

# PDFs and TMDs from Functional Methods

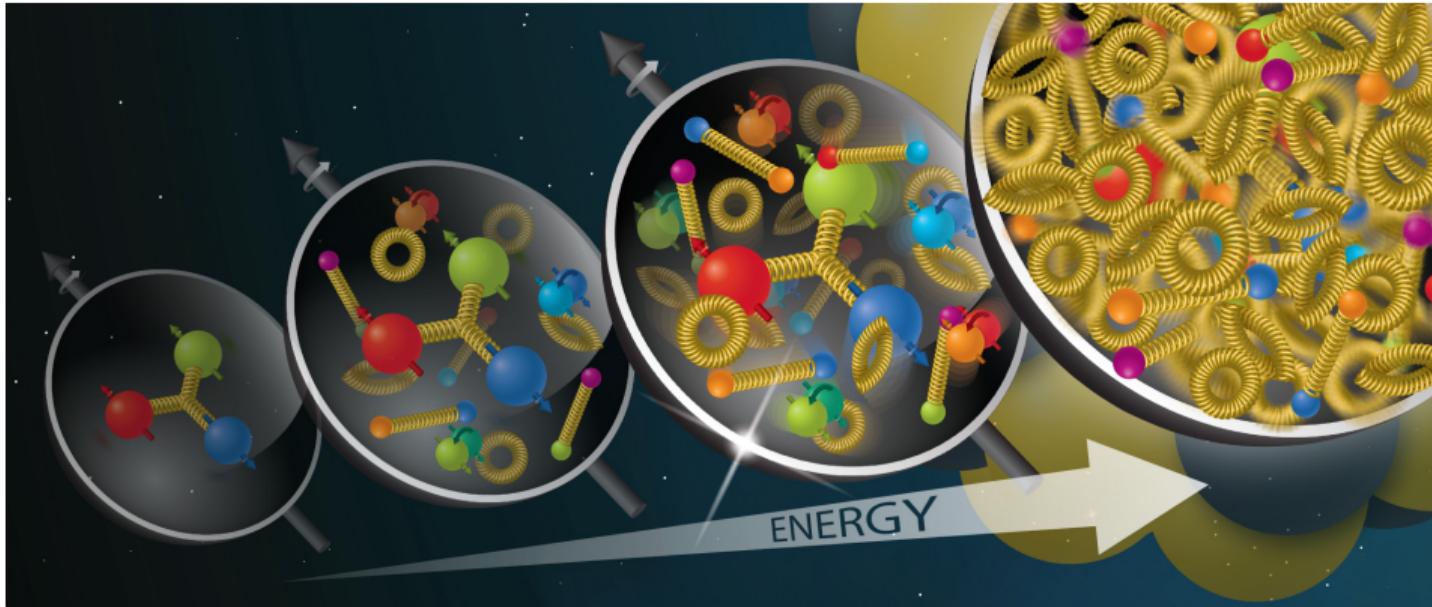
Eduardo Ferreira

*with Gernot Eichmann & Alfred Stadler*

PSR 2024 – July 4, 2024

# Partonic Structure Functions

- PDFs, TMDs, etc... describe this picture:



(Picture source: EIC website)

- Different energy scales probe different aspects of the hadron.
- Contribution from **valence** and **sea** partons.

1

# Non-perturbative methods

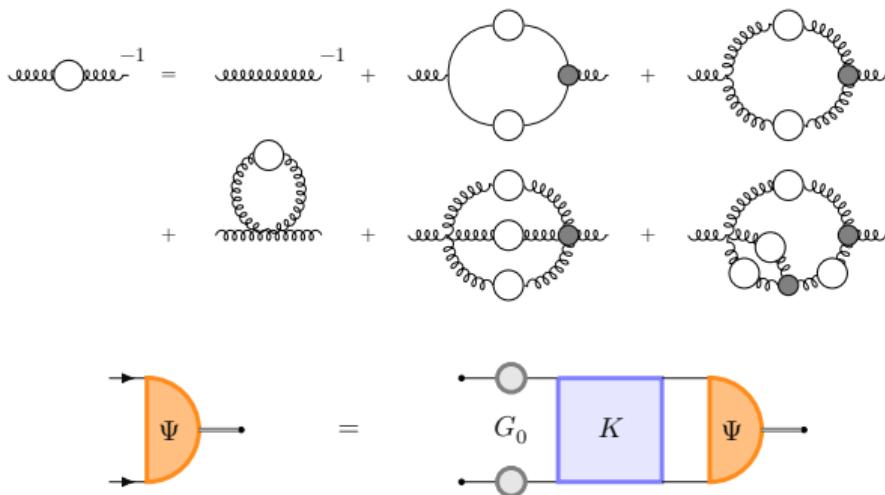
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2 PDFs/TMDs

3 Four-point function

# Theoretical tools

## ■ FUNctional Methods



- Amplitude analysis
- Effective theories –  $\chi$ EFT
- Pheno models

## ■ Lattice QCD

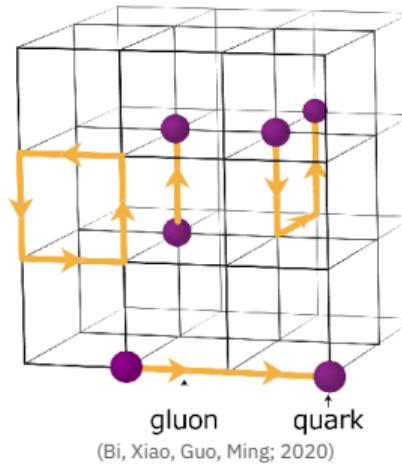


Figure: “Minecraft QCD”  
(Bi, Xiao, Guo, Ming; 2020)

# Dyson-Schwinger Equations

- Quantum equations of motion
- We can relate  $n$ -point functions with each other:

$$\text{---} \circ \text{---}^{-1} = \text{---}^{-1} + \text{---} \circ \text{---} + \text{---} \circ \text{---}$$

- The **full inverse quark propagator** is the inverse bare propagator plus self-energy  $\Sigma$
- $\Sigma$  depends on gluon and  $q\bar{q}g$  vertex

# Dyson-Schwinger Equations

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$$\text{---} \circ \text{---}^{-1} = \text{---} \text{---} \text{---}^{-1} + \text{---} \text{---} \text{---} \text{---} \circ \text{---} + \text{---} \text{---} \text{---} \text{---} \text{---} \circ \text{---}$$
$$+ \text{---} \text{---} \text{---} \text{---} \text{---} \text{---} \circ \text{---} + \text{---} \text{---} \text{---} \text{---} \text{---} \text{---} \text{---} \circ \text{---} + \text{---} \text{---} \text{---} \text{---} \text{---} \text{---} \text{---} \text{---} \circ \text{---}$$

- Coupled equations:  $n$ -point function depends on higher order correlations.

# Truncation

*Is the name of the game*

- Calculations only possible with truncations



$$\frac{\alpha(k^2)}{k^2} = \pi\eta^7 \left(\frac{k^2}{\Lambda^2}\right)^2 e^{-\eta^2 \frac{k^2}{\Lambda^2}} + \frac{2\pi\gamma_m \left(1 - e^{-\frac{k^2}{\Lambda^2 t}}\right)}{\ln \left[e^2 - 1 + \left(1 + \frac{k^2}{\Lambda_{QCD}^2}\right)^2\right]}$$

## Rainbow-ladder truncation

(Maris, Roberts; 1997) (Maris, Tandy; 1999) (Maris, Tandy; 2000)

$$\text{---} \bullet \text{---}^{-1} = \text{---} \text{---}^{-1} + C_F \text{---} \text{---}$$

$$\text{---} \bullet \text{---}^{-1} = \text{---} \text{---}^{-1} + C_A \text{---} \text{---}$$

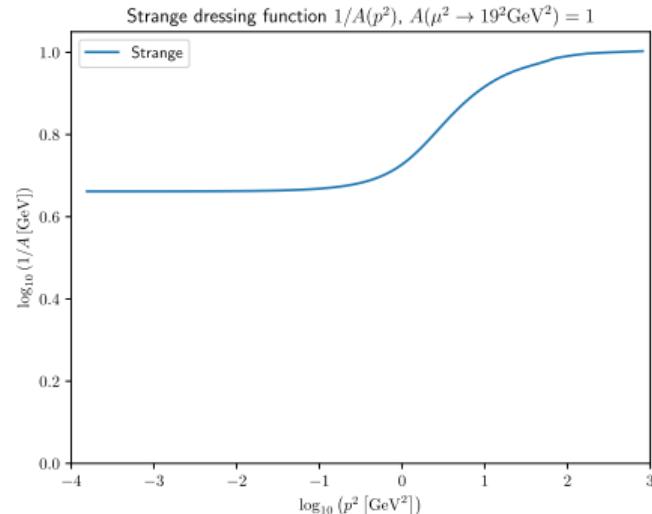
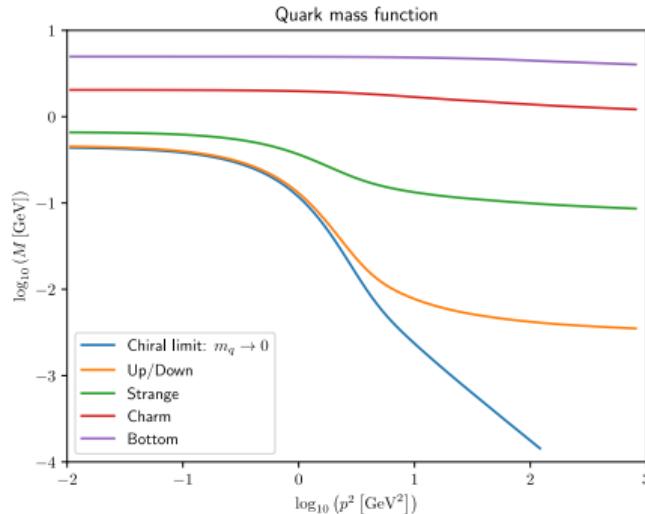
$$\text{---} \bullet \text{---}^{-1} = \text{---} \text{---}^{-1} + TN_f \text{---} \text{---} + C_A \left( \text{---} \text{---} + \text{---} \text{---} \right)$$

$$\text{---} \bullet \text{---} = \text{---} \text{---} + \frac{1}{2} C_A^2 \left( \text{---} \text{---} + \text{---} \text{---} \right)$$

(Alkofer, Zierler; 2023)

# Truncation Is the name of the game

- Calculations only possible with truncations



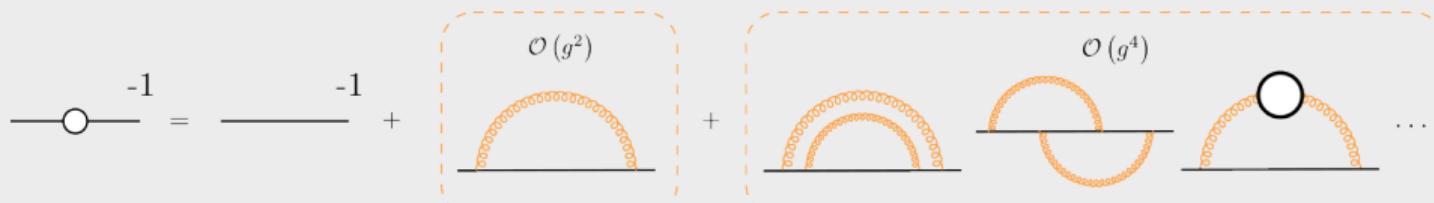
$$S(p) = \sigma_s(p^2)\mathbb{1} - i\sigma_v(p^2)\not{p} = \frac{1}{A(p^2)} \frac{-i\not{p} + M(p^2)}{p^2 + M(p^2)^2}$$

(Numerical results by: Raúl Torres)

# BUT! It's not all-orders perturbation theory

- Non-perturbative physics  $\gg$  all-orders perturbation theory

D $\chi$ SB is purely non-perturbative!



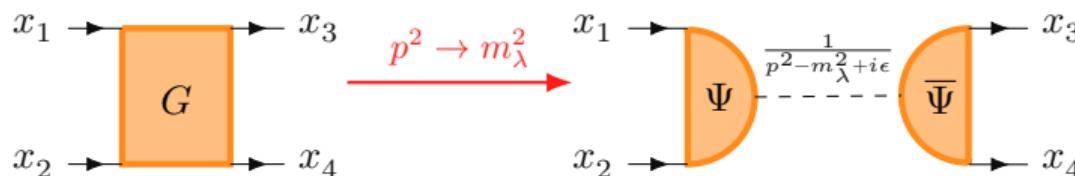
Series expansion:  $\frac{1}{1-x} = f(x) \approx 1 + x + x^2 + x^3 + \dots, \quad |x| < 1$

Exact result:  $\frac{1}{1-x} = f(x) = 1 + xf(x) = 1 + x + x^2f(x) \dots, \quad \text{for all } x$

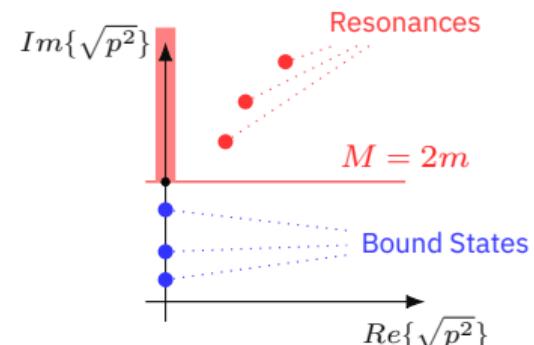
# Bethe-Salpeter Equation *How singularities relate to bound-states*

- The Bethe-Salpeter Wavefunction (BSWF) appears as the residue of a correlation function  $G(p)$ :

$$\Psi(x, P) = \langle 0 | T\phi(0)\phi(x) | P \rangle \quad \Psi(k, P) = \int d^4x e^{-ik \cdot x} \Psi(x, P)$$

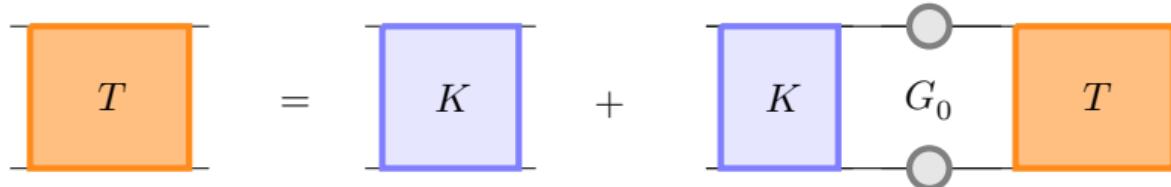


- Poles in correlation functions encode the theory's bound-state spectrum
- Position on the  $p^2$  complex plane indicates their nature: bound-state, resonance, ...
- Euclidean metric!

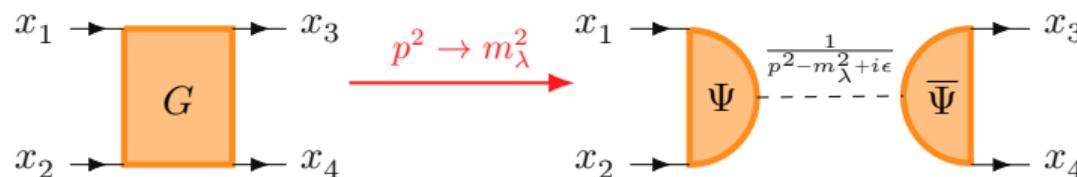


# Bethe-Salpeter Equation *Self-consistent bound-state equation*

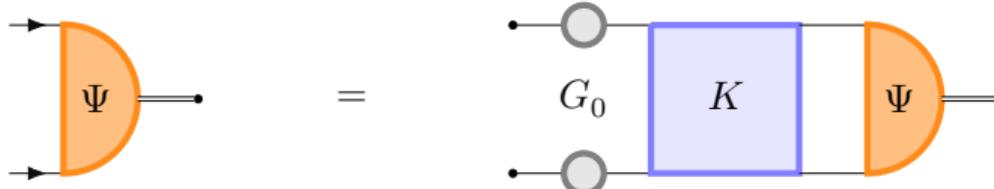
- Take the scattering equation:  $G = G_0 + G_0 T G_0 \Leftrightarrow T = K + K G_0 T$



&



yields

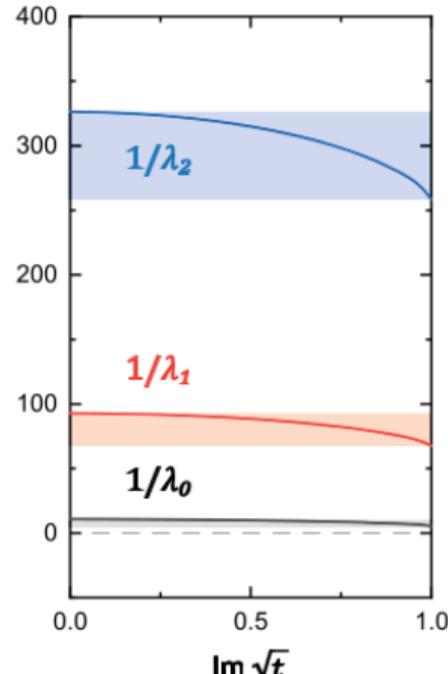


# Bethe-Salpeter Equation *Eigenvalue/vector equation*

- The BSE is usually solved as an eigenvalue equation:

$$\lambda_i(P^2)\psi_i = \mathbf{KG}_0\psi_i$$

- States found when  $\lambda_i(P^2) \equiv 1$ .
- Mass is given by  $P^2 = -M^2$
- Eigenvalue/vector spectrum gives ground state and excited states.
- Like DSE, analytic structure is important, and **contour deformations** may be needed!



QFT “version” of the Schrödinger Equation

3 Non-perturbative methods

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## PDFs/TMDs

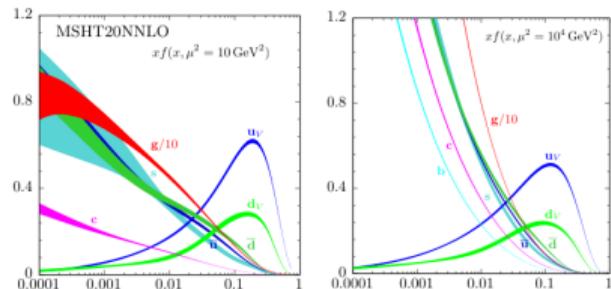
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3 Four-point function

# Hadrons on the Light Front

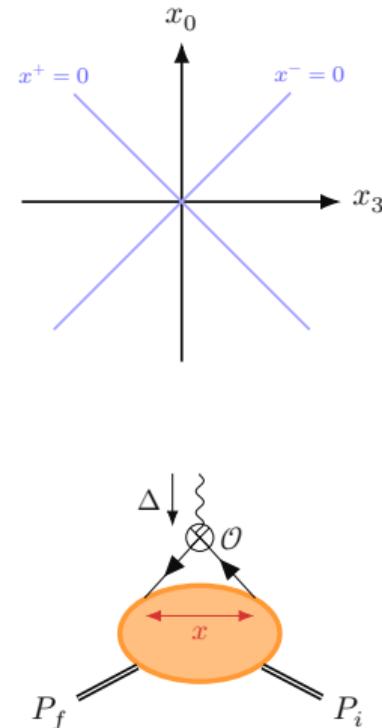
**Goal:** Use DSE/BSE to study hadrons on the light front,  $x^+ = 0$ .

- Natural frame for defining parton distribution functions: PDFs, TMDs, ...



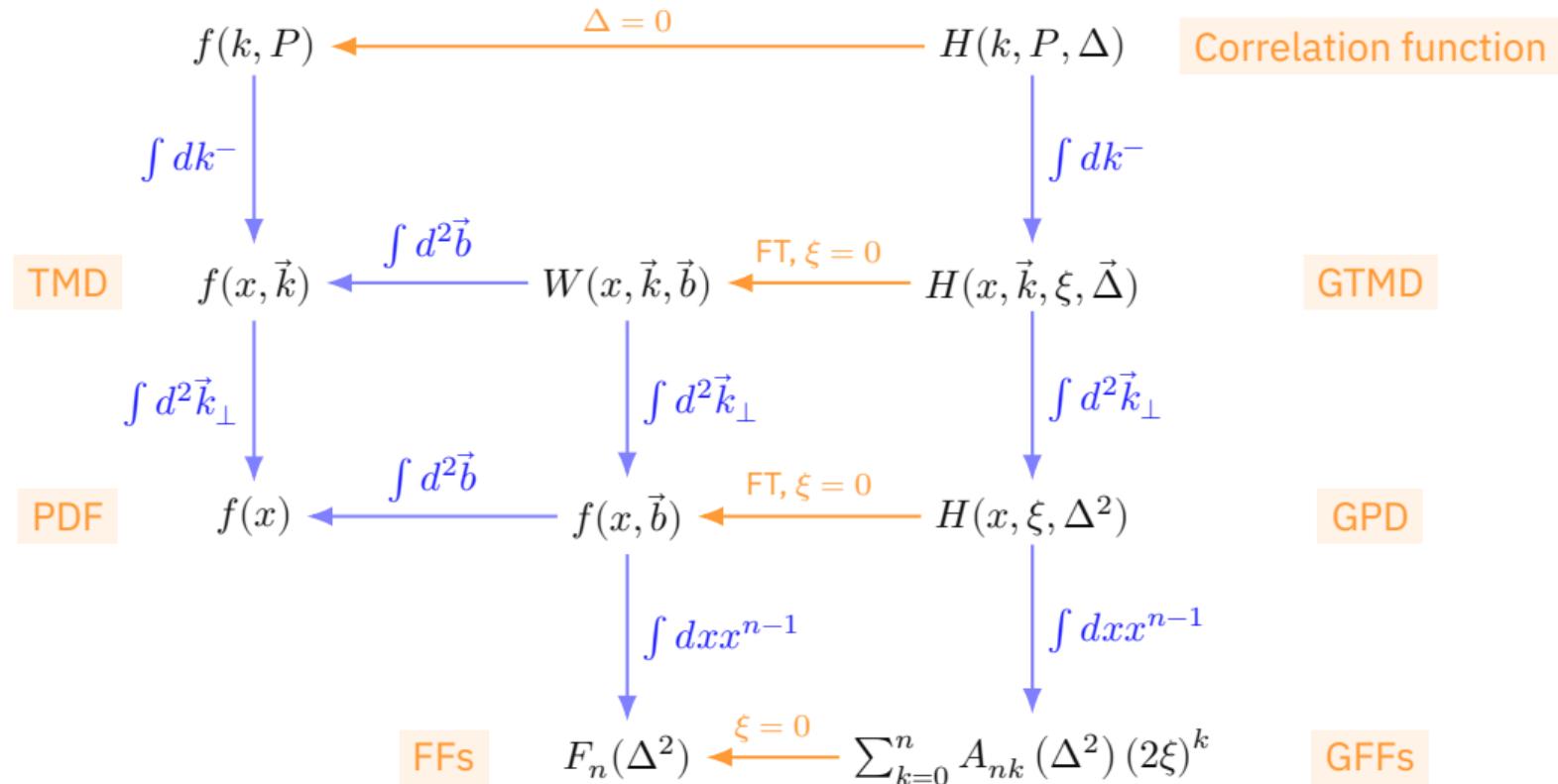
- Future: COMPASS/AMBER @ CERN  
EIC @ Brookhaven National Laboratory.

(AMBER: arXiv:1808.00848)  
(EIC: arXiv:2305.14572)



$$\langle P_f | T\Phi(x)\mathcal{O}\Phi(0) | P_i \rangle$$

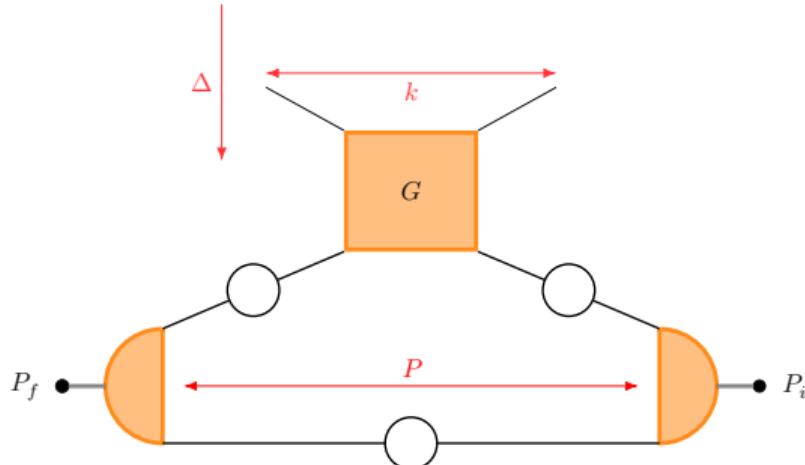
# Hadronic quantities



(Lorce, Pasquini, Vanderhaeghen; 2011), (Picture adapted from: Diehl, 2016), (Diehl, 2003), (Meißner, Goeke, Metz, Schlegel; 2008), (Meißner, Metz, Schlegel; 2009)

# Our Goal

- **Main Goal:** Get partonic distribution functions from hadron-hadron correlations via **FUNctional Methods**



- $G$  is the four-point quark correlation function, calculated with scattering equation.
- The quark propagator is calculated via quark DSE.
- The BSWF is calculated via the meson BSE.

(Mezrag; 2015), (Diehl, Gousset; 1998), (Tiburzi, Miller; 2003),  
(Mezrag, Chang, Moutarde, Roberts, Rodríguez-Quintero, Sabatié, Schmidt; 2015),  
many others, ...

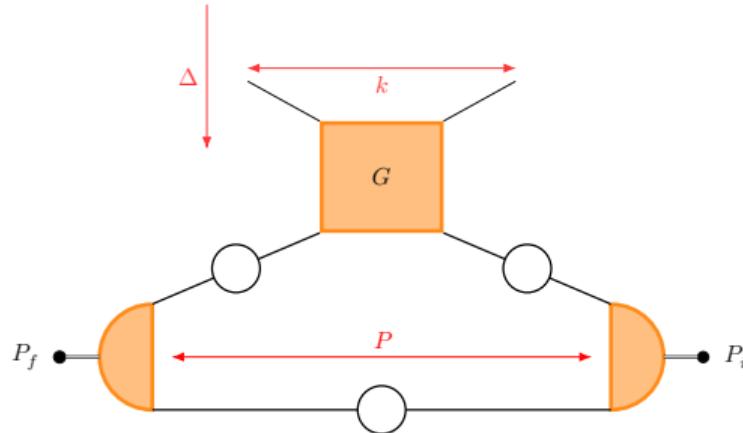
$$\mathcal{G}^{[\Gamma]}(P, k, \Delta) = \frac{1}{2} \text{Tr} \left[ \int dk^- \int \frac{d^4 z}{2\pi^4} e^{ik^- z} \langle P_f | \bar{\psi}(z) \mathcal{W} \Gamma \psi(0) | P_i \rangle \right]$$

- Partonic distributions are calculated by integrating the correlator in  $k^-$  and taking appropriate traces.

# Triangle Diagram

- We start by solving a simple model, and gradually build up the complexity of the calculation.

- Hadrons are on-shell:  $P^2 = -M^2$



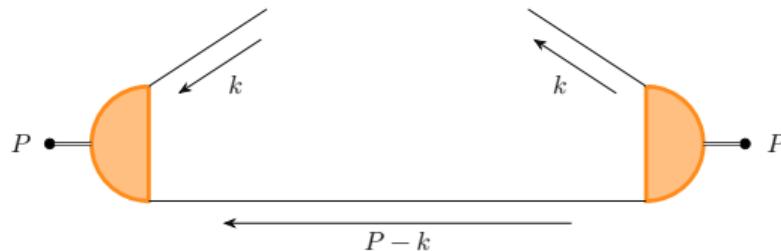
$$\Delta = \sqrt{\Delta^2} \begin{pmatrix} 0 \\ 0 \\ 1 \\ 0 \end{pmatrix} \quad P = 2m\sqrt{t} \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$

$$k = xP + M\sqrt{R} \begin{pmatrix} 0 \\ 0 \\ \sqrt{1-Z^2} \\ Z \end{pmatrix}$$

# Triangle Diagram

- We start by solving a simple model, and gradually build up the complexity of the calculation.

- Hadrons are on-shell:  $P^2 = -M^2$
- **Forward limit:**  $\Delta \rightarrow 0$  – We get the PDFs and TMDs



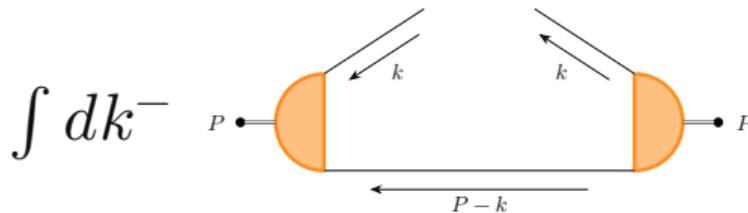
$$P = 2m\sqrt{t} \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} \quad k = xP + M\sqrt{R} \begin{pmatrix} 0 \\ 0 \\ \sqrt{1-Z^2} \\ Z \end{pmatrix}$$

- Steps:

1. Calculate BSWF;
2. Calculate the correlator;
3. Project to the light-front;

# Light-front Projection

- The TMD is defined as:



- Need the BSWF in  $Z \in (-\infty, \infty)$ .
- Use Padé approximants:

$$f(Z) = \frac{a_0 + a_1 Z + a_2 Z^2 + \dots + a_N Z^N}{1 + b_1 Z + b_2 Z^2 + \dots + b_M Z^M}$$

(L. Schlessinger, 1968) (Tripolt et al., 2019) (D. Binosi, R-A. Tripolt; 2019)

- In our kinematic variables:

$$k^- \propto -2i\sqrt{R}Z$$

- $N - M$  is fixed, can control behaviour at very large  $Z$ .

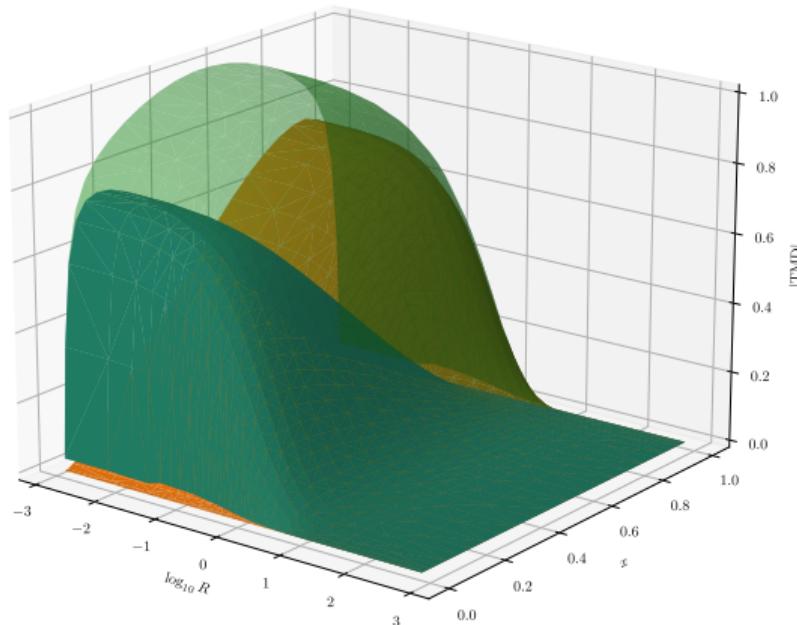
## Definition of the TMD

$$\text{TMD}(R, x) \propto -2i\sqrt{R} \int_{-\infty}^{\infty} dZ \mathcal{G}(R, Z, x)$$

(Eichmann, EF, Stadler; 2022)

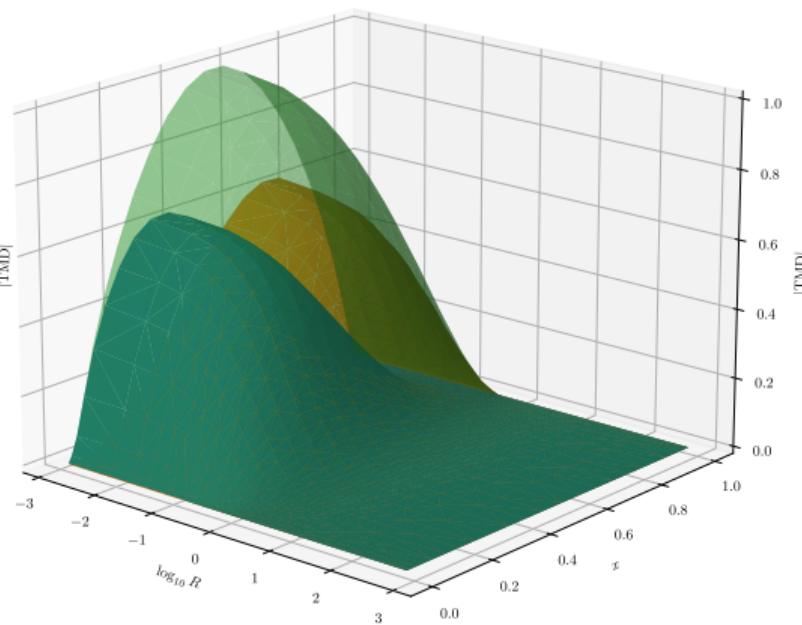
# TMD: Some results

■  $\gamma = 1.5, \beta = 4, c = 1$



■ Upper line open

■  $\gamma = 1, \beta = 1, c = 1$

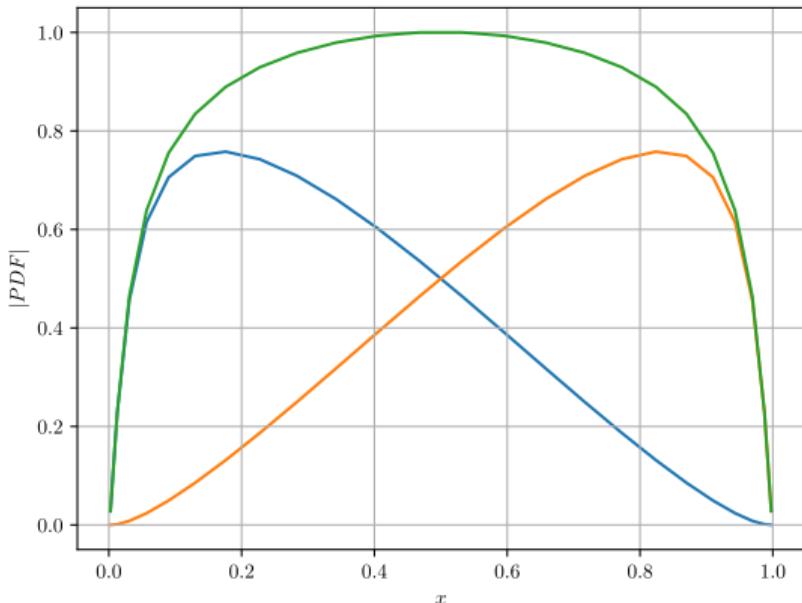


■ Lower line open

■ Sum of both

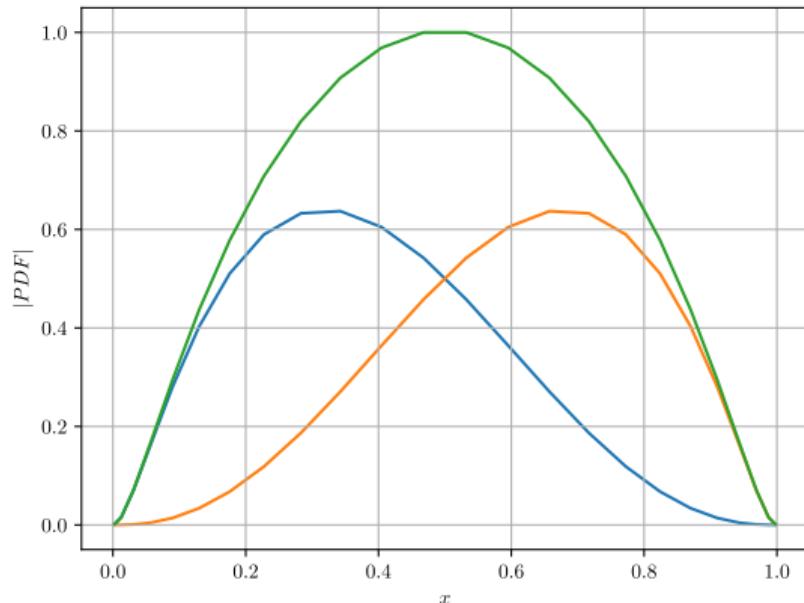
# PDFs

■  $\sqrt{t} = 0.33i, \beta = 4, c = 1$



■ Upper line open

■  $\sqrt{t} = 0.5i, \beta = 1, c = 1$

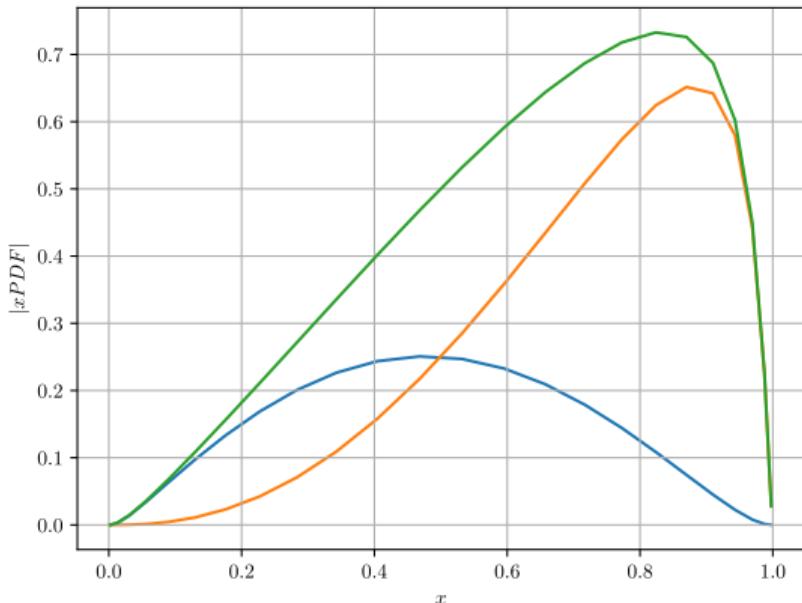


■ Lower line open

■ Sum of both

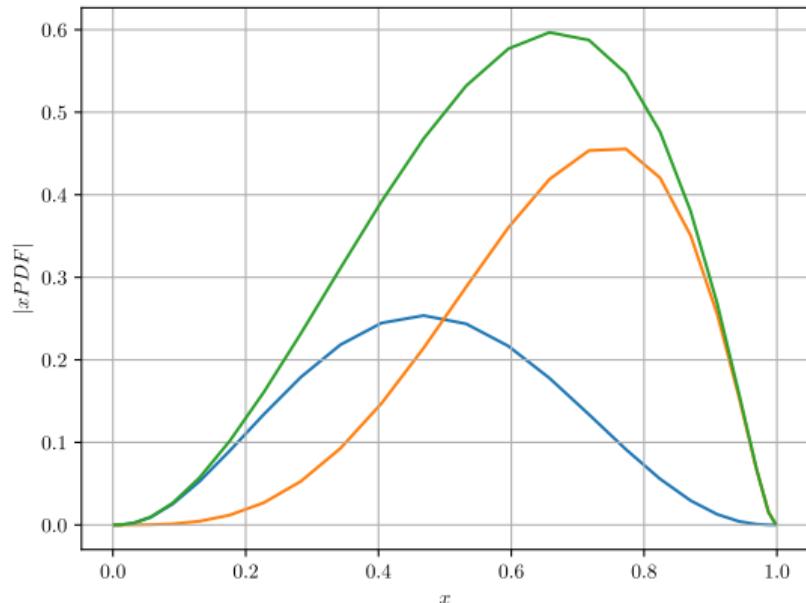
# PDFs

■  $\sqrt{t} = 0.33i, \beta = 4, c = 1$



■ Upper line open

■  $\sqrt{t} = 0.5i, \beta = 1, c = 1$



■ Sum of both

■ Non-perturbative methods

■ PDFs/TMDs

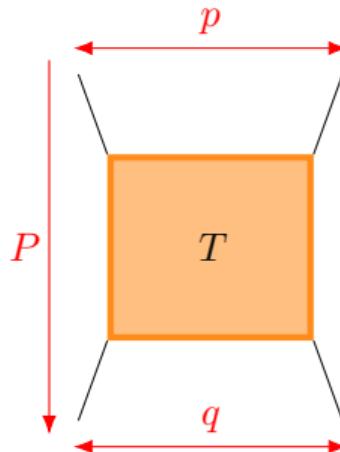
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# **Four-point function**

# 4-point function

- 4-point function determined from scattering equation:

$$\mathbf{G} = \mathbf{G}_0 + \mathbf{G}_0 \mathbf{T} \mathbf{G}_0 \implies \mathbf{T} = \mathbf{K} + \mathbf{K} \mathbf{G}_0 \mathbf{T} \implies \mathbf{T} = (\mathbb{1} - \mathbf{K} \mathbf{G}_0)^{-1} \mathbf{K}.$$



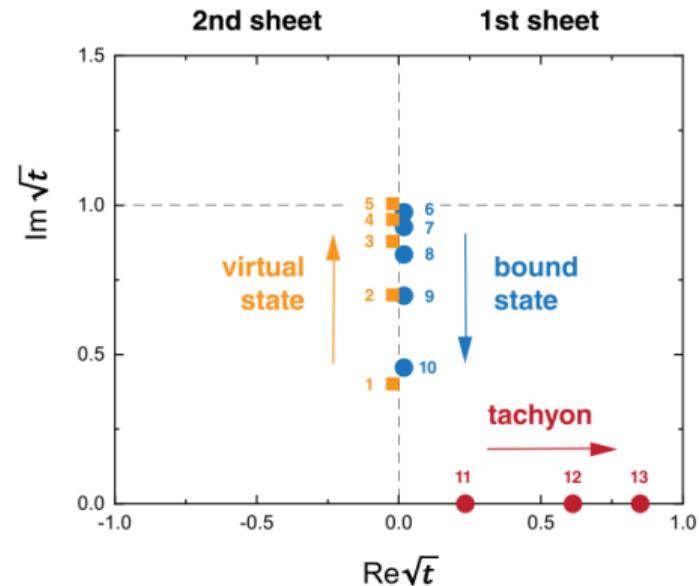
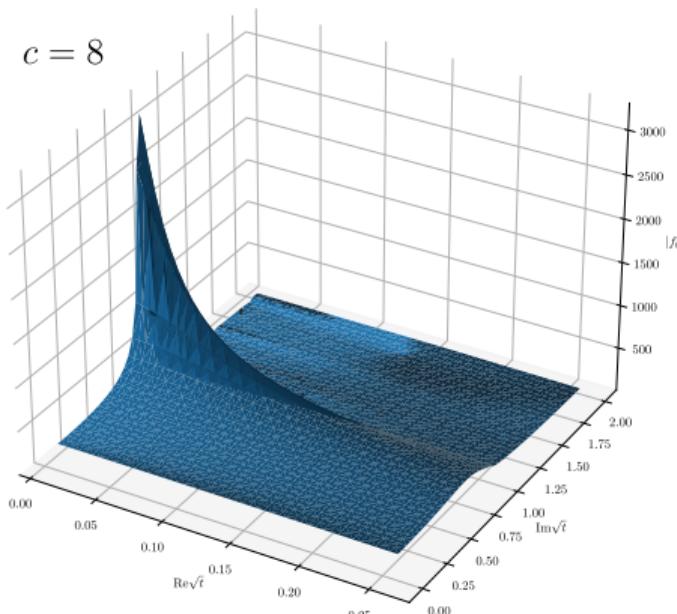
$$\begin{aligned} T(t, X, R, Z, Y, Q) &= K(X, R, p \cdot q) \\ &+ \frac{1}{2} \frac{m^4}{2\pi^4} \int_0^\infty dx x \int_{-1}^1 dz \sqrt{1-z^2} G_0(x, z, t) \\ &\times \int_{-1}^1 dy \int_0^{2\pi} d\Psi K(X, x, k \cdot q) T(t, x, y, z, R, Q) \end{aligned}$$

- Fully off-shell: 6 Lorentz invariants
  - 3 radial:  $X, t, R;$
  - 3 angular:  $Y, Z, Q;$

- Same  $G_0$  and  $K$  as in the BSE
- Solved numerically as linear system
- Contains *all* dynamics of 2  $\phi$  particles
  - **Must produce bound state poles dynamically!**

# 4-point function

- $T$  is made fully on-shell
- Partial-wave expansion shows bound-state pole in the first Riemann sheet

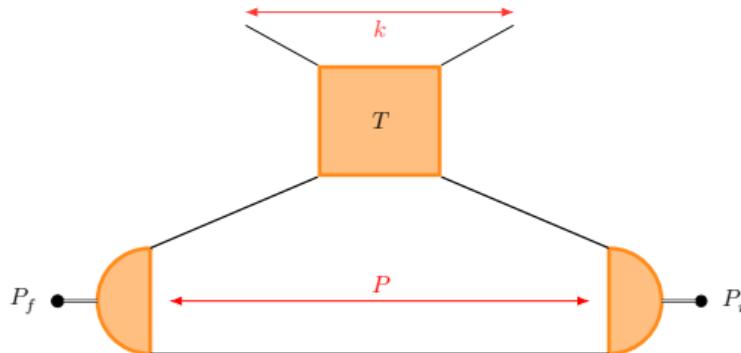


(Eichmann, Duarte, Peña, Stadler; 2019)

- Describes both long-range and short-range  $qq$  dynamics.

# Full triangle

- Can we use the full 4-point function? – **YES!**

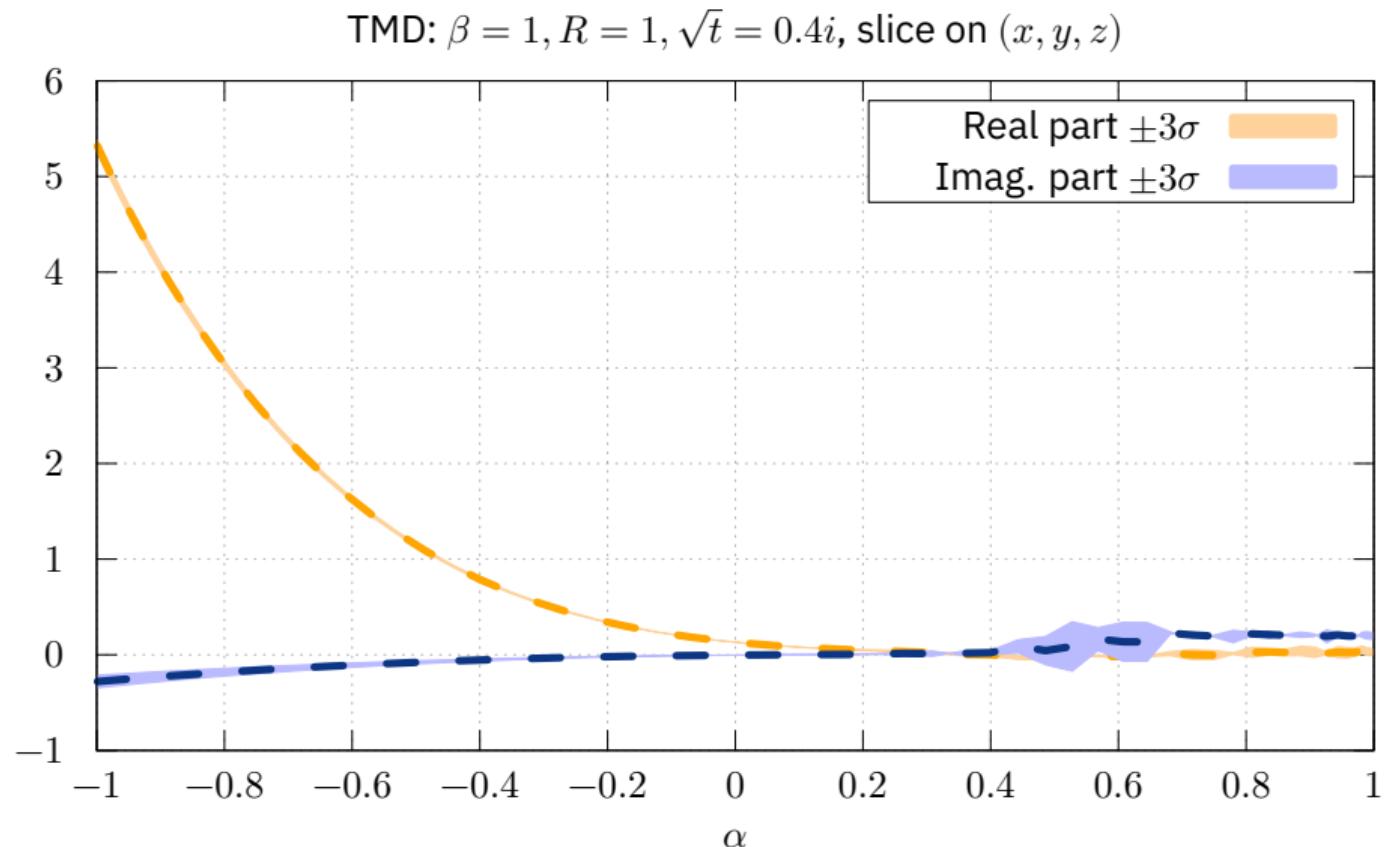


1. Calculate 4-point function
2. Calculate BSE
3. Do the loop integration
4. Project to LF
5. Do statistics on extrapolation

- We solve one triangle for each point in the  $R, x$  grid – **HPC Needed!**
  - $N_x \times N_R \times N_Z$  4-point functions!

# Results

## Preliminary



# Conclusions

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- We have a method for TMDs/PDFs from first-principles self-consistent functional methods
- We have incorporated BSE solutions
- We have implemented contour deformations – resonances!
- We have made considerable progress in the inclusion of the 4-point function

Backup

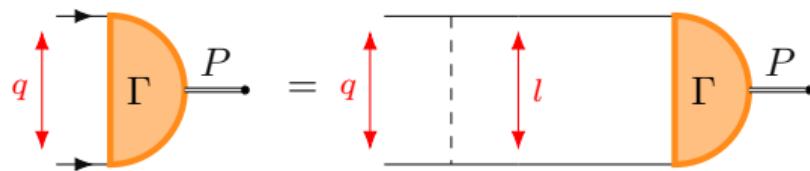
# Scalar toy model

## Bethe-Salpeter Equation

- Scalar model:

- $\phi$  of mass  $m$
- $\chi$  of mass  $\mu$

- BS amplitude for bound-state of two  $\phi$ :



$$q = \frac{\alpha}{2}P + m\sqrt{\rho} \begin{pmatrix} 0 \\ \times \\ \times \\ \times \end{pmatrix} \quad l = \frac{\alpha}{2}P + m\sqrt{l} \begin{pmatrix} \times \\ \times \\ \times \\ \times \end{pmatrix}$$

- The BSWF is a function of the kinematic invariants:

$$-M^2 = P^2 \quad \frac{q^2}{m^2} \propto \rho \quad z \propto \hat{k} \cdot \hat{P}$$

$$\begin{aligned} \psi(\rho, z, x) &= \frac{m^4}{(2\pi)^3} \frac{1}{2} \int_0^\infty dl \, l \\ &\times \int_{-1}^1 dz_l \sqrt{1 - z_l^2} \mathbf{G}_0(l, z_l, \alpha) \\ &\times \int_{-1}^1 dy_l \mathbf{K}(\rho, z, l, z_l, y_l) \psi(l, z_l, \alpha) \end{aligned}$$

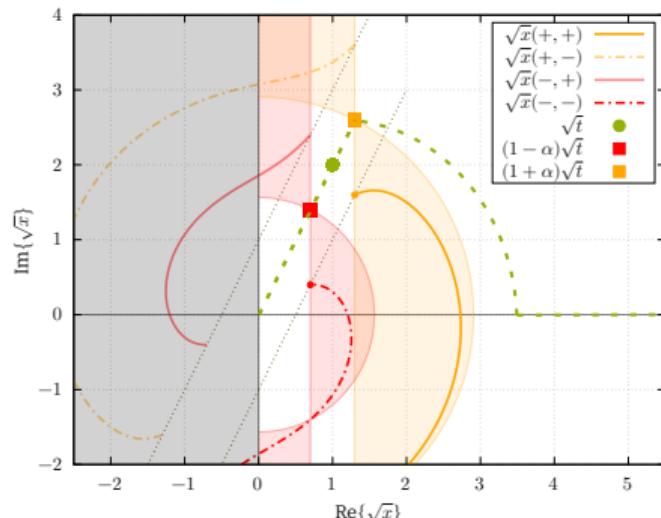
# Analytic Structure of BSE

$$\operatorname{Im} \sqrt{t} > \min\left(\frac{1}{1 \pm \alpha}\right)$$

$$\mathbf{G}_0^{-1} = (l_+^2 + m^2)(l_-^2 + m^2)$$

- Branch cuts in complex  $\sqrt{t}$  plane

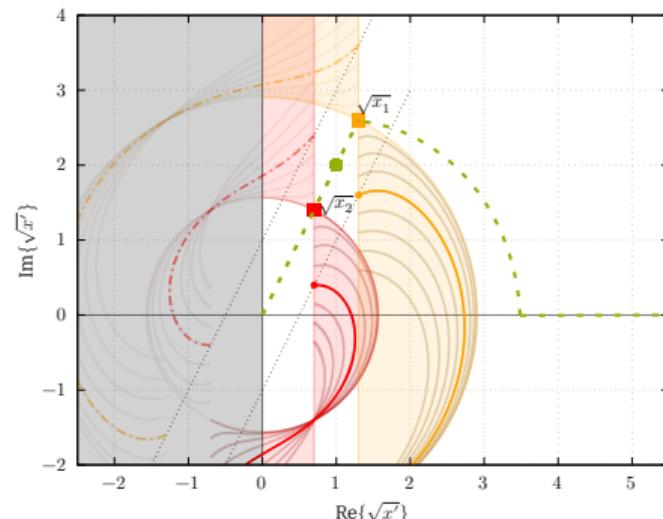
$$\sqrt{t}_{\pm}^{\lambda} = \mp(1 \pm \alpha)\sqrt{t} \left[ z_l + i\lambda \sqrt{1 - z_l^2 + \frac{1}{(1 \pm \alpha)^2 t}} \right]$$



$$\mathbf{K}^{-1} = (q - l)^2 + \mu^2$$

- Branch cuts in complex  $\sqrt{t}$  plane depend on path taken:

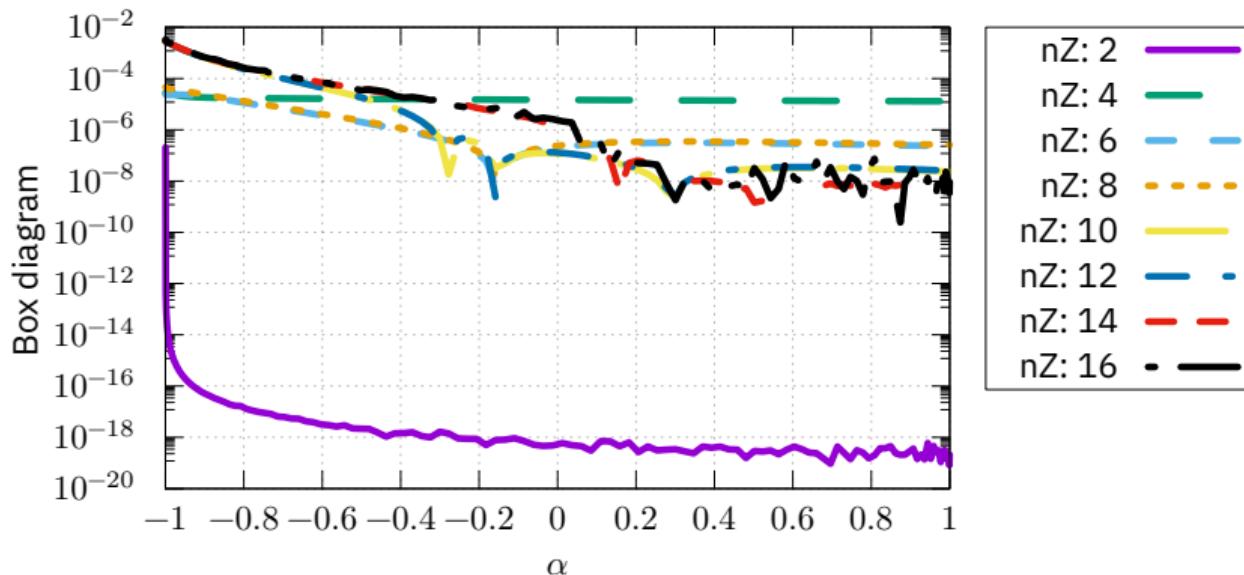
$$\sqrt{t} = \sqrt{\rho} \left( \Omega \pm i \sqrt{1 - \Omega^2 + \frac{\beta^2}{\rho}} \right)$$



# Extrapolation (can be) unreliable

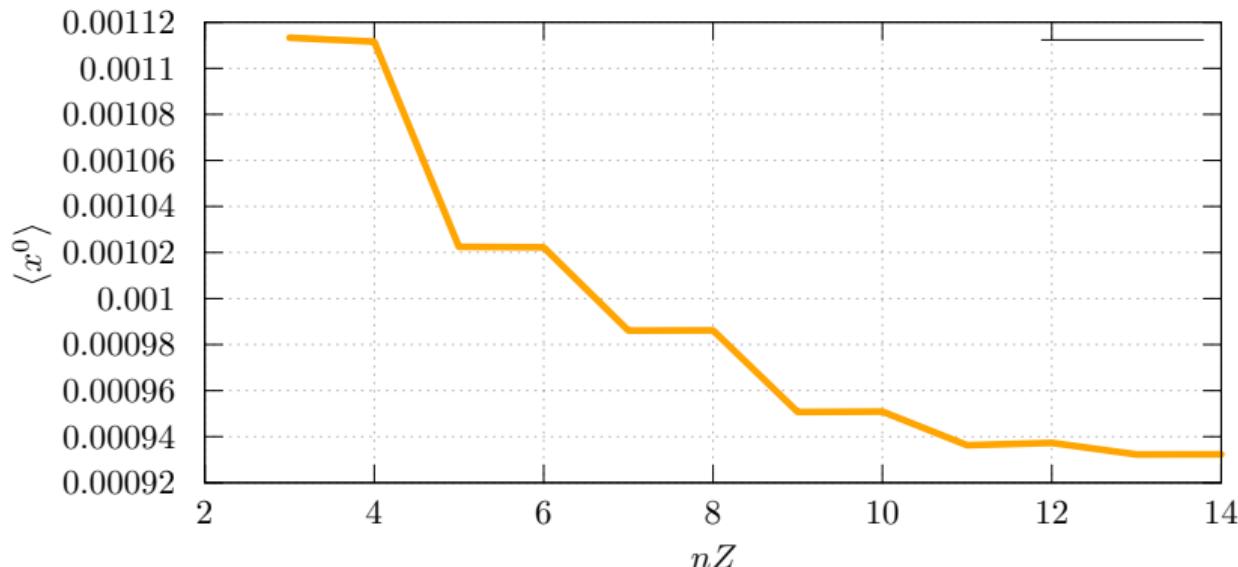
- Schlessinger method is fast but limited – at most  $\propto \frac{1}{Z}$ 
  - Coefficients must be very accurately calculated
- Different input points  $\Rightarrow$  different output!

$$R(Z) = \frac{f_0}{1 + \frac{a_0(Z - Z_0)}{1 + \frac{a_1(Z - Z_1)}{1 + \dots}}}$$



# Use Padé Approximants

- In this case we know the analytic dependence – can improve extrapolation
- Uncertainty **massively** reduced – automate search for best parameters
  - Use statistics to “approximate” uncertainty



- Figure of merit: **First Mellin moment**  $\langle x^0 \rangle = \int d\alpha f(\alpha)$ .

# Use Padé Approximants

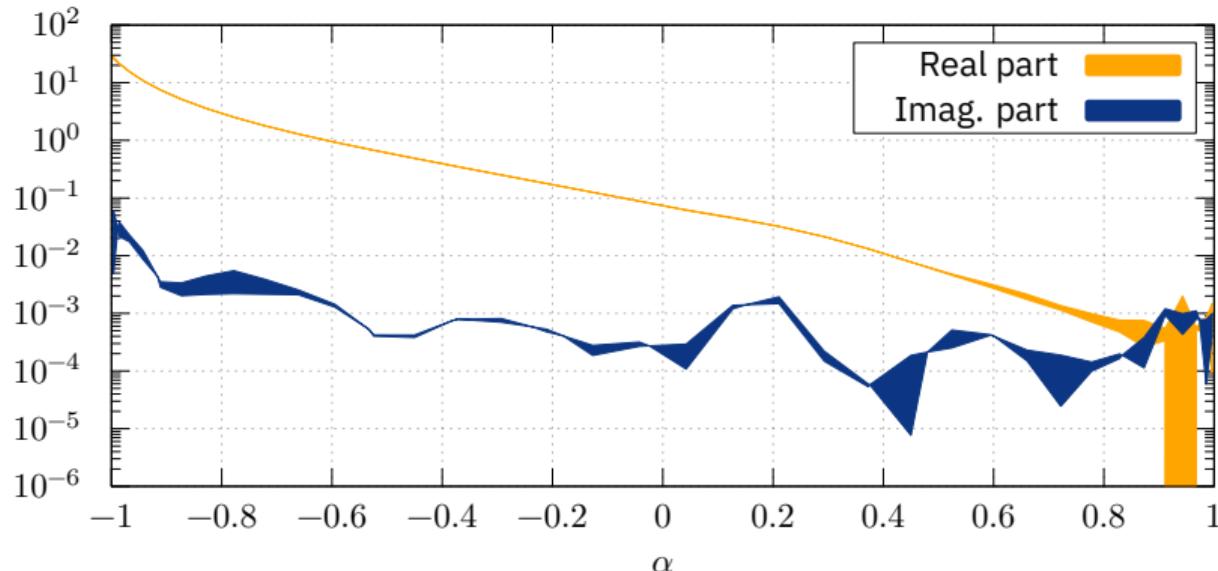
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0) (0.855494,0) (0.952481,0) (0.899179,0)
[main::main()]
0) (0.791917,0) (0.841999,0) (0.922319,0)
[main::main()]
0) (0.76608,0) (0.958332,0) (0.829141,0)
[main::main()]
0) (0.730918,0) (0.883785,0) (0.984933,0)
[main::main()]
0) (0.78723,0) (0.821294,0) (0.927566,0)
[main::main()]
0) (0.759677,0) (0.940532,0) (0.958998,0)
[main::main()]
0) (0.912295,0) (0.928642,0) (-0.786102,0)
[main::main()]
0) (0.856692,0) (0.835676,0) (0.123163,0)
[main::main()]
0) (0.761718,0) (0.811805,0) (0.355118,0)
[main::main()]
0) (0.704273,0) (0.777318,0) (0.571327,0)
[main::main()]
0) (0.642733,0) (0.608136,0) (0.428164,0)
[main::main()]
0) (0.581043,0) (0.428975,0) (0.17211,0)
[main::main()]
0) (0.627808,0) (0.561273,0) (0.316577,0)
[main::main()]
0) (0.54071,0) (0.328953,0) (0.279375,0)
[main::main()]
0) (0.64143,0) (0.184463,0) (0.279375,0)
[main::main()]
0) (0.427782,0) (0.132344,0) (0.427782,0)
[main::main()]
0) (0.556484,0) (0.419112,0) (0.190868,0)
[main::main()]
0) (0.298888,0) (0.286554,0) (0.286554,0)
[main::main()]
0) (0.550732,0) (0.408883,0) (0.111661,0)
[main::main()]
0) (0.389683,0) (0.111661,0) (0.111661,0)
[main::main()]
0) (0.591231,0) (0.388676,0) (0.388676,0)
[main::main()]
0) (0.50868,0) (0.468123,0) (0.148025,0)
[main::main()]
0) (0.554563,0) (0.328618,0) (0.1704687,0)
[main::main()]
0) (0.528205,0) (0.0710795,0) (0.418727,0)
[main::main()]
0) (0.517209,0) (0.188441,0) (0.378999,0)
[main::main()]
0) (0.6020596,0) (0.398628,0) (0.126602,0)
[main::main()]
0) (0.116918,0) (0.292988,0) (0.116918,0)
[main::main()]
0) (0.626998,0) (0.304783,0) (0.143185,0)
[main::main()]
0) (0.535192,0) (0.0845858,0) (0.0429227,0)
[main::main()]
0) (0.388719,0) (0.0429227,0) (0.388719,0)
[main::main()]
0) (0.118805,0) (0.0355118,0) (0.571327,0)
```

- Generate large collection of  $Z$  points ( $\approx 50$ ).

# Use Padé Approximants

- In this case we know the analytic dependence – can improve extrapolation
- Uncertainty **massively** reduced – automate search for best parameters
  - Use statistics to “approximate” uncertainty



- Print the results as  $\bar{f}(\alpha) \pm \sigma_f(\alpha)$ .