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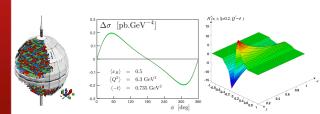


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Prospects on nucleon tomography



Colloquium of the Department of Fundamental Research (NCBJ) \mid Hervé MOUTARDE

universitė

Nov. 18, 2019

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QCD large distance dynamics from the hadron structure viewpoint.



Nucleon Tomography

■ Lattice QCD clearly showed that the mass of hadrons is generated by the **interaction**, not by the quark masses.

Motivation Mass without mass



Phenomenology

Physical content

Analogy

Fit status

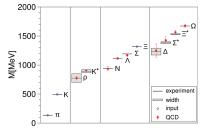
Framework

Design Architecture

Ergonomics Features

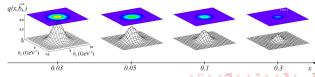
Releases





Durr et al., Science **322**, 1224 (2008)

Can we map the location of mass inside a hadron?





Nucleon structure and its observational consequences.



Nucleon Tomography

Motivation

Mass without mass

Analogy

Phenomenology

Physical content

Fit status

Neural network fits

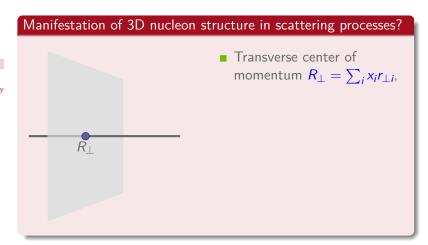
Framework

Design Architecture

Ergonomics

Features

Releases





Nucleon structure and its observational consequences.



Nucleon Tomography

Motivation

Mass without mass

Analogy Phenomenology

Physical content

DVCS data

Fit status

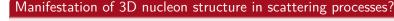
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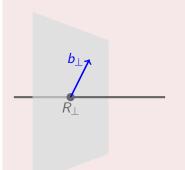
Design Architecture

Ergonomics

Features

Releases





- Transverse center of momentum $R_{\perp} = \sum_{i} x_{i} r_{\perp i}$,
- Impact parameter b_{\perp} ,



Nucleon structure and its observational consequences.



Nucleon Tomography

Motivation

Mass without mass

Analogy

Phenomenology

Physical content

Fit status

Neural network fits

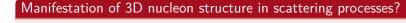
Framework

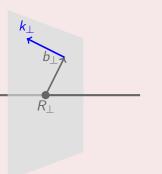
Design Architecture

Freonomics

Features

Releases





- Transverse center of momentum $R_{\perp} = \sum_{i} x_{i} r_{\perp i}$,
- Impact parameter b_{\perp} ,
- Transverse momentum k_{\perp} ,



Nucleon structure and its observational consequences.



Nucleon Tomography

Motivation

Mass without mass

Phenomenology

Physical content

Analogy

Fit status

Neural network fits

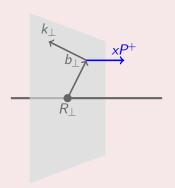
Framework

Design Architecture

Ergonomics

Features Releases





- Transverse center of momentum $R_{\perp} = \sum_{i} x_{i} r_{\perp i}$,
- Impact parameter b_{\perp} ,
- lacksquare Transverse momentum k_{\perp} ,
- Longitudinal momentum xP^+



Nucleon structure and its observational consequences.



Nucleon Tomography

Motivation

Mass without mass

Analogy

Phenomenology

Physical content

Fit status

Neural network fits

Framework

Design

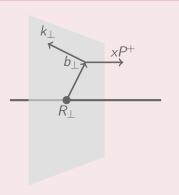
Architecture

Features

Releases

Conclusion

Manifestation of 3D nucleon structure in scattering processes?



- Transverse center of momentum $R_{\perp} = \sum_{i} x_{i} r_{\perp i}$,
- Impact parameter b_{\perp} ,
- lacksquare Transverse momentum k_{\perp} ,
- Longitudinal momentum xP^+ .
- What is the distorsion brought by spin?



Imaging the origin of mass. Identification of underlying mechanisms from parton distributions.



Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content DVCS data

Fit status Neural network fits

Framework

Design

Architecture Ergonomics

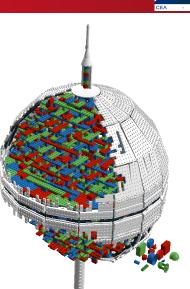
Features

Releases

Conclusion

How can we recover the wellknown characterics of the nucleon from the properties of its colored building blocks?

> Mass? Spin? Charge?





Imaging the origin of mass.

Identification of underlying mechanisms from parton distributions.



Nucleon Tomography

Motivation

Mass without mass

Nucleon structure

Analogy

Phenomenology

Physical content

DVCS data

Fit status

Neural network fits

Framework

Design Architecture

Ergonomics

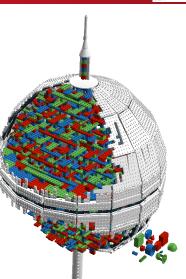
Features Releases

Conclusion

How can we recover the well-known characterics of the nucleon from the properties of its **colored building blocks**?

Mass? Spin? Charge?

What are the relevant effective degrees of freedom and effective interaction at large distance?



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Imaging the origin of mass. Identification of underlying mechanisms from parton distributions.

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Nucleon Tomography

Motivation

Mass without mass

Nucleon structure
Analogy

Phenomenology

Physical content

Fit status

Framework

Design Architecture Ergonomics

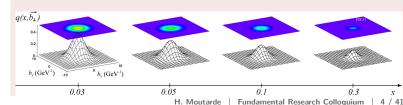
Features

Releases

Conclusion

Structuring questions for the hadron physics community

- QCD mechanisms behind the origin of mass in the visible universe?
- Cartography of interactions giving its mass to the nucleon?
- Pressure and density profiles of the nucleon as a continuous medium?
- Localization of quarks and gluons inside the nucleon?
- Possible impact on **initial state** of pp or pA scattering?







Nucleon Tomography

Motivation

Mass without mass Nucleon structure

Analogy

Phenomenology

Physical content DVCS data

Fit status

Framework

Design

Architecture Ergonomics

Features

Releases

Conclusion

Colorless proton



Colorful components









Nucleon Tomography

Motivation

Mass without mass
Nucleon structure
Analogy

Phenomenology

Physical content

DVCS data

Neural network fits

Framework

Architecture

Ergonomics

Features Releases

Conclusion

"Can one hear the shape of a drum?"



Kac, Am. Math. Mon. **74**, 1 (1966)





Nucleon Tomography

Motivation

Mass without mass
Nucleon structure
Analogy

Phenomenology

Physical content DVCS data

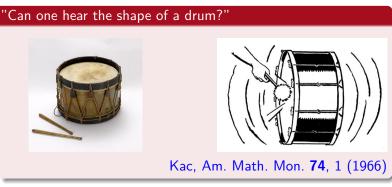
Fit status

Framework

Design Architecture

Freonomics

Features Releases







Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content
DVCS data
Fit status
Neural network fits

Framework

Design Architecture

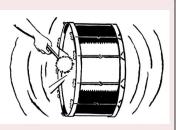
Ergonomics Features

Releases

Conclusion

"Can one hear the shape of a drum?"





Kac, Am. Math. Mon. **74**, 1 (1966)

In quantitative terms



■ Dirichlet problem for the Laplacian:

$$\Delta u + \lambda u = 0$$
 and $u_{|\partial\Omega} = 0$



Can one hear the shape of a proton? Harmonics and patterns.



Nucleon Tomography

Motivation

Mass without mass Nucleon structure

Analogy

Phenomenology

Physical content DVCS data

Fit status

Neural network fits

Framework

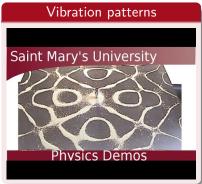
Design Architecture

Ergonomics

Features Releases

Conclusion

Vibration patterns



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Can one hear the shape of a proton? Harmonics and patterns.



Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content

Fit status

Neural network fits

Framework

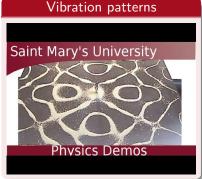
Design Architecture Ergonomics

Features

Releases

Conclusion

Vibration patterns



What about the proton?

- "Hit" the proton, e.g. with a virtual photon:
- "Listen" to the distribution of produced particles:
- "Measure" harmonics:



Can one hear the shape of a proton? Harmonics and patterns.



Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology Physical content

DVCS data Fit status

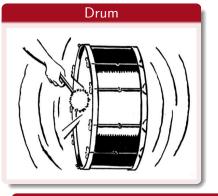
Neural network fits

Pramework
Design
Architecture

Ergonomics Features

Releases

Conclusion





What about the proton?

- "Hit" the proton, e.g. with a virtual photon: hard
- "Listen" to the distribution of produced particles: exclusive
- "Measure" harmonics: amplitudes (Compton form factors)





Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content
DVCS data
Fit status

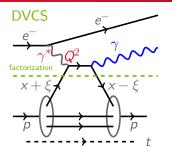
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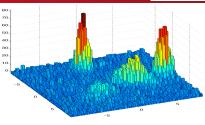
Framework

Design Architecture

Ergonomics

Features Releases









Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content

Fit status

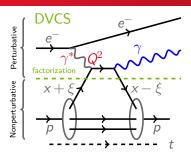
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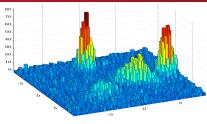
Framework

Design Architecture

Ergonomics

Features Releases









Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content

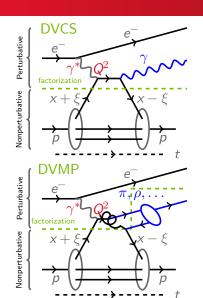
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Neural network fits

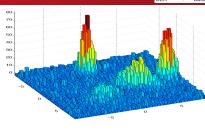
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Design Architecture

Ergonomics Features

Features Releases











Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content DVCS data

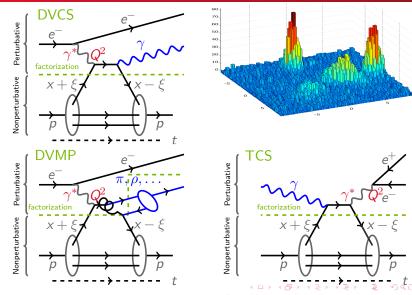
Fit status
Neural network fits

Neural network

Framework

Architecture Ergonomics

Features Releases







Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content DVCS data

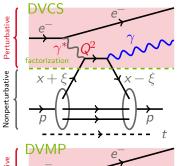
Fit status

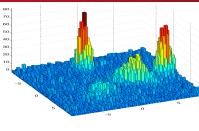
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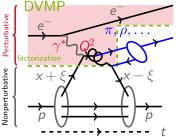
Framework

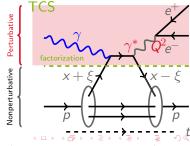
Design Architecture Ergonomics

Features Releases













Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content
DVCS data
Fit status

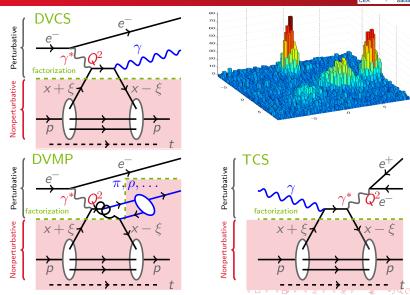
Neural network fits

Framework

Design Architecture

Ergonomics Features

Releases





Exclusive processes amplitudes. The sound of DVCS: Compton Form Factors (CFFs).



Nucleon Tomography

Motivation Mass without mass Nucleon structure Analogy

Nucleon structure Analogy Phenomenology Physical content

DVCS data

Neural network fits

Framework

Architecture Ergonomics Features Releases

Conclusion

Bjorken regime : large Q^2 and fixed $xB \simeq 2\xi/(1+\xi)$

- Partonic interpretation relies on **factorization theorems**.
- All-order proofs for DVCS, TCS and some DVMP.
- GPDs depend on a (arbitrary) factorization scale μ_F .
- **Consistency** requires the study of **different channels**.
- GPDs enter DVCS through **Compton Form Factors** :

$$\mathcal{F}(\xi, t, Q^2) = \int_{-1}^1 dx \, C\left(x, \xi, \alpha_S(\mu_F), \frac{Q}{\mu_F}\right) F(x, \xi, t, \mu_F)$$

for a given GPD F.

 \blacksquare CFF $\mathcal F$ is a **complex function**.

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Phenomenology





Nucleon Tomography

- Correlation of the longitudinal momentum and the transverse position of a parton in a hadron.
- DVCS recognized as the cleanest channel to access GPDs.

Motivation Mass without mass

Nucleon structure
Analogy

Phenomenology

Physical content

DVCS data

Fit status

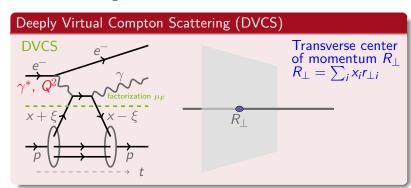
Neural network fits

Framework

Design

Architecture Ergonomics

Features Releases







Nucleon Tomography

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Motivation Mass without mass

Mass without mas Nucleon structure Analogy

Phenomenology

Physical content

DVCS data

Fit status

Neural network fits

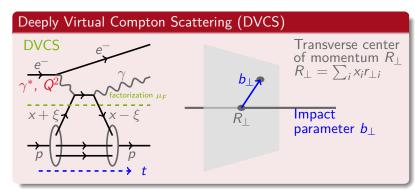
Framework

Design Architecture

Ergonomics

Features

Releases







Nucleon Tomography

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Motivation Mass without mass

Nucleon structure Analogy

Phenomenology

Physical content

DVCS data

Fit status

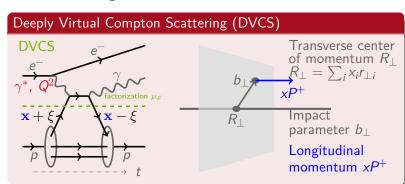
Neural network fits

Framework

Design

Architecture Ergonomics

Features Releases







10 / 41

Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content

DVCS data

Fit status

Neural network fits

Framework

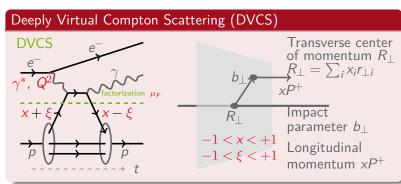
Design Architecture

Ergonomics Features

Releases

Conclusion

- Correlation of the longitudinal momentum and the transverse position of a parton in a hadron.
- DVCS recognized as the cleanest channel to access GPDs.



24 GPDs $F^i(x, \xi, t, \mu_F)$ for each parton type i = g, u, d, ... for leading and sub-leading twists.





Nucleon Tomography

■ Probabilistic interpretation of Fourier transform of $GPD(x, \xi = 0, t)$ in transverse plane.

$$\rho(\mathbf{x},b_{\perp},\lambda,\lambda_{N}) = \frac{1}{2} \left[\mathbf{H}(\mathbf{x},0,b_{\perp}^{2}) + \frac{b_{\perp}^{j} \epsilon_{ji} S_{\perp}^{i}}{M} \frac{\partial \mathbf{E}}{\partial b_{\perp}^{2}} (\mathbf{x},0,b_{\perp}^{2}) \right]$$
Analogy

$$+\lambda\lambda_N\tilde{\boldsymbol{H}}(\mathbf{x},0,b_\perp^2)$$

Phenomenology

Physical content

Fit status

Framework

Design Architecture

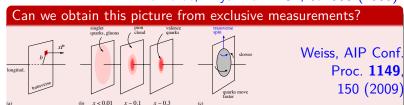
Ergonomics Features

Releases

Conclusion

Notations: quark helicity λ , nucleon longitudinal polarization λ_N and nucleon transverse spin S_{\perp} .

Burkardt, Phys. Rev. **D62**, 071503 (2000)







Nucleon Tomography

■ **Probabilistic interpretation** of Fourier transform of $GPD(x, \xi = 0, t)$ in **transverse plane**.

Motivation Mass without mass

$$\rho(\mathbf{x}, b_{\perp}, \lambda, \lambda_{N}) = \frac{1}{2} \left[\mathbf{H}(\mathbf{x}, 0, b_{\perp}^{2}) + \frac{b_{\perp}^{j} \epsilon_{ji} S_{\perp}^{i}}{M} \frac{\partial \mathbf{E}}{\partial b_{\perp}^{2}} (\mathbf{x}, 0, b_{\perp}^{2}) \right]$$

Nucleon structure Analogy

$$+\lambda\lambda_{N}\tilde{\boldsymbol{H}}(x,0,b_{\perp}^{2})$$

Phenomenology

In the status λ is a Notations λ in the limit λ in the local polynomial Notation λ is a Notation λ in the local network fits λ in the local network λ in the

Physical content

Framework

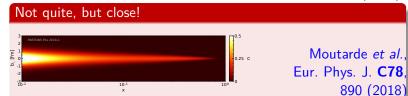
polarization λ_N and nucleon transverse spin S_{\perp} . Burkardt, Phys. Rev. **D62**, 071503 (2000)

Design

Architecture Ergonomics

Features Releases

Conclusion



H. Moutarde



Nucleon

Anatomy of hadrons. GPDs, 3D hadron imaging, and beyond (3/4).



Tomography

Motivation

Nucleon structure Analogy

Phenomenology Physical content

DVCS data Fit status Neural network fits

Framework Design

Architecture Ergonomics Features

Releases Conclusion

 Most general structure of matrix element of energy momentum tensor between nucleon states:

$$\left\langle N, P + \frac{\Delta}{2} \middle| T^{\mu\nu} \middle| N, P - \frac{\Delta}{2} \right\rangle = \bar{u} \left(P + \frac{\Delta}{2} \right) \left[A(t) \gamma^{(\mu} P^{\nu)} + B(t) P^{(\mu} i \sigma^{\nu) \lambda} \frac{\Delta_{\lambda}}{2M} + \frac{C(t)}{M} (\Delta^{\mu} \Delta^{\nu} - \Delta^{2} \eta^{\mu\nu}) \right] u \left(P - \frac{\Delta}{2} \right)$$

with $t = \Delta^2$.

Key observation: link between GPDs and gravitational form factors

$$\int dx x \mathbf{H}^{q}(x, \xi, t) = \mathbf{A}^{q}(t) + 4\xi^{2} \mathbf{C}^{q}(t)$$
$$\int dx x \mathbf{E}^{q}(x, \xi, t) = \mathbf{B}^{q}(t) - 4\xi^{2} \mathbf{C}^{q}(t)$$





Nucleon

Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content

DVCS data Fit status Neural network fits

Framework Design

Architecture Ergonomics Features

Releases Conclusion

Spin sum rule:

 $\int dx x (H^{q}(x,\xi,0) + E^{q}(x,\xi,0)) = A^{q}(0) + B^{q}(0) = 2J^{q}$ Ji, Phys. Rev. Lett. **78**, 610 (1997)

Shear and **pressure** of a hadron considered as a continuous medium:

$$\langle N | T^{ij}(\vec{r}) | N \rangle N = s(r) \left(\frac{r^{i}r^{j}}{\vec{r}^{2}} - \frac{1}{3}\delta^{ij} \right) + p(r)\delta^{ij}$$

Polyakov and Shuvaev, hep-ph/0207153

Energy density, tangential and radial pressures of a hadron considered as a continuous medium. Lorcé et al., Eur. Phys. J. C79, 89 (2019)



Spin-0 Generalized Parton Distribution.



Nucleon **Tomography**

Phenomenology

Physical content

DVCS data Fit status

Neural network fits

Framework Design

Architecture Ergonomics Features

Releases

Conclusion

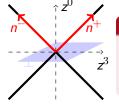
Definition and simple properties.



 $H_{\pi}^{q}(x,\xi,t) =$

$$\frac{1}{2} \int \frac{\mathrm{d}z^{-}}{2\pi} e^{ixP^{+}z^{-}} \left\langle \pi, P + \frac{\Delta}{2} \middle| \bar{q} \left(-\frac{z}{2} \right) \gamma^{+} q \left(\frac{z}{2} \right) \middle| \pi, P - \frac{\Delta}{2} \right\rangle_{\substack{z^{+}=0\\z_{\perp}=0}}$$

with $t = \Delta^2$ and $\xi = -\Delta^+/(2P^+)$.



References

Müller et al., Fortschr. Phys. 42, 101 (1994) Ji, Phys. Rev. Lett. 78, 610 (1997)

Radyushkin, Phys. Lett. **B380**, 417 (1996)

$$H^{q}(x,0,0) = q(x)$$

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Nucleon Tomography

Spin-0 Generalized Parton Distribution. Definition and simple properties.

References



Motivation

Mass without mass Nucleon structure

Analogy

Phenomenology

Physical content

DVCS data Fit status

Neural network fits Framework

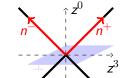
Design Architecture

Ergonomics Features



$$\frac{1}{2} \int \frac{\mathrm{d}z^{-}}{2\pi} \, \mathrm{e}^{\mathrm{i}xP^{+}z^{-}} \left\langle \pi, P + \frac{\Delta}{2} \middle| \bar{q} \left(-\frac{z}{2} \right) \gamma^{+} q \left(\frac{z}{2} \right) \middle| \pi, P - \frac{\Delta}{2} \right\rangle_{\substack{z^{+}=0\\z_{\perp}=0}}$$

 $H_{\pi}^{q}(x,\xi,t) =$



Conclusion

Releases

PDF forward limit

Form factor sum rule

Form factor **sum rul**
$$r+1$$

 $dx H^{q}(x, \xi, t) = F_1^{q}(t)$ H. Moutarde | Fundamental Research Colloquium

Müller et al., Fortschr. Phys. 42, 101 (1994)

Radyushkin, Phys. Lett. **B380**, 417 (1996)

Ji, Phys. Rev. Lett. **78**, 610 (1997)



Spin-0 Generalized Parton Distribution. Definition and simple properties.



Tomography

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content

DVCS data Fit status

Neural network fits

Framework

Design Architecture Ergonomics

Features Releases

Conclusion

$H_{\pi}^{q}(x,\xi,t) =$ Nucleon $\frac{1}{2} \int \frac{\mathrm{d}z^{-}}{2\pi} e^{ixP^{+}z^{-}} \left\langle \pi, P + \frac{\Delta}{2} \middle| \bar{q} \left(-\frac{z}{2} \right) \gamma^{+} q \left(\frac{z}{2} \right) \middle| \pi, P - \frac{\Delta}{2} \right\rangle_{z^{+}=0}$

with
$$t=\Delta^2$$
 and $\xi=-\Delta^+/(2P^+)$. References

References

Müller et al., Fortschr. Phys. 42, 101 (1994)

Ji, Phys. Rev. Lett. 78, 610 (1997) Radyushkin, Phys. Lett. **B380**, 417 (1996)

- PDF forward limit
- Form factor sum rule
- \blacksquare H^q is an **even function** of ξ from time-reversal invariance.



Nucleon Tomography

Spin-0 Generalized Parton Distribution. Definition and simple properties.



Mass without mass Nucleon structure

Analogy

Phenomenology

Physical content

DVCS data

Fit status Neural network fits

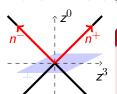
Framework Design

Architecture

Ergonomics Features Releases

Conclusion

$\frac{1}{2} \int \frac{\mathrm{d}z^{-}}{2\pi} e^{ixP^{+}z^{-}} \left\langle \pi, P + \frac{\Delta}{2} \middle| \bar{q} \left(-\frac{z}{2} \right) \gamma^{+} q \left(\frac{z}{2} \right) \middle| \pi, P - \frac{\Delta}{2} \right\rangle_{z^{+}=0}$ Motivation



 $H_{\pi}^{q}(x,\xi,t) =$



$$\left\langle \pi, P + \frac{\Delta}{2} \right\rangle$$

with $t = \Delta^2$ and $\xi = -\Delta^+/(2P^+)$.



References





Müller et al., Fortschr. Phys. 42, 101 (1994)

Radyushkin, Phys. Lett. **B380**, 417 (1996)

- Ji, Phys. Rev. Lett. 78, 610 (1997)
- PDF forward limit
- Form factor sum rule
- H^q is an **even function** of ξ from time-reversal invariance. H^q is real from hermiticity and time-reversal invariance.





Nucleon Tomography

Polynomiality

Motivation

Analogy

Mass without mass Nucleon structure

Phenomenology

Physical content

DVCS data

Fit status

Neural network fits

Framework

Design

Architecture Ergonomics

Features

Releases

$$\int_{-1}^{+1} dx x^n H^q(x, \xi, t) = \text{polynomial in } \xi$$





Nucleon Tomography

Polynomiality

Lorentz covariance

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content

DVCS data

Fit status

Neural network fits

Framework

Design

Architecture Ergonomics

Features

Releases





Nucleon **Tomography**

Polynomiality

Lorentz covariance

Motivation

Analogy

Positivity Mass without mass

Nucleon structure Phenomenology

Physical content

DVCS data

Fit status Neural network fits

Framework

Design

Architecture Ergonomics

Features Releases

$$H^{q}(x,\xi,t) \leq \sqrt{q\left(\frac{x+\xi}{1+\xi}\right)q\left(\frac{x-\xi}{1-\xi}\right)}$$





Nucleon **Tomography**

Polynomiality

Lorentz covariance

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content DVCS data

Fit status Neural network fits

Framework

Design Architecture

Ergonomics

Features

Releases Conclusion

Positivity

Positivity of Hilbert space norm





Nucleon **Tomography**

Polynomiality

Positivity

Lorentz covariance

Motivation Mass without mass

Nucleon structure Analogy

Phenomenology

DVCS data Fit status

Physical content

Neural network fits

Framework Design

Architecture Ergonomics Features

Releases Conclusion

Positivity of Hilbert space norm

 H^q has support $x \in [-1, +1]$.

15 / 41





Nucleon Tomography

Polynomiality

Lorentz covariance

Motivation Mass without mass

Analogy

Mass without mass
Nucleon structure

Positivity

Positivity of Hilbert space norm

Phenomenology Physical content

■ H^q has support $x \in [-1, +1]$.

DVCS data Fit status Neural network fits

Relativistic quantum mechanics

Framework

Design Architecture Ergonomics Features

Releases





Nucleon Tomography

Polynomiality

Lorentz covariance

Motivation

Mass without mass

Positivity

Positivity of Hilbert space norm

Phenomenology Physical content

Nucleon structure Analogy

■ H^q has support $x \in [-1, +1]$.

Relativistic quantum mechanics

DVCS data

Fit status Neural network fits **Soft pion theorem** (pion target)

Framework

Soft pion theorem (pion target

Design Architecture

$$H^{q}(x, \xi = 1, t = 0) = \frac{1}{2}\phi_{\pi}^{q}\left(\frac{1+x}{2}\right)$$

Ergonomics Features Releases

Conclusion

Fundamental Research Colloquium





Nucleon Tomography

Polynomiality

Lorentz covariance

Motivation

Mass without mass

Analogy

Positivity

Positivity of Hilbert space norm

Phenomenology Physical content

■ H^q has support $x \in [-1, +1]$.

Relativistic quantum mechanics

DVCS data

Fit status Neural network fits

■ **Soft pion theorem** (pion target)

Framework

Soft pion theorem (pion tal

Architecture Ergonomics Features Dynamical chiral symmetry breaking

Releases





Nucleon Tomography

Polynomiality

Lorentz covariance

Motivation

Mass without mass

Nucleon structure

Analogy

Positivity

Positivity of Hilbert space norm

Phenomenology Physical content

DVCS data
Fit status

■ H^q has support $x \in [-1, +1]$.

Relativistic quantum mechanics

Framework

Design

Architecture

Features
Releases
Conclusion

Soft pion theorem (pion target)

Dynamical chiral symmetry breaking

Implementing these theoretical constraints is non-trivial!

- Fulfilling polynomiality and positivity: solved in 2017.
- Reduction to form factors or PDFs underway.



Mechanical properties of hadrons. From the nucleon to compact stars (1/3).



Nucleon **Tomography**

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content DVCS data

Fit status Neural network fits

Framework Design

Architecture Ergonomics

Features Releases

Conclusion

■ Matrix element in the Breit frame (a = q, g):

$$\left\langle \frac{\Delta}{2} \left| \mathcal{T}_{\mathsf{a}}^{\mu\nu}(0) \right| - \frac{\Delta}{2} \right\rangle = M \left\{ \eta^{\mu 0} \eta^{\nu 0} \left[A_{\mathsf{a}}(t) + \frac{t}{4 M^2} B_{\mathsf{a}}(t) \right] \right\}$$

$$+\eta^{\mu\nu}\left[\bar{C}_{a}(t)-\frac{t}{M^{2}}C_{a}(t)\right]+\frac{\Delta^{\mu}\Delta^{\nu}}{M^{2}}C_{a}(t)\right\}$$

Anisotropic fluid in relativistic hydrodynamics:

$$\Theta^{\mu\nu}(\vec{r}) = \left[\varepsilon(r) + p_t(r)\right] u^{\mu} u^{\nu} - p_t(r) \eta^{\mu\nu} + \left[p_r(r) - p_t(r)\right] \chi^{\mu} \chi^{\nu}$$

where u^{μ} and $\chi^{\mu} = x^{\mu}/r$.

Define **isotropic pressure** and **pressure anisotropy**:

$$p(r) = \frac{p_r(r) + 2 p_t(r)}{3}$$

$$s(r) = p_r(r) - p_t(r)$$



4 D > 4 A >



Mechanical properties of hadrons. From the nucleon to compact stars (2/3).



17 / 41

Nucleon Tomography

Write dictionary between quantum and fluid pictures:

Motivation
$$\varepsilon_a(r)$$

$$\int d^3 \vec{\Delta} = i \vec{\Delta} \cdot \vec{r} \left(A(r) + \vec{c}(r) + \vec{c}(r) \right)$$

Fit status Neural network fits

Framework Design Architecture Ergonomics

Features Releases Conclusion

 $\frac{p_{a}(r)}{M} = \int \frac{\mathrm{d}^{3} \Delta}{(2\pi)^{3}} e^{-i\vec{\Delta} \cdot \vec{r}} \left\{ -\bar{C}_{a}(t) + \frac{2}{3} \frac{t}{M^{2}} C_{a}(t) \right\}$

 $\frac{s_{a}(r)}{M} = \int \frac{\mathrm{d}^{3}\vec{\Delta}}{(2\pi)^{3}} e^{-i\vec{\Delta} \cdot \vec{r}} \left\{ -\frac{4}{r^{2}} \frac{t^{-1/2}}{M^{2}} \frac{\mathrm{d}^{2}}{\mathrm{d}t^{2}} \left(t^{5/2} C_{a}(t)\right) \right\}$

$\frac{\varepsilon_{\mathsf{a}}(r)}{M} = \int \frac{\mathrm{d}^3 \Delta}{(2\pi)^3} \, e^{-i\vec{\Delta} \cdot \vec{r}} \left\{ A_{\mathsf{a}}(t) + \bar{C}_{\mathsf{a}}(t) + \frac{t}{4M^2} \left[B_{\mathsf{a}}(t) - 4C_{\mathsf{a}}(t) \right] \right\}$



 $\frac{p_{t,a}(r)}{M} = \int \frac{\mathrm{d}^3 \vec{\Delta}}{(2\pi)^3} e^{-i\vec{\Delta} \cdot \vec{r}} \left\{ -\bar{C}_a(t) + \frac{4}{r^2} \frac{t^{-1/2}}{M^2} \frac{\mathrm{d}}{\mathrm{d}t} \left[t \frac{\mathrm{d}}{\mathrm{d}t} \left(t^{3/2} C_a(t) \right) \right] \right\}$

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Mechanical properties of hadrons. From the nucleon to compact stars (3/3).





■ Evaluate orders of magnitude with naive multiple model:

Motivation Mass without mass

Nucleon structure Analogy

Phenomenology

Physical content

DVCS data Fit status

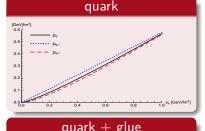
Neural network fits

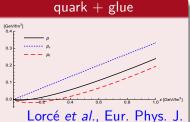
Framework

Design Architecture Ergonomics

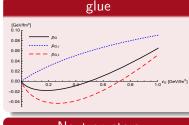
Features Releases

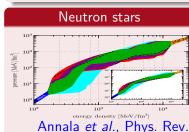
Conclusion





C79, 89 (2019)





Lett. **120**, 172703 (2018)





Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content

DVCS data

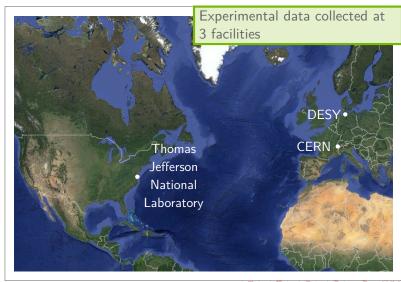
Neural network fits

Framework

Design Architecture

Ergonomics

Features Releases







Nucleon Tomography

Motivation

Mass without mass

Nucleon structure

Phenomenology

Physical content

Analogy

Fit status

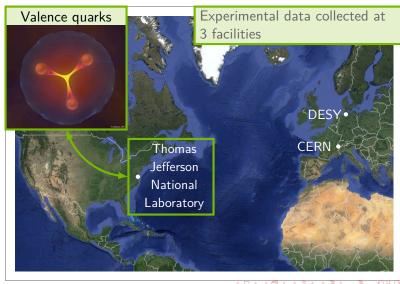
Neural network fits

Framework

Architecture

Ergonomics

Features Releases







Nucleon Tomography

Motivation Mass without mass

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content

DVCS data

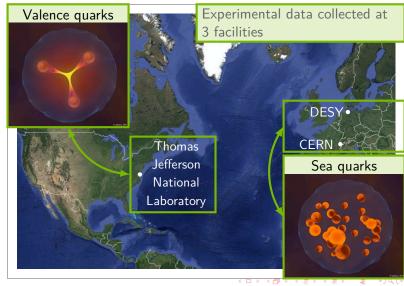
Neural network fits

Framework

Design Architecture

Ergonomics

Features Releases







Nucleon Tomography

Motivation Mass without mass

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content

DVCS data

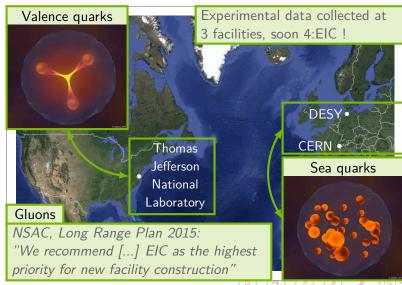
Fit status Neural network fits

Framework

Design Architecture

Ergonomics

Features Releases







Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content

DVCS data Fit status

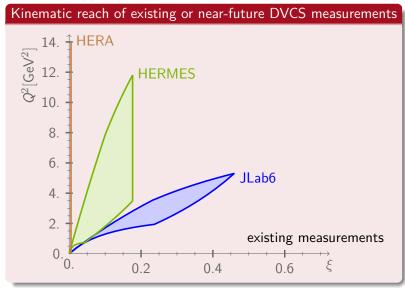
Neural network fits

Framework

Design Architecture

Ergonomics Features

Releases







Nucleon Tomography

Motivation Mass without mass Nucleon structure

Analogy

Phenomenology Physical content

DVCS data

Fit status

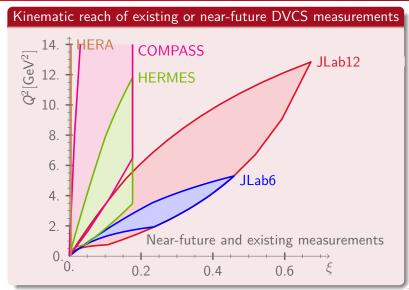
Framework

Design

Architecture

Features

Releases







Nucleon Tomography

Motivation Mass without mass Nucleon structure

Phenomenology

Physical content

DVCS data

Analogy

Neural network fits

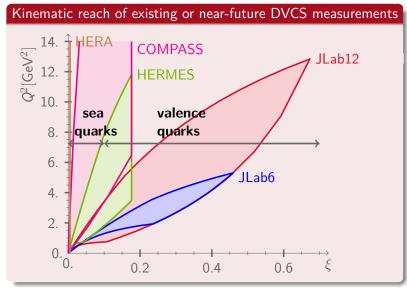
Framework

Design

Architecture

Features

Releases







Nucleon **Tomography**

Motivation Mass without mass Nucleon structure

Phenomenology

Analogy Physical content

DVCS data

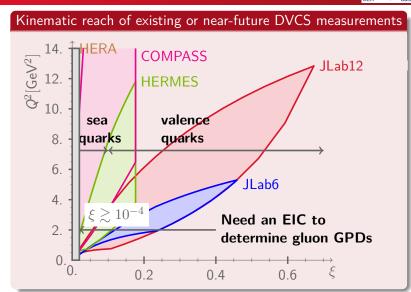
Fit status

Neural network fits

Framework Design

Architecture Ergonomics

Features Releases







Nucleon Tomography

Motivation Mass without mass Nucleon structure

Phenomenology

Physical content

DVCS data

Analogy

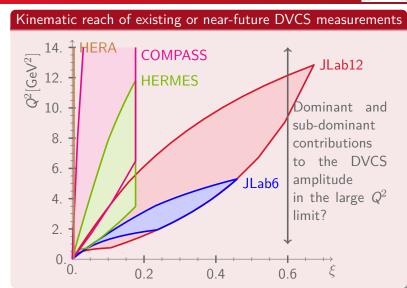
Fit status

iveural network

Framework Design

Architecture Ergonomics

Features Releases





DVCS analysis and fits. No global GPD fit has been obtained so far.



Nucleon Tomography

Motivation Mass without mass Nucleon structure Analogy

Phenomenology

Physical content

DVCS data Fit status

i it status

Neural network fits

Framework

Architecture

Features Releases

Conclusion

■ GPD fits **only in the small** x_B **region** with a **flexible** parameterization (kinematic simplifications).

- Global fits of CFFs in the sea, valence and glue regions.
- Some GPD models with non-flexible parameterizations adjusted to experimental DVCS or DVMP data.

Kumerički et al., Eur. Phys. J. A52, 157 (2016)

The situation is being improved!

■ GPD parameterizations satisfying *a priori* all theoretical constraints on GPDs.

Chouika et al., Eur. Phys. J. C77, 906 (2017)

Computing framework to go beyond leading order and leading twist analysis.

Berthou et al., Eur. Phys. J. C78, 478 (2018)



Neural network global fit of CFFs. All existing sets except $d^4\sigma_{\text{TIII}}^-$ from Hall A (2015-17).



Nucleon	No.	Collab.	Collab. Year Ref. Observable		ble	Kinematic dependence	No. of points used / all	
Tomography	1		A_{LU}^+		φ	10 / 10		
	2		2006	41	$A_C^{\cos i\phi}$	i = 1	t	4/4
	3		2008	42	$A_C^{\cos i\phi}$ $A_C^{\sin(\phi-\phi_S)\cos i\phi}$	i=0,1	x_{Bj}	18 / 24
Motivation					$A_{UT,DVCS}^{\sin(\phi-\phi_S)\cos i\phi}$ $A_{UT,I}^{\sin(\phi-\phi_S)\cos i\phi}$	i = 0 i = 0, 1		
Mass without mass					$A_{UT,I}^{\cos(\phi-\phi_S)\sin i\phi}$	i = 1		
Nucleon structure	4		2009	43	$A_{LU,I}^{\sin i\phi}$	i=1,2	x_{Bj}	35 / 42
Analogy					$A_{LU,DVCS}^{\sin i\phi}$	i = 1		
Phenomenology	5		2010	44	$A_C^{\cos i\phi}$ $A_C^{+,\sin i\phi}$ A_U^{L}	i = 0, 1, 2, 3 i = 1, 2, 3	$x_{ m Bj}$	18 / 24
Physical content	6		2011	45	$A_{LL}^{+,\cos i\phi}$ $A^{\cos(\phi-\phi_S)\cos i\phi}$	i = 0, 1, 2 i = 0, 1	m-	24 / 32
DVCS data	0		2011	40	$A_{LT,DVCS}^{\cos(\phi-\phi_S)\cos i\phi}$ $A_{LT,DVCS}^{\sin(\phi-\phi_S)\sin i\phi}$	i = 0, 1 i = 1	x_{Bj}	24 / 32
Fit status					$A_{I,T,I}^{\cos(\phi-\phi_S)\cos i\phi}$	i = 0, 1, 2		
Neural network fits					$A_{LT,I}^{\sin(\phi-\phi_S)\sin i\phi}$	i=1,2		
Framework	7		2012	46	$A_{LU,\mathrm{DVCS}}^{\sin i\phi}$ $A_{LU,\mathrm{DVCS}}^{\sin i\phi}$ $A_{C}^{\cos i\phi}$	i = 1, 2 i = 1 i = 0, 1, 2, 3	x_{Bj}	35 / 42
Design	8	CLAS	2001	47	$A_{LU}^{-,\sin i\phi}$	i = 0, 1, 2, 3 i = 1, 2	_	0 / 2
Architecture	9		2006	48	A_{UL}^{LU} , sin $i\phi$	i = 1, 2	_	2 / 2
Ergonomics	10		2008	49	A_{LU}^-		ϕ	283 / 737
Features	11		2009	50	A_{LU}^-		φ	22 / 33
Releases	12 13		2015 2015	$\frac{51}{52}$	$A_{LU}^{-}, A_{UL}^{-}, A_{LL}^{-}$ $d^{4}\sigma_{UU}^{-}$		ϕ ϕ	311 / 497 1333 / 1933
	14	Hall A	2015	34	$\Delta d^4 \sigma_{LU}^-$		φ	228 / 228
Conclusion	15		2017	35	$\Delta d^4 \sigma_{LU}^-$		ϕ	276 / 358
	16	COMPASS	2018	36	$d^3\sigma_{UU}^{\pm}$		t	2 / 4
	17	ZEUS	2009	37	$d^3\sigma_{UU}^+$		t	4 / 4
	18 19	H1	2005 2009	38 39	$d^3\sigma_{UU}^+$		t	7 / 8 12 / 12
	19		2009	39	$d^3\sigma_{UU}^{\pm}$		t	12 / 12

Moutarde *et al.*, Eur. Phys. J. **C79**, 614 (2019)



A selection of results.

2600+ measurements of 30 observables published during 2001-17.



Nucleon **Tomography**

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

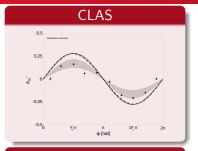
Physical content DVCS data Fit status

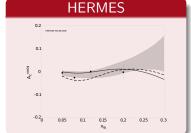
Neural network fits

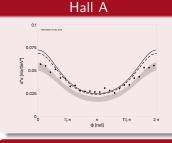
Framework Design

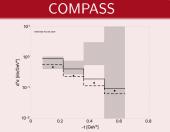
Architecture Ergonomics

Features Releases











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Nucleon **Tomography**

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content DVCS data

Fit status

Neural network fits

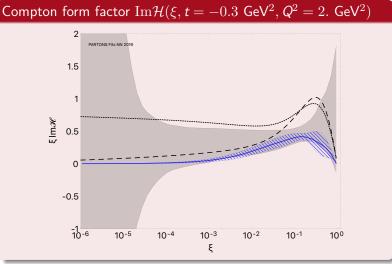
Framework

Design Architecture

Ergonomics

Features Releases

Conclusion



Moutarde et al., Eur. Phys. J. C79, 614 (2019)



A selection of results.

2600+ measurements of 30 observables published during 2001-17.



Nucleon Tomography

Motivation

Mass without mass Nucleon structure

Phenomenology

Physical content

Fit status

Analogy

Neural network fits

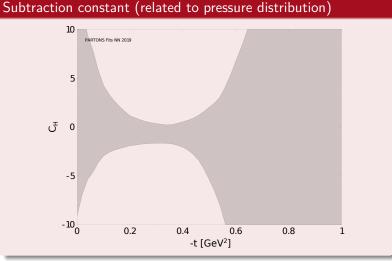
Framework

Design Architecture

Ergonomics Features

Releases

Conclusion



Moutarde et al., Eur. Phys. J. C79, 614 (2019)



Software for the phenomenology of GPDs. Different questions to be answered with the same tools.



Nucleon **Tomography**

Motivation Mass without mass Nucleon structure

Phenomenology

Physical content DVCS data

Fit status

Analogy

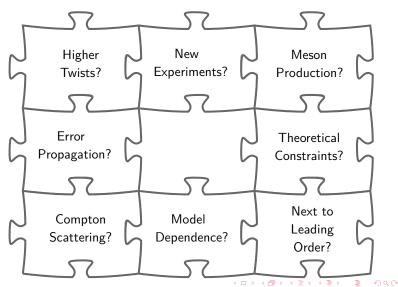
Neural network fits

Framework

Design Architecture

Ergonomics Features

Releases





Software for the phenomenology of GPDs. Different questions to be answered with the same tools.



Nucleon **Tomography**

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content DVCS data

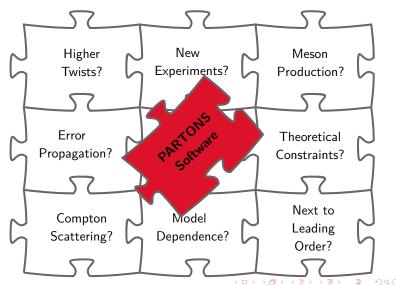
Fit status

Neural network fits Framework

Design

Architecture Ergonomics

Features Releases



The PARTONS framework



PARtonic Tomography Of Nucleon Software





Nucleon Tomography

Motivation

Mass without mass

Nucleon structure

Phenomenology

Analogy

Physical content
DVCS data
Fit status

Neural network fits

Framework

Design Architecture

Ergonomics Features Releases

Conclusion

Experimental data and phenomenology

Computation of amplitudes

First principles and fundamental parameters

Full processes

Small distance contributions

Large distance contributions

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Nucleon Tomography

Motivation

Mass without mass

Nucleon structure

Phenomenology

Analogy

Physical content
DVCS data
Fit status
Neural network fits

Framework

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Design Architecture

Ergonomics Features Releases

Conclusion

Experimental data and phenomenology

Computation of amplitudes

First principles and fundamental parameters

Full processes

Small distance contributions

Large distance contributions

Fundamental Research Colloquium





Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content DVCS data

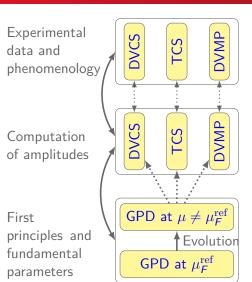
Fit status

Framework

Design Architecture

Ergonomics Features

Releases







Nucleon **Tomography**

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content DVCS data

Fit status

Neural network fits

Framework

Design Architecture

Ergonomics Features

Releases

Conclusion

Experimental data and phenomenology Need for

modularity

Computation of amplitudes

First principles and fundamental parameters

DVMP **DVCS TCS**

GPD at $\mu \neq \mu_F^{\text{ref}}$ Evolution GPD at μ_{F}^{ref}

- Many observables.
- Kinematic reach.
- Perturbative approximations.
- Physical models.
 - Fits.
- Numerical methods.
- Accuracy and speed.



Nucleon

A computing framework for physics. Done: tests, benchmarking, documentation, tutorials.



Tomography

Motivation Mass without mass

Nucleon structure Analogy

Phenomenology

Physical content DVCS data Fit status

Neural network fits

Framework

Design Architecture

Ergonomics Features Releases

Conclusion

 Long-term perspective: future users or developers (when EIC takes data!) will not have taken part to the design.

Design.

■ 3 stages:

- 2 Integration and validation.
- 3 Benchmarking and production.
- 1 new physical development = 1 new module.
- Aggregate knowledge and know-how in a robust long-term solution:
 - Models.
 - Measurements.
 - Numerical techniques.
 - Validation.
- What *can* be automated *will be* automated.
- Flexible software architecture.

B. Berthou *et al.*, Eur. Phys. **J. C78**, 478 (2018)

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Modularity. Inheritance, standardized inputs and outputs.



Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content DVCS data

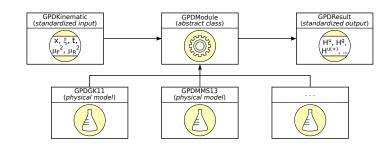
Fit status

Framework

Design

Architecture Ergonomics

Features Releases



- Steps of logic sequence in parent class.
- Model description and related mathematical methods in daughter class.



Flexibility.

Example: implementation of new coefficient functions.



Nucleon **Tomography**

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content DVCS data

Fit status

Neural network fits

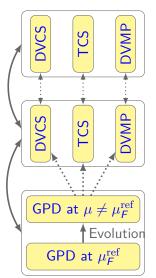
Framework

Design

Architecture Ergonomics

Features Releases

Conclusion



 A DVCS coefficient function module generically outputs a complex number when provided $(\xi, t, Q^2, \mu_F^2, \mu_P^2)$.

ConvolCoeffFunctionModule.h

- virtual std::complex<double> compute(double xi. double t. double Q2. double MuF2. double MuR2, GPDType::Type gpdType) = 0;
 - This module can be anything:
 - Constant CFFs for local fits.
 - CFFs for massless quarks.
 - CFFs for heavy guarks.
 - CFFs with TMC.



Systematic studies made easy.

A faster and safer way to GPD phenomenology... and beyond?!



Nucleon **Tomography**

Motivation

Analogy

Automation allows...:

to run numerous computations with various physical assumptions,

- to run nonregression tests.
- to perform **fits** with various models.
- physicists to focus on physics!

Nucleon structure Phenomenology

Mass without mass

Physical content DVCS data

Fit status

Neural network fits Framework

Design Architecture

Ergonomics

Features Releases

Conclusion

Without PARTONS



With PARTONS





Systematic studies made fast. What can be done from scratch in about 1 hour.



Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content

DVCS data Fit status

Neural network fits

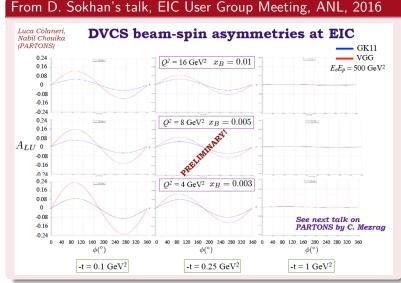
Framework

Design

Architecture

Features

Releases





GPD or CFF fits (1/4). Local fit of CFFs.



Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content

DVCS data Fit status

Neural network fits

Neural network fit

Framework

Design

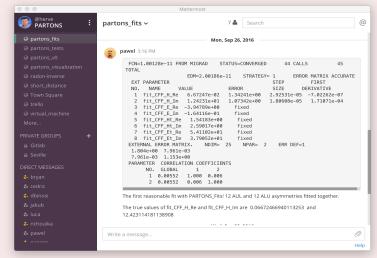
Architecture Ergonomics

Features

Releases

Conclusion

First local fit of pseudo DVCS data, Sep. 26^{th} , 2016





GPD or CFF fits (2/4). Global fit of CFFs using a analytic parameterization.

Isla CEA - Saciay

Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology Physical content

DVCS data

Neural network fits

Framework

Design Architecture

Ergonomics Features

Releases

Conclusion

Parametric global fit of JLab DVCS data, Apr. $5^{ m th}$, 2017

RESULTS

• Kinematic cuts $Q^2 > 1.5 \text{ GeV}^2$ (where we can rely on LO approximation)

-t / Q² < 0.25 (where we can rely on GPD factorization)

■ χ^2 / ndf 3272.6 / (3433 - 7) ≈ 0.96

 $\hspace{0.5in} \blacksquare \hspace{0.5in} \text{Free parameters} \hspace{0.5in} \mathtt{a}_{Hsea}, \hspace{0.5in} \mathtt{a}_{\tilde{H}sea}, \hspace{0.5in} \mathtt{c}_{sub}, \hspace{0.5in} \mathtt{a}_{sub}, \hspace{0.5in} \mathtt{N}_{E}, \hspace{0.5in} \mathtt{N}_{\tilde{E}} \\$

χ² / ndf per data set

[1] Phys. Rev. C 92, 055202 (2015) [2] Phys. Rev. Lett. 115, 212003 (2015)

[3] Phys. Rev. D 91, 052014 (2015)

Experiment	Reference	Observables	N points all	N points selected		chi2 / ndf
Hall A	[1] KINX2	σUU	120	120	135.0	1.19
Hall A	[1] KINX2	ΔσLU	120	120	98.9	0.88
Hall A	[1] KINX3	σUU	108	108	274.8	2.72
Hall A	[1] KINX3	ΔσLU	108	108	107.3	1.06
CLAS	[2]	σUU	1933	1333	1089.2	0.82
CLAS	[2]	ΔσLU	1933	1333	1171.9	0.88
CLAS	[3]	AUL, ALU, ALL	498	305	338.1	1.13
Pawel Sznaider			NS 2017			12



GPD or CFF fits (3/4). Global fit of CFFs using neural networks.

NEURAL NETWORK



Nucleon Tomography

Motivation

Mass without mass Nucleon structure

Phenomenology

Physical content

Fit status

Analogy

Neural network fits

Framework

Design

Architecture

Features

Releases

Conclusion

Neural network global fit of CLAS asymmetries, May $31^{\rm st}$, 2017



- Our very first attempt to use NN technique → proof of feasibility
- Genetic algorithm (GA) to learn NN
- NN and GA libraries by PARTONS group
- Very simple design of NN
- CLAS asymmetry data only



GPD or CFF fits (4/4). From local to global fits in 8 months!



Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology Physical content

DVCS data Fit status

Neural network fits

Framework

Design Architecture Ergonomics

Features

Releases

Conclusion

- PARTONS architecture allows focusing on parameterization and fitting engine.
- The same machinery is used for local and global fits.
- Fast and constant progress since the first fits.

Moutarde et al., Eur. Phys. J. C78, 890 (2018)

Moutarde et al., Eur. Phys. J. C79, 614 (2019)

■ More to come in the near future: GPD fits, impact studies!



Open source release. Publicly available on CEA GitLab server.



Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content DVCS data

Fit status

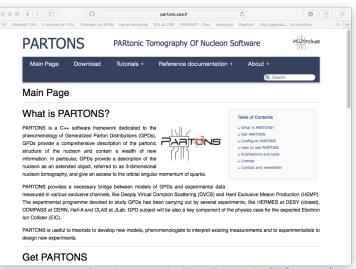
Neural network fits

Framework

Design Architecture

Ergonomics

Features Releases





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Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology Physical content

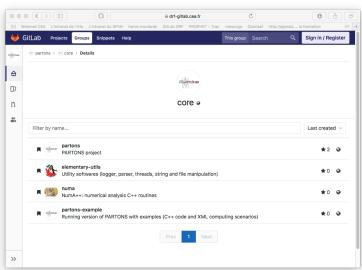
DVCS data Fit status

Neural network fits

Framework

Architecture

Features





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Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content

Fit status

Neural network fits

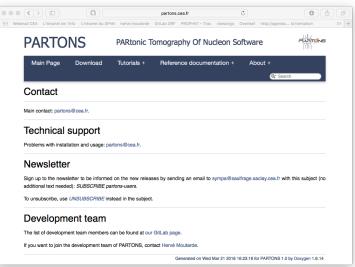
Framework

Design Architecture

Ergonomics

Features Releases

Conclusion



Berthou et al., Eur. Phys. J. C78, 478 (2018)



Within the next four years. Virtual Access Infrastructure 3DPartons in STRONG-2020.



Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology Physical content

DVCS data
Fit status
Neural network fits

Framework

Design Architecture Ergonomics

Features Releases

Conclusion

Work Package objectives

- Aggregate, improve and homogenize existing codes written by independent groups from the GPD and TMD communities: ensure interoperability.
- Maintain and release robust, flexible, validated and up-to-date open source codes to the 3D hadron structure community: foster progress.
- Provide documentation, technical assistance and perform nonregression tests: facilitate dissemination.
- Promote Open Data and Open Science: build on previous research and get new results faster.





Within the next four years. Virtual Access Infrastructure 3DPartons in STRONG-2020.



Nucleon Tomography

Motivation

Mass without mass Nucleon structure Analogy

Phenomenology Physical content

DVCS data
Fit status
Neural network fits

Framework

Design Architecture Ergonomics

Features Releases

Conclusion

Work Package tasks

- Flexible software architecture for GPD and TMD codes, elaborating on existing libraries and benefiting from experience from the PDF community.
- Generic MC event generators for GPDs and TMDs.
- Associated tools to compare theoretical calculations to experimental data.
- 3DPartons workshops and training schools.
- Webpage, software forge and mailing lists.
- Interact with relevant Work Packages of STRONG-2020.





Conclusion and prospects. Putting all the pieces together.



Nucleon Tomography

Motivation

Nucleon structure Analogy

Phenomenology Physical content

DVCS data Fit status

Neural network fits

Design Architecture

Ergonomics Features Releases

Conclusion

We now have tools to systematically relate these models to experimental data in multi-channel analysis.

- We now have an **operating engine** for global CFF fits.
- We revisit the **mechanical properties** of hadrons to assess how much we can learn from GPD extractions.
- We can now build generic GPD models satisfying a priori all theoretical constraints.

New studies become possible!

- Global GPD fits.
- Energy-momentum structure of hadrons.
- Quantitative impact of nonperturbative QCD ingredients on 3D hadron structure studies.
 - GPD and TMD studies in a common framework.
 - ... And probably much more!



PARTONS joins the EU effort towards EIC!



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Motivation

Mass without mass Nucleon structure Analogy

Phenomenology

Physical content DVCS data

Fit status

Neural network fits

Framework

Design

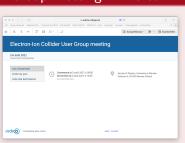
Architecture Ergonomics

Features Releases

Previous European EIC User Group Meeting in Paris



Next European EIC User Group Meeting in Warsaw



Fundamental Research Colloquium

Commissariat à l'énergie atomique et aux énergies alternatives
Centre de Saclay 91191 Gif-sur-Yvette Cedex

Etablissement public à caractère industriel et commercial R.C.S. Paris B 775 68

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