

Hadronic structure studies within 3DPartons

Valerio Bertone

IRFU, CEA, Université Paris-Saclay

université
PARIS-SACLAY



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Outline

VA2-Virtual Access to 3DPartons (3DPartons)

 Virtual Access

3DPartons gives access to open-source code necessary for high precision phenomenology in the field of 3D hadron structure, with a specific emphasis on generalized parton distributions (GPDs) and transverse momentum dependent parton distributions (TMDs).

 Discussed in this talk:

-  The deconvolution problem: extracting GPDs from Compton form factors
-  The evolution of GPDs
-  Extracting TMDs from data
-  Extraction of light-hadron fragmentation functions to NNLO
-  Reconstructing generalised TMDs (GTMDs) to high accuracy

The deconvolution problem

- Given an arbitrarily large and precise set of data for the Compton form factor (CFF) \mathcal{H} and the convolution formula:

$$\mathcal{H}(\xi, Q^2) = \int_{-1}^1 \frac{dx}{\xi} T\left(\frac{x}{\xi}, \frac{Q^2}{\mu}, \alpha_s(\mu^2)\right) H(x, \xi, \mu^2) \equiv [T \otimes H](\xi, Q)$$

- where T is perturbatively known to some fixed order and H is a GPD, can we uniquely extract H ? This often goes under the name of *deconvolution problem*.
- Since the dependence on x of the GPD is integrated over, one may superficially expect that it is *not* possible to extract the GPD H uniquely.
- While this argument has long been advocated and proven to tree level, it was also believed that evolution effects may provide a handle on H .
- Indeed, a general answer to this question (valid to any perturbative order and scale) requires considering evolution effects provided by the solution of:

$$\frac{dH(x, \xi, \mu^2)}{d \ln \mu^2} = [P \otimes H](x, \xi, \mu^2)$$

- Clearly, the evolution entangles x , ξ , and μ^2 and may potentially allow one to find a unique solution to the deconvolution problem.

The deconvolution problem

- 🍏 In [*Phys.Rev.D* 103 (2021) 11, 114019] we have addressed the deconvolution problem accounting for NLO corrections in T and evolution effects for H .
- 🍏 The striking result of this paper is that it is possible to identify non-trivial GPDs with *arbitrarily small* imprint on the CFFs: the **shadow GPDs**.
- 🍏 This is the *explicit proof* that the deconvolution problem has no unique solution.
- 🍏 The shadow GPD is constructed in double-distribution (DD) space:

$$H_{\text{shadow}}(x, \xi) = \int_{-1}^1 d\beta \int_{-1+|y|}^{1+|y|} d\alpha \delta(x - \beta - \xi\alpha) F_{\text{shadow}}(\alpha, \beta)$$

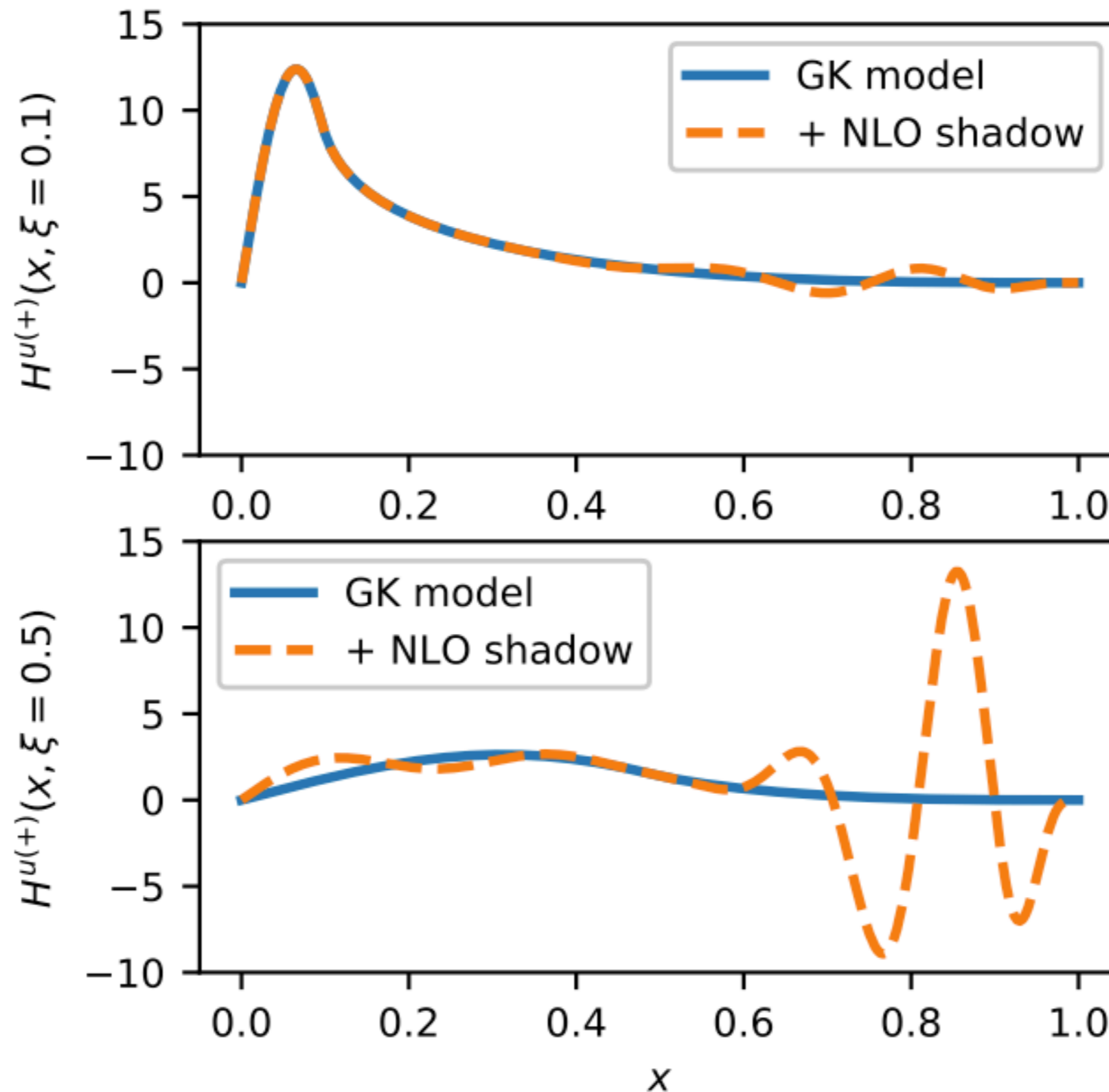
- 🍏 Accounting for polynomiality of GPDs, we approximate F_{shadow} as:

$$F_{\text{shadow}}(\alpha, \beta) = \sum_{\substack{m+n \leq N \\ m \text{ even} \\ n \text{ odd}}} c_{mn} \alpha^m \beta^n$$

- 🍏 We then require that the CFF at some scale vanishes as well as $H(x,0) = 0$ in a way that it does not affect the forward limit of the GPD (*i.e.* the PDF).
- 🍏 The final result is an **underconstrained** problem that admits infinitely many solutions provided that the degree N is large enough.

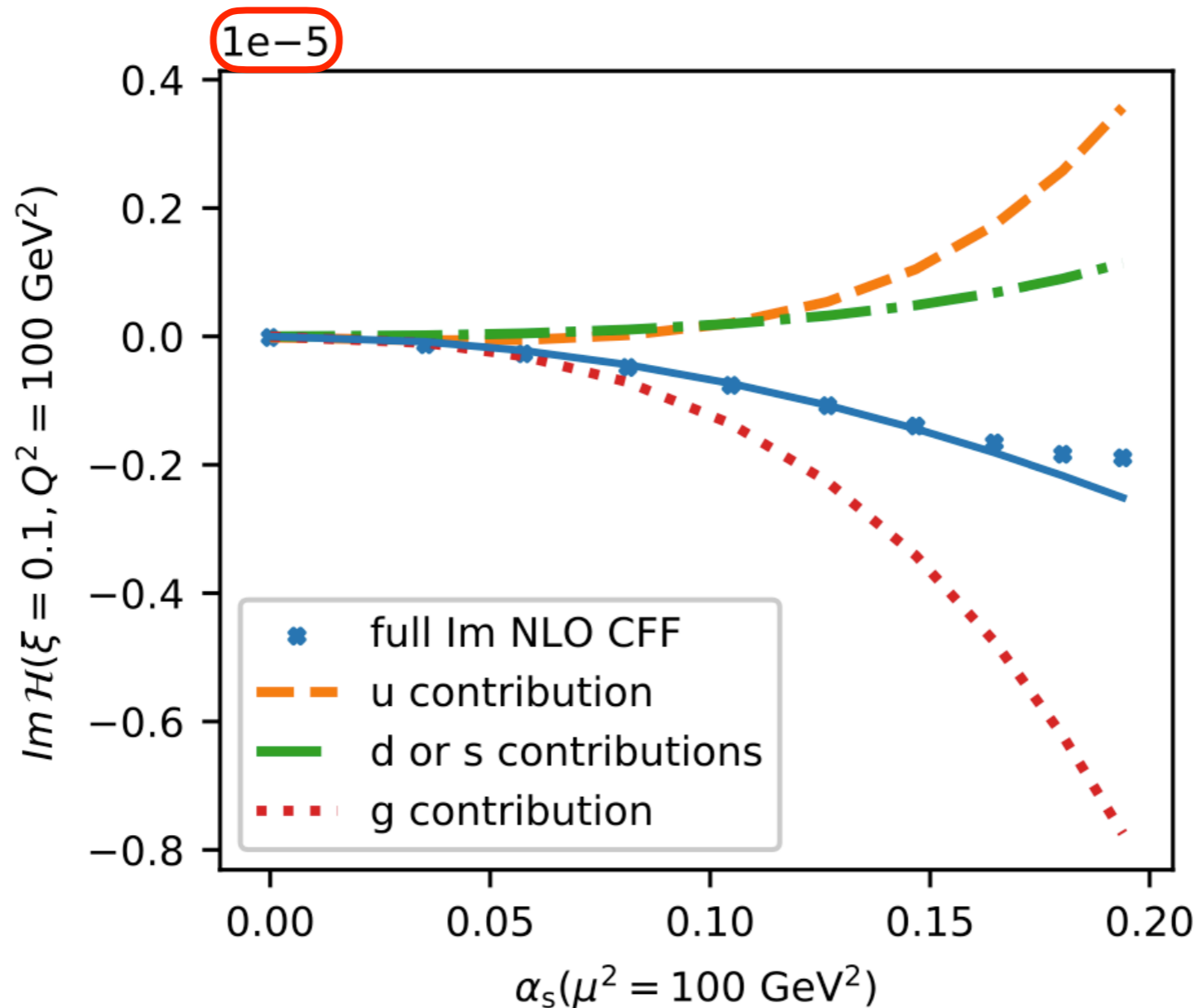
The deconvolution problem

🍏 The effect of a possible shadow GPD on the GK model:



The deconvolution problem

- 🍏 NLO CFF generated by a shadow GPD evolved from $\mu_0^2 = 1 \text{ GeV}^2$ to $\mu^2 = 100 \text{ GeV}^2$:
 - 🍏 it scales quadratically with $\alpha_s(\mu^2 = 100 \text{ GeV}^2)$ as expected,
 - 🍏 it is $\mathcal{O}(10^{-5})$, *i.e.* negligible w.r.t. a typical physical value.



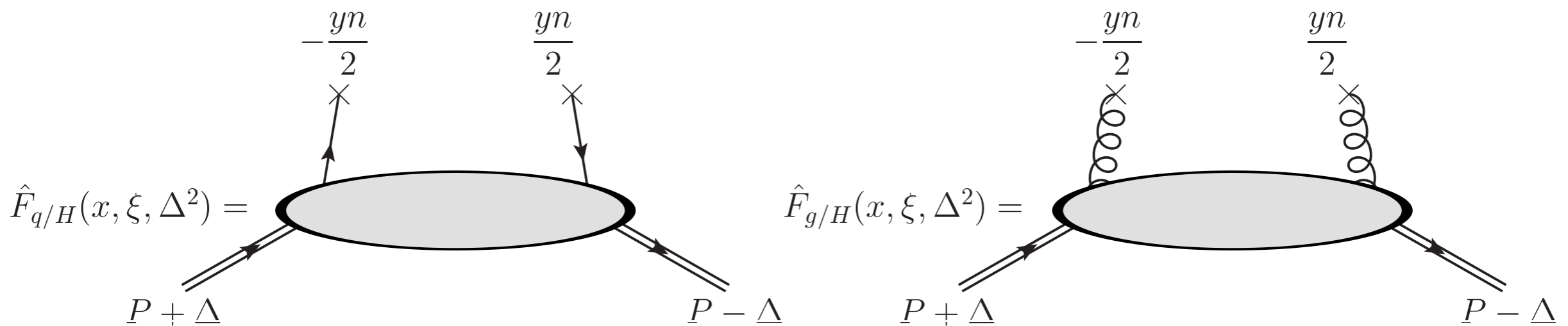
- 🍏 Numerical results obtained with **PARTONS** interfaced to **APFEL++**, both developed within the VA2 work package.

GPD evolution

🍏 GPDs admit and operator definition that in **light-cone gauge** ($n \cdot A = 0$) reads:

$$\hat{F}_{q/H}(x, \xi, \Delta^2) = \frac{1}{\sqrt{1-\xi^2}} \int \frac{dy}{2\pi} e^{-ix(n \cdot P)y} \left\langle P - \Delta \left| \bar{\psi}_q \left(\frac{yn}{2} \right) \frac{\not{n}}{2} \psi_q \left(-\frac{yn}{2} \right) \right| P + \Delta \right\rangle \quad \xi = \frac{\Delta^+}{P^+}$$

$$\hat{F}_{g/H}(x, \xi, \Delta^2) = -x(n \cdot P) \int \frac{dy}{2\pi} e^{-ix(n \cdot P)y} \left\langle P - \Delta \left| A_a^\alpha \left(\frac{yn}{2} \right) A_{a\alpha} \left(-\frac{yn}{2} \right) \right| P + \Delta \right\rangle$$



🍏 UV divergences arise that need to be renormalised leading to evolution equations:

$$\frac{dF^\pm(x, \xi, \mu)}{d \ln \mu^2} = \int_x^\infty \frac{dy}{y} \mathcal{P}^\pm \left(y, \frac{\xi}{x} \right) F^\pm \left(\frac{x}{y}, \xi, \mu \right)$$

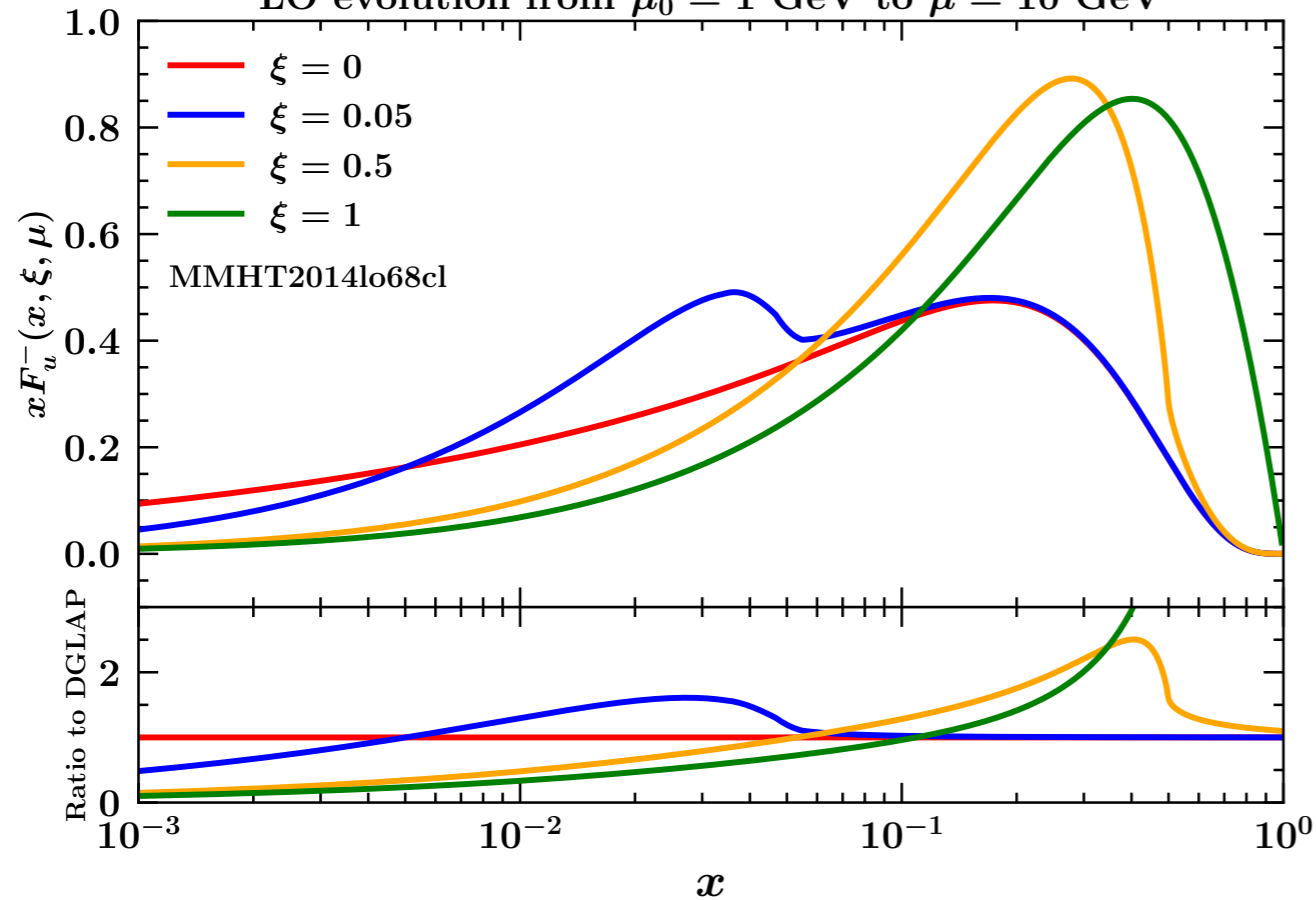
$$F^- = F_{q/H} - F_{\bar{q}/H}$$

$$F^+ = \left(\begin{array}{c} \sum_{q=1}^{n_f} F_{q/H} + F_{\bar{q}/H} \\ F_{g/H} \end{array} \right)$$

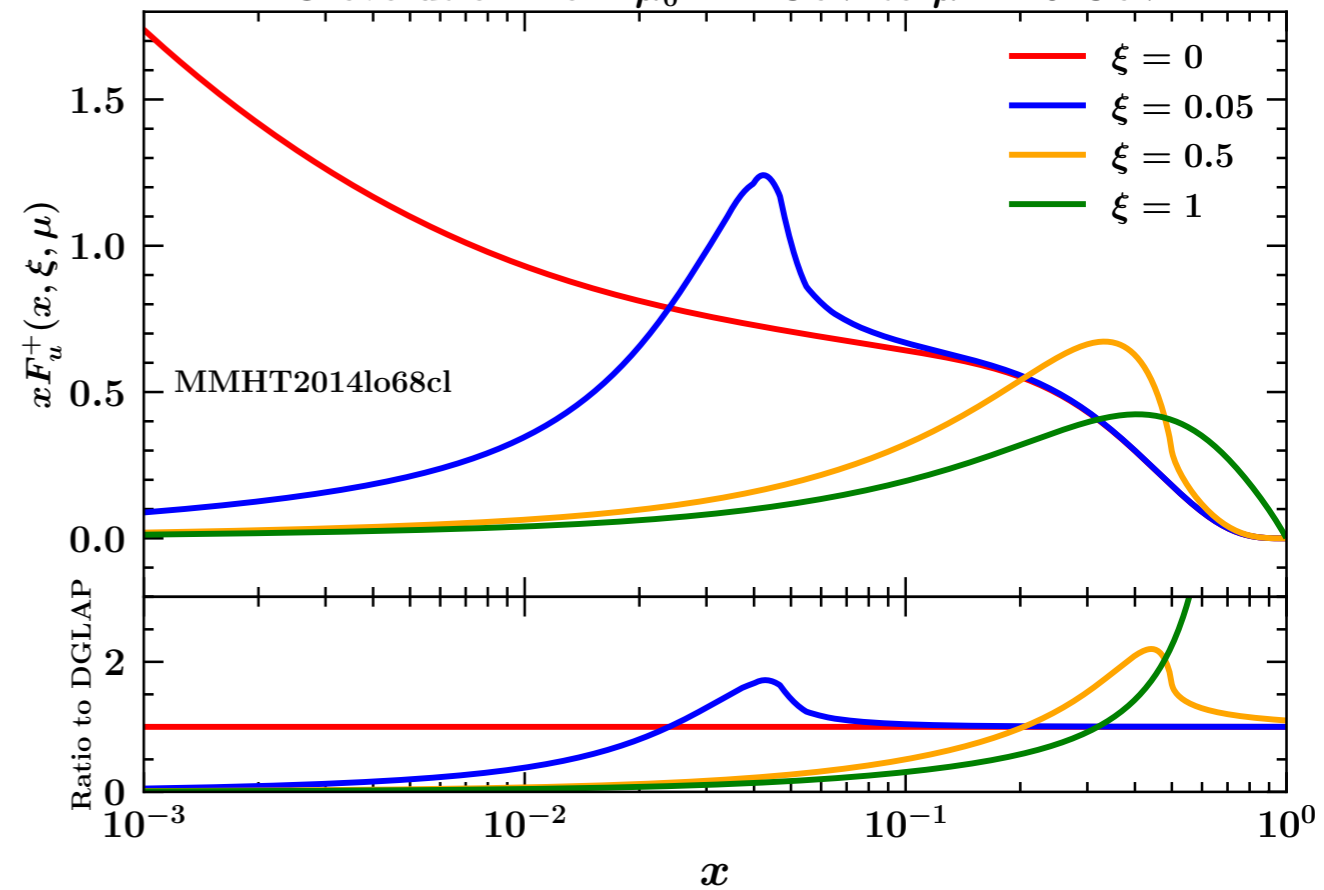
🍏 In [[Eur.Phys.J.C 82 \(2022\) 10, 888](#)] we have (re)computed the one-loop evolution kernels \mathcal{P}^\pm and provided a public implementation in **PARTONS** through **APFEL++**.⁷

GPD evolution

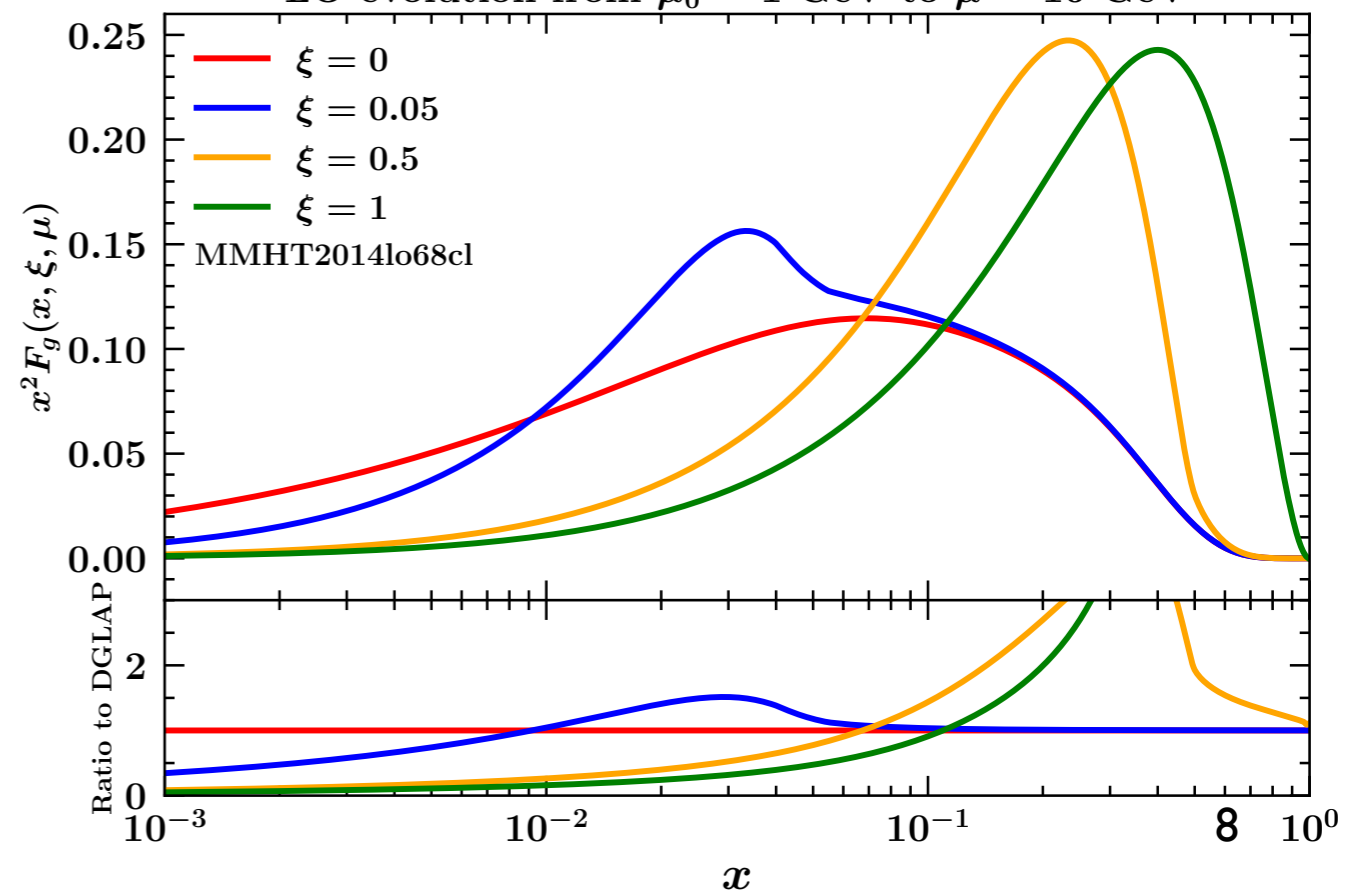
LO evolution from $\mu_0 = 1$ GeV to $\mu = 10$ GeV



LO evolution from $\mu_0 = 1$ GeV to $\mu = 10$ GeV



LO evolution from $\mu_0 = 1$ GeV to $\mu = 10$ GeV



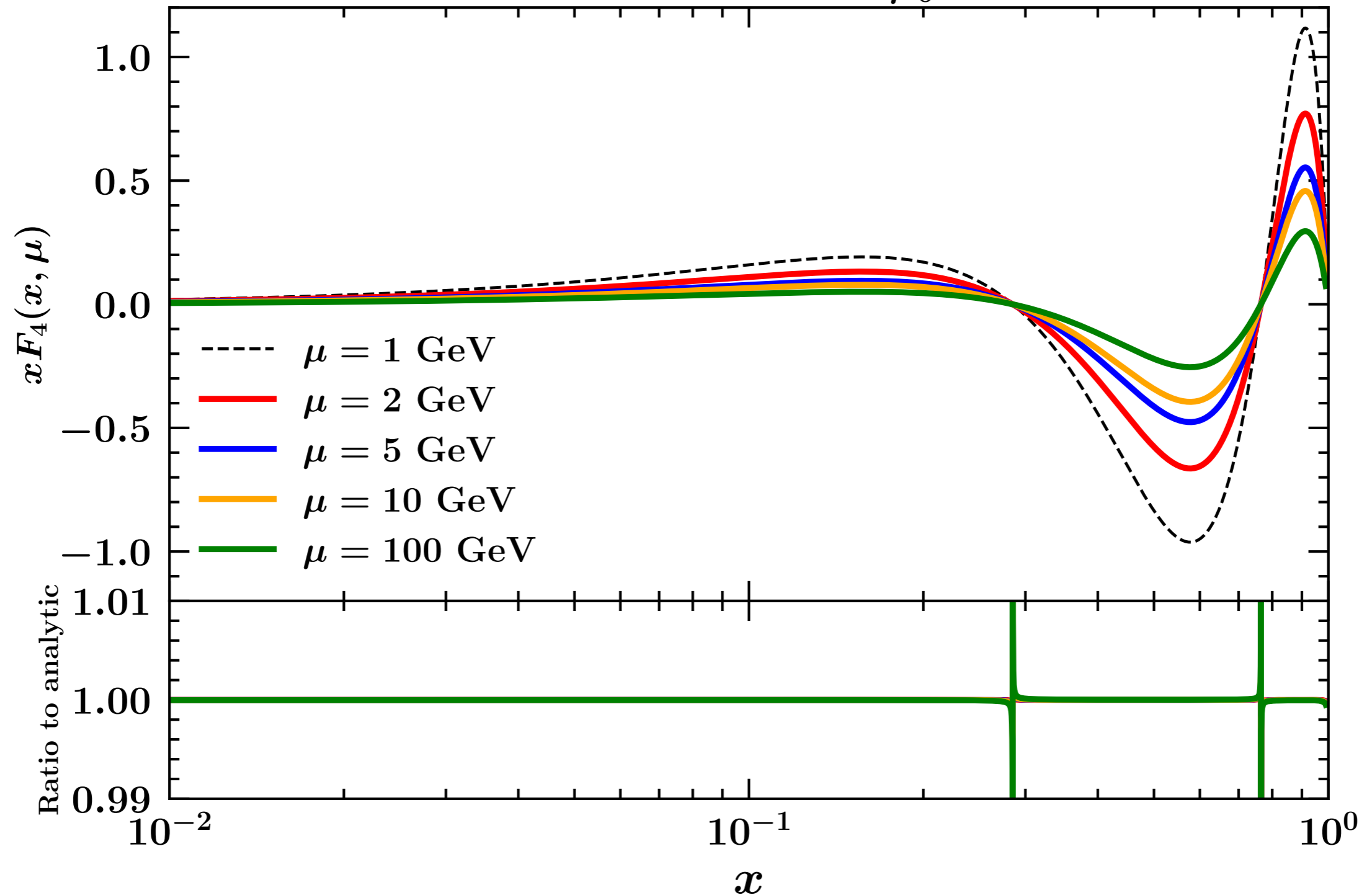
🍎 **DGLAP limit** reproduced within 10^{-5} relative accuracy.

🍎 GPD evolution may significantly deviate from DGLAP evolution.

🍎 The evolution generates a cusp at $x = \xi$ but the distribution remains **continuous** at this point.

GPD evolution

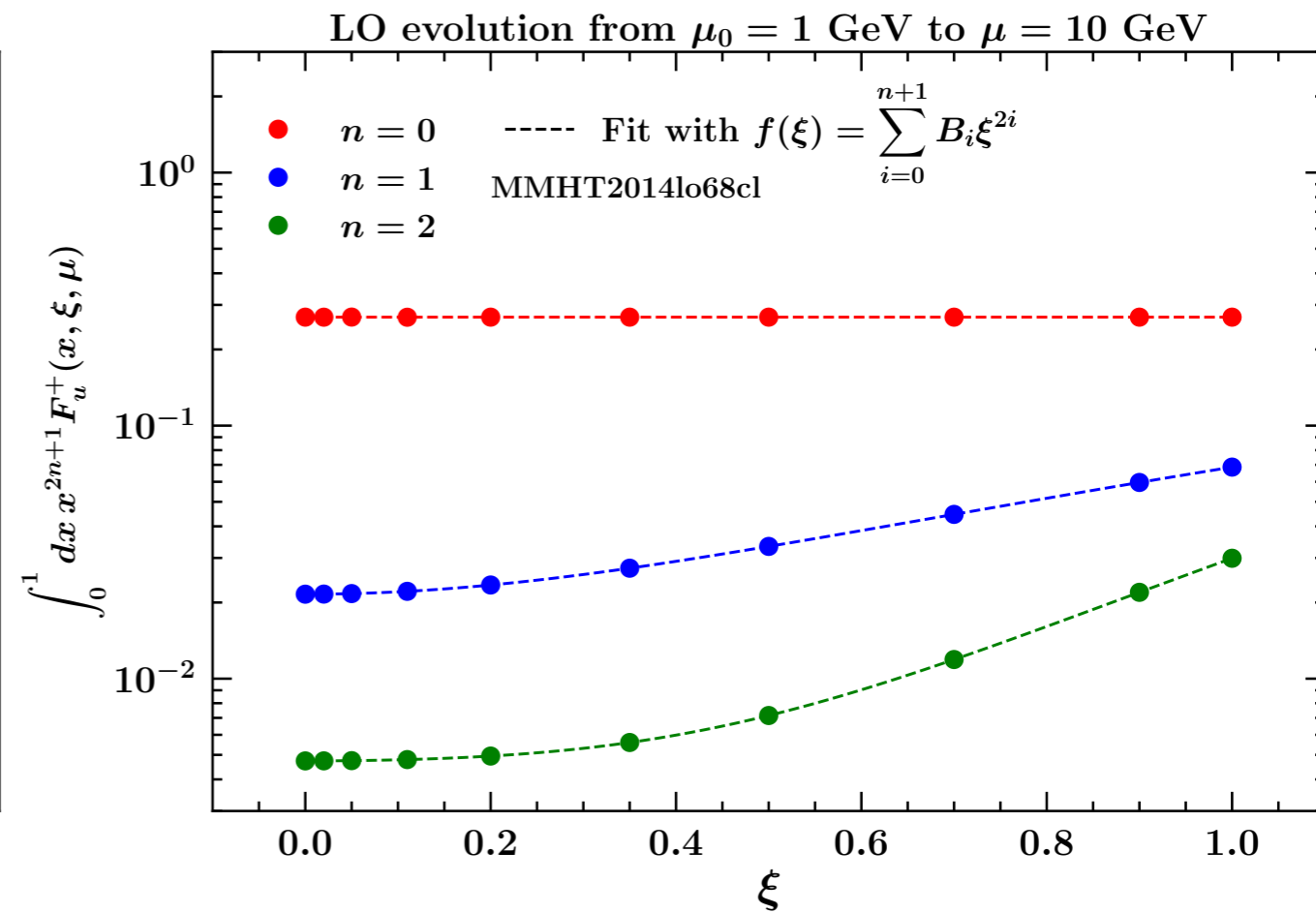
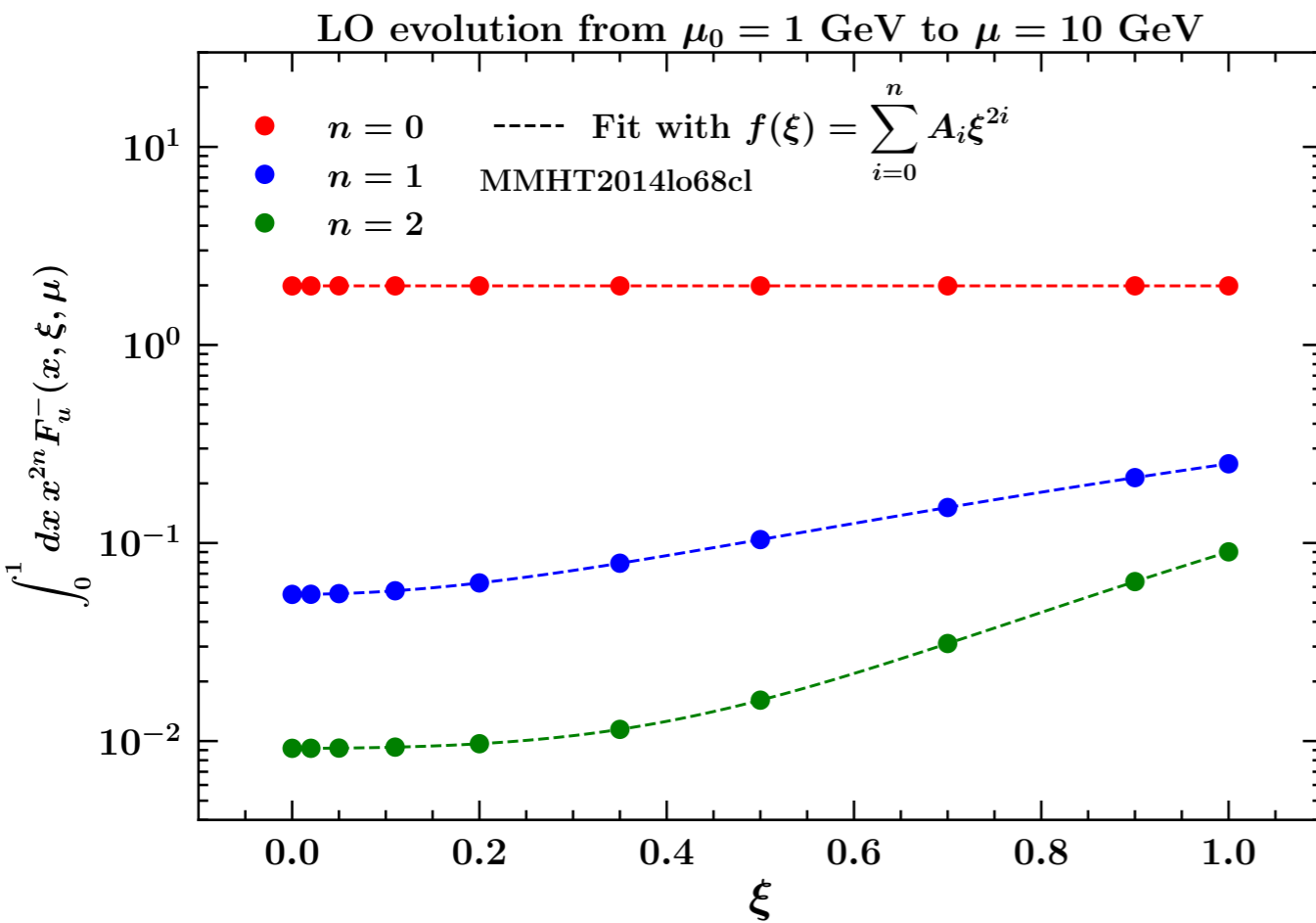
LO ERBL evolution from $\mu_0 = 1$ GeV



🍏 **ERBL limit** reproduced within less than 10^{-5} relative accuracy,

🍏 Same accuracy for **higher-degree** Gegenbauer polynomials.

GPD evolution



🍏 **First moment** for both singlet and non-singlet is **constant** in ξ :

🍏 this was expected and the expectation is very nicely fulfilled.

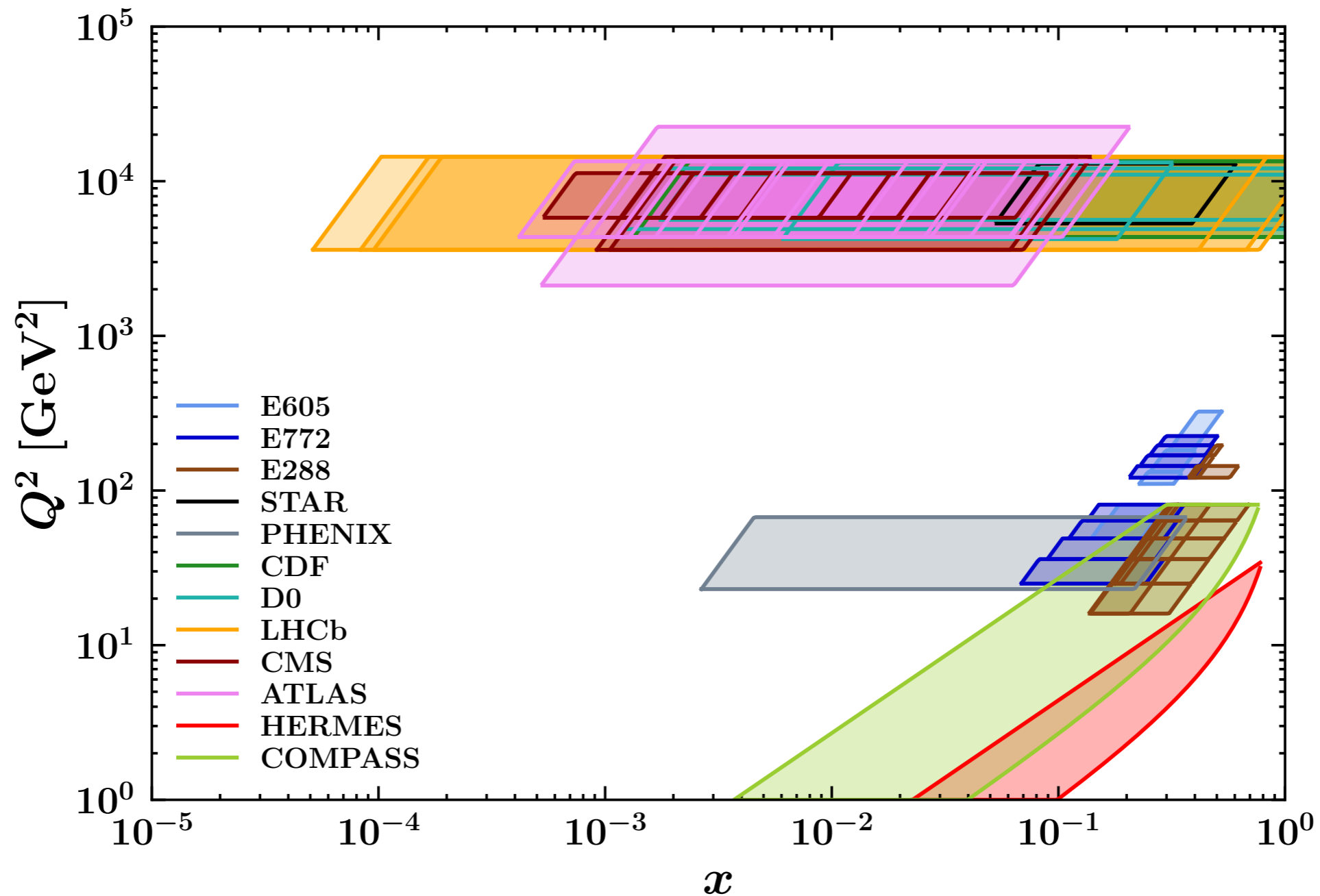
🍏 **Second and third moments** follow the expected power laws:

🍏 including odd-power terms in the fit gives coefficients very close to zero.

TMD extractions (1)



In [\[arXiv:2206.07598\]](https://arxiv.org/abs/2206.07598) (just accepted for publication in JHEP) we have carried out a *global* extraction of unpolarised TMD PDFs of the proton and TMD FFs of light hadrons from Drell-Yan and SIDIS data at N³LL accuracy.



A total of **2031** fitted data points (comparable to a fit of PDFs).



Extraction performed using the **NangaParbat** framework.

TMD extractions (1)

🍏 A generic cross section in TMD factorisation is computed as:

$$\frac{d\sigma}{dq_T} \propto H(Q) \int_0^\infty db_T b_T J_0(q_T b_T) F^{(1)}(x_1, b_T, Q) F^{(2)}(x_2, b_T, Q) + \mathcal{O}\left(\frac{q_T^2}{Q^2}\right)$$

🍏 Limited to $q_T \ll Q$. The single TMD reads:

$$F(x, b_T, Q) = f_{\text{NP}}(x, b_T, Q) R(Q \leftarrow 1/b_*(b_T)) [C \otimes f](x, 1/b_*(b_T))$$

🍏 with $b_*(b_T) \xrightarrow{b_T \rightarrow \infty} b_{\text{max}} \simeq 1 \text{ GeV}$ and $b_*(b_T) \xrightarrow{b_T \rightarrow 0} Q$.

🍏 Determine the f_{NP} 's (one for TMD PDFs and one form TMD FFs) from data:

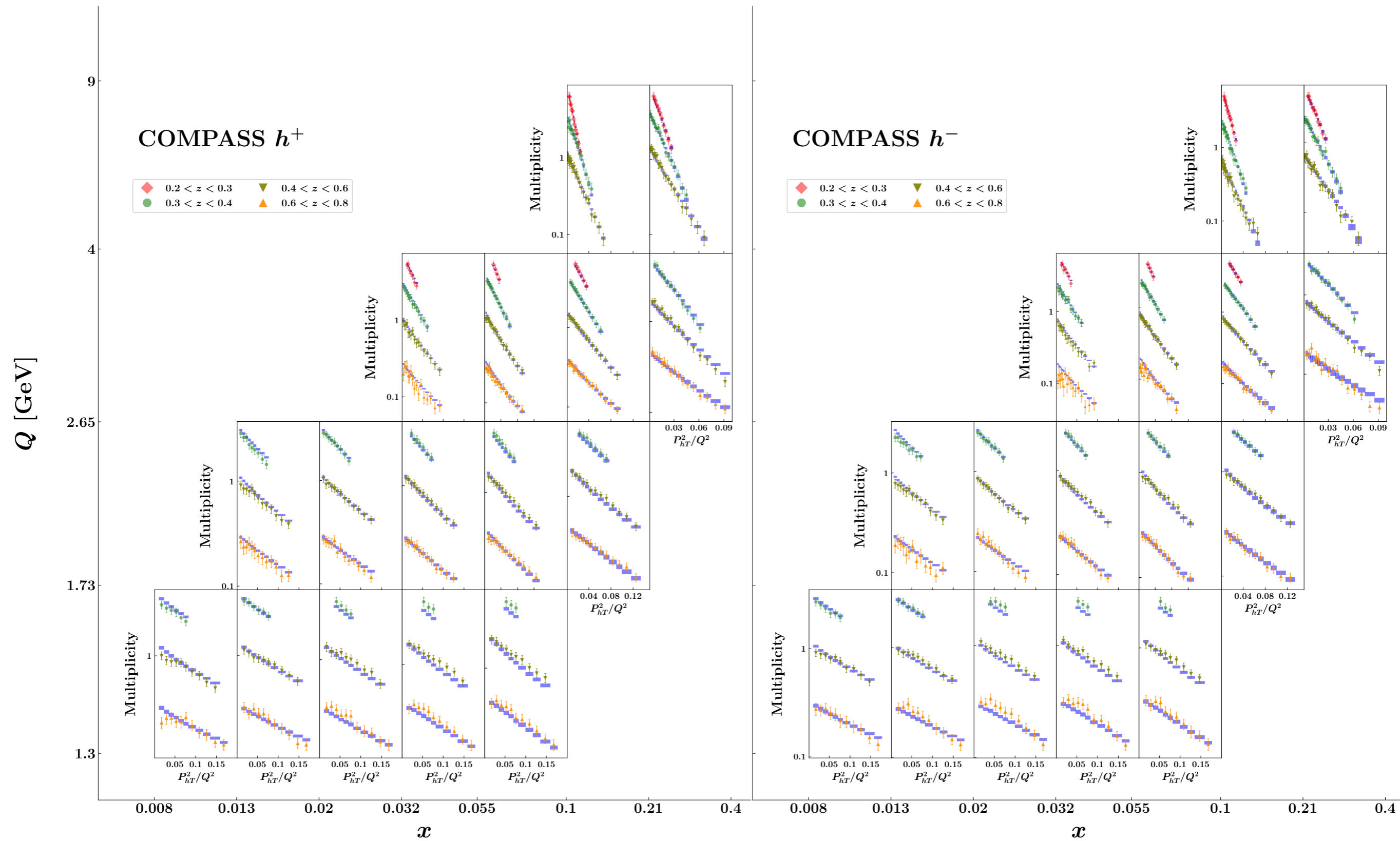
$$f_{\text{NP}}^{\text{PDF}}(x, b_T, Q) = \frac{g_1(x) e^{-g_1(x) \frac{b_T^2}{4}} + \lambda^2 g_{1B}^2(x) \left[1 - g_{1B}(x) \frac{b_T^2}{4}\right] e^{-g_{1B}(x) \frac{b_T^2}{4}} + \lambda_2^2 g_{1C}(x) e^{-g_{1C}(x) \frac{b_T^2}{4}}}{g_1(x) + \lambda^2 g_{1B}^2(x) + \lambda_2^2 g_{1C}(x)} \left(\frac{Q}{Q_0^2}\right)^{\frac{g_K(b_T)}{2}}$$

$$f_{\text{NP}}^{\text{FF}}(x, b_T, Q) = \frac{g_3(x) e^{-g_3(x) \frac{b_T^2}{4x^2}} + \frac{\lambda_F}{x^2} g_{3B}^2(x) \left[1 - g_{3B}(x) \frac{b_T^2}{4x^2}\right] e^{-g_{3B}(z) \frac{b_T^2}{4x^2}}}{g_3(x) + \frac{\lambda_F}{x^2} g_{3B}^2(x)} \left(\frac{Q}{Q_0^2}\right)^{\frac{g_K(b_T)}{2}}$$

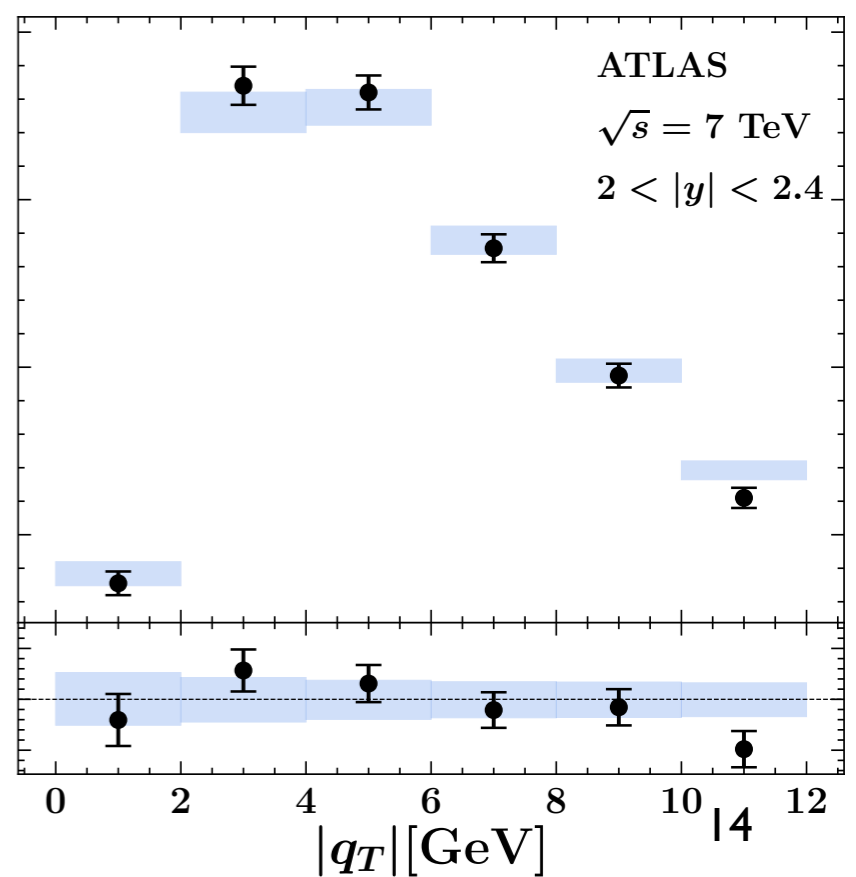
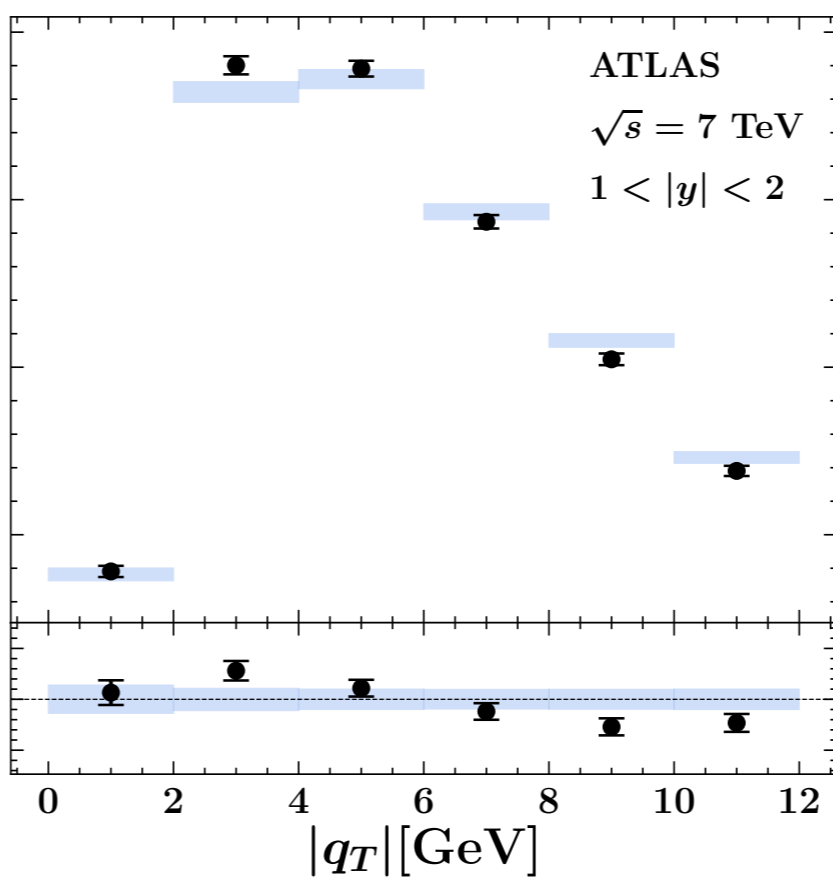
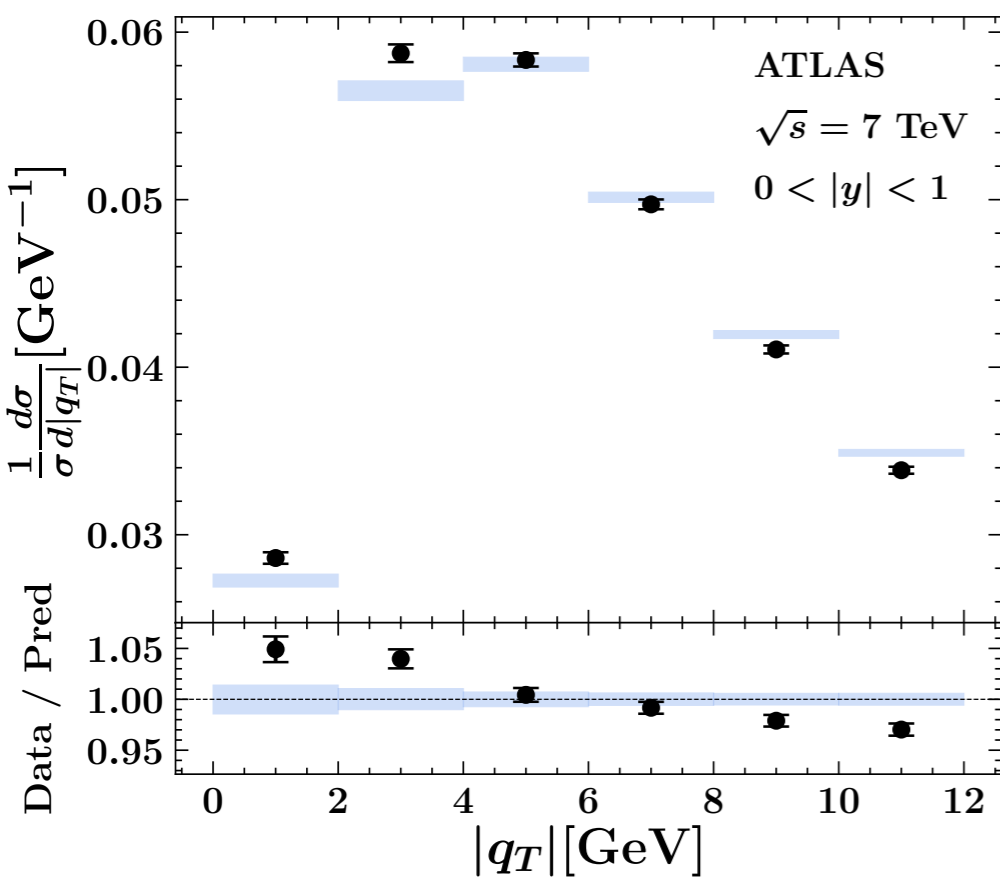
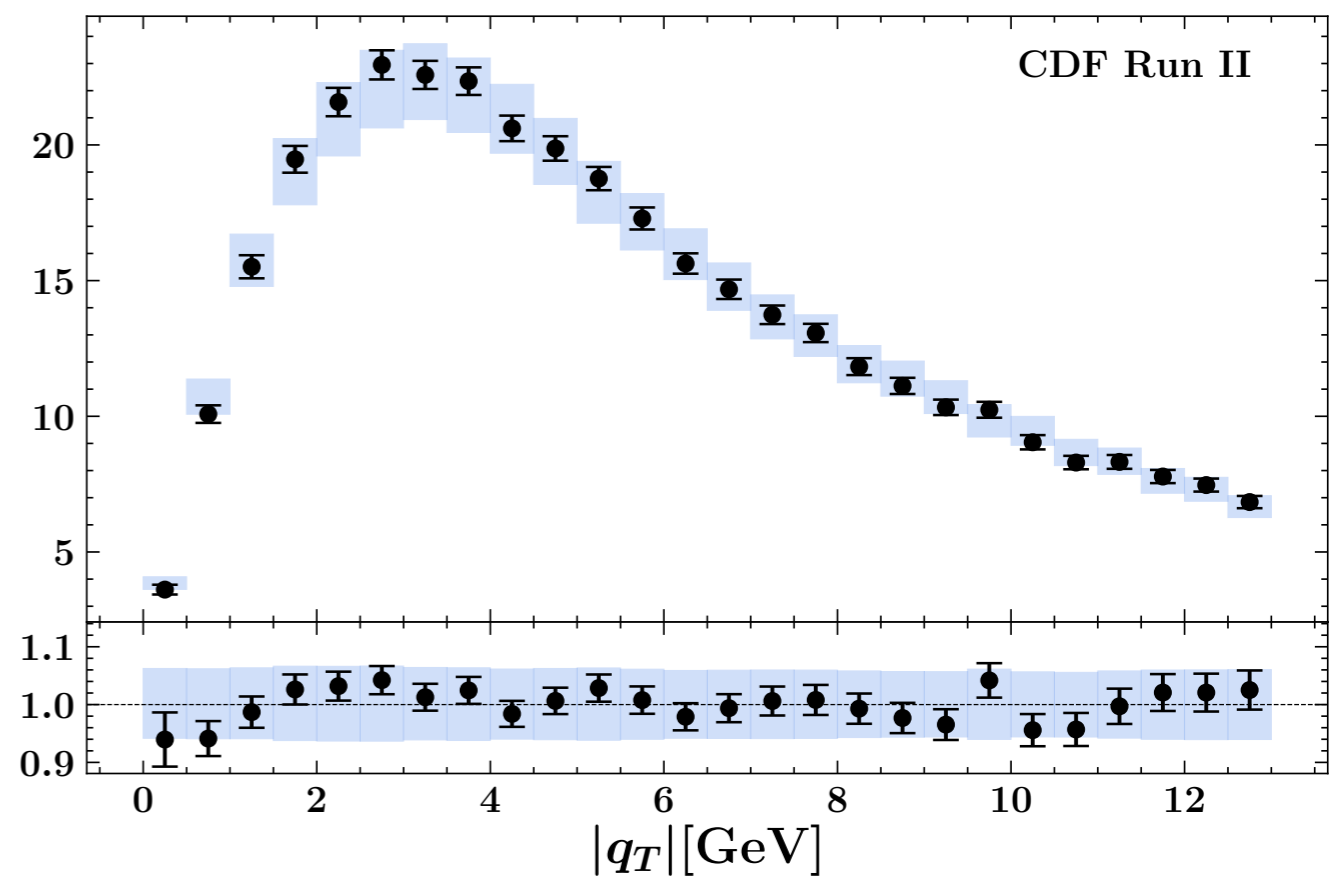
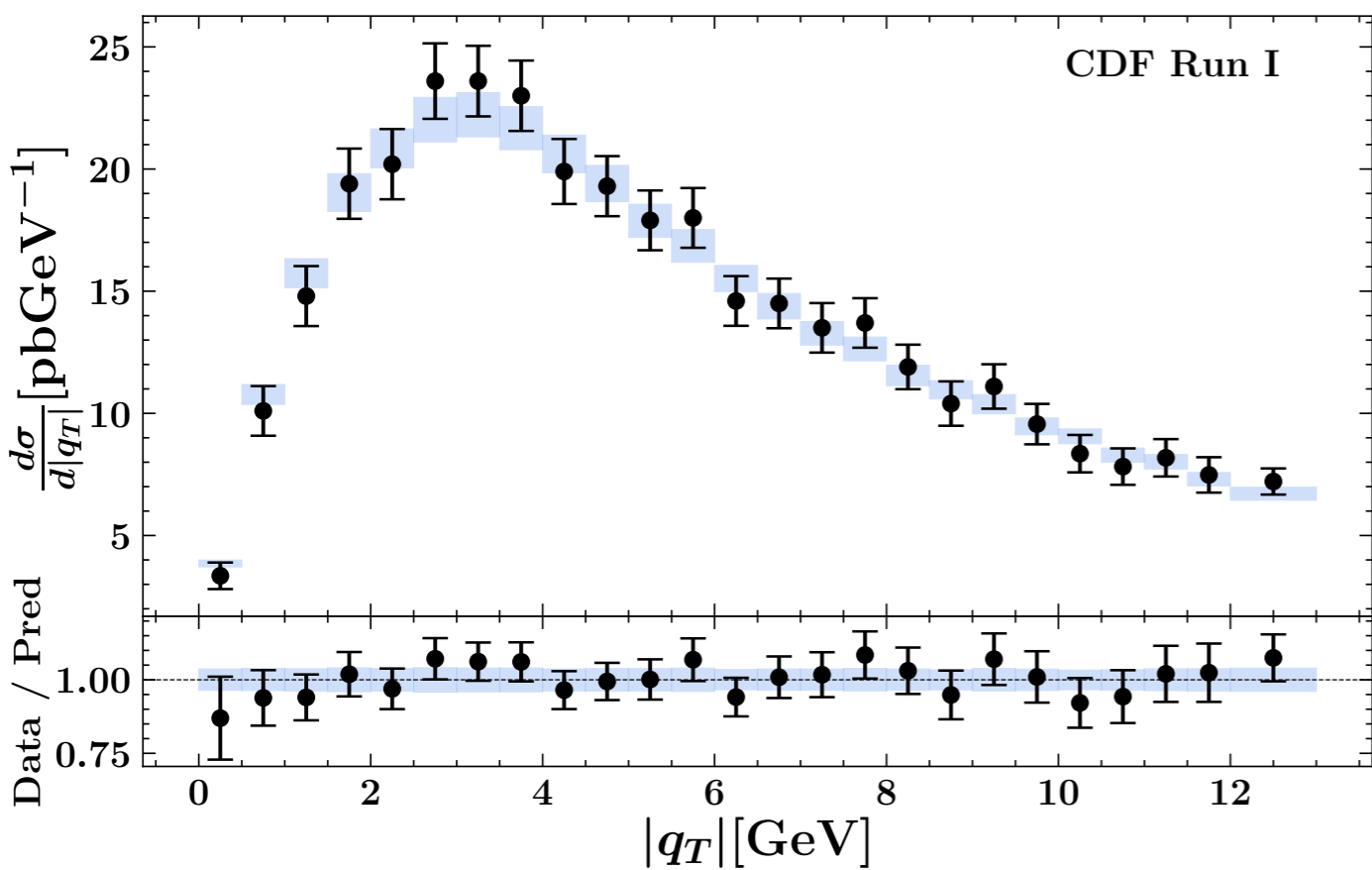
$$g_{\{1,1B,1C\}}(x) = N_{\{1,1B,1C\}} \frac{x^{\sigma_{\{1,2,3\}}} (1-x)^{\alpha_{\{1,2,3\}}^2}}{\hat{x}^{\sigma_{\{1,2,3\}}} (1-\hat{x})^{\alpha_{\{1,2,3\}}^2}} \quad g_{\{3,3B\}}(x) = N_{\{3,3B\}} \frac{(x^{\beta_{\{1,2\}}} + \delta_{\{1,2\}}^2)(1-x)^{\gamma_{\{1,2\}}^2}}{(\hat{x}^{\beta_{\{1,2\}}} + \delta_{\{1,2\}}^2)(1-\hat{z})^{\gamma_{\{1,2\}}^2}}$$

🍏 A total of **21 free parameters**. Excellent fit quality $\chi^2/N_{d.o.f.} = 1.06$.

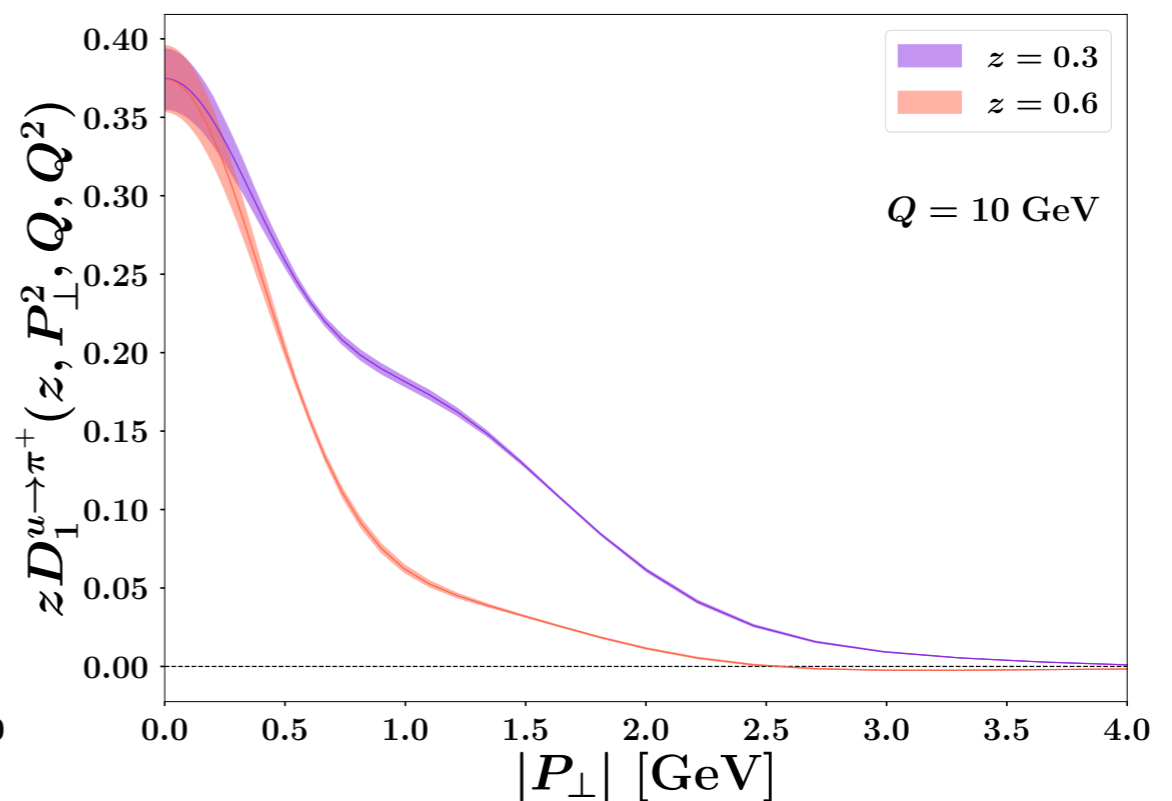
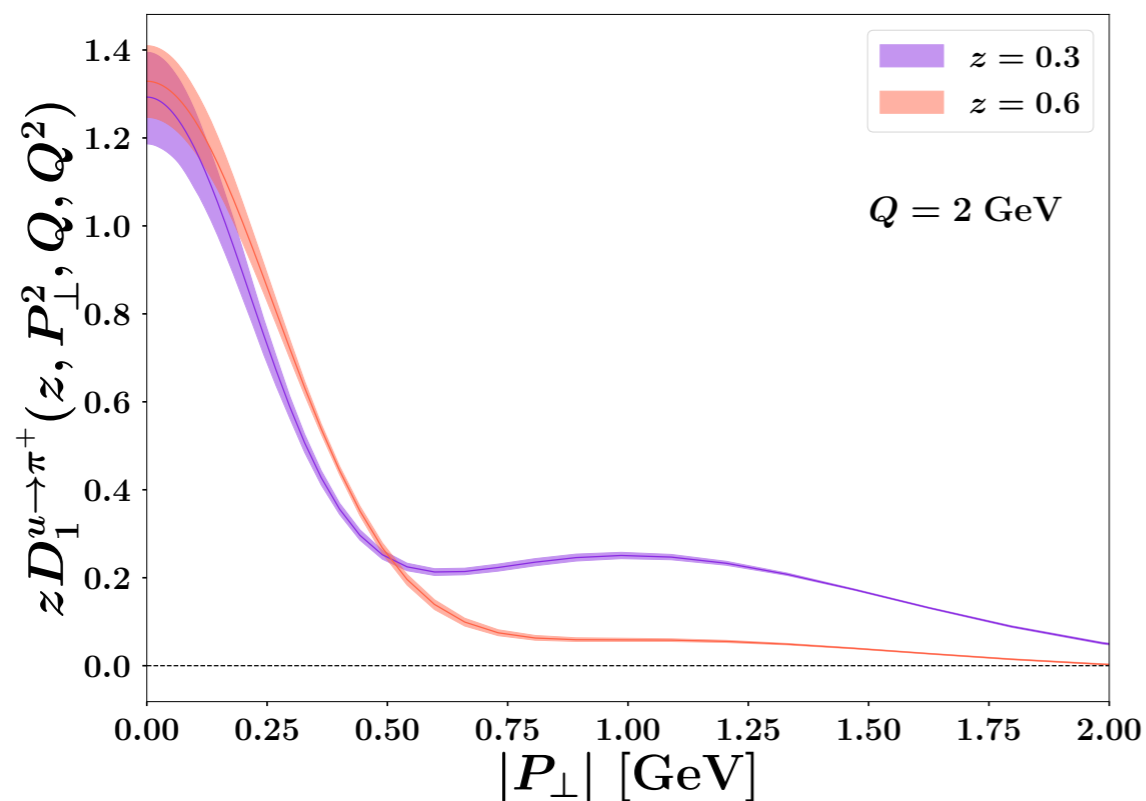
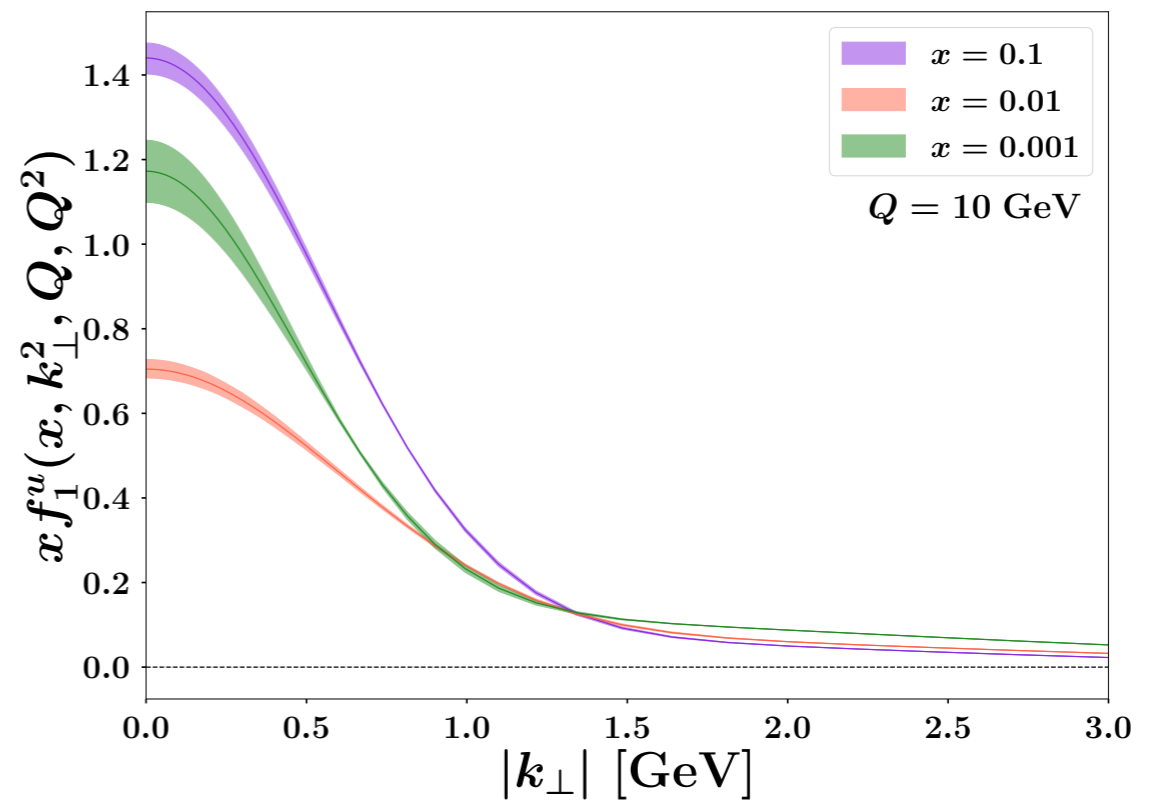
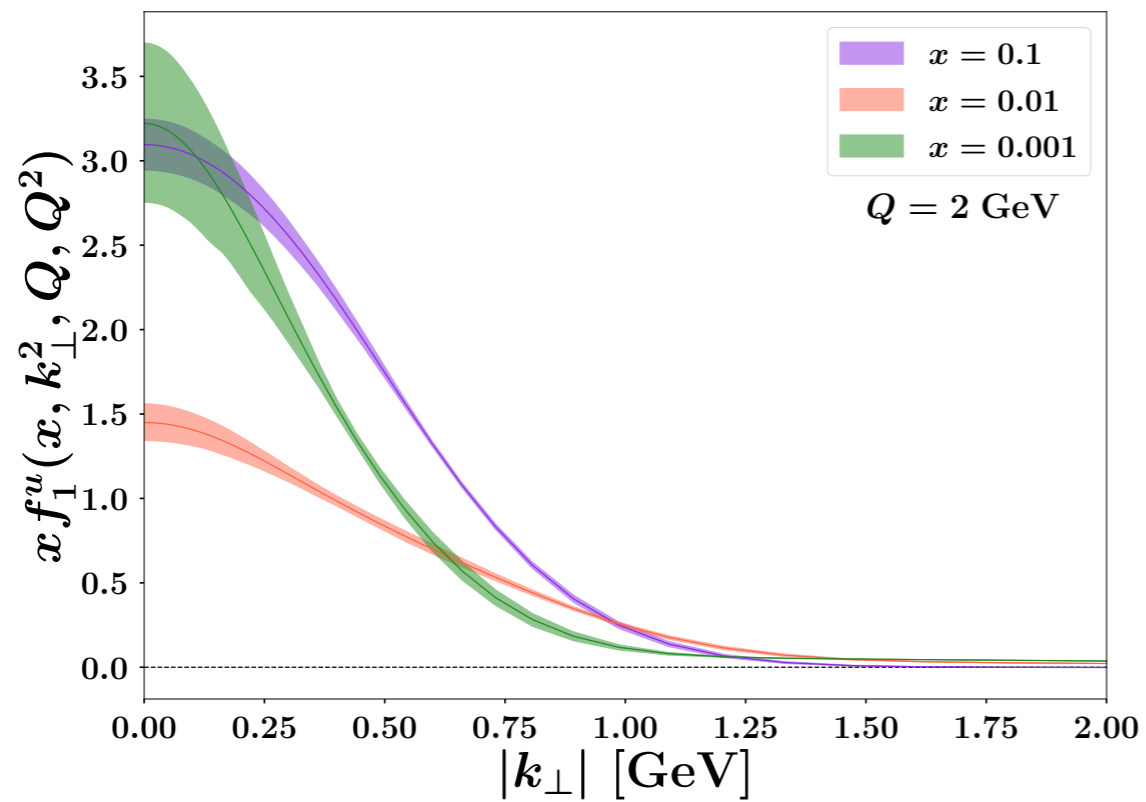
TMD extractions (1)



TMD extractions (1)



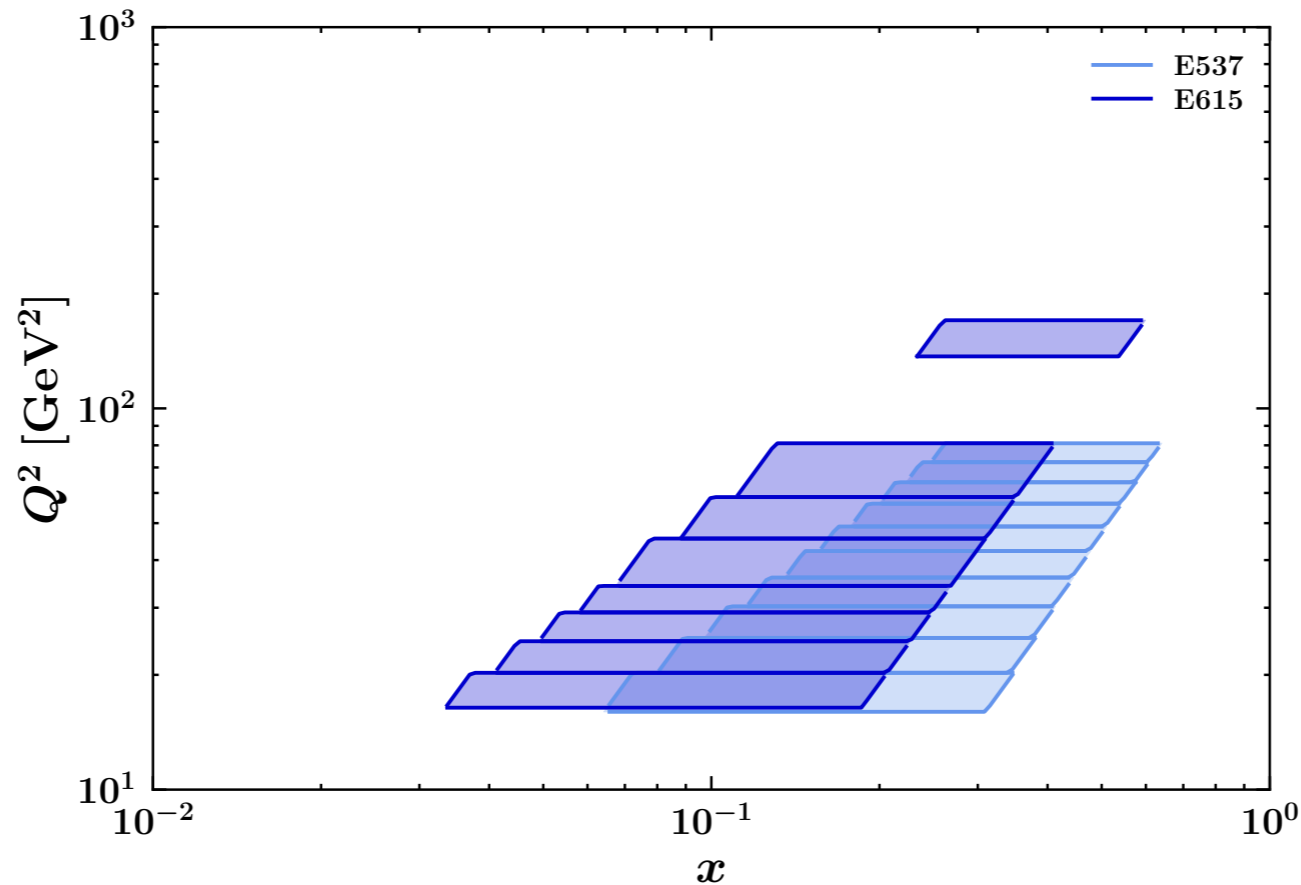
TMD extractions (1)



TMD sets released through **TMDlib**.

TMD extractions (2)

- 🍏 An analogous technology can be used to extract the TMD PDFs of the **pion**.
- 🍏 We again used **NangaParbat** to extract these TMDs and published the result in [\[arXiv:2010.01733\]](https://arxiv.org/abs/2010.01733).



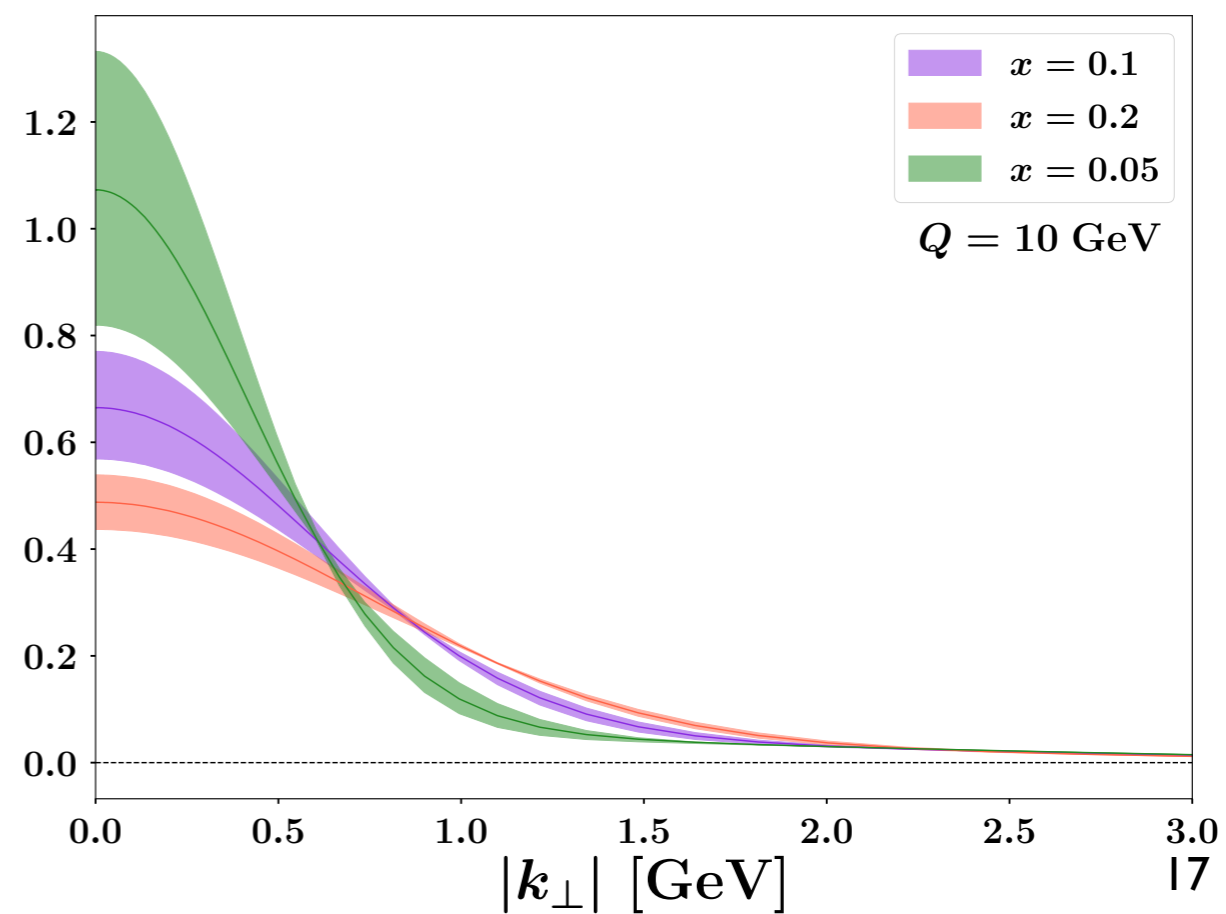
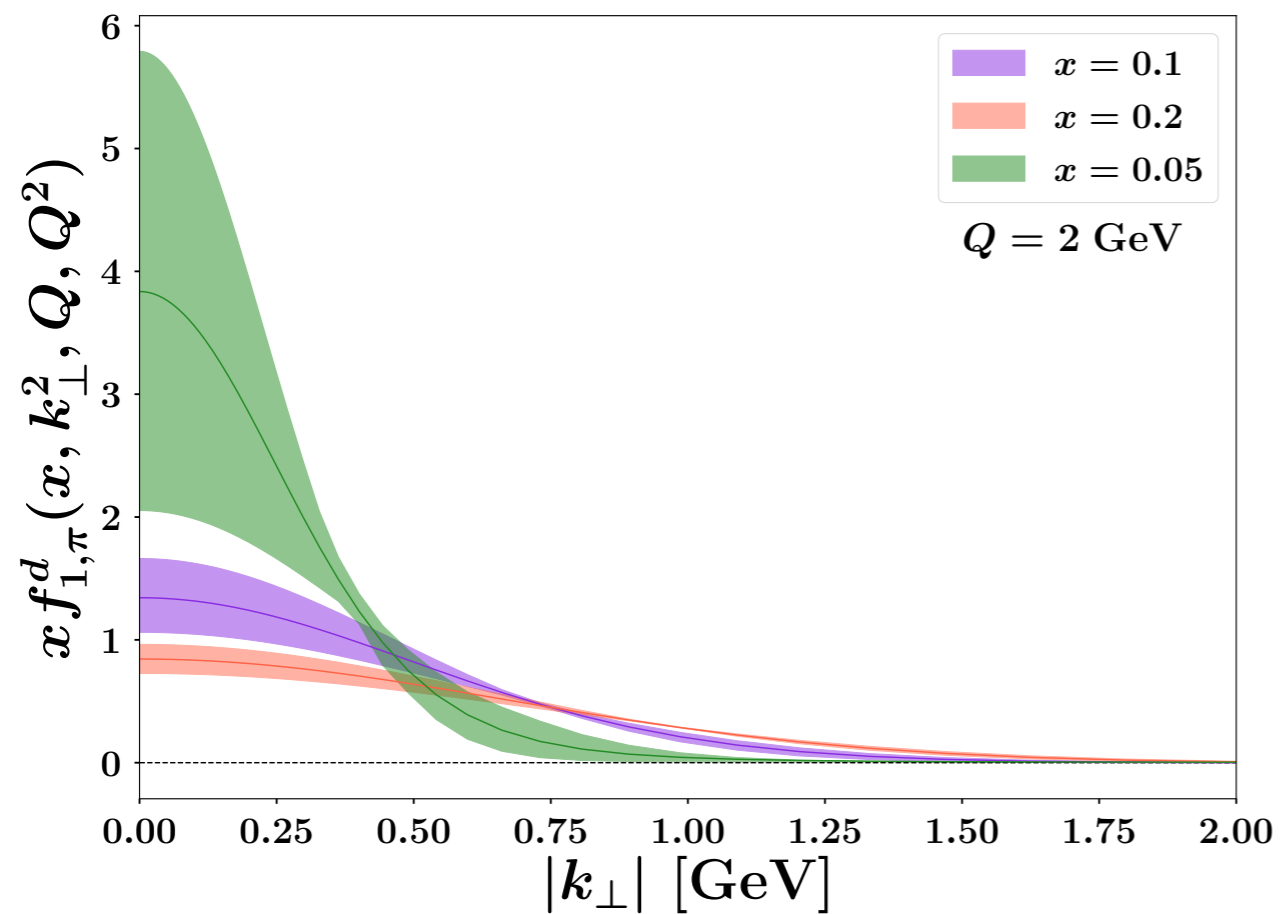
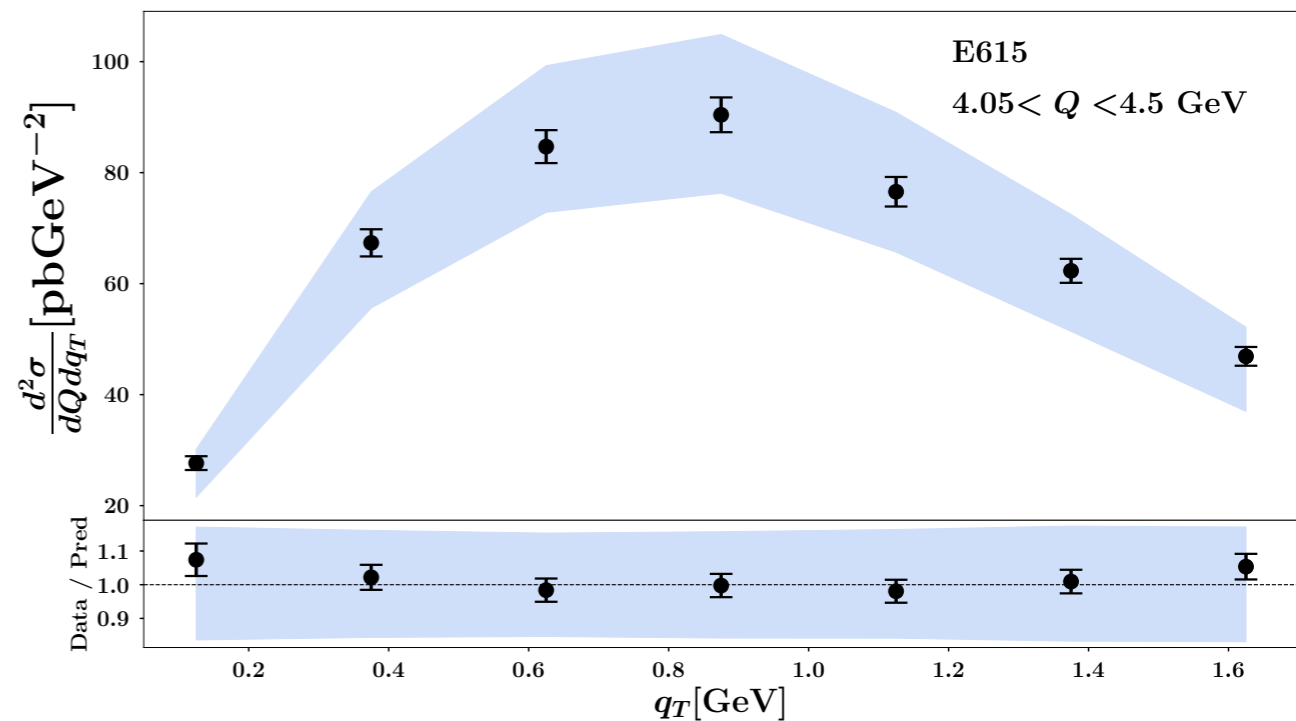
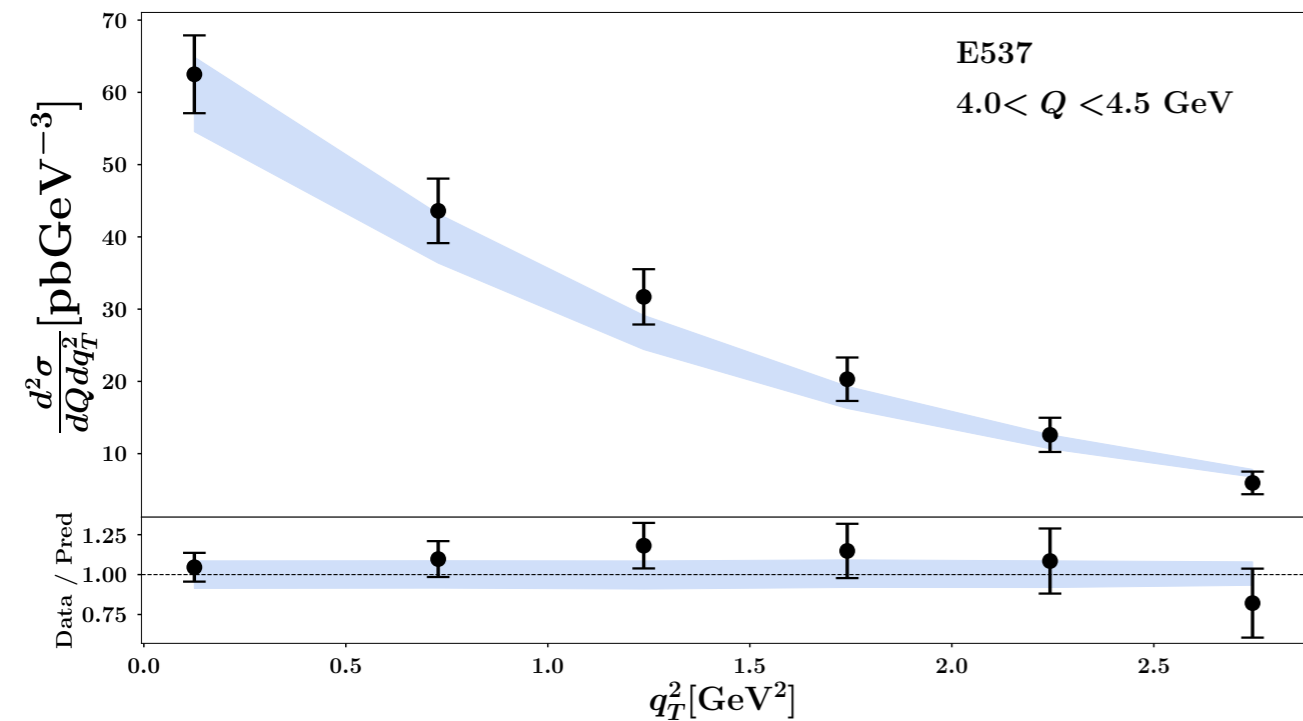
- 🍏 Much reduced data coverage as compared to proton data: only 138 data points.
- 🍏 Old fixed target data sets from FNAL.

- 🍏 Simpler functional form for f_{NP}^π :

$$f_{\text{NP}}^\pi(x, b_T, Q) = g_{1\pi}(x) e^{-g_{1C}(x) \frac{b_T^2}{4}} \left(\frac{Q}{Q_0^2} \right)^{\frac{g_K(b_T)}{2}} \quad g_{1\pi}(x) = N_{1\pi} \frac{x^{\sigma_\pi} (1-x)^{\alpha_\pi^2}}{\hat{x}^{\sigma_\pi} (1-\hat{x})^{\alpha_\pi^2}}$$

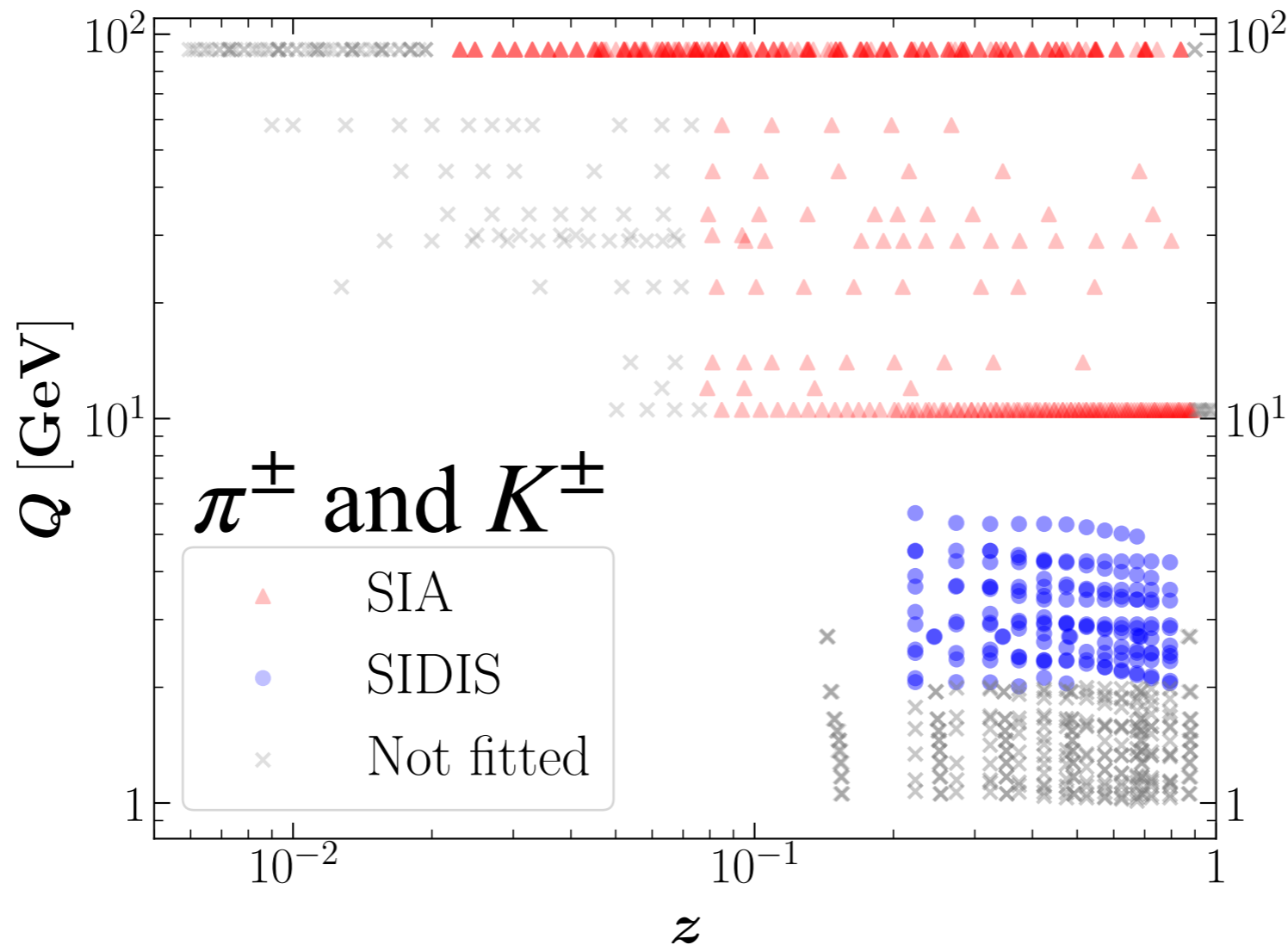
- 🍏 Only 3 free parameters.

TMD extractions (2)



Light-hadron FFs at NNLO

- 🍏 In [[Phys.Lett.B 834 \(2022\) 137456](#)] we have extracted π^\pm and K^\pm fragmentation functions (FFs) from a broad set of single-inclusive e^+e^- annihilation (SIA) and SIDIS data at NNLO accuracy (the first ever FF sets made public at this order).

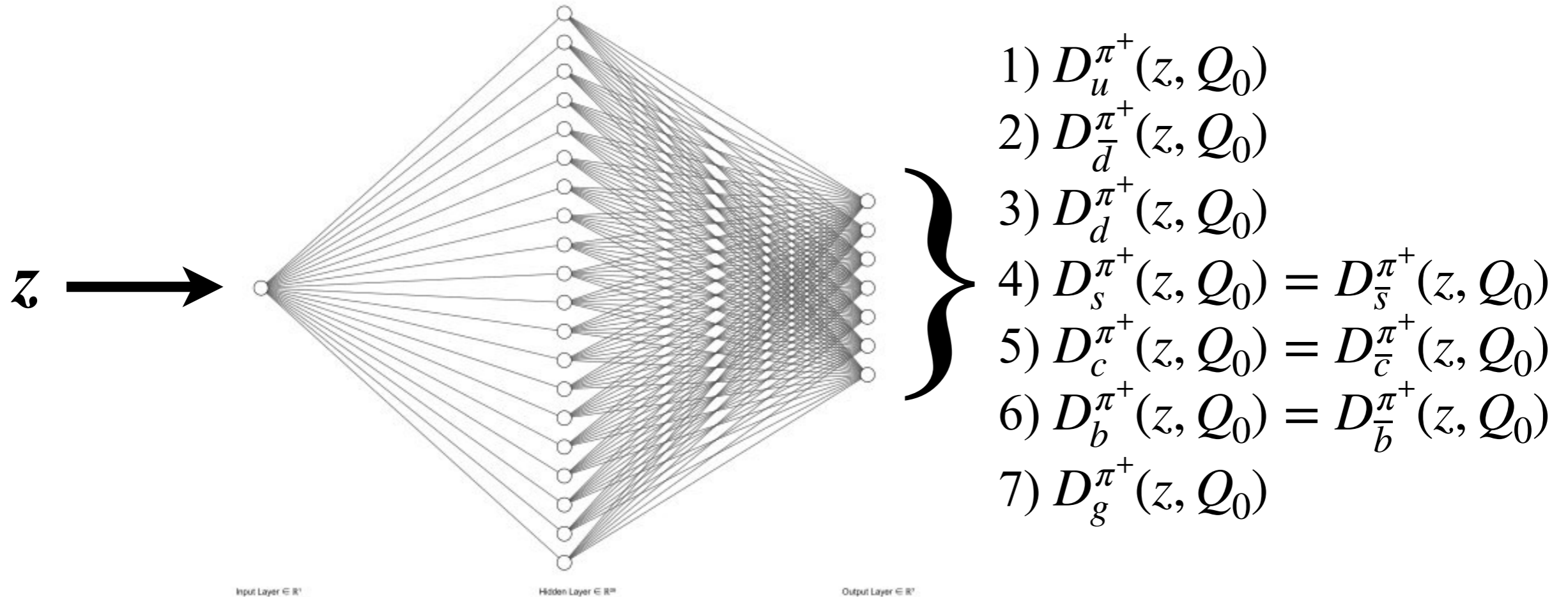


- 🍏 Around 700 data points for **SIA and SIDIS** for both pions and kaons.
- 🍏 **No pp data:** NNLO corrections not known yet.
- 🍏 **Wide coverage** in z and Q .
- 🍏 Extraction performed using the **MontBlanc** framework.

Light-hadron FFs at NNLO

🍏 **All** fitted FFs are parameterised using a **single** NN:

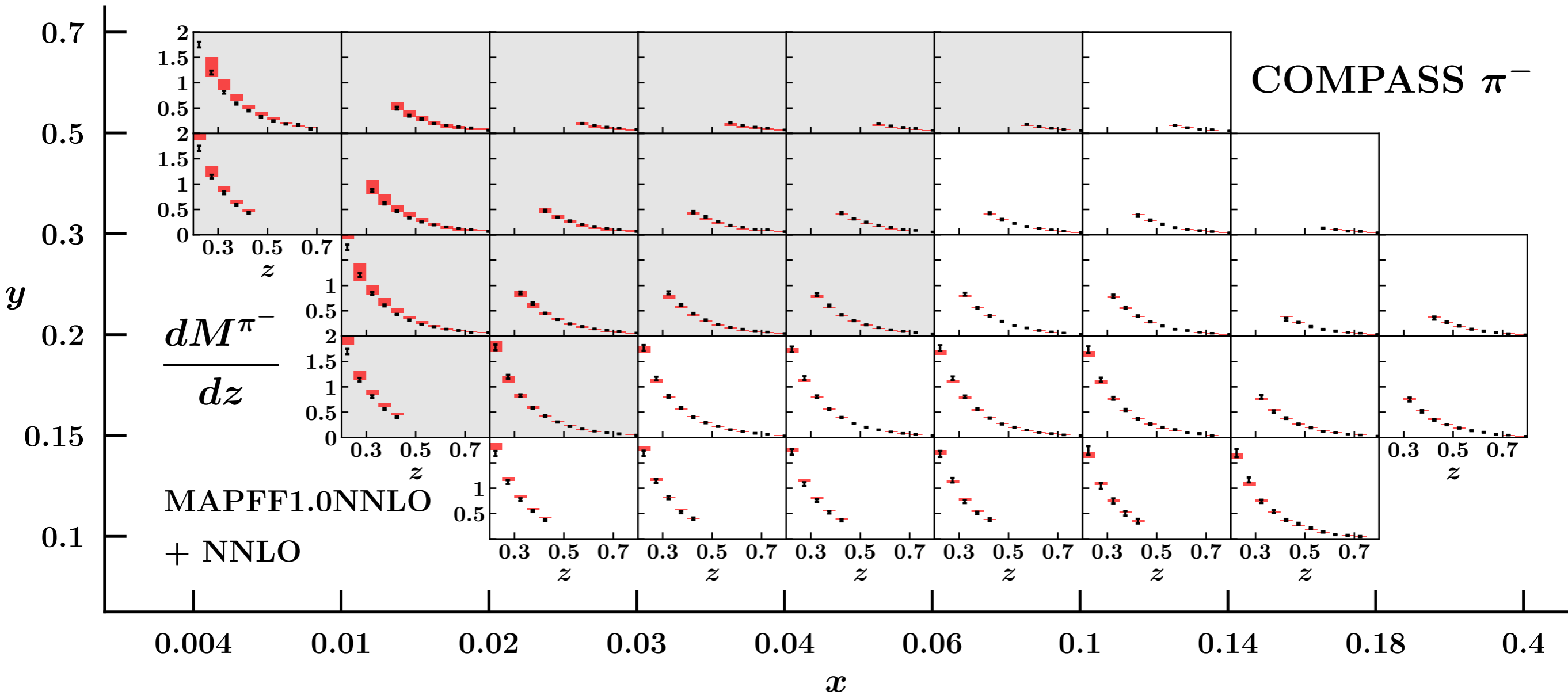
🍏 architecture 1-20-7 (187 free parameters).



🍏 We exploit the ability to compute the **analytic derivatives** of any NN w.r.t. its free parameters using the **NNAD** library. [R.Abdul Khalek, V. Bertone, arXiv:2005.07039]

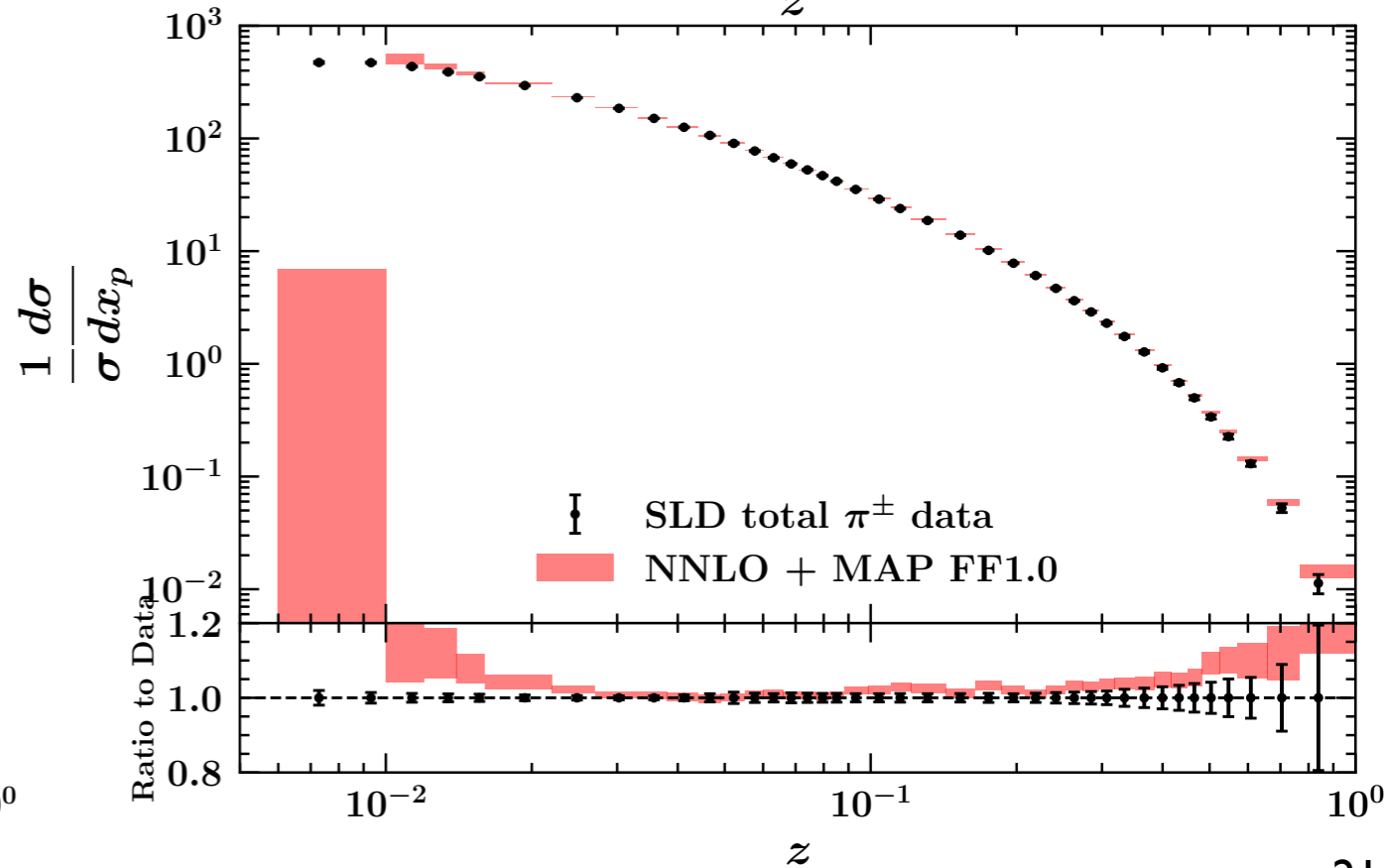
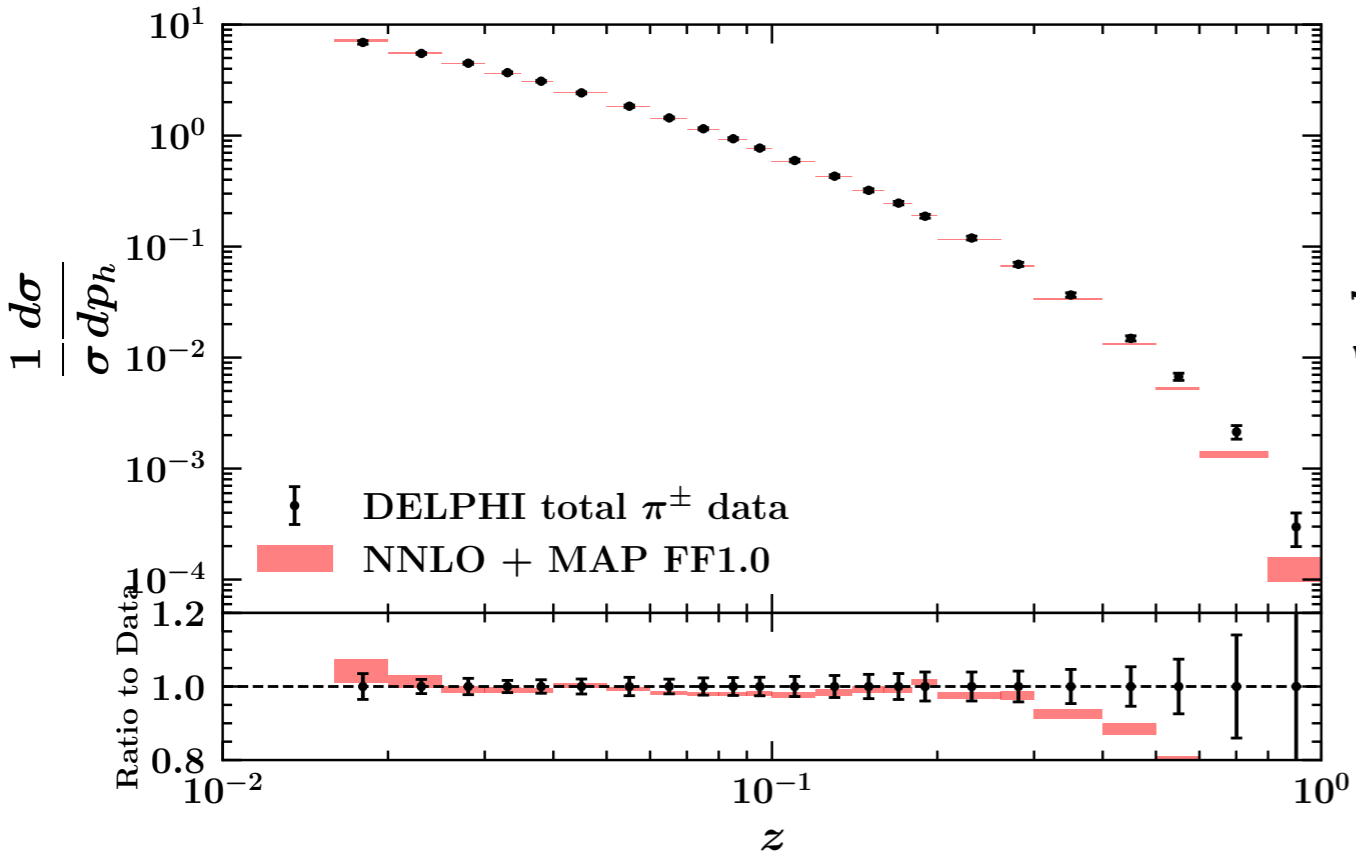
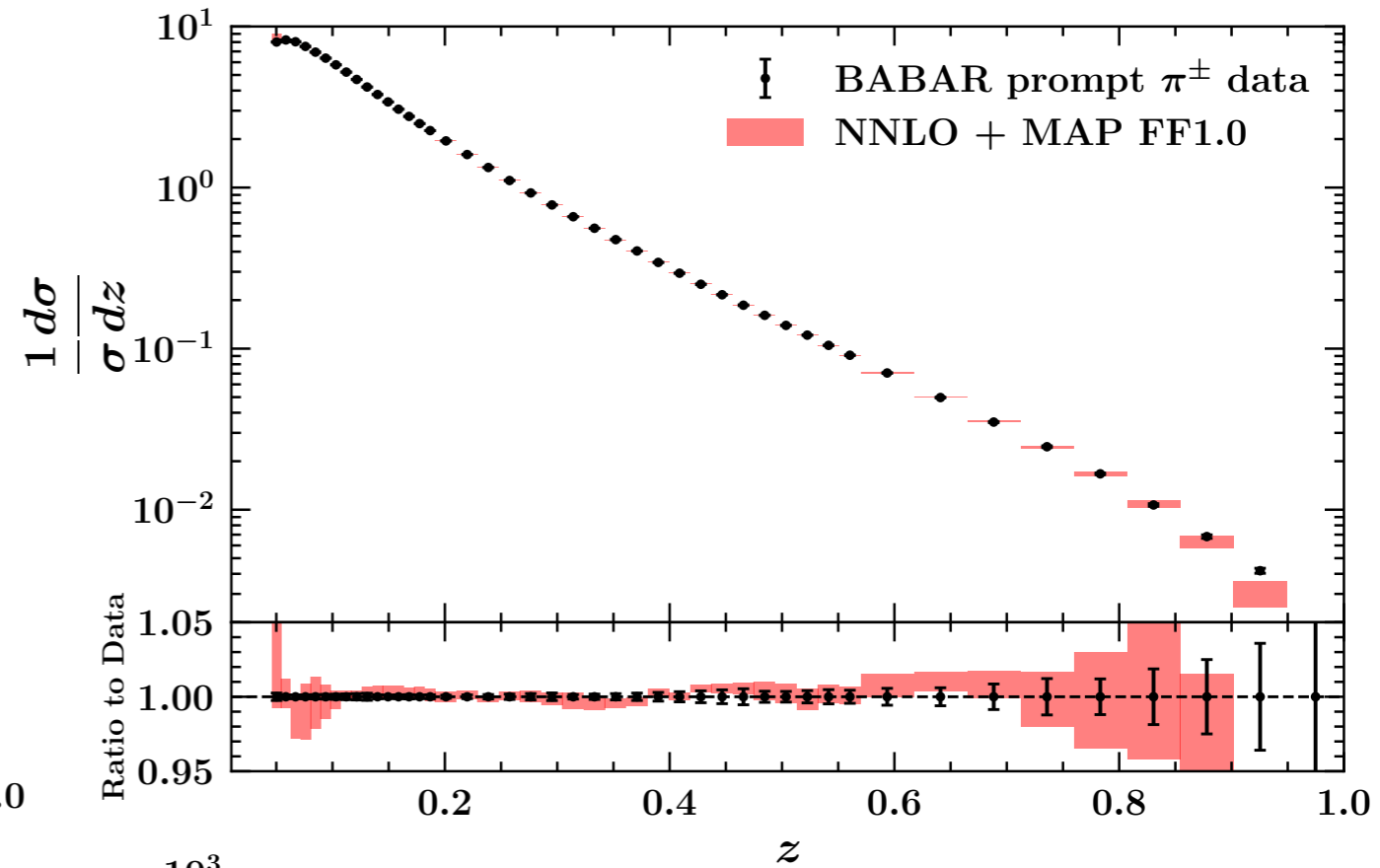
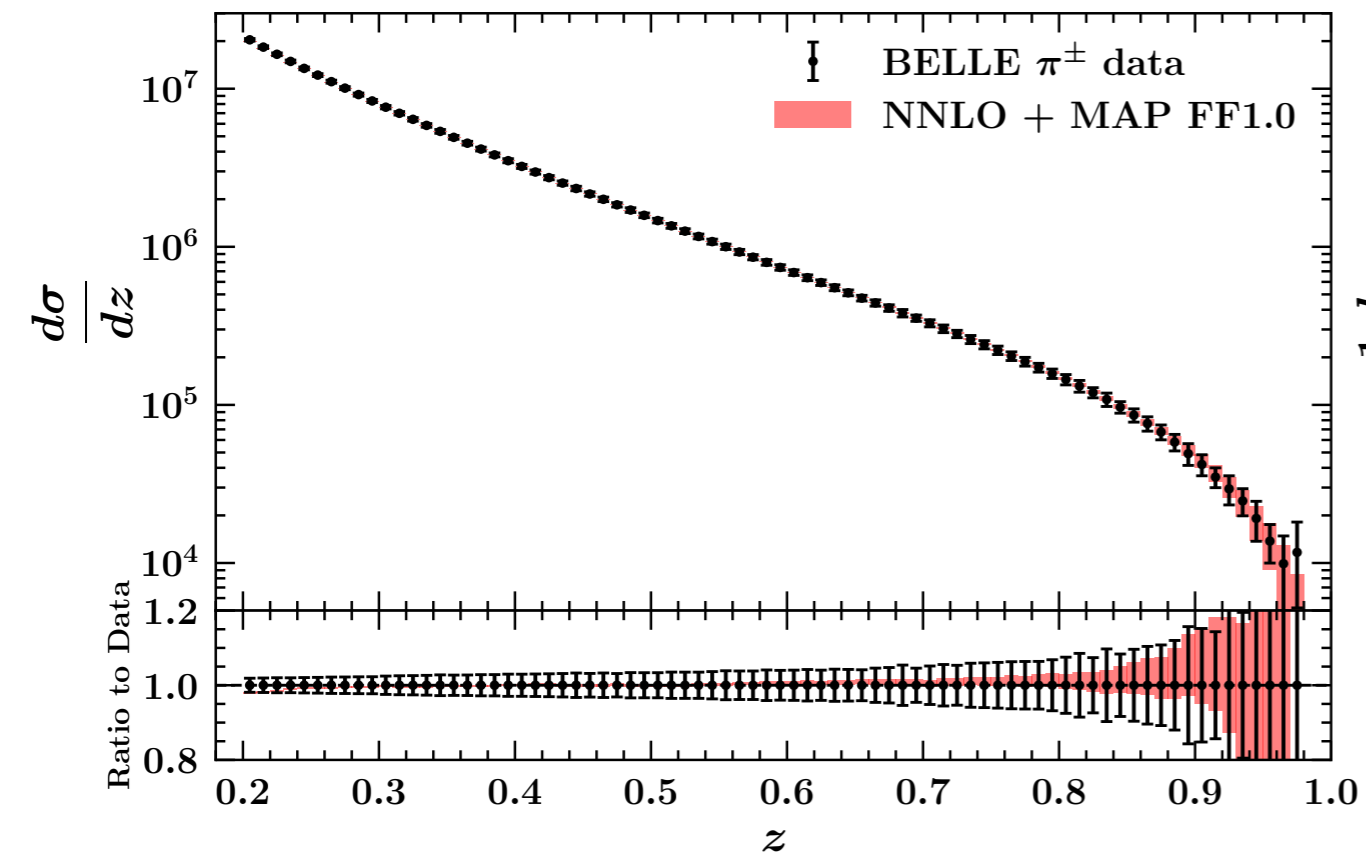
🍏 This enormously simplifies the task of the minimiser in that the gradient of the χ^2 can be computed analytically (as opposed to numerical or automatic derivatives).

Light-hadron FFs at NNLO



- Good description of the data included in the fit:
 - even for bins that are not included because of kinematic cuts.

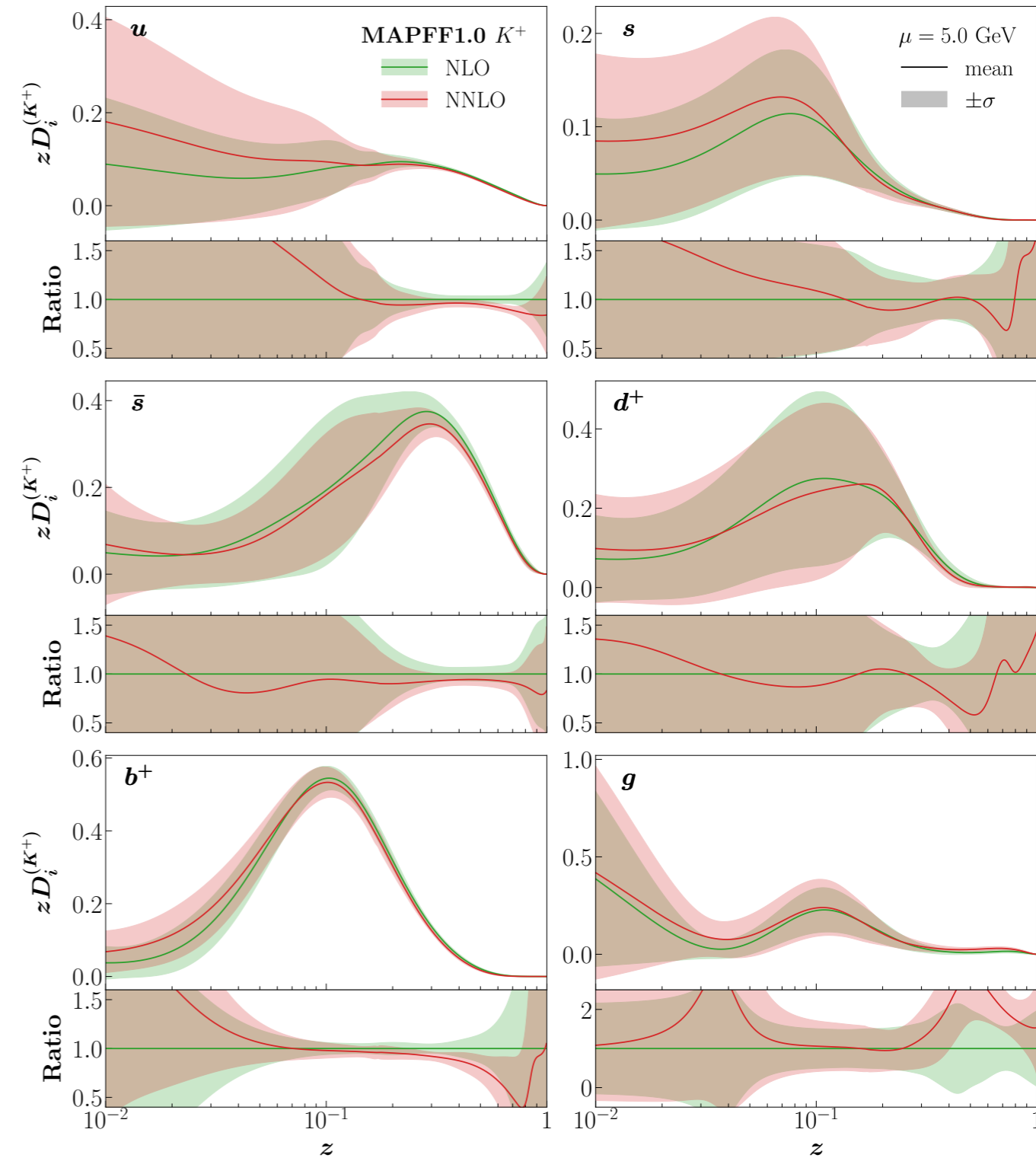
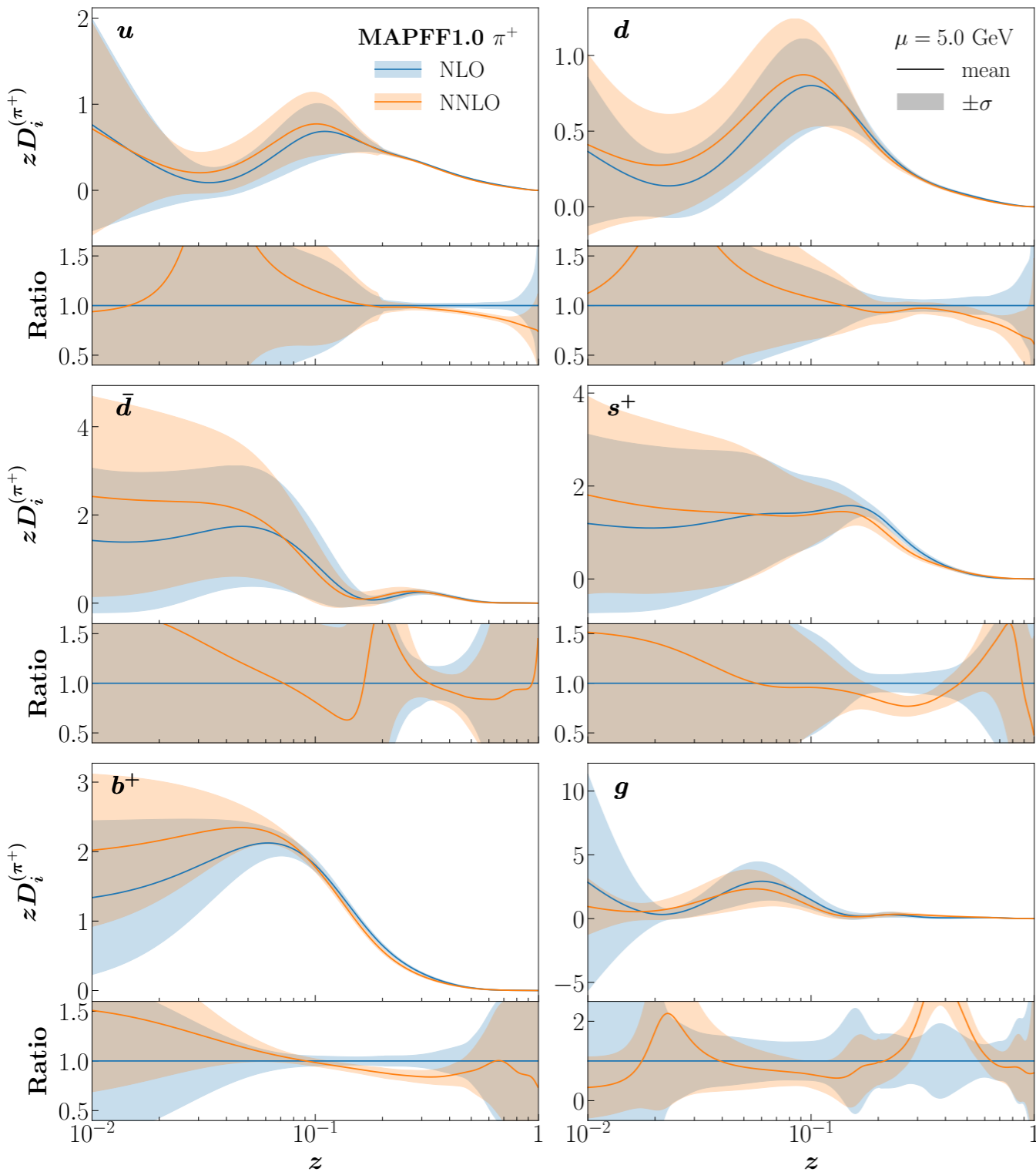
Light-hadron FFs at NNLO



Light-hadron FFs at NNLO

Pions

Kaons



Reconstructing GTMDs

🍏 In [\[arXiv:2207.09526\]](https://arxiv.org/abs/2207.09526) (recently accepted for publication in EPJC) I have computed the so-called GTMD matching functions at one-loop accuracy.

🍏 The unpolarised GTMD correlator can be decomposed as: [Meißner, Metz, Schlegel \[JHEP 08 \(2009\) 056\]](#)

$$\mathcal{F}_{i/H} = \frac{1}{2M} \bar{u}(P_{\text{out}}) \left[F_{1,1}^i + \frac{i\sigma^{\mathbf{k}_T n}}{n \cdot P} F_{1,2}^i + \frac{i\sigma^{\Delta_T n}}{n \cdot P} F_{1,3}^i + \frac{i\sigma^{\mathbf{k}_T \Delta_T}}{M^2} F_{1,4}^i \right] u(P_{\text{in}})$$

🍏 Each function $F_{1,l}^i$ is complex and can be decomposed into a real and an imaginary part:

$$F_{1,l}^i = F_{1,l}^{i,e} + iF_{1,l}^{i,o} \quad F_{1,l}^{i,e}, F_{1,l}^{i,o} \in \mathbb{R}$$

🍏 $F_{1,1}^{i,e}$ for $b_T \simeq 0$ and $Q \simeq 1/b_T$ is related to the GPDs H_j and E_j as follows:

$$F_{1,1}^{i,e}(x, \xi, b_T, t, Q) \Big|_{b_T \simeq 0} = \mathcal{C}_{i/j}(x, \xi, b_T, Q) \otimes_x \left[(1 - \xi^2) H_j(x, \xi, t, Q) - \xi^2 E_j(x, \xi, t, Q) \right]$$

🍏 Moreover, the forward limit of $F_{1,1}^{i,e}$ is the unpolarised TMD $f_{1,i}$:

$$\lim_{\xi, t \rightarrow 0} F_{1,1}^{i,e}(x, \xi, b_T, t, \mu, \zeta) = f_{1,i}(x, b_T, \mu, \zeta)$$

🍏 As for TMDs, the value of $F_{1,1}^{i,e}$ for any values of b_T and Q is achieved by introducing a non-perturbative function (f_{NP}) and solving appropriate evolution equations:

🍏 f_{NP} is (mostly) the same as that of TMDs,

🍏 also the evolution equations closely follow those for TMDs and can thus be solved analogously.

Reconstructing GTMDs

🍏 A numerical code to compute $F_{1,1}^{i,e}$ as briefly described above is public at:

<https://github.com/vbertone/GTMDMatching>

🍏 and is based on a combination VA2 public codes.

🍏 **PARTONS** for the handling of GPDs:

🍏 the Goloskokov-Kroll (GK) model for the GPDs H_j and E_j has been used.

🍏 **NangaParbat** for the handling of TMDs:

🍏 the PV19 [[JHEP 07 \(2020\) 117](#)] determination of f_{NP} along with the b_* function.

🍏 **APFEL++** is used for:

🍏 the numerical computation of the **convolutions**,

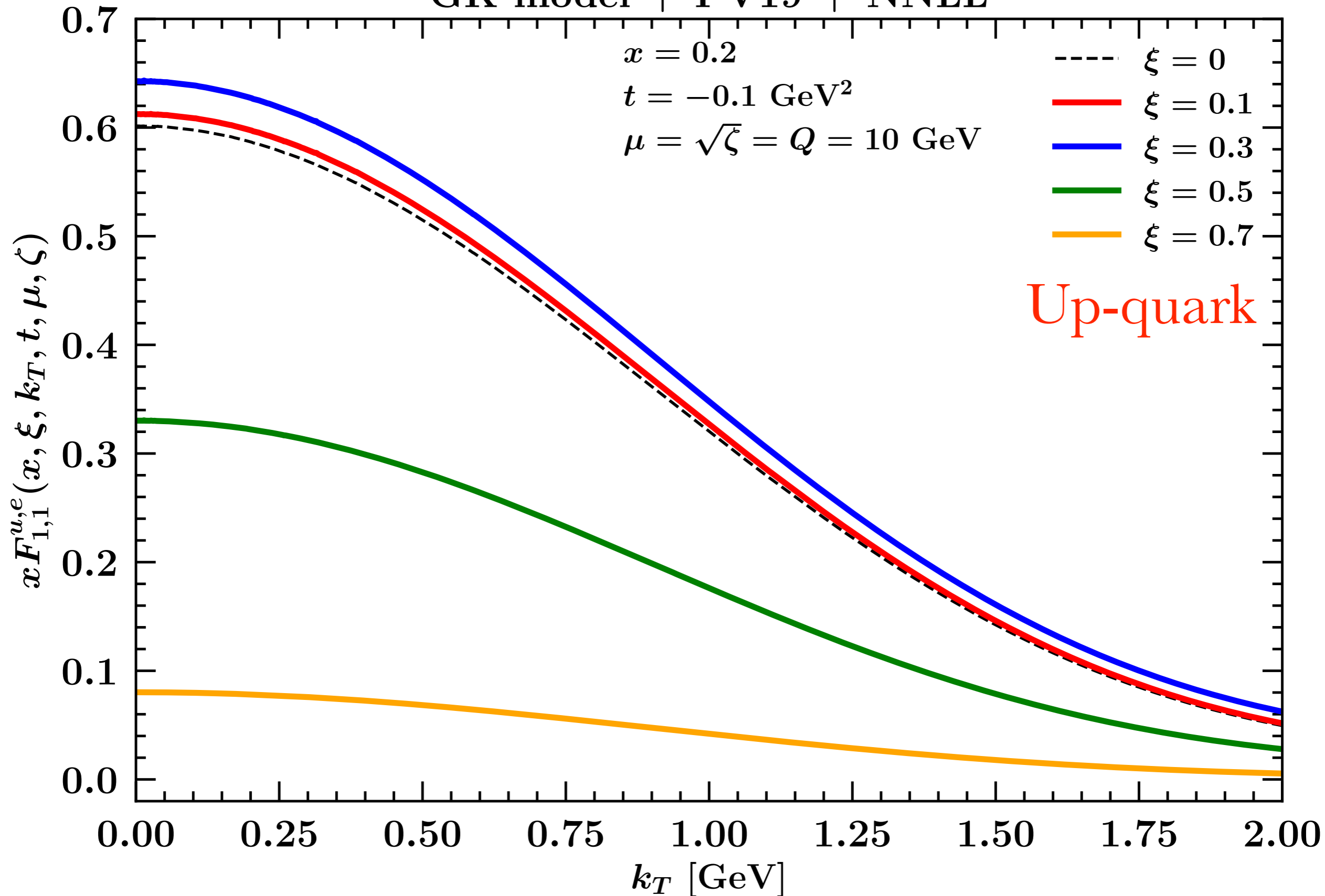
🍏 the **collinear evolution of GPDs**,

🍏 the computation of the **Sudakov form factor**,

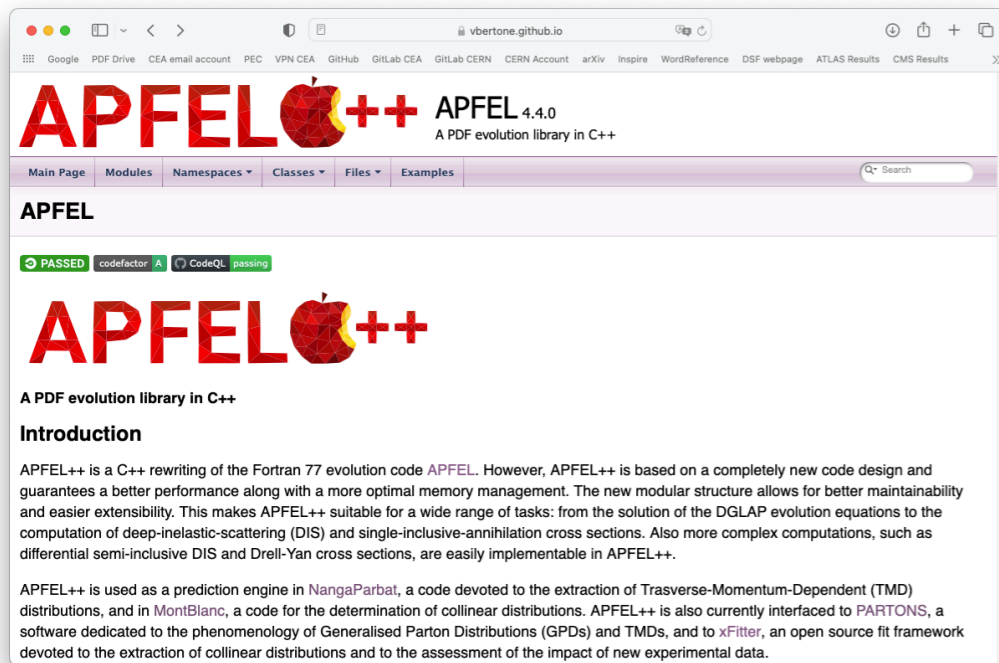
🍏 the **inverse Fourier transform**.

Reconstructing GTMDs

GK model + PV19 + NNLL



The VA2 tools



APFEL++ APFEL 4.4.0
A PDF evolution library in C++

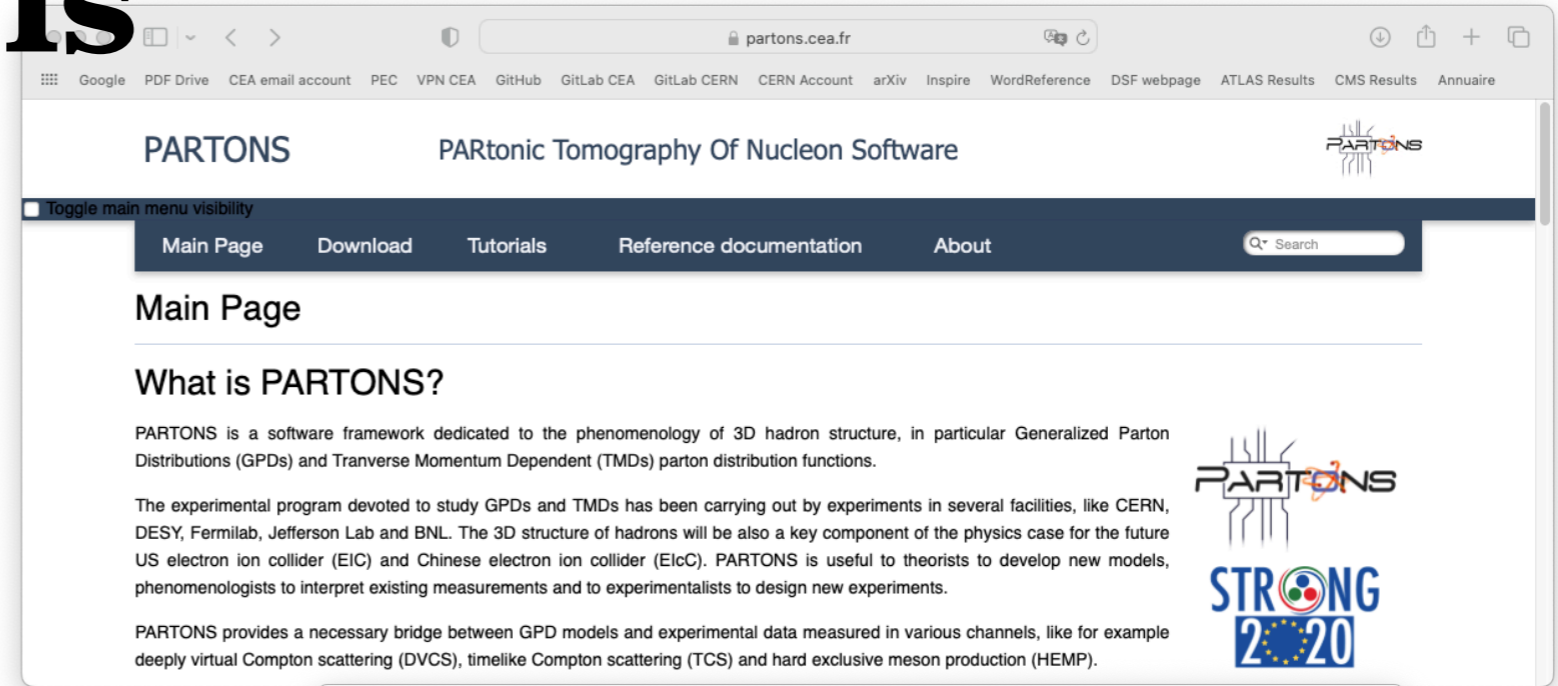
Navigation: Main Page, Modules, Namespaces, Classes, Files, Examples

Quality: PASSED, codefactor A, CodeQL passing

Introduction

APFEL++ is a C++ rewriting of the Fortran 77 evolution code APFEL. However, APFEL++ is based on a completely new code design and guarantees a better performance along with a more optimal memory management. The new modular structure allows for better maintainability and easier extensibility. This makes APFEL++ suitable for a wide range of tasks: from the solution of the DGLAP evolution equations to the computation of deep-inelastic-scattering (DIS) and single-inclusive-annihilation cross sections. Also more complex computations, such as differential semi-inclusive DIS and Drell-Yan cross sections, are easily implementable in APFEL++.

APFEL++ is used as a prediction engine in NangaParbat, a code devoted to the extraction of Transverse-Momentum-Dependent (TMD) distributions, and in MontBlanc, a code for the determination of collinear distributions. APFEL++ is also currently interfaced to PARTONS, a software dedicated to the phenomenology of Generalised Parton Distributions (GPDs) and TMDs, and to xFitter, an open source fit framework devoted to the extraction of collinear distributions and to the assessment of the impact of new experimental data.



PARTONS PARTonic Tomography Of Nucleon Software

Navigation: Main Page, Download, Tutorials, Reference documentation, About

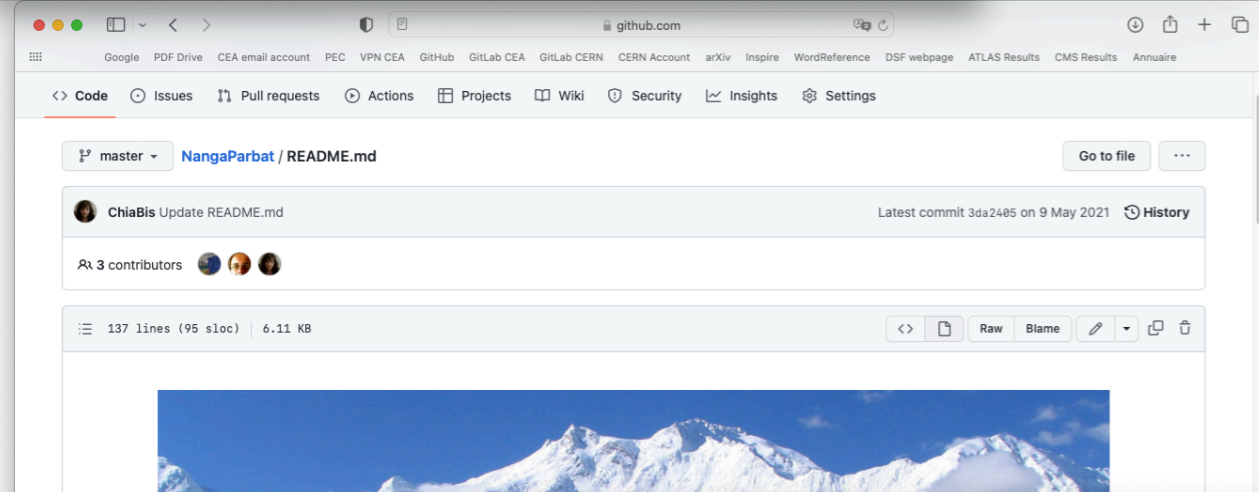

Main Page

What is PARTONS?

PARTONS is a software framework dedicated to the phenomenology of 3D hadron structure, in particular Generalized Parton Distributions (GPDs) and Transverse Momentum Dependent (TMDs) parton distribution functions.

The experimental program devoted to study GPDs and TMDs has been carrying out by experiments in several facilities, like CERN, DESY, Fermilab, Jefferson Lab and BNL. The 3D structure of hadrons will be also a key component of the physics case for the future US electron ion collider (EIC) and Chinese electron ion collider (EicC). PARTONS is useful to theorists to develop new models, phenomenologists to interpret existing measurements and to experimentalists to design new experiments.

PARTONS provides a necessary bridge between GPD models and experimental data measured in various channels, like for example deeply virtual Compton scattering (DVCS), timelike Compton scattering (TCS) and hard exclusive meson production (HEMP).

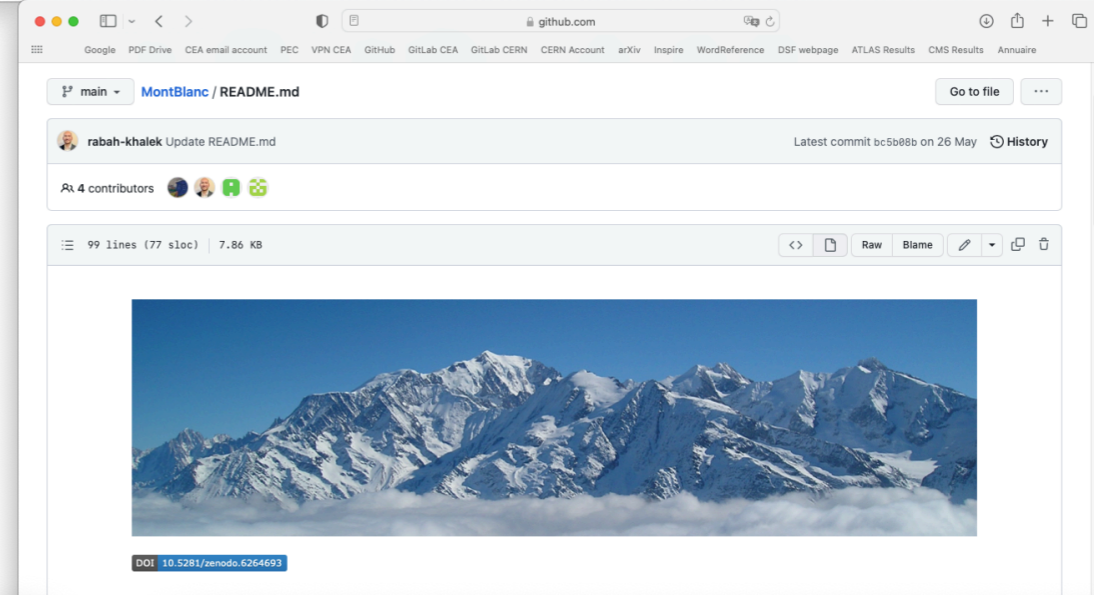
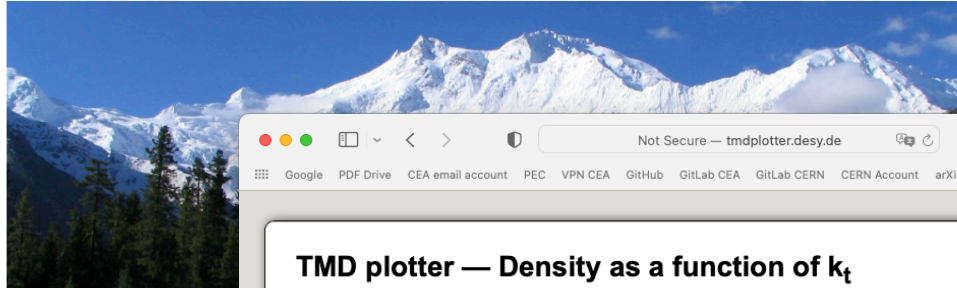


GitHub repository: NangaParbat / README.md

Latest commit: ChiaBis Update README.md on 9 May 2021

Contributors: 3

137 lines (95 sloc) | 6.11 KB

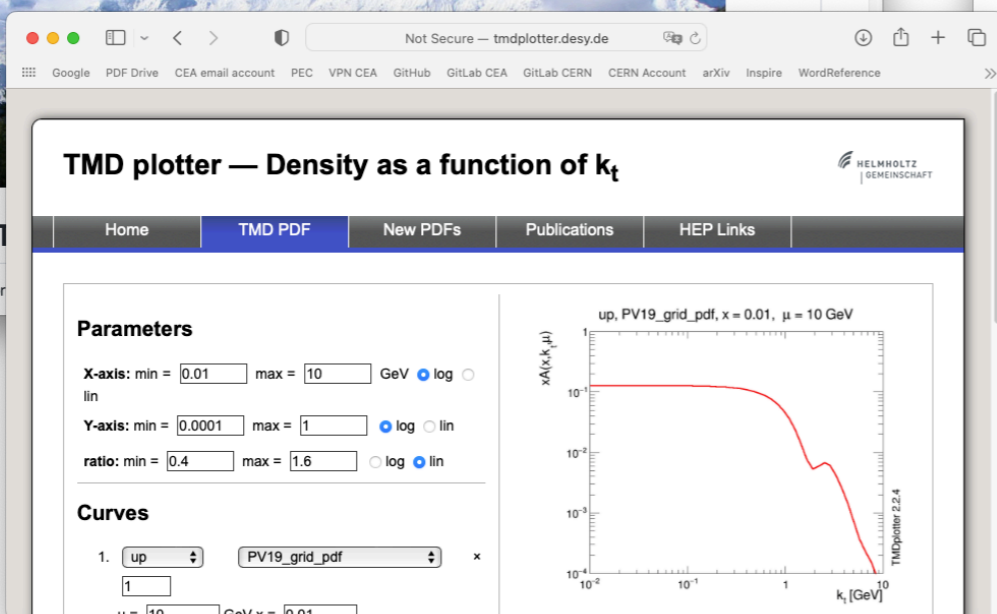



GitHub repository: MontBlanc / README.md

Latest commit: rabah-khalek Update README.md on 26 May

Contributors: 4

99 lines (77 sloc) | 7.86 KB



TMD plotter — Density as a function of k_t

Navigation: Home, TMD PDF, New PDFs, Publications, HEP Links

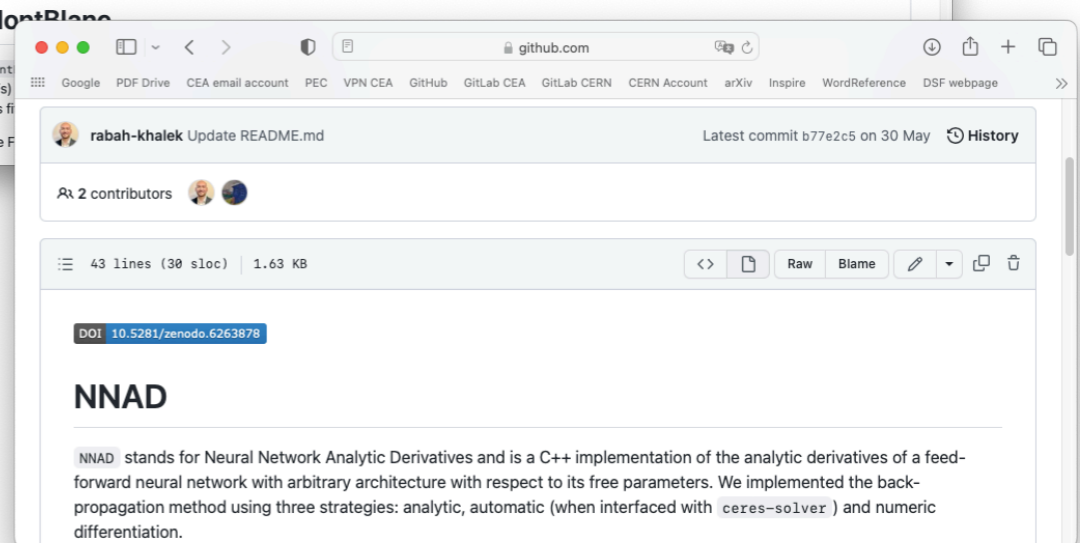
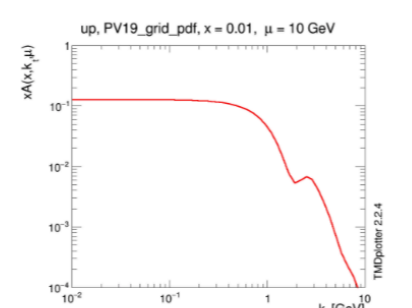
Parameters:

- X-axis: min = 0.01, max = 10 GeV (log)
- lin
- Y-axis: min = 0.0001, max = 1 (log)
- lin
- ratio: min = 0.4, max = 1.6 (log)
- lin

Curves:

- up, PV19_grid_pdf

$\mu = 10$ GeV, $x = 0.01$



GitHub repository: NNAD / README.md

Latest commit: rabah-khalek Update README.md on 30 May

Contributors: 2

43 lines (30 sloc) | 1.63 KB

NNAD

NNAD stands for Neural Network Analytic Derivatives and is a C++ implementation of the analytic derivatives of a feed-forward neural network with arbitrary architecture with respect to its free parameters. We implemented the back-propagation method using three strategies: analytic, automatic (when interfaced with `ceres-solver`) and numeric differentiation.

Conclusions

- 🍏 The VA2 work package (3DPartons) has worked in full swing:
 - 🍏 many relevant physics results (only a selection involving myself mentioned here),
 - 🍏 much numerical and open-source infrastructure has been developed,
 - 🍏 existing codes are being interfaced to create a seamless framework (see the example of PARTONS - APFEL - LHAPDF),
 - 🍏 most of the involved codes are written in C++, that guarantees performance, modularity, and maintainability, also python wrappers are being developed
 - 🍏 APFEL and LHAPDF have a python interface and one is being developed also for PARTONS.
- 🍏 These developments are having a tangible impact on different experimental physics programmes:
 - 🍏 the determination of the W mass at the LHC,
 - 🍏 the preparatory work in view of the EIC,
 - 🍏 physics at JLab,
 - 🍏 ...

Back up

Numerical setup

- 🍏 The evolution kernels for *unpolarised* evolution that we have recomputed are now implemented in **APFEL++** and available through **PARTONS** allowing for LO GPD evolution in momentum space.
- 🍏 The remarkable properties of the evolution kernels allowed us to obtain for the first time a stable numerical implementation over the full range $\mathbf{0 \leq \xi \leq 1}$:
 - 🍏 first numerical check that both the **DGLAP** and **ERBL** limits are recovered,
 - 🍏 first numerical check of **polynomiality** conservation.
- 🍏 Numerical tests mostly use the MMHT14 PDF set at LO as an initial-scale set of distributions evolved from 1 to 10 GeV for the first time in the **variable-flavour-number scheme**, *i.e.* accounting for heavy-quark-threshold crossing.
- 🍏 Tests have also been performed using more realistic GPD models such as the Goloskokov-Kroll model [*Eur.Phys.J.C* 53 (2008) 367-384] based on the Radyushkin double-distribution ansatz [*Phys.Lett.B* 449 (1999) 81-88].

The ERBL limit

🍏 The limit $\xi \rightarrow 1$ ($\kappa \rightarrow 1/x$) we should reproduce the **ERBL equation**.

🍏 It is well known that in this limit **Gegenbauer polynomials** decouple upon LO evolution, such that:

$$F_{2n}(x, \mu_0) = (1 - x^2) C_{2n}^{(3/2)}(x) \quad \Rightarrow \quad F_{2n}(x, \mu) = \exp \left[\frac{V_{2n}^{[0]}}{4\pi} \int_{\mu_0}^{\mu} d \ln \mu^2 \alpha_s(\mu) \right] F_{2n}(x, \mu_0)$$

🍏 where the kernels $V_{2n}^{[0]}$ can be read off, for example, from [Brodsky, Lepage, *Phys.Rev.D* 22 (1980) 2157] or [Efremov, Radyushkin, *Phys.Lett.B* 94 (1980) 245-250].

🍏 We have compared this expectation with the numerical results for GPD evolution by setting $\kappa = 1/x$ and using a Gegenbauer polynomial as an initial-scale GPD.

Conformal-space evolution

🍏 In order to check that LO GPD evolution ($\xi \neq 0$) in conformal space is diagonal in a **realistic** case, we have considered the RDDA:

$$H_q(x, \xi, \mu_0) = \int_{\Omega} d\beta d\alpha \delta(x - \beta - \xi\alpha) q(|\beta|) \pi(\beta, \alpha)$$

with:

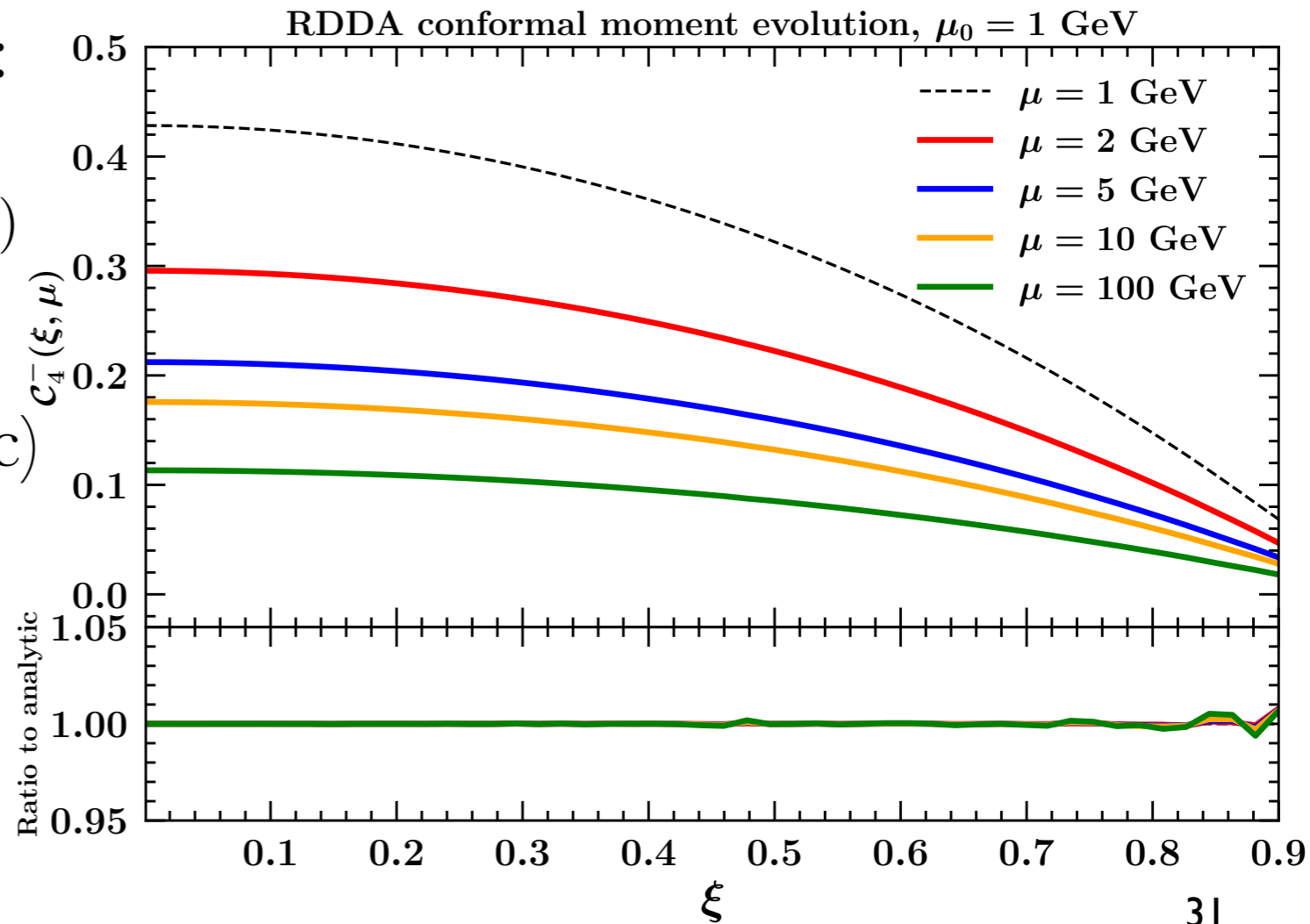
$$q(x) = \frac{35}{32} x^{-1/2} (1-x)^3, \quad \pi(\beta, \alpha) = \frac{3}{4} \frac{((1-|\beta|)^2 - \alpha^2)}{(1-|\beta|)^3}$$

We have evolved the 4th moment:

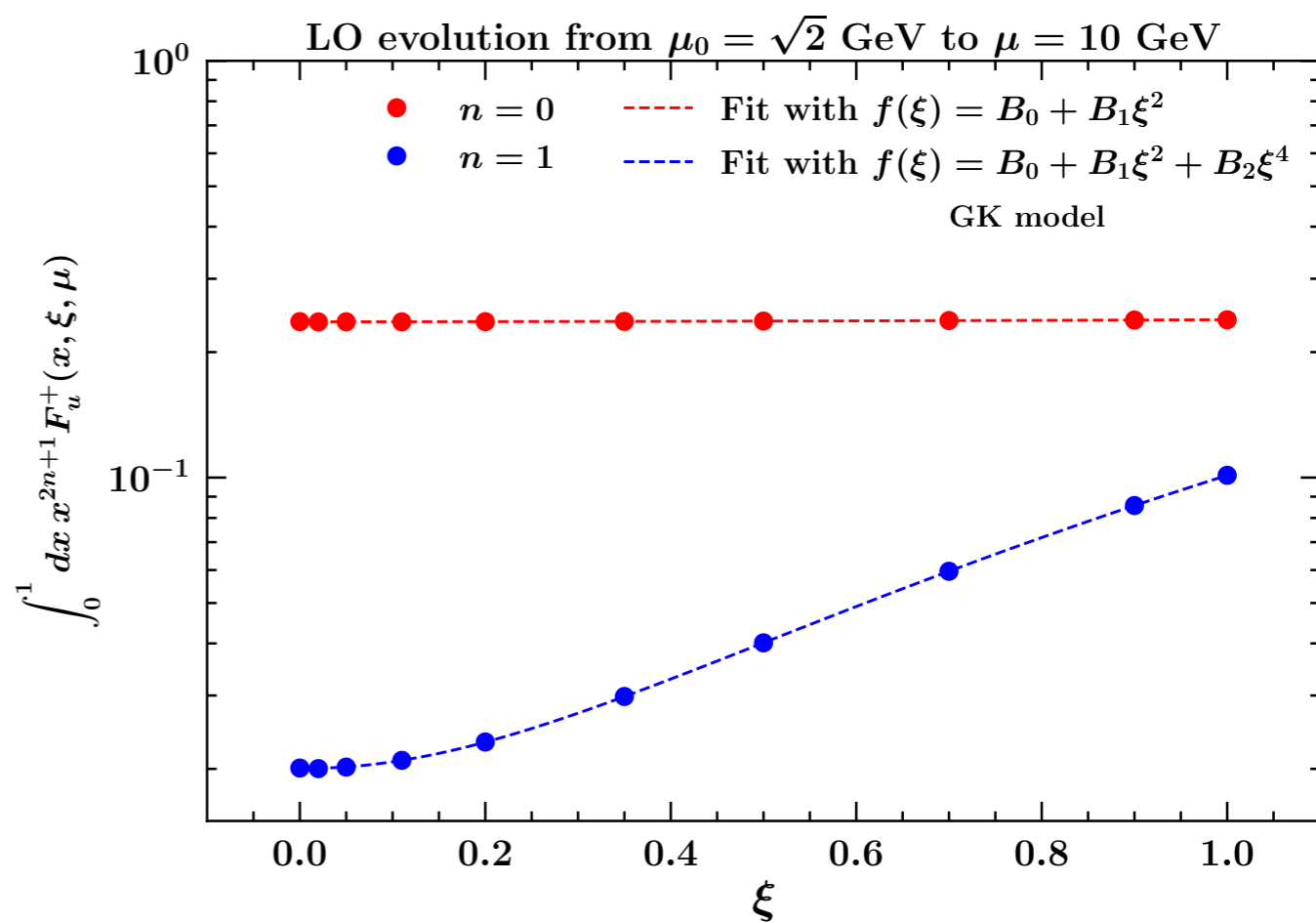
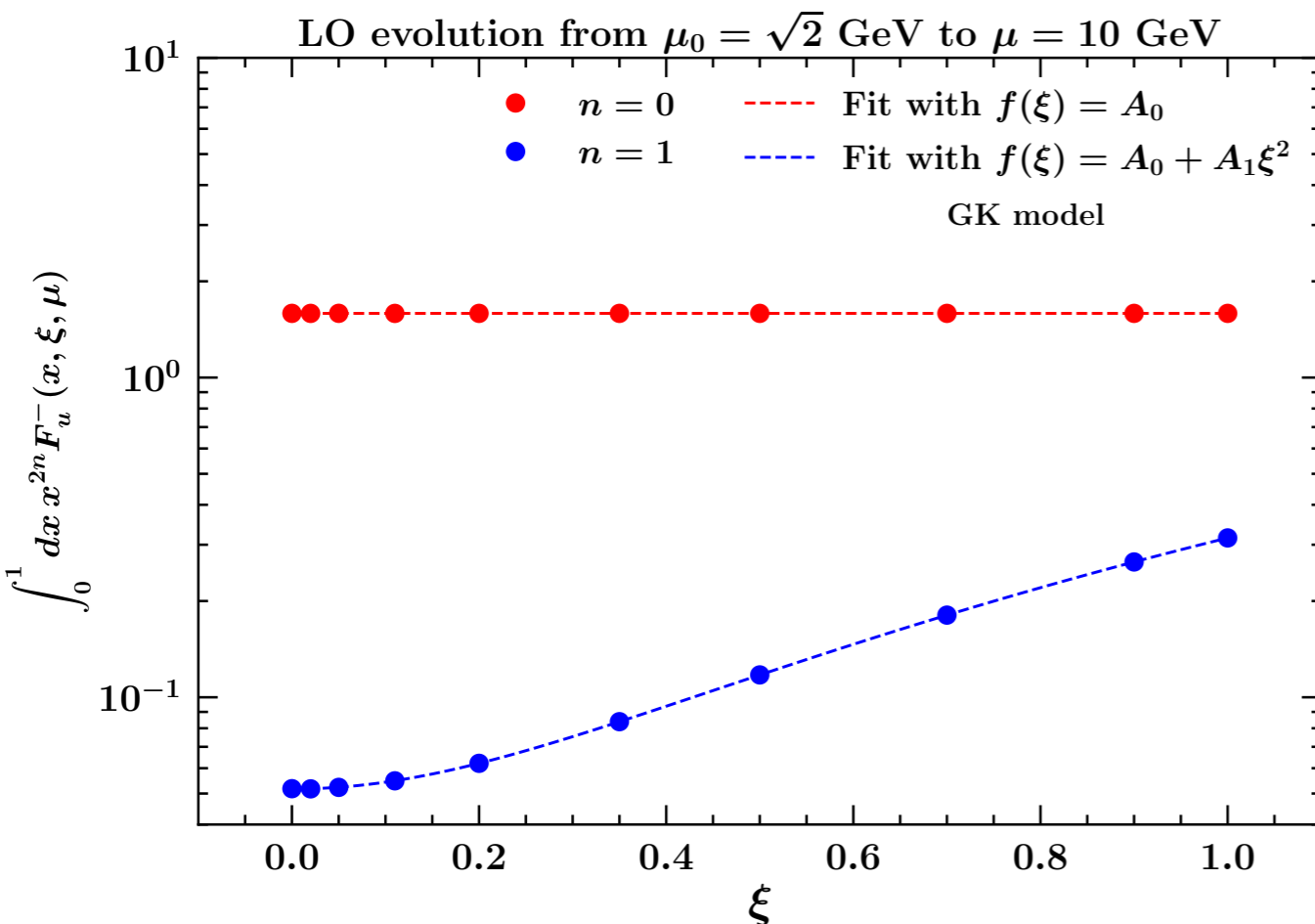
$$C_4^-(\xi, \mu) = \xi^4 \int_{-1}^1 dx C_4^{(3/2)}\left(\frac{x}{\xi}\right) H_q(x, \xi, \mu)$$

from $\mu_0 = 1$ GeV using the (analytic) conformal-space evolution and the (numerical) momentum-space evolution.

we found excellent agreement.



Polynomiality



🍏 **First moment** for both singlet and non-singlet is indeed **constant** in ξ :

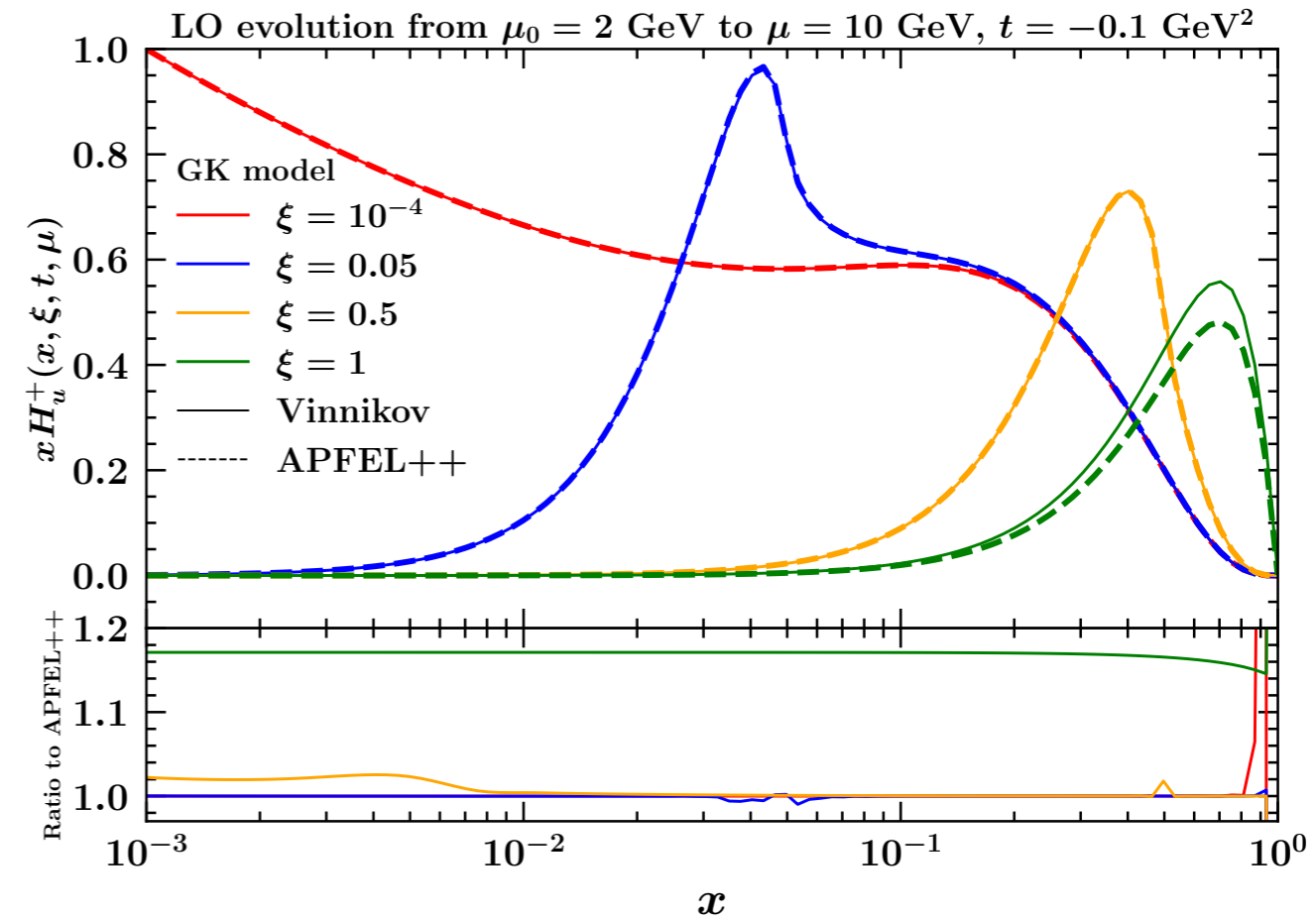
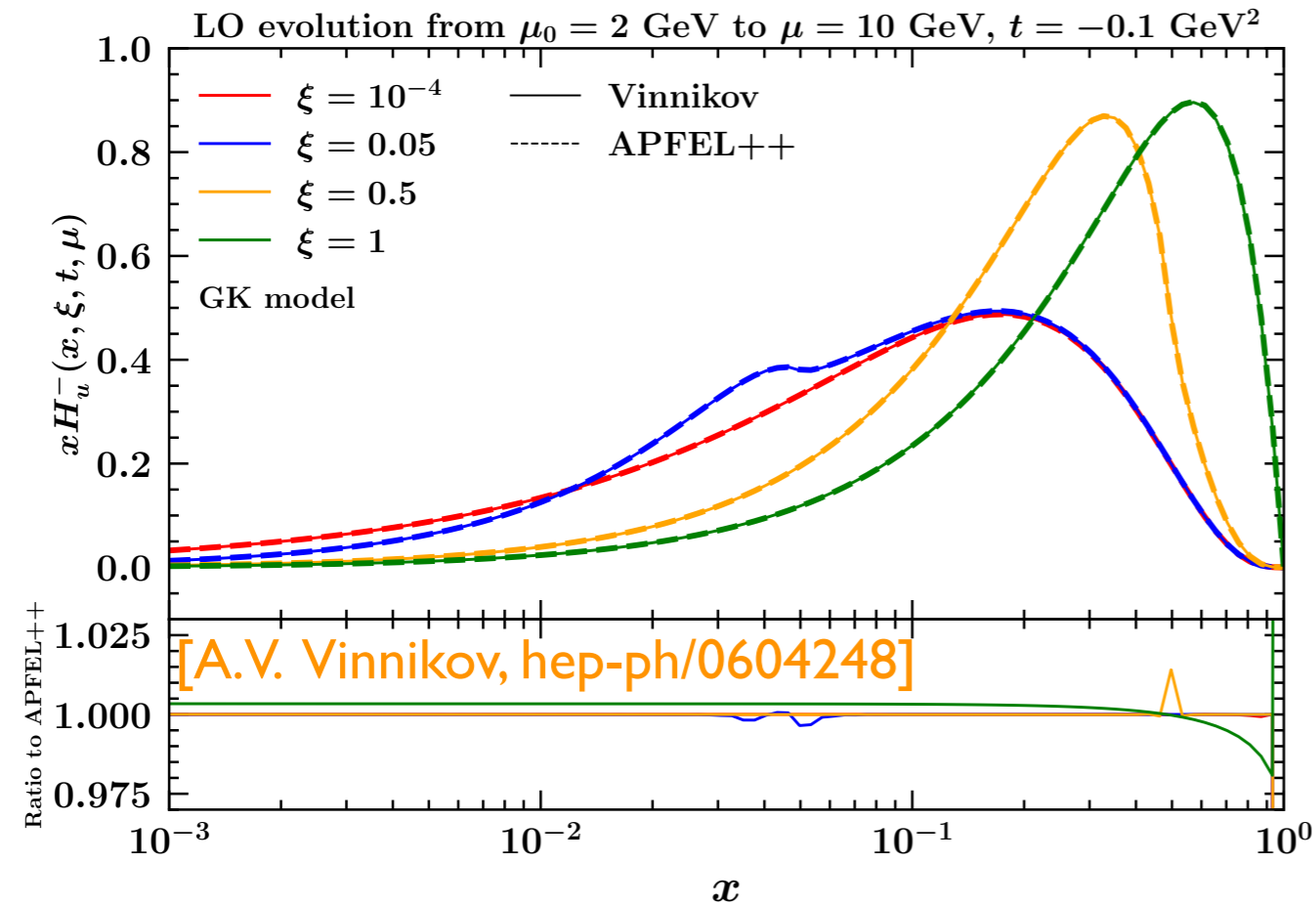
🍏 this was expected and the expectation is very nicely fulfilled.

🍏 **Second and third moments** follow the expected law:

🍏 including odd-power terms in the fit gives coefficients very close to zero.

🍏 B_{n+1} in the singlet is consistently found to be compatible with zero (no D-term)₃₂

APFEL vs. Vinnikov's code



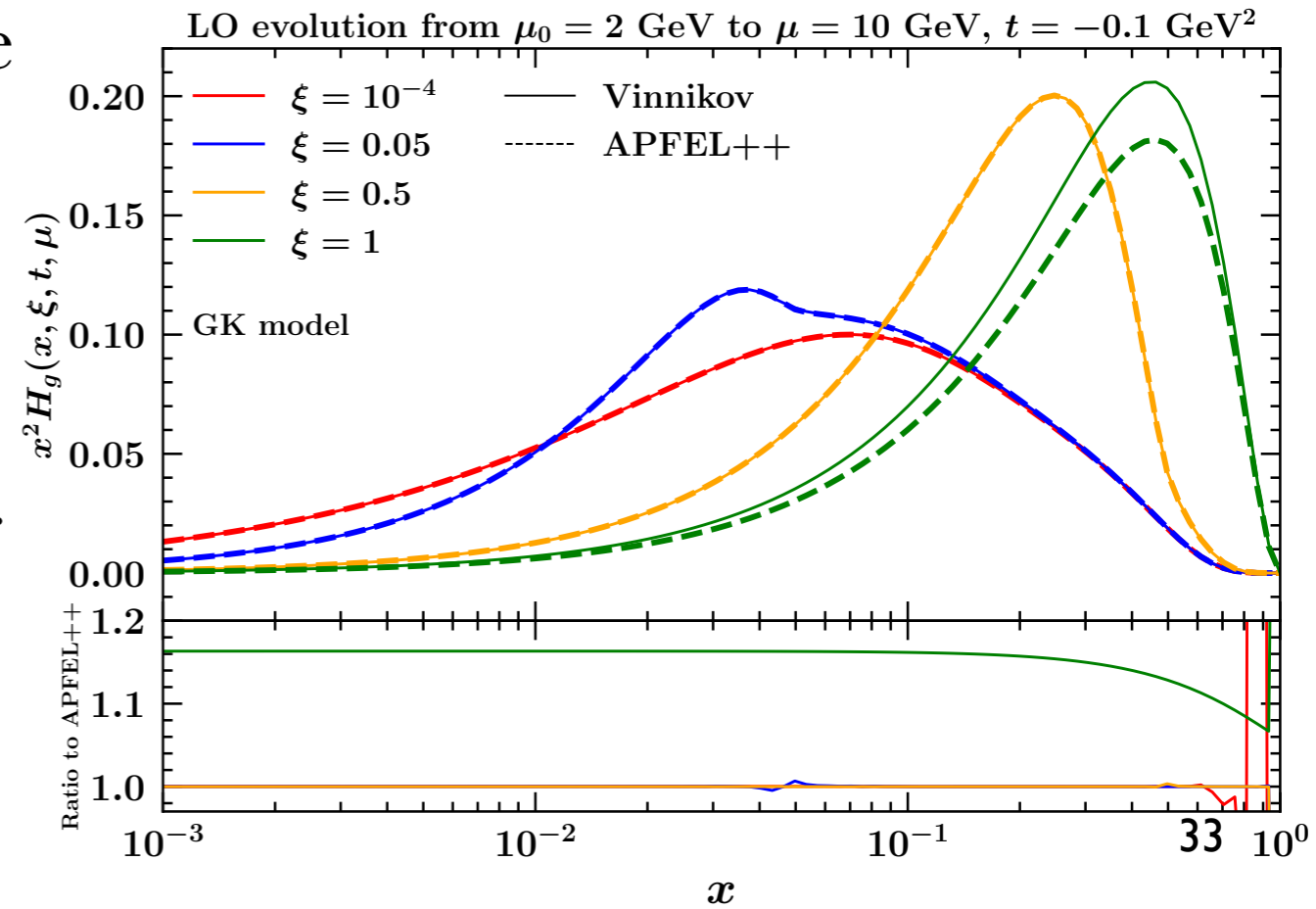
🍏 **Excellent agreement** between the two code for $\xi \lesssim 0.6$.

🍏 Agreement deteriorates for $\xi \gtrsim 0.6$:

🍏 discrepancy larger for the singlets ($\sim 20\%$) than for the non-singlet ($\sim 1\%$).

🍏 possible numerical instabilities of Vinnikov's code?

🍏 Inability to check the ERBL limit.



Polynomiality

🍏 GPD evolution should preserve **polynomiality**.

[Xiang-Dong Ji, *J.Phys.G* 24 (1998) 1181-1205] [A.V. Radyushkin, *Phys.Lett.B* 449 (1999) 81-88]

🍏 The following relations for the Mellin moments must hold at **all scales**:

$$\int_0^1 dx x^{2n} F_q^-(x, \xi, \mu) = \sum_{k=0}^n A_k(\mu) \xi^{2k}$$

$$\int_0^1 dx x^{2n+1} F_q^+(x, \xi, \mu) = \sum_{k=0}^{n+1} B_k(\mu) \xi^{2k}$$

🍏 Polynomiality predicts that the first moment ($n = 0$) of the *non-singlet* distribution is **constant** in ξ .

🍏 The coefficient of the ξ^{2n+2} term of the *singlet* (D-term) is absent in our initial conditions and it is *not* generated by evolution, so that also the first moment of the singlet is expected to be **constant** in ξ .

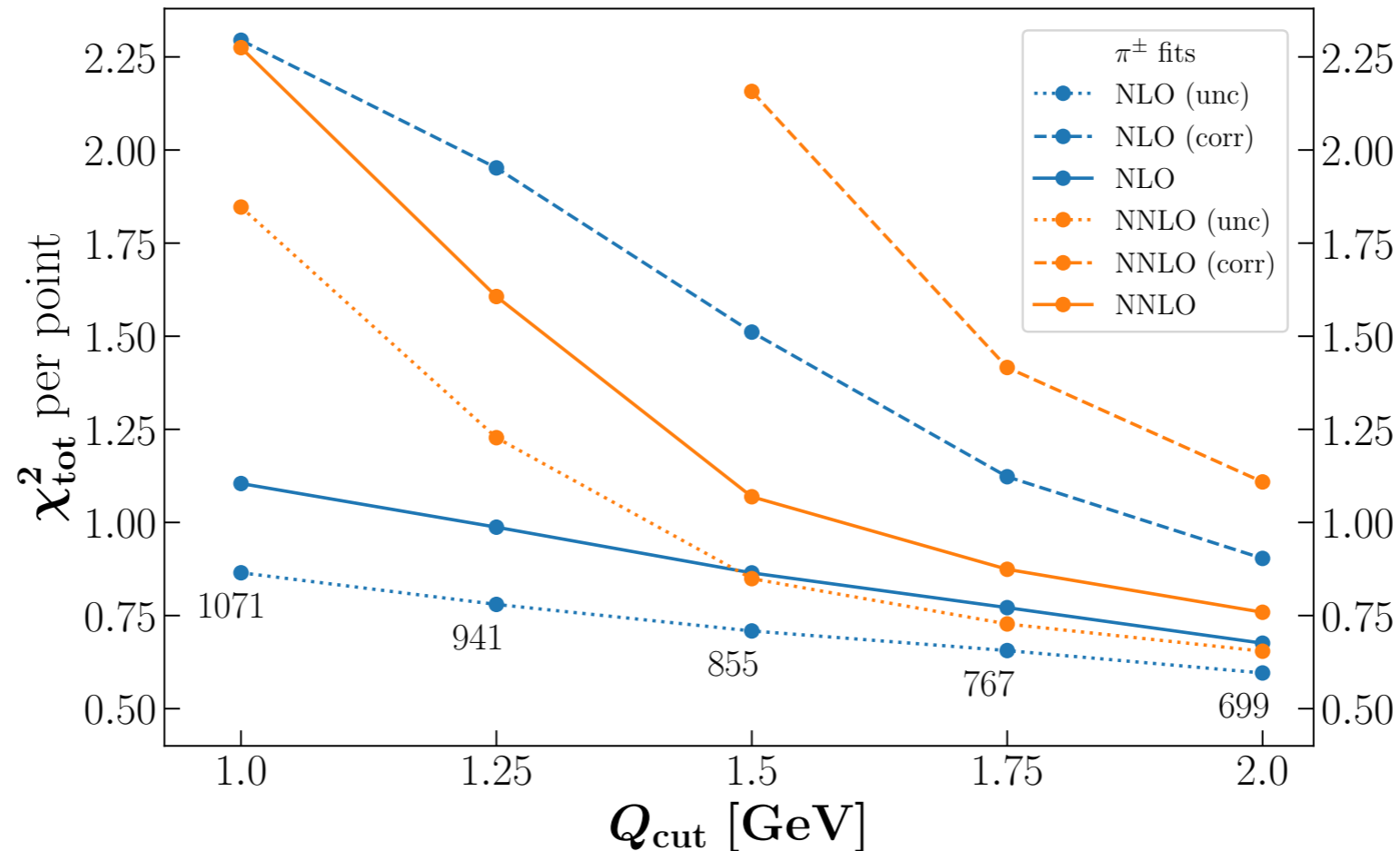
🍏 For the other values of n one can just **fit** the behaviour in ξ and check that it follows the **expected power law**.

FFs at NNLO

NLO vs. NNLO

- While both MAPFF1.0 and BDSS confirm that COMPASS high- Q data is better described by NNLO, it is not clear as yet where NNLO starts doing better than NLO.

Bertone *et al.* [arXiv:2204.10331]



Borsa *et al.* [arXiv:2202.05060]

Experiment	$Q^2 \geq 1.5 \text{ GeV}^2$			$Q^2 \geq 2.0 \text{ GeV}^2$			$Q^2 \geq 2.3 \text{ GeV}^2$			$Q^2 \geq 3.0 \text{ GeV}^2$		
	#data	NLO	NNLO	#data	NLO	NNLO	#data	NLO	NNLO	#data	NLO	NNLO
SIA	288	1.05	0.96	288	0.91	0.87	288	0.90	0.91	288	0.93	0.86
COMPASS	510	0.98	1.14	456	0.91	1.04	446	0.91	0.92	376	0.94	0.93
HERMES	224	2.24	2.27	160	2.40	2.08	128	2.71	2.35	96	2.75	2.26
TOTAL	1022	1.27	1.33	904	1.17	1.17	862	1.17	1.13	760	1.16	1.07

Reconstructing GTMDs

- 🍏 We can evolve $F_{1,1}^{i,e}$ to *any* scale by solving the evolution equations:
- 🍏 $\mathcal{O}(\alpha_s)$ matching functions allow us to reach **NNLL accuracy**. Anomalous dimensions (that coincide with the TMD ones) need to be evaluated accordingly.
- 🍏 Extrapolation to large $|\mathbf{k}_T|$ is obtained *a la* CSS, *i.e.* by means of a b_* prescription:

$$b_*(b_T) = \frac{b_0}{Q} \left(\frac{1 - \exp\left(-\frac{b_T^4 Q^4}{b_0^4}\right)}{1 - \exp\left(-\frac{b_T^4}{b_0^4}\right)} \right)^{\frac{1}{4}}$$

- 🍏 and introducing an appropriate non-perturbative function f_{NP} . The final result is:

$$\begin{aligned} F_{1,1}^{i,e}(x, \xi, b_T, t, \mu, \zeta) &= \mathcal{C}_{i/j}(x, \kappa, b_*, \mu_{b_*}, \mu_{b_*}^2) \otimes_x \left[(1 - \xi^2) H_j(x, \xi, t, \mu_{b_*}) - \xi^2 E_j(x, \xi, t, \mu_{b_*}) \right] \\ &\times R_i \left[(\mu, \zeta) \leftarrow (\mu_{b_*}, \mu_{b_*}^2) \right] \\ &\times f_{\text{NP}}(x, b_T, (1 - \xi^2)\zeta) \end{aligned}$$

- 🍏 The evolution operator (or Sudakov form factor) is given by:

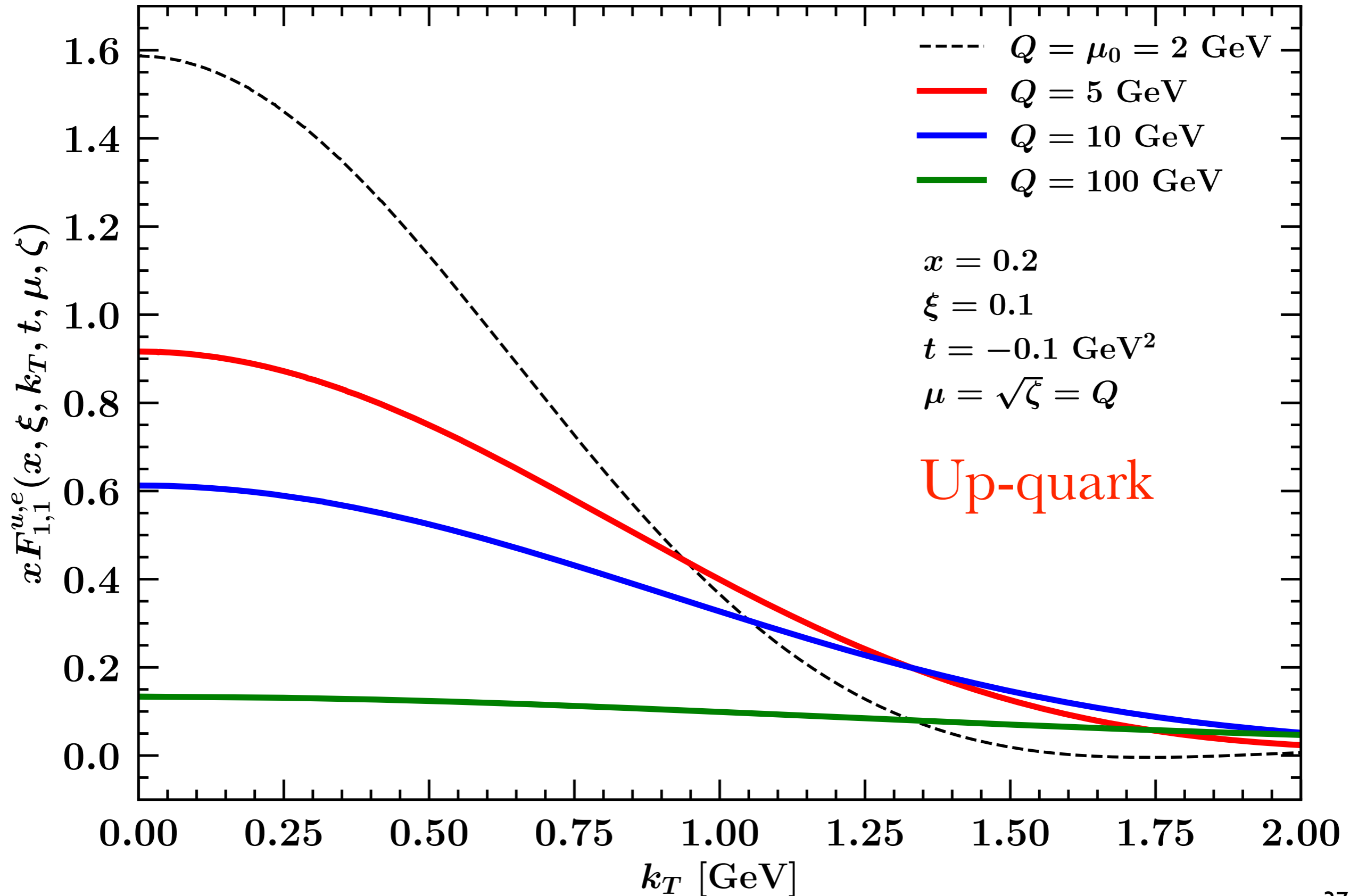
$$R_i = \exp \left\{ K_i(b_*, \mu_{b_*}) \ln \frac{\sqrt{(1 - \xi^2)\zeta}}{\mu_{b_*}} + \int_{\mu_{b_*}}^{\mu} \frac{d\mu'}{\mu'} \left[\gamma_{F,i}(\alpha_s(\mu')) - \gamma_{K,i}(\alpha_s(\mu')) \ln \frac{\sqrt{(1 - \xi^2)\zeta}}{\mu'} \right] \right\}$$

- 🍏 Finally the GTMDs in \mathbf{k}_T space are obtained by inverse Fourier transform:

$$F_{1,1}^{i,e}(x, \xi, k_T, t, \mu, \zeta) = \frac{1}{2\pi} \int_0^\infty db_T b_T J_0(k_T b_T) F_{1,1}^{i,e}(x, \xi, b_T, t, \mu, \zeta)$$

Reconstructing GTMDs

GK model + PV19 + NNLL



Reconstructing GTMDs

GK model + PV19 + NNLL

