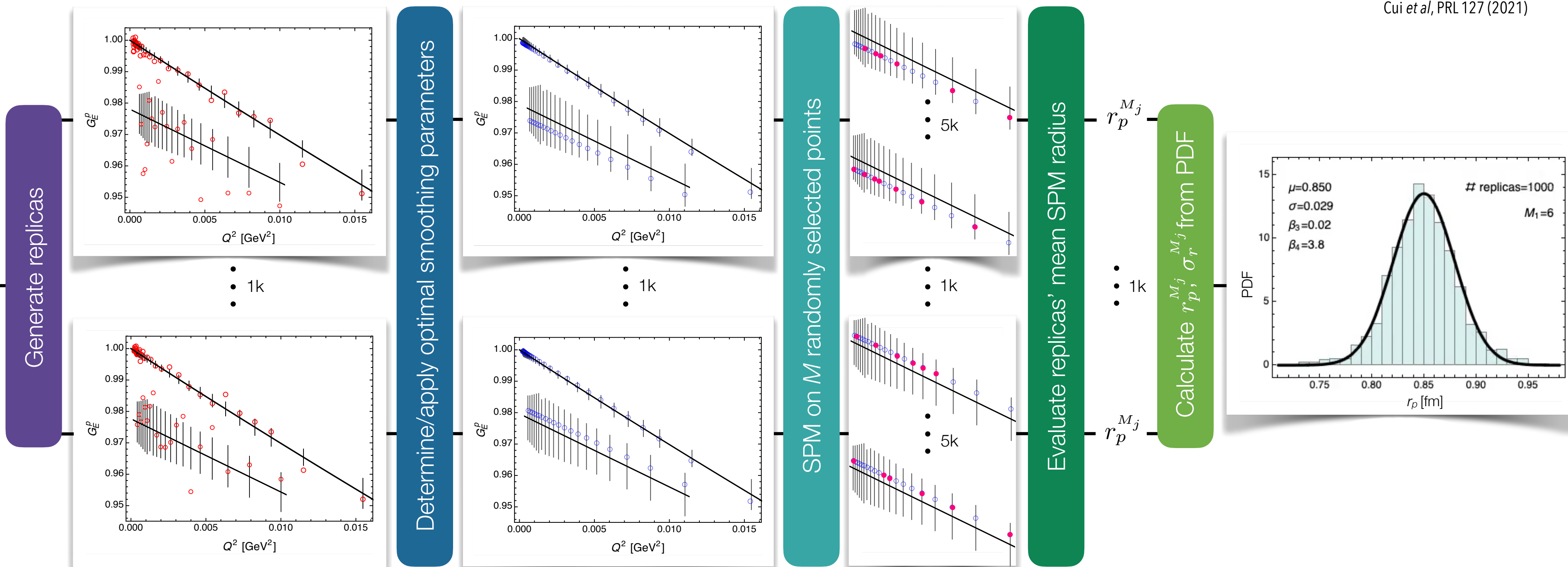


Cui et al, PRL 127 (2021)

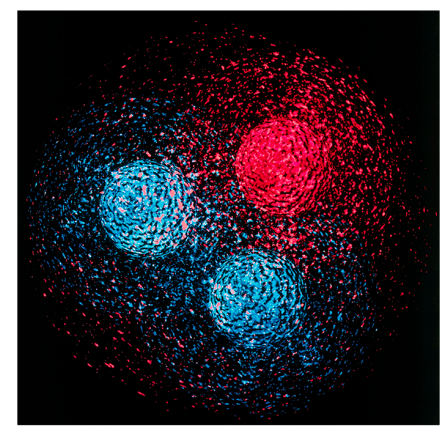


EXAMPLE: 1.1 GeV kinematics

DIPOLE, M=6



Proton CHARGE RADIUS



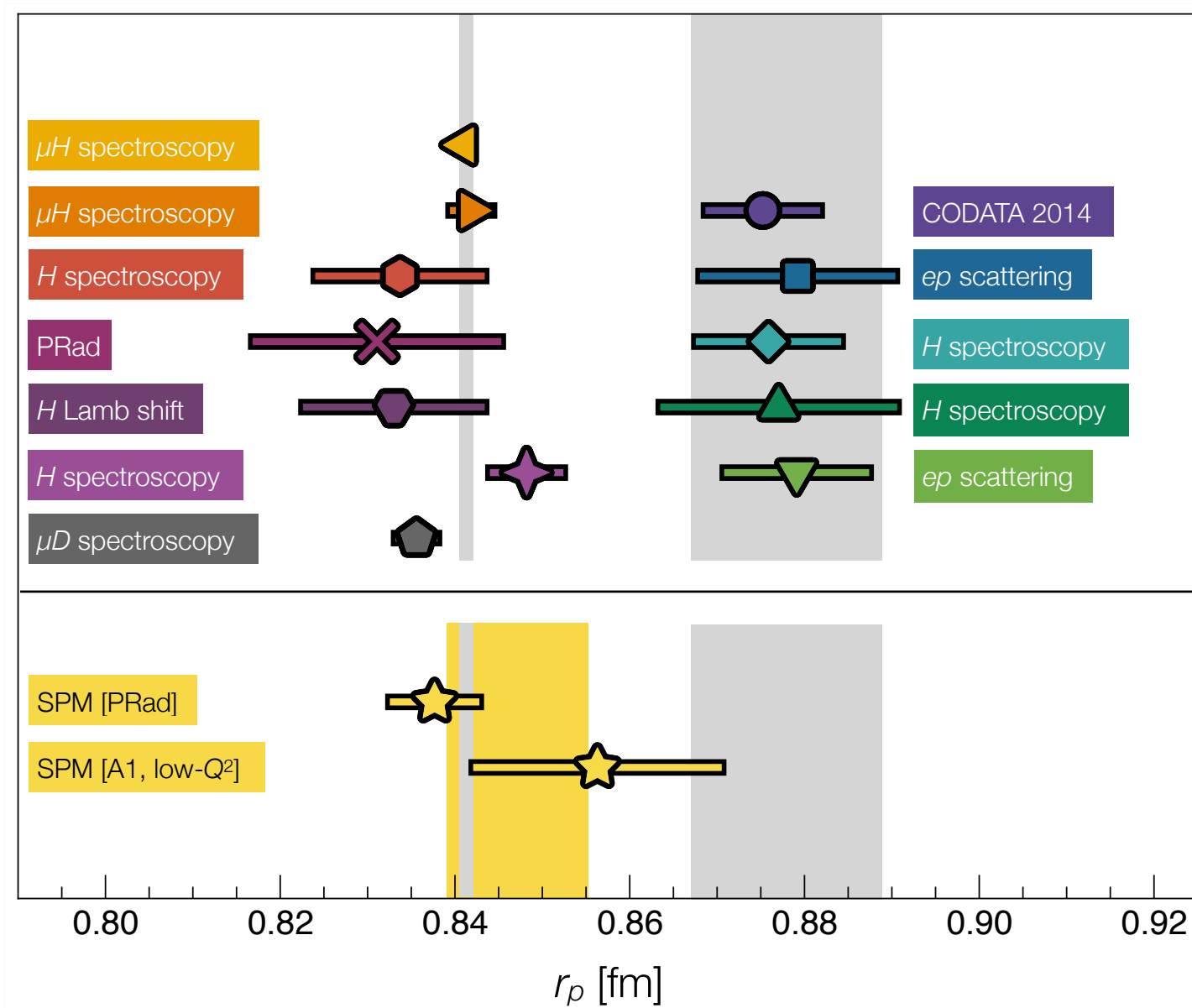
1 PRad DATA

$$r_p^{\text{PRad}} = 0.838 \pm 0.005_{\text{stat}} \text{ [fm]}$$

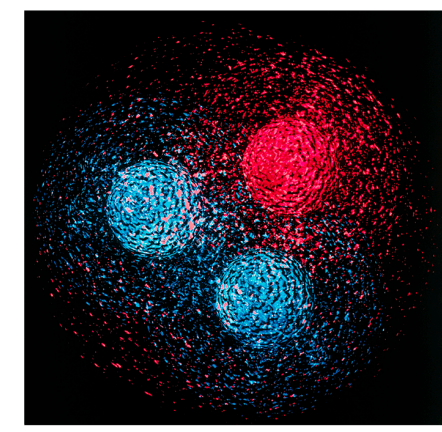
2 A1 DATA

$$r_p^{\text{A1-low}Q^2} = 0.856 \pm 0.014_{\text{stat}} \text{ [fm]}$$

Cui et al, PRL 127 (2021)



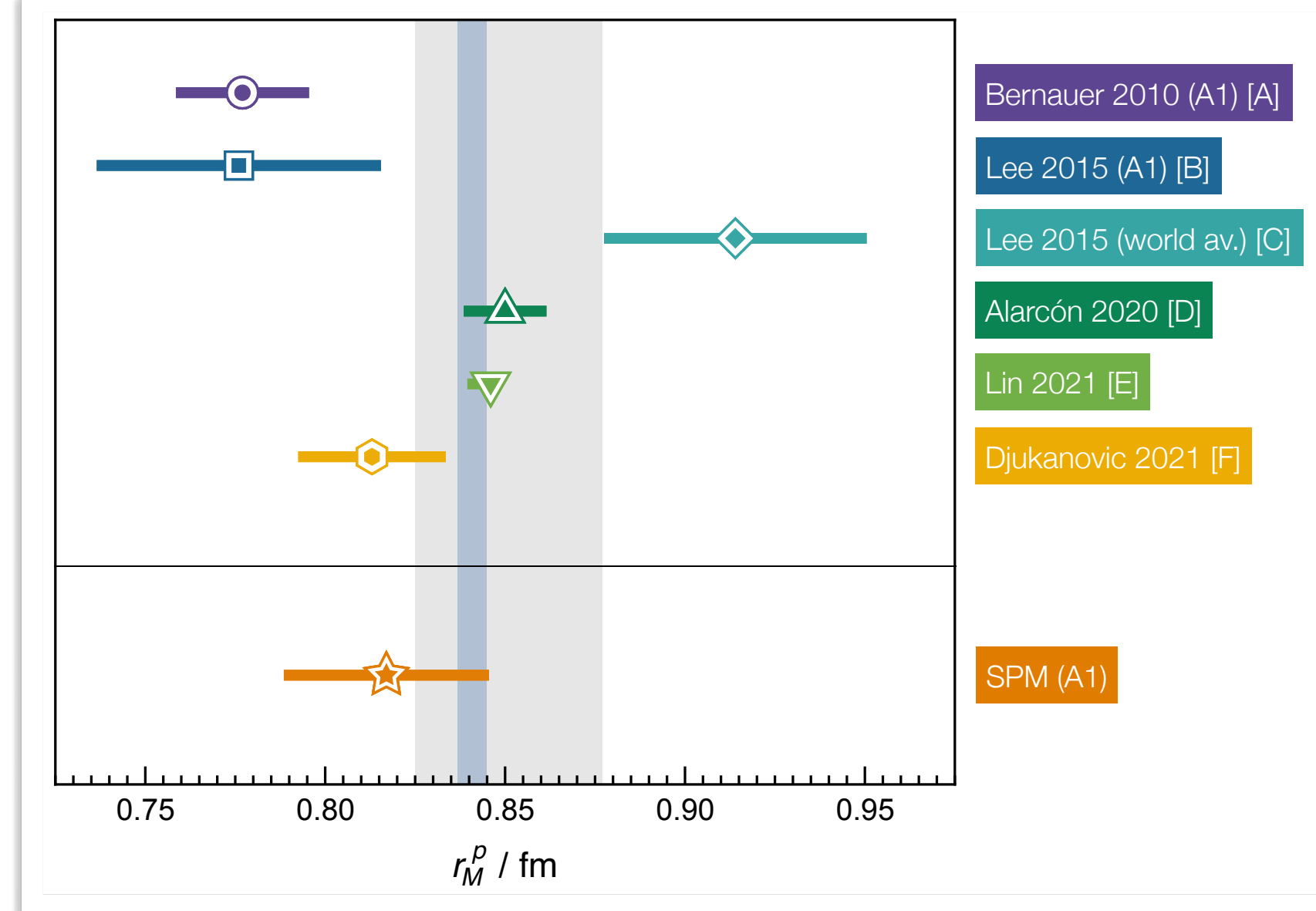
Proton PAULI RADIUS



1 A1 DATA

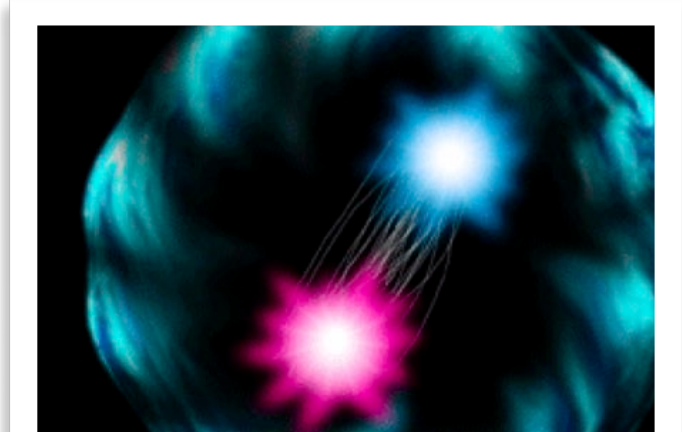
$$r_M^{\text{A1}} = 0.817 \pm 0.027_{\text{stat}} \text{ [fm]}$$

Cui et al, 2109.08768



$$\frac{F_2'(0)}{\kappa_p} \simeq F_1'(0) - \frac{\mu_p}{4m_p^2}$$

Pion CHARGE RADIUS

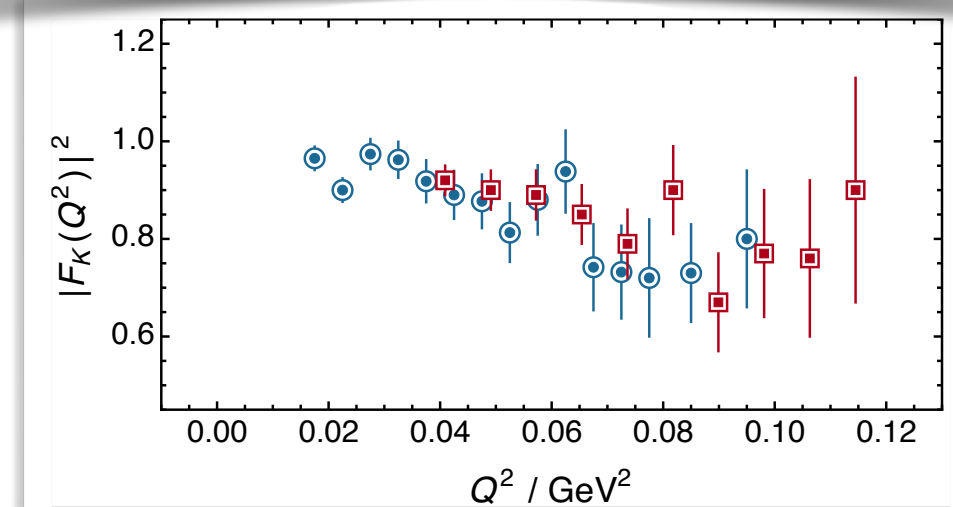
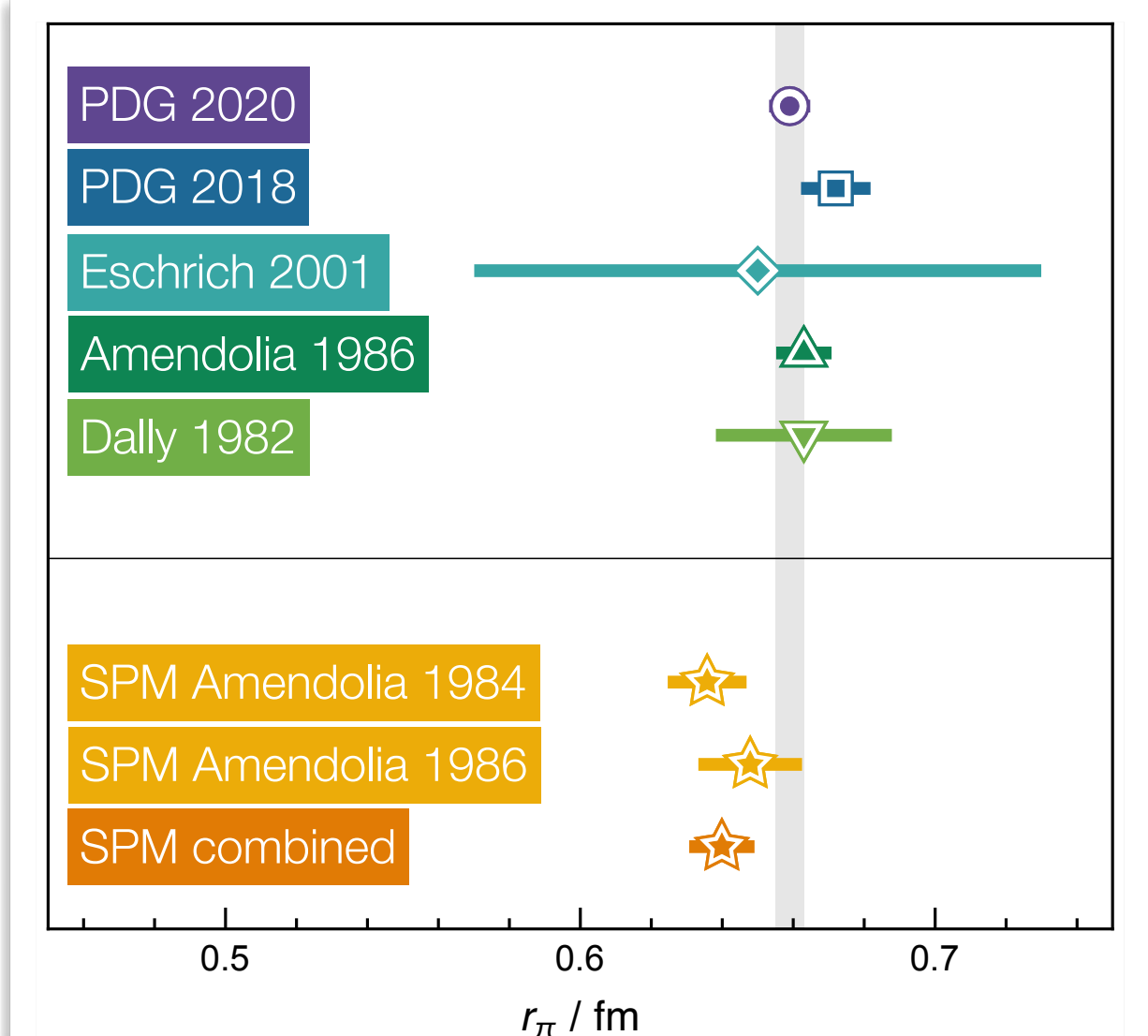


1 NA7 DATA

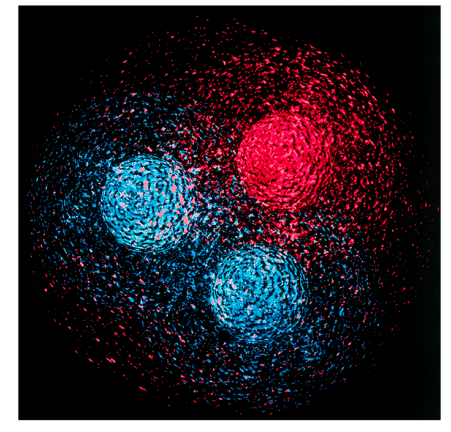
$$r_\pi^{\text{NA7-86}} = 0.648 \pm 0.013_{\text{stat}} \text{ [fm]}$$

$$r_\pi^{\text{NA7-84}} = 0.636 \pm 0.009_{\text{stat}} \text{ [fm]}$$

Cui et al, PLB 822 (2021)



Proton STRUCTURE



MARATHON dataset
21 data equally spaced in x
 $0.225 \leq x \leq 0.825$

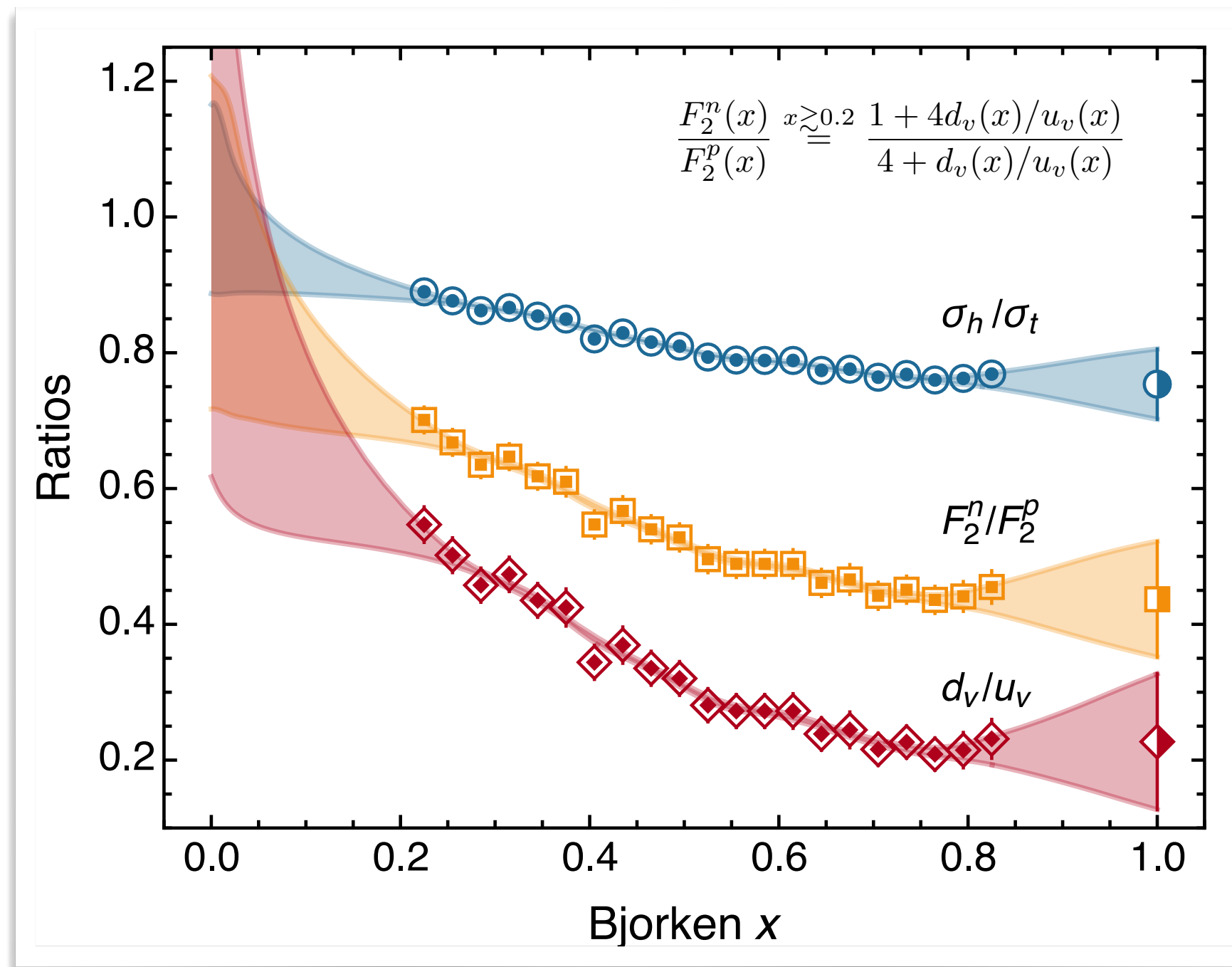
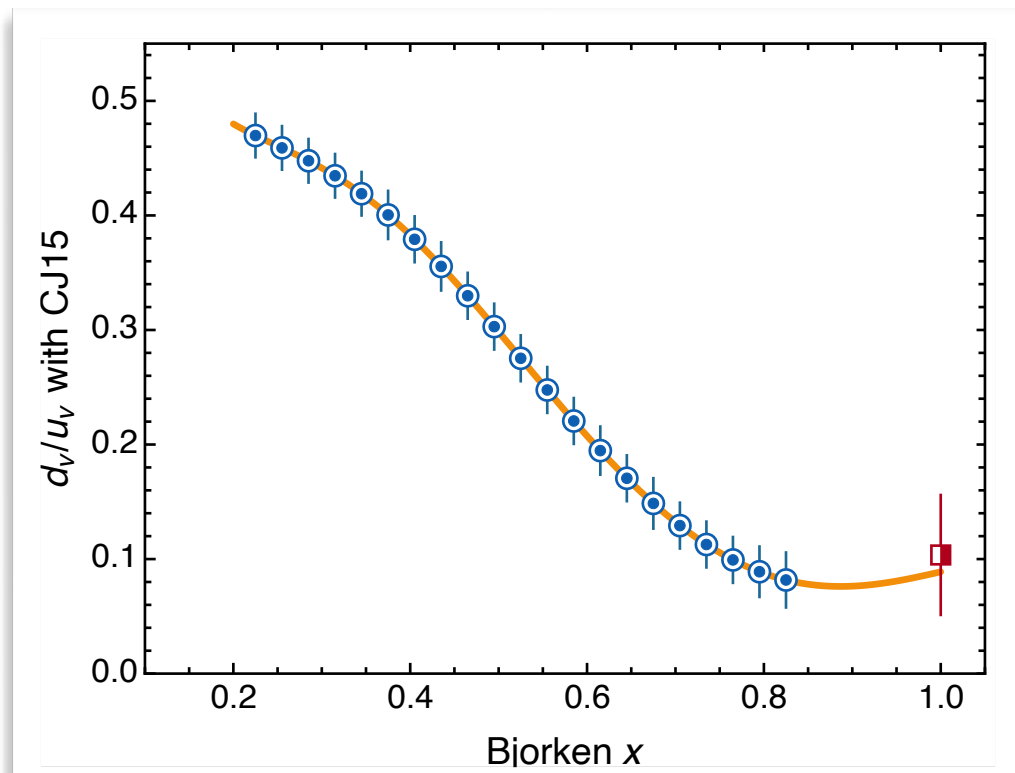
SPM set up

replicas **1,000** ×
interpolators/replica **1,000** ×
 M_j **10** =

total interpolators **10,000,000**

Validation (CJ15)

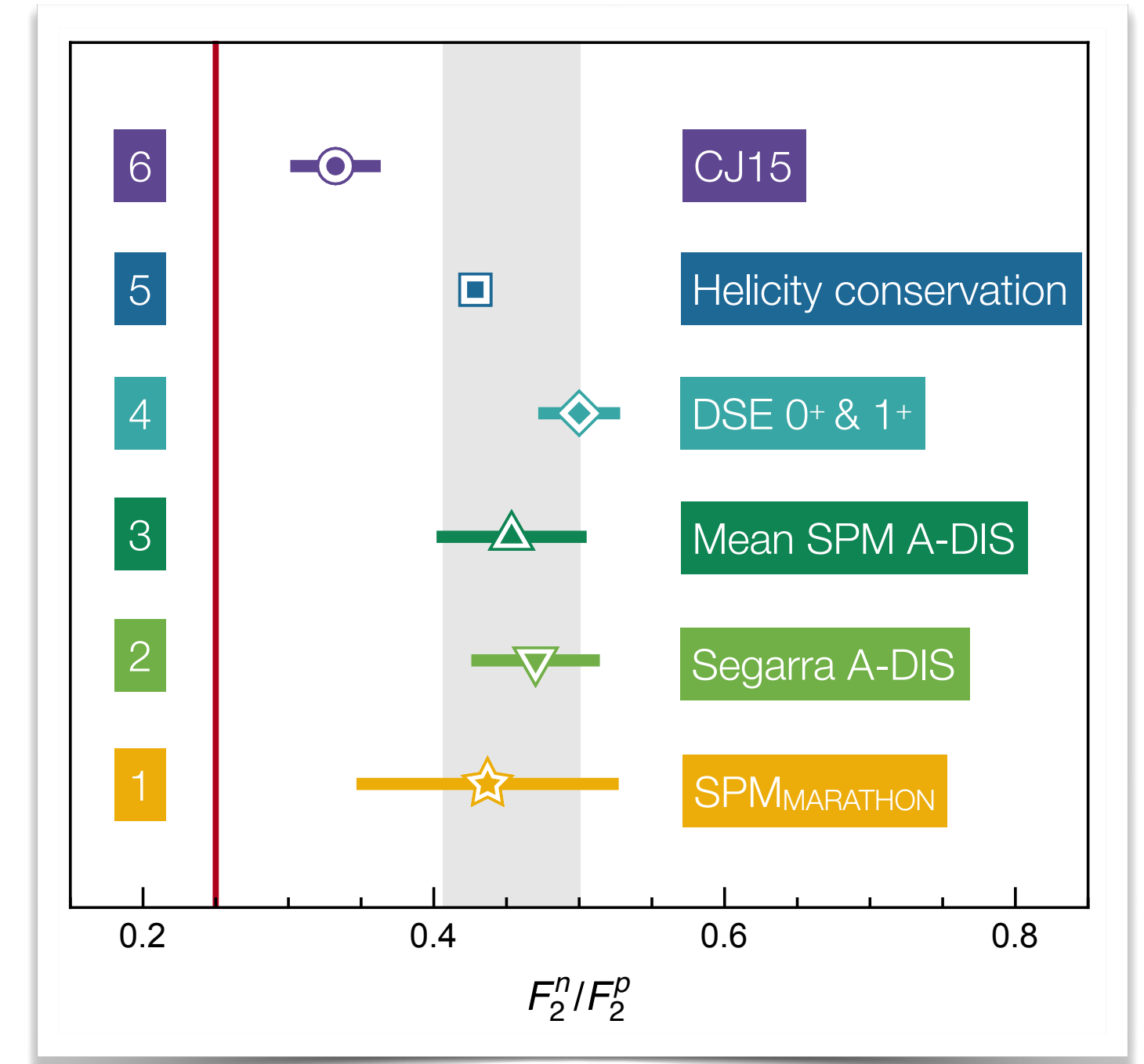
$$\lim_{x \rightarrow 1} d_v^{\text{CJ15}} / u_v^{\text{CJ15}} \stackrel{\text{SPM}}{=} 0.10(5) \quad \text{exp: } 0.09(3)$$



$$\left. \begin{matrix} \sigma_h/\sigma_t \\ F_2^n/F_2^p \\ d_v/u_v \end{matrix} \right\} \stackrel{x \simeq 0}{=} \begin{cases} 1.026 \pm 0.139, \\ 0.962 \pm 0.245, \\ 1.323 \pm 0.706. \end{cases} \quad \text{consistent with sea-quark dominance at low } x$$

$$\left. \begin{matrix} \sigma_h/\sigma_t \\ F_2^n/F_2^p \\ d_v/u_v \end{matrix} \right\} \stackrel{x \simeq 1}{=} \begin{cases} 0.754 \pm 0.052, \\ 0.437 \pm 0.085, \\ 0.227 \pm 0.100. \end{cases} \quad \begin{aligned} \frac{F_2^n(x)}{F_2^p(x)} &= \frac{2\mathcal{R}_{ht} - F_2^h/F_2^t}{2F_2^h/F_2^t - \mathcal{R}_{ht}} \\ \mathcal{R}_{ht}(x=1) &= 1.019(13) \end{aligned}$$

$$\lim_{x \rightarrow 1} \frac{d_v(x)}{u_v(x)} = 0.230 \pm 0.057$$



$$F_2^n/F_2^p \Big|_{x \rightarrow 1}^{\text{SPM \& DIS-A}} = 0.454 \pm 0.047$$

Lower Nachtmann bound excluded at a fairly high level of accuracy: Ψ cannot contain scalar-diquarks only!

Agrees with:

Farrar et al, PRL 35 (1975); Brodsky et al, NPB 441 (1995)

uncorrelated SU(4) spin-flavour proton wave function and helicity conservation in high- Q^2 interactions

Poincaré-covariant Faddeev equation approach giving rise to scalar and vector diquarks correlations with dynamically determined strengths

Segovia et al, FBS 55 (2014); Xu et al, PRD 92 (2015)