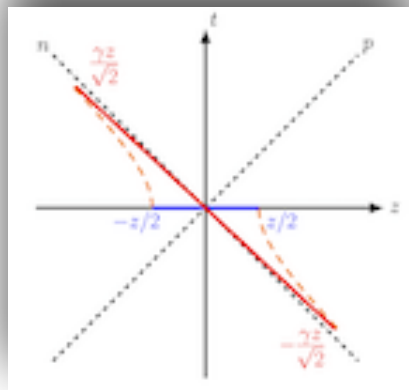


Controlling Systematic Uncertainties in Lattice QCD Calculations of Proton Unpolarized GPDs



Presented by -

Arif Sarkar

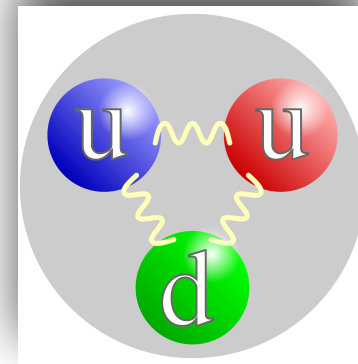
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In collaboration with:

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Outline:

1. Core Framework & Lattice Setup
 - A. Generalized Parton Distributions (GPDs) and Mellin Moments
 - B. Lattice extraction of Bare Nucleon Matrix Elements
2. Renormalization & Matching
 - A. Non-perturbative Renormalization: The Double Ratio Scheme
 - B. Short-Distance Factorization & Light-Cone Matching
3. t -dependent Parameterization via Artificial Neural Networks
4. Control of Systematics: Excited State Contamination

A work in progress...

Generalized Parton Distributions:

GPDs are important in the context of the study of strong interactions. These are non perturbative in nature.

GPDs for quarks are defined on light cone coordinates ($z^+ = 0$ & $z_\perp = 0$) as:

$$\begin{aligned}\tilde{F}(x, t, \xi) &= \frac{1}{2} \int \frac{dz^-}{2\pi} e^{-ixP^+z^-} \langle p_2 | \bar{\psi}^q(-\frac{z^-}{2}) \gamma^+ \mathcal{W}(-\frac{z^-}{2}, \frac{z^-}{2}) \psi^q(\frac{z^-}{2}) | p_1 \rangle \\ &= \frac{1}{2P^+} \bar{u}(p_2) \left(\gamma^+ H^q(x, t, \xi) + \frac{i\sigma^{+\nu} \Delta_\nu}{2M} E^q(x, t, \xi) \right) u(p_1)\end{aligned}$$

With $P = \frac{p_2^+ + p_1^+}{2}$, $\Delta = \vec{p}_f - \vec{p}_i$, $\xi = -\Delta^+/2P^+$ and $z^\pm = \frac{1}{\sqrt{2}}(z^0 \pm z^3)$

Local OPE of Non-Local Quark Bilinears

Leading twist (twist 2) expansion of non-local matrix elements — local operators

$$\bar{q}\left(-\frac{z}{2}\right)\gamma^\rho\mathcal{W}\left(-\frac{z}{2},\frac{z}{2}\right)q\left(\frac{z}{2}\right)=\sum_{j=0}^{\infty}\frac{1}{j!}z^{\mu_1}z^{\mu_2}\dots z^{\mu_j}\left[\bar{q}(0)\gamma^\rho\overleftrightarrow{D}_{\mu_1}\overleftrightarrow{D}_{\mu_2}\dots\overleftrightarrow{D}_{\mu_j}q(0)\right]$$

The Mellin Moments: $\mathcal{M}(\xi, t) = \int_{-1}^1 dx x^m H(x, \xi, t)$

The Mellin Moment are related to the local operator matrix elements via:



Robert Hjalmar Mellin
1854 - 1933

Local OPE of Non-Local Quark Bilinears

Leading twist (twist 2) expansion of non-local matrix elements — local operators

$$\bar{q} \left(-\frac{z}{2} \right) \gamma^\rho \mathcal{W} \left(-\frac{z}{2}, \frac{z}{2} \right) q \left(\frac{z}{2} \right) = \sum_{j=0}^{\infty} \frac{1}{j!} z^{\mu_1} z^{\mu_2} \dots z^{\mu_j} \left[\bar{q}(0) \gamma^\rho \overleftrightarrow{D}_{\mu_1} \overleftrightarrow{D}_{\mu_2} \dots \overleftrightarrow{D}_{\mu_j} q(0) \right]$$

The Mellin Moment are related to the local operator matrix elements via:

$$\langle p_2 | \bar{\psi}^q(0) \gamma^\rho \overleftrightarrow{D}^{\mu_1} \dots \overleftrightarrow{D}^{\mu_j} \psi^q(0) | p_1 \rangle = P^{\{\rho} \sum_{i=1}^{m+1} P^{\mu_1} \dots P^{\mu_{i-1}} \Delta^{\mu_i} \dots \Delta^{\mu_m} \} A_{i,m}^q(t) + \Delta^\rho \Delta^{\mu_1} \dots \Delta^{\mu_m} C_{m+1}^q(t)$$

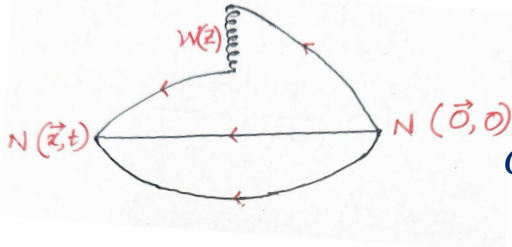
Generalized Form Factors

Polynomiality Relation:

$$\int_{-1}^1 dx x^m H^q(x, \xi = 0, t) = \sum_{\substack{k=0 \\ \text{even}}}^m A_{m+1,k}^q(t), \quad \int_{-1}^1 dx x^m E^q(x, \xi = 0, t) = \sum_{\substack{k=0 \\ \text{even}}}^m B_{m+1,k}^q(t)$$

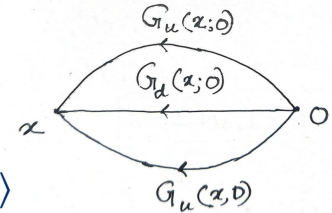
Towards Lattice calculations of (bare) GPDs:

The nucleon two- and three- point functions can be calculated from lattice:



$$C^{2pt}(\Gamma, t) = \Gamma_{\alpha, \beta} \sum_{\vec{x}} e^{-i\vec{P} \cdot \vec{x}} \langle 0 | N_{\alpha}(\vec{x}, t) \bar{N}_{\beta}(\vec{x}, 0) | 0 \rangle$$

$$C^{3pt}(\Gamma, z, p_f, p_i, t, \tau) = \Gamma_{\alpha\beta} \sum_{\vec{x}, \vec{y}} \langle 0 | N_{\alpha}(\vec{x}, t) \mathcal{O}(\vec{y}, \tau; z) \bar{N}_{\beta}(\vec{0}, 0) | 0 \rangle$$



Ref: K. Cichy and M. Constantinou, Adv. High Energy Phys. (2019) 3036904

The Matrix Elements (MEs) are calculated by using ratio:

$$R_{\mu}(\Gamma_{\kappa}, z, P_f, P_i; t_s, \tau) = \frac{C_{\mu}^{3pt}(\Gamma_{\kappa}, z, P_f, P_i; t_s, \tau)}{C^{2pt}(\Gamma_0, P_f; t_s)} \sqrt{\frac{C^{2pt}(\Gamma_0, P_i, t_s - \tau) C^{2pt}(\Gamma_0, P_f, \tau) C^{2pt}(\Gamma_0, P_f, t_s)}{C^{2pt}(\Gamma_0, P_f, t_s - \tau) C^{2pt}(\Gamma_0, P_i, \tau) C^{2pt}(\Gamma_0, P_i, t_s)}}$$

1. **Ground State Dominant MEs** $\rightarrow 0 \ll \tau \ll t_s$
2. From here MEs of GPD are calculated via Lorentz invariant Decompositions as mentioned in Eqn. 73 to 88 of: *S. Bhattacharya et al., PRD 106 (2022) 114512.*

Lattice calculations of GPDs - Asymmetric Frame



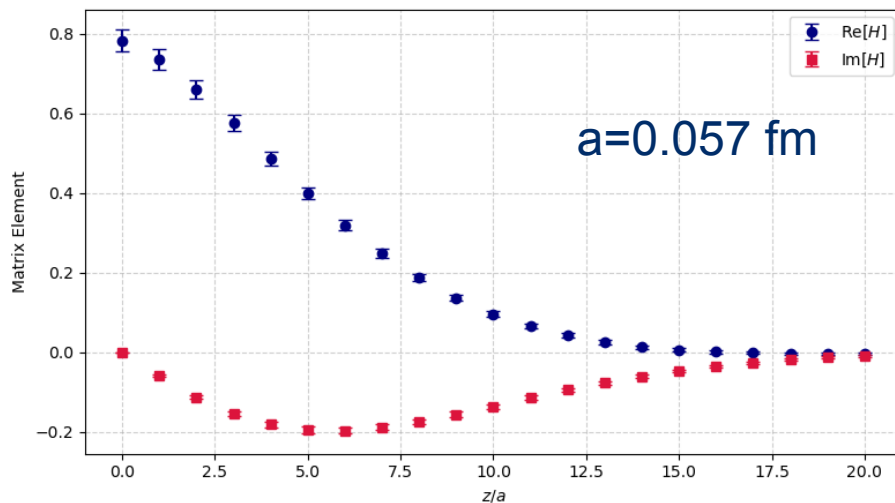
The MEs are calculated in asymmetric frame, defined as:

$$p_f = \vec{P} = (0, 0, P_3) \quad p_i = \vec{P} - \vec{\Delta} = (\Delta_1, \Delta_2, P_3) \quad \text{Zero-skewness} \rightarrow \Delta_3 = 0$$

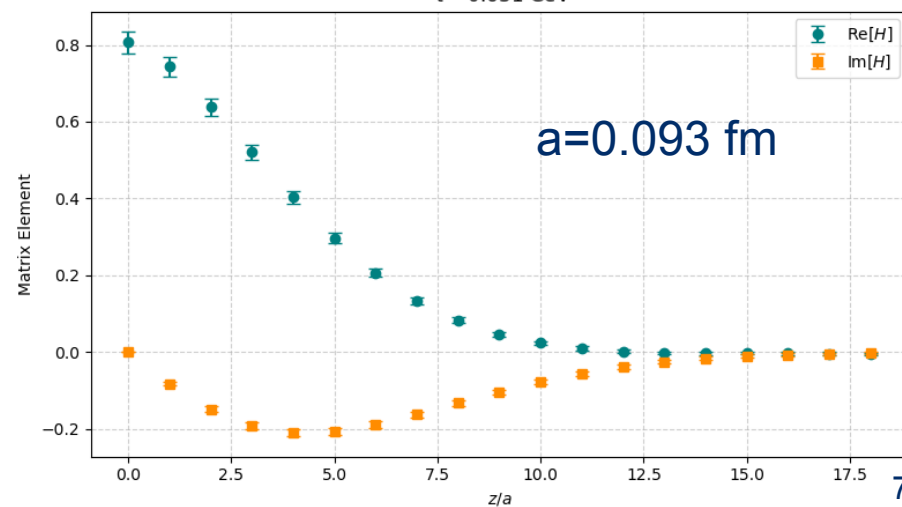
MEs for several t -values can be computed at once \rightarrow computational advantage over symmetric frame.

$$-t = \vec{\Delta}^2 - [E(p_f) - E(p_i)]^2$$

cD Ensemble Bare MEs (P2, Q2)
 $-t = 0.741 \text{ GeV}^2$



cA Ensemble Bare MEs (P2, Q2)
 $-t = 0.631 \text{ GeV}^2$



Double Ratio Scheme Renormalization:

bare MEs exhibit UV divergences:

$$h^B(z, P_3, a) = e^{\delta m|z|} Z(a) h^R(z, P_3, a)$$

Ref: X. Ji et al., PRL 120 (2018) 112001

The **RG invariant**, double ratio:

$$\mathcal{M}(z, P_3, P_3^0) = \frac{h^B(z, P_3, a) h^B(0, P_3^0, a)}{h^B(z, P_3^0, a) h^B(0, P_3, a)}$$

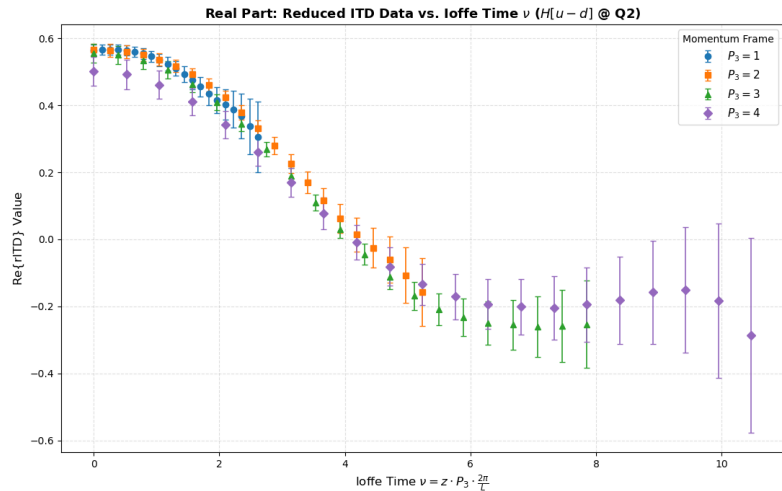
This Double Ratio ME is named as reduced **ioffee-time** distribution (rITD) —

- Lorentz invariant $\rightarrow \mathcal{M}(z, P_3, P_3^0) \sim \mathcal{M}(z^2, zP_3, zP_3^0) = \mathcal{M}(z^2, \nu, \nu_0)$
- Cancels some **discretization effects** and **higher twist effects**.

Refs

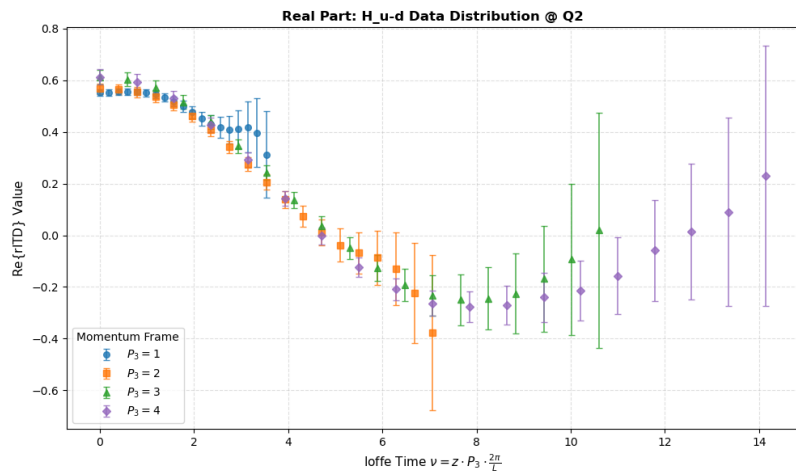
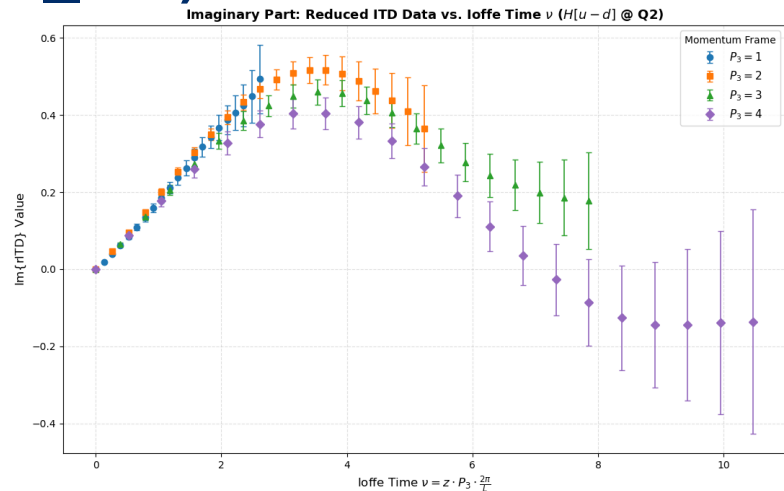
1. H. Dutrieux et al. (HadStruc Collaboration), JHEP. 2024, 162 (2024)
2. X. Gao et al., PRD 102 (2020) 094513

rITDs at different lattice spacing (H_u-d)



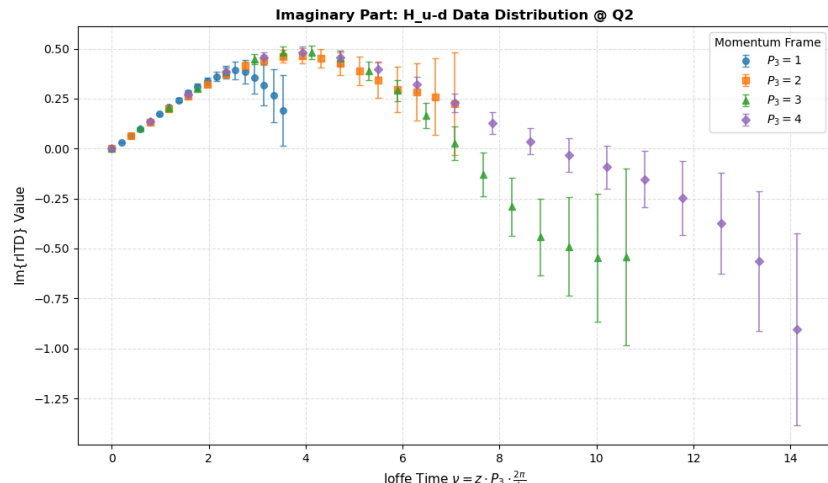
$a = 0.057$ fm

P_3
 0.4448 GeV
 0.8896 GeV
 1.3344 GeV
 1.7793 GeV



$a = 0.093$ fm

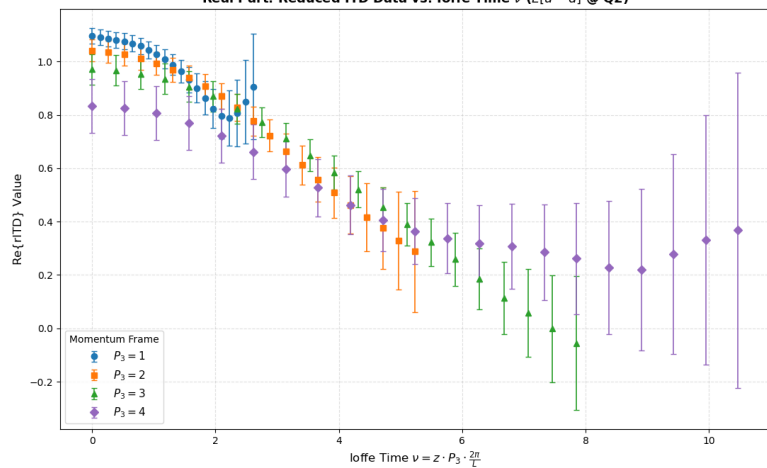
P_3
 0.4089 GeV
 0.8179 GeV
 1.2268 GeV
 1.6358 GeV



rITDs at different lattice spacing (E_{u-d})

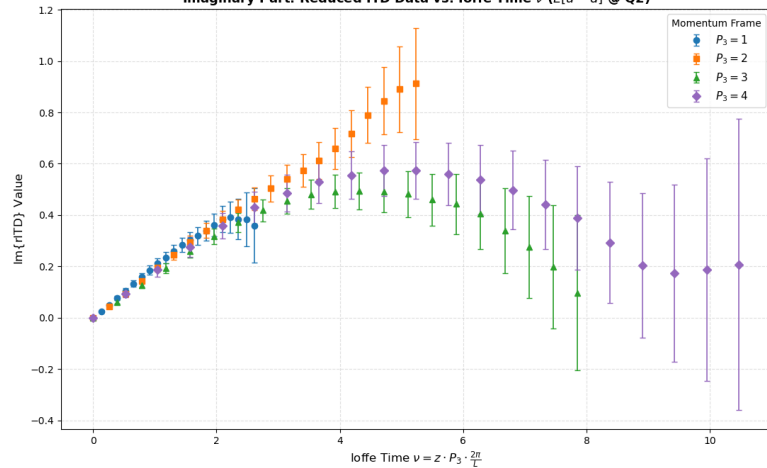


Real Part: Reduced ITD Data vs. Ioffe Time ν (E_{u-d}) @ Q2

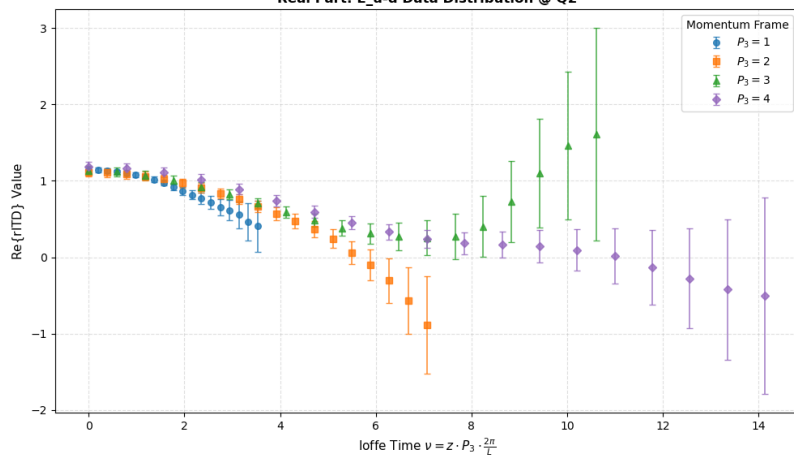


$A = 0.057$ fm

Imaginary Part: Reduced ITD Data vs. Ioffe Time ν (E_{u-d}) @ Q2

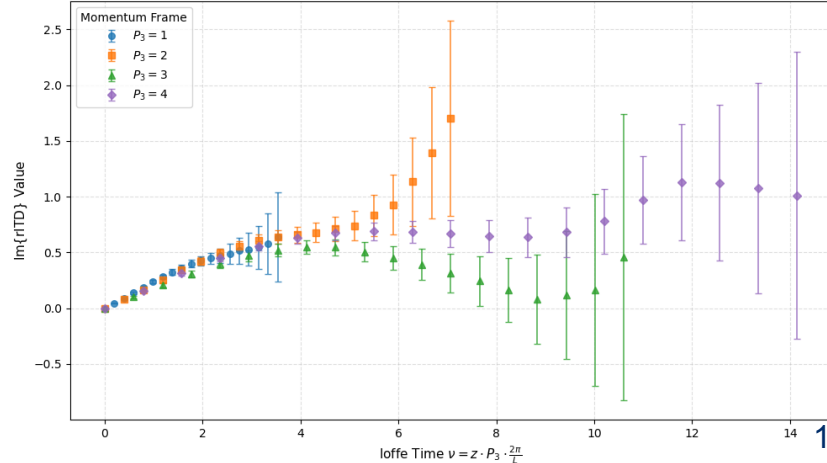


Real Part: E_{u-d} Data Distribution @ Q2



$A = 0.093$ fm

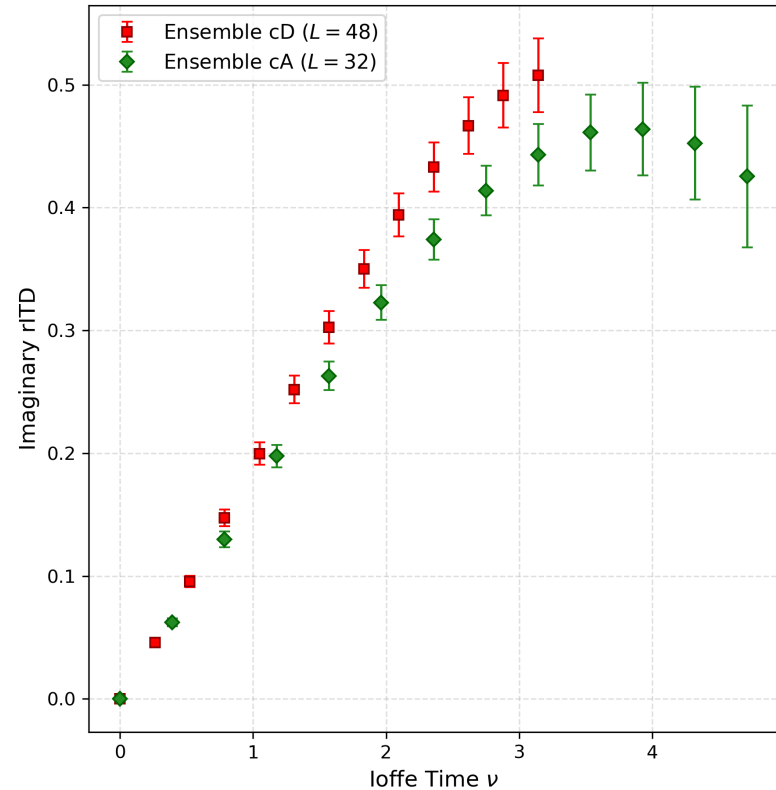
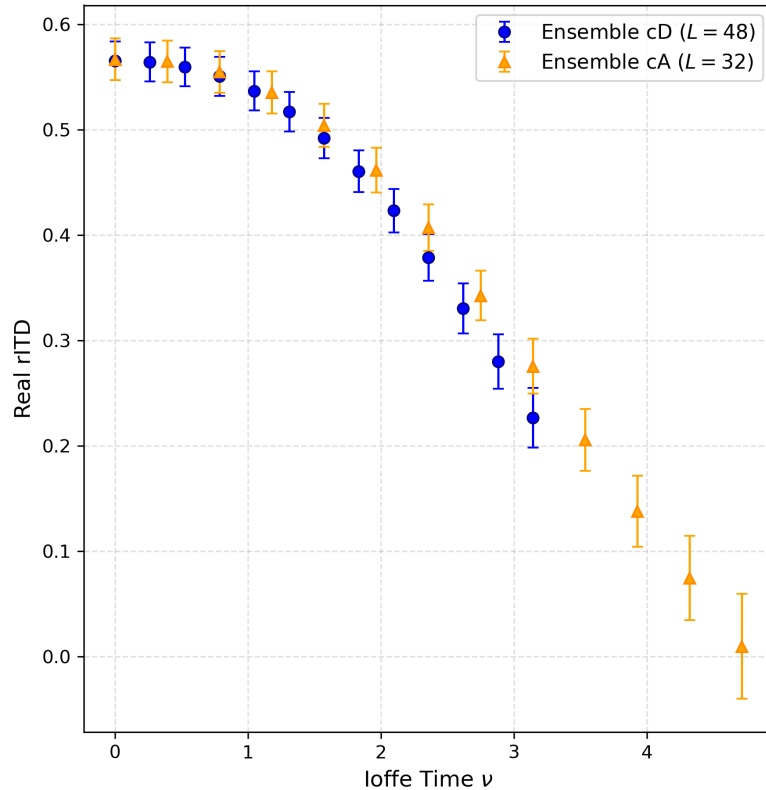
Imaginary Part: E_{u-d} Data Distribution @ Q2



Comparison of rITDs at different lattice spacing (H_u-d)



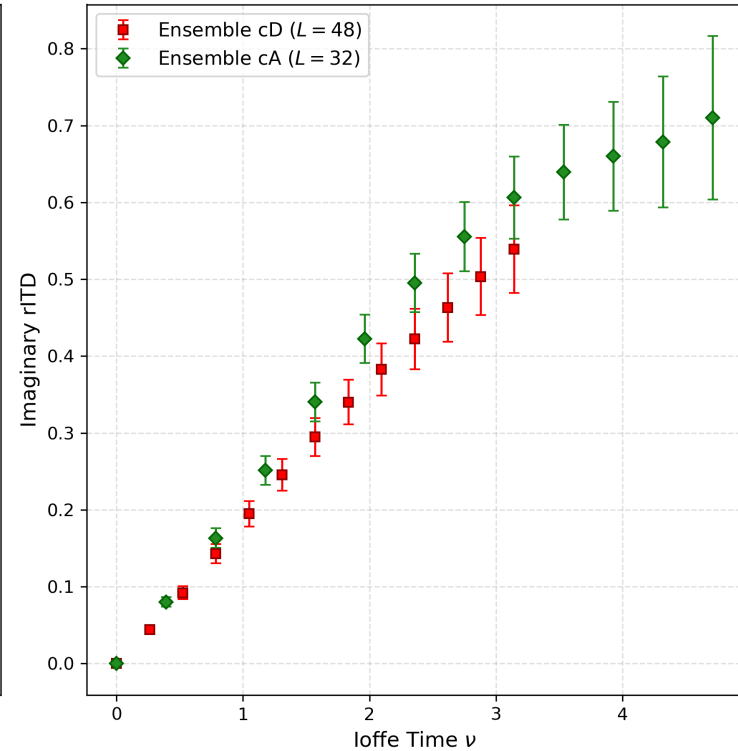
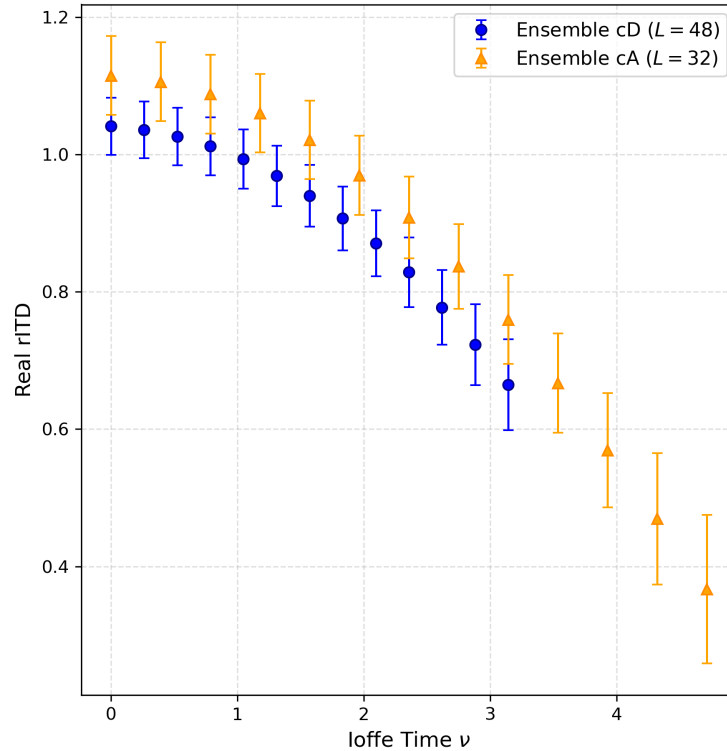
Ensemble Comparison H_u-d: $\Delta = (+2, 0, 0)$ $P_3 = 2$



Comparison of rITDs at different lattice spacing (E_u-d)



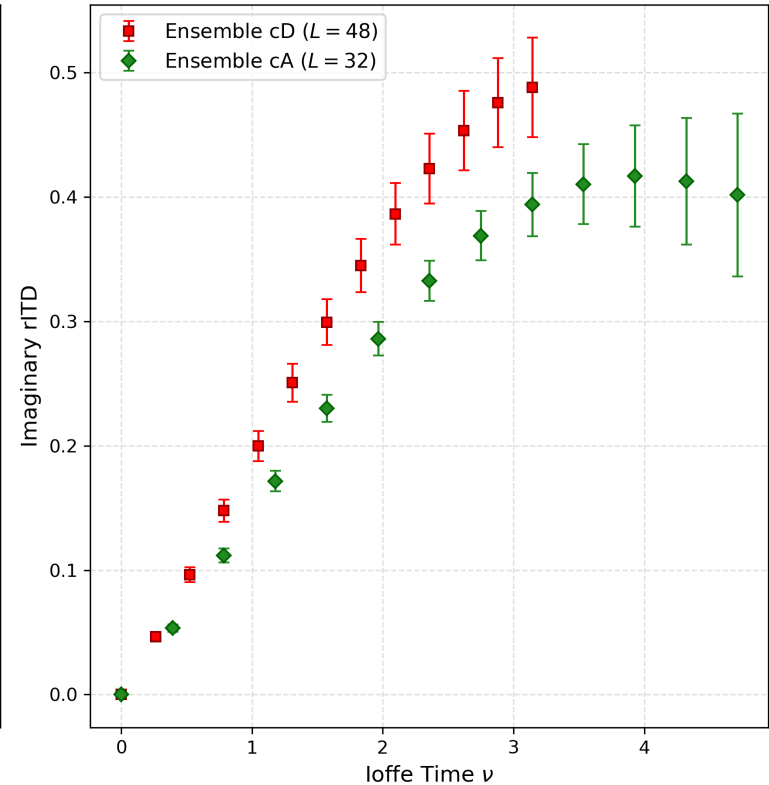
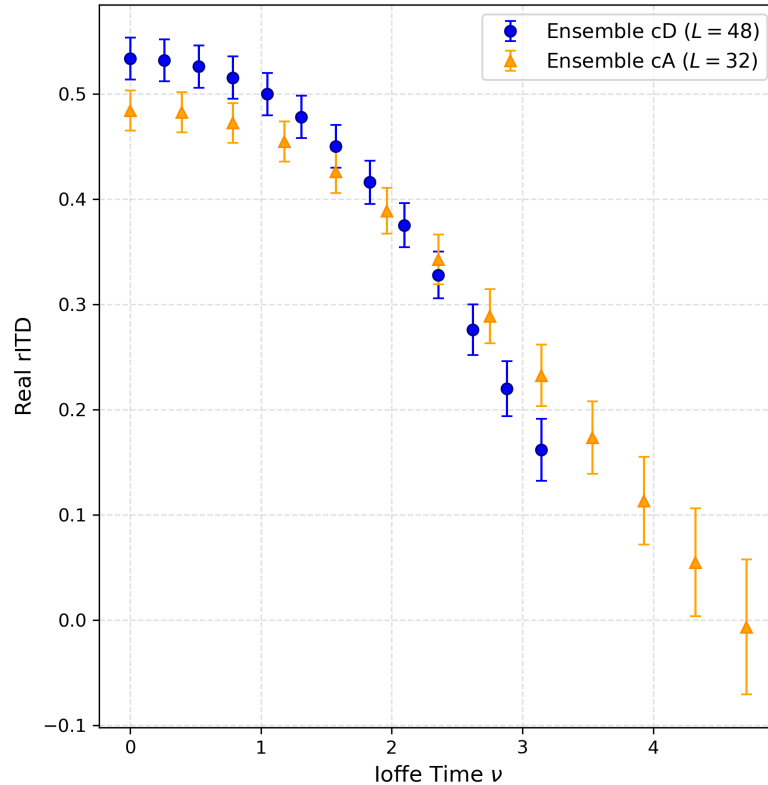
Ensemble Comparison E_u-d: $\Delta = (+2, 0, 0)$ $P_3 = 2$



Comparison of rITDs at different lattice spacing (H_u-d)



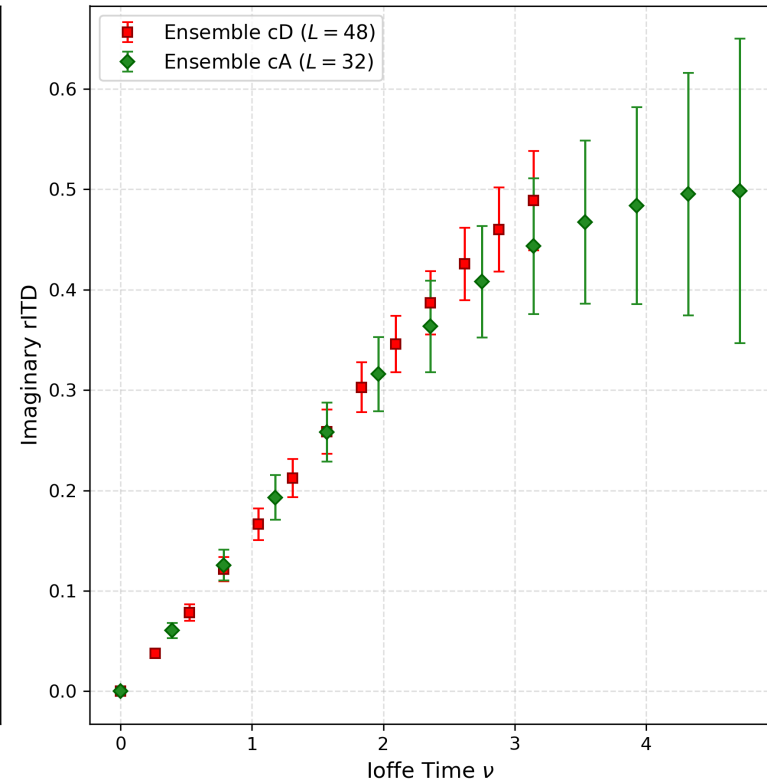
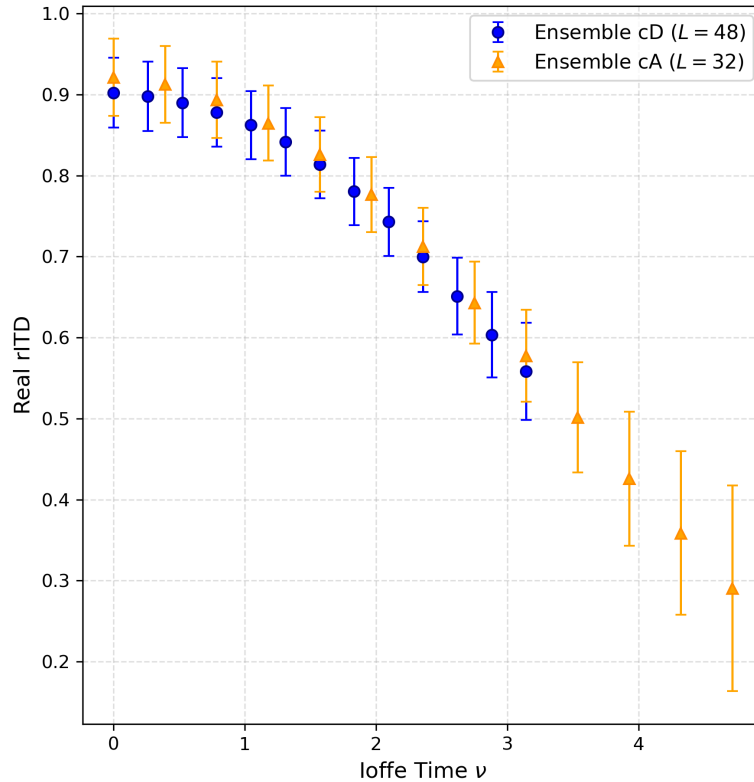
Ensemble Comparison H_u-d: $\Delta = (+1, +2, 0)$ $P_3 = 2$



Comparison of rITDs at different lattice spacing (E_u-d)



Ensemble Comparison E_u-d: $\Delta = (+1, +2, 0)$ $P_3 = 2$



Matching to Light-Cone Coordinates:

Matching to the light-cone is performed at the level of moments. the rITD moments are:

$$\mathcal{M}(\nu, \xi, t, z^2) = \sum_{n=0}^{\infty} \frac{(i\nu)^n}{n!} \mathcal{M}_{n+1}(\xi, t, z^2)$$

The rITD moments are matched to the matched ITD moments via perturbative kernel $C_{n+1, n-k}(\alpha_s, z^2 \mu^2)$:

$$(-i)^n \mathcal{M}_{n+1}(\xi, t, z^2) = (-i)^n \sum_{k=0}^n \bar{F}_{k+1}(\xi, t, \mu) \xi^{n-k} C_{n+1, n-k}(\alpha_s, z^2 \mu^2) + \mathcal{O}(z^2 \Lambda_{QCD}^2)$$

—known as **Short Distance Factorization**.

- My calculation involves $\xi = 0$ and fixed (next leading) order matching.
- Target scale = 2 GeV and $\alpha_s = 0.393945$

Refs

1. H. Dutrieux et al. (HadStruc Collaboration), JHEP. 2024, 162 (2024)
2. M. Bhat et al., PRD 103 (2021) 034510.

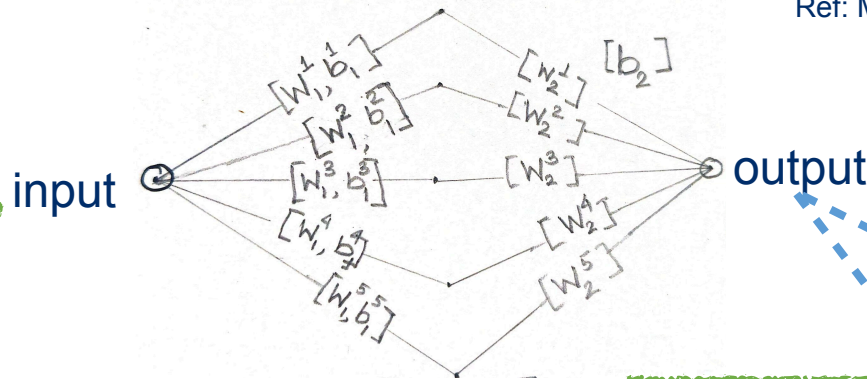
Can we construct Matched ITD Moments as a **Function** of t ?

The moments of Matched ITDs in my calculations are represented using **Artificial Neural Networks (ANNs)**.

Ref: M.-H. Chu et al., JHEP 2026 (2026) 210

$$Q_{xy} = \left(x \cdot \frac{2\pi}{aL}, y \cdot \frac{2\pi}{aL}, 0 \right)$$

$$P_z = z \cdot \frac{2\pi}{aL}$$



hidden
(5 neurones)

$$\begin{bmatrix} t_{11}(\mathcal{Q}1, P1) & t_{12}(\mathcal{Q}2, P1) & \dots & t_{18}(\mathcal{Q}22, P1) \\ t_{21}(\mathcal{Q}1, P2) & t_{22}(\mathcal{Q}2, P2) & \dots & \\ t_{31}(\mathcal{Q}1, P3) & \vdots & & \\ t_{41}(\mathcal{Q}1, P4) & & & t_{48}(\mathcal{Q}22, P4) \end{bmatrix}$$

$$\text{Re} [\overline{\mathcal{M}}(\nu, \xi, t, \mu^2)] = \overline{F}_1(\nu, \xi, t; \mu^2) + \frac{\nu^2}{2} \overline{F}_3(\nu, \xi, t; \mu^2)$$

$$\text{Im} [\overline{\mathcal{M}}(\nu, \xi, t, \mu^2)] = \nu \overline{F}_2(\nu, \xi, t; \mu^2) - \frac{\nu^3}{6} \overline{F}_4(\nu, \xi, t; \mu^2)$$

4 such networks corresponding to $\overline{F}_1(t)$, $\overline{F}_2(t)$, $\overline{F}_3(t)$, $\overline{F}_4(t)$.

How does ANN work? Why Should I use it?



$$\text{The Architecture: } ANN(t) = \sum_{j=1}^5 W_2^j \cdot \tanh [W_1^i t + b_1^i]^j + b_2$$

Random guess values of $[W_1, b_1, W_2, b_2]$ and $t \rightarrow$ ANN prediction \rightarrow calculate χ^2 :

$$\chi^2(W_1, b_1, W_2, b_2) = \sum_{n=0}^{n_{data}} \frac{[\mathcal{M}_{ANN}(t | W_1, b_1, W_2, b_2) - \mathcal{M}_{data}(t)]^2}{\sigma_{\mathcal{M}}^2}$$

Gradient Based Optimizers (Adam) is used to train the parameters such that χ^2 is minimum.

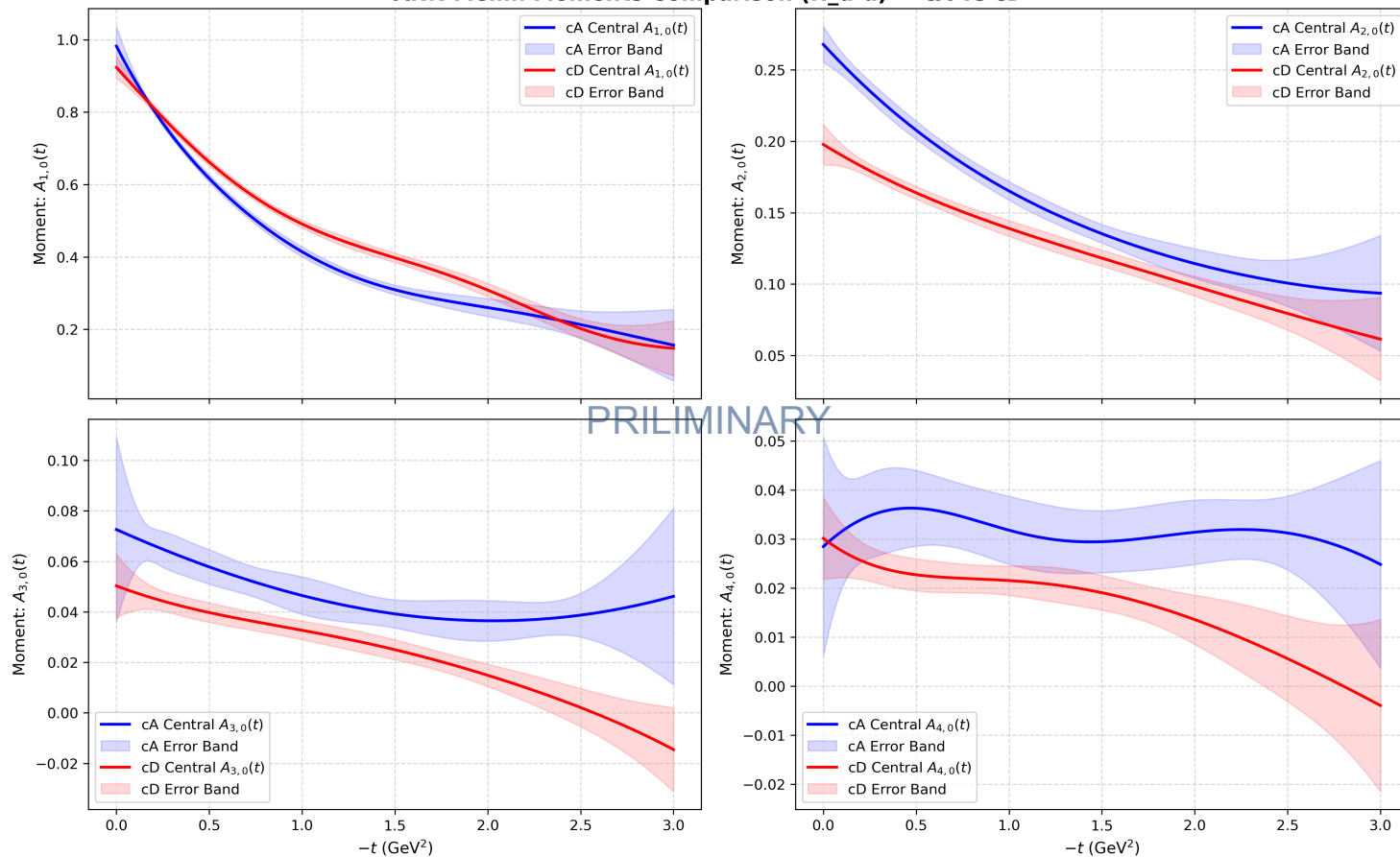
$$\theta_{n+1} \rightarrow \theta_n - \eta \nabla \chi^2(\theta_n)$$

1. Once the parameters are trained, ANN becomes a “**continuous**” function of t .
2. Takes **full t-dependence into account**, (\rightarrow Manuel's talk).
3. **Model Independent** — highly suitable for taking forward limit.

Comparison of Moments for different lattice spacings



ANN Mellin Moments Comparison (H_{u-d}) — cA vs cD



Forward Limit of H GPD,
u-d flavor

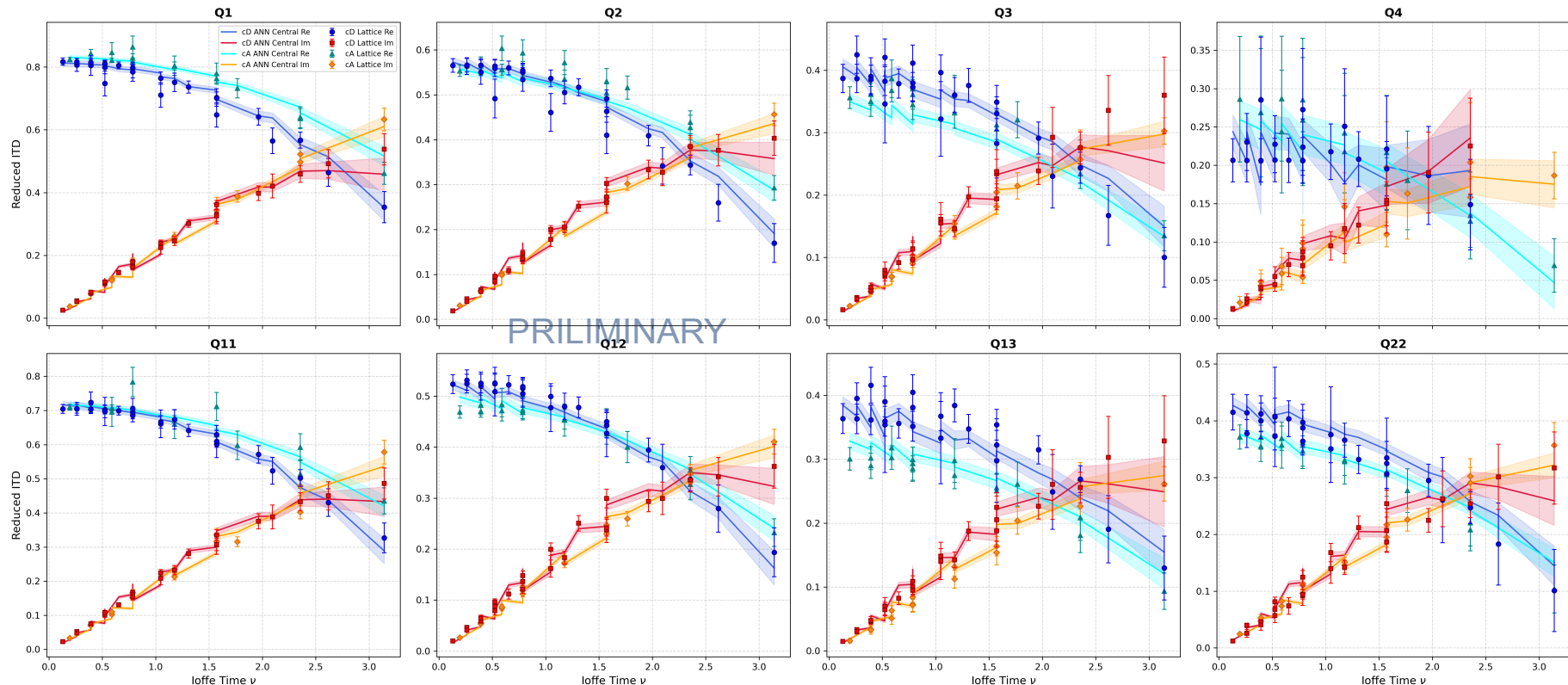
	$A_{10}(0)$	$A_{20}(0)$
[5]	0.982 (47)	0.267 (13)
0.093 fm (cA)	0.982 (54)	0.268 (12)
0.057 fm (cD)	0.923 (30)	0.198 (14)

ETMC' 23: S. Bhattacharya et al.,
PRD 108 (2023) 014507

Fitting data with ANNs (lattice spacings [0.093fm, 0.057fm])



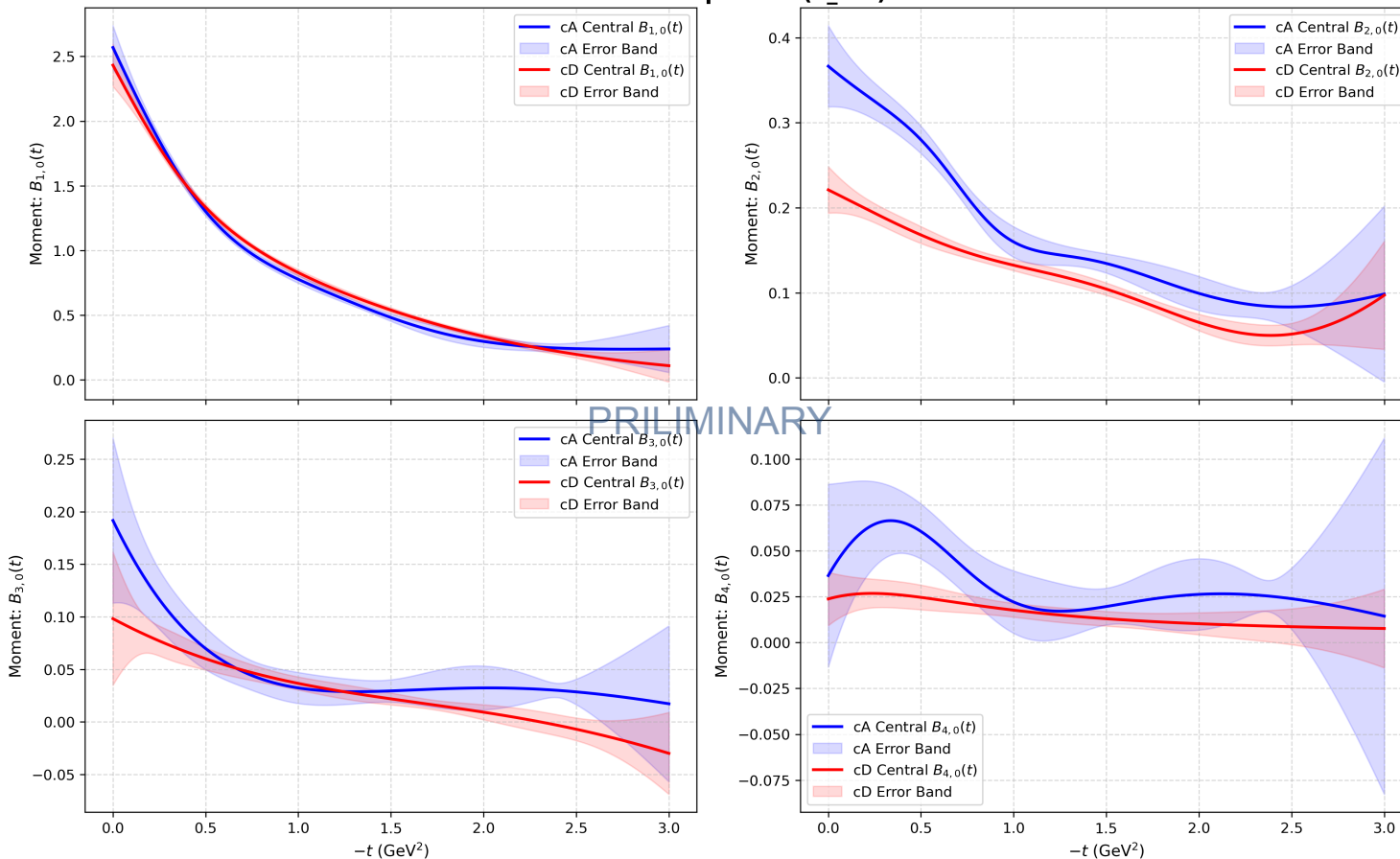
GPD H_{u-d} Ensemble Comparison: cD ($Z_{max} = 6$) vs cA ($Z_{max} = 4$)



Comparison of Moments for different lattice spacings



ANN Mellin Moments Comparison (E_{u-d}) – cA vs cD



Forward Limit of E GPD,
u-d flavor

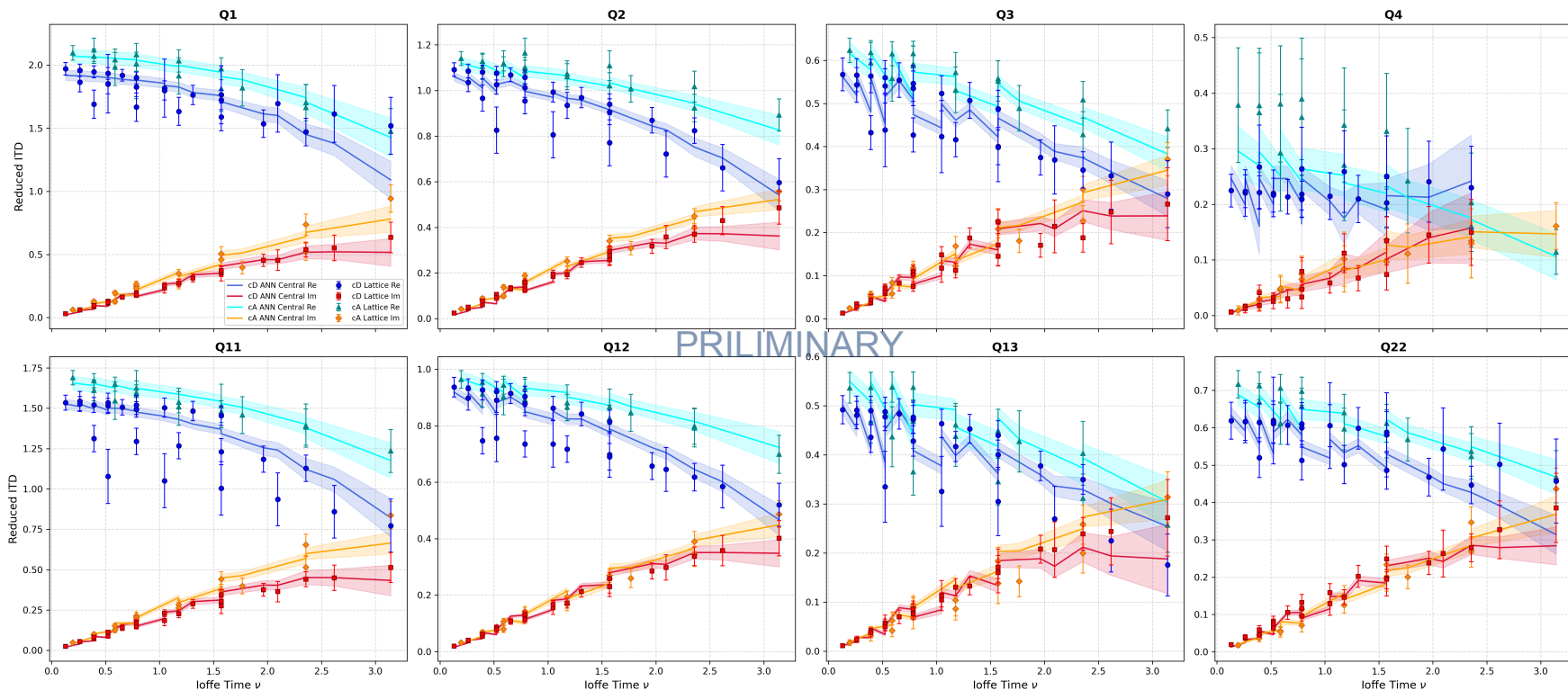
	$B_{10}(0)$	$B_{20}(0)$
[5]	2.540 (221)	0.295 (38)
0.093 fm (cA)	2.569 (169)	0.366 (47)
0.057 fm (cD)	2.432 (164)	0.221 (27)

ETMC' 23: S. Bhattacharya et al.,
PRD 108 (2023) 014507

Fitting data with ANNs (lattice spacings [0.093fm, 0.057fm])



GPD E_u-d Ensemble Comparison: cD ($z_{max} = 6$) vs cA ($z_{max} = 4$)



— Excited State Analysis —

Fitting Ansatz for Excited State Analysis:

The functional forms of two- and three-point correlations —

$$C^{2pt}(\vec{P}, t) = |Z_0|^2 e^{-E_0 t} \left(1 + |f_{10}|^2 e^{-\Delta E} \right) + \text{terms}(n \geq 2)$$

$$C^{3pt}(\vec{P}, t) = |Z_0|^2 e^{-E_0 t} \left(\langle 0 | \mathcal{O} | 0 \rangle + f_{10} \langle 1 | \mathcal{O} | 0 \rangle e^{-\Delta E(t-\tau)} + f_{10}^\dagger \langle 1 | \mathcal{O} | 0 \rangle e^{-\Delta E \tau} + |f_{10}|^2 \langle 1 | \mathcal{O} | 1 \rangle e^{-\Delta E t} \right) + \text{terms}(n \geq 2)$$

The lattice calculated **PDF**/GPD matrix elements are complex numbers. Hence, the fitting function for ratio has also real and imaginary parts:

$$\text{Re} [\text{Ratio}(P_3, z, t_s)] = \frac{\text{Re}(A_{00}) + \text{Re}(B) [e^{-\Delta E(t_s-\tau)} + e^{-\Delta E \tau}] + |f_{10}|^2 \text{Re}(A_{11}) e^{-\Delta E t_s}}{1 + |f_{10}|^2 e^{-\Delta E t_s}}$$

$$\text{Im} [\text{Ratio}(P_3, z, t_s)] = \frac{\text{Im}(A_{00}) + \text{Im}(B) [e^{-\Delta E(t_s-\tau)} - e^{-\Delta E \tau}] + |f_{10}|^2 \text{Im}(A_{11}) e^{-\Delta E t_s}}{1 + |f_{10}|^2 e^{-\Delta E t_s}}$$

With $\text{Re}(A_{00}) + i \text{Im}(A_{00}) = \langle 0 | \mathcal{O} | 0 \rangle$, $\text{Re}(B) + i \text{Im}(B) = f_{10} \langle 1 | \mathcal{O} | 0 \rangle$, $\text{Re}(A_{11}) + i \text{Im}(A_{11}) = \langle 1 | \mathcal{O} | 1 \rangle$

Extracting information out of C^{2pt}

Fitting Window Determination: it is essential to determine the interval $[t_{min}, t_{max}]$, where contributions are dominated by the nucleon ground state and the first excited state.



Perform Correlated Fits

Extracting information out of C^{2pt} - *The Correlated Fits:*

Step 1: Generalized χ^2 : $\chi^2(\theta) = [C_{\text{ansatz}}^{2pt}(t|\theta) - C_{\text{data}}^{2pt}(t)]^T [Cov^{2pt}]^{-1} [C_{\text{ansatz}}^{2pt}(t|\theta) - C_{\text{data}}^{2pt}(t)]$

$$Cov^{2pt} = \frac{1}{n-1} \sum_{i=1}^n (\vec{c}_i - E[\vec{c}])(\vec{c}_i - E[\vec{c}])^T$$

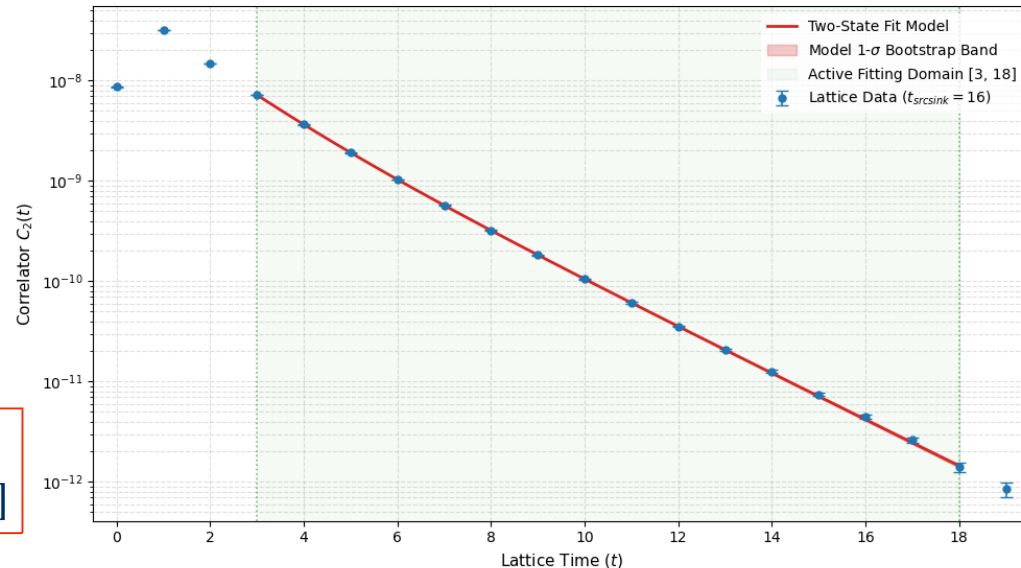
Ref: M. Bruno and R. Sommer, Comput. Phys. Commun. 285, 108643 (2023)

Includes correlation between the data from t_i to t_j for $i \neq j$ and $i = j$.

Step 2: Generalized Chi2, is minimised with an optimizer (such as Nelder-Mead).

Reduced χ^2 normalizes to ~ 1 , for appropriate fitting windows, such as $t \in [3, 18]$

Two-Point Correlator: Fit Model vs. Raw Data ($t_{\text{srcsink}} = 16$)



Towards *Two State Fits* of the Ratio:

Parameters estimated from fitting C^{2pt} should be unique and stable —

$$|Z_0|^2, E_0, |f_{10}|^2, \Delta E = 2.00744 \times 10^{-8} \pm 1.10384 \times 10^{-9}, 0.53013 \pm 0.00506, 2.40937 \pm 0.14965, 0.38644 \pm 0.01328$$

(measured in lattice unites, $a = 0.057$ fm)

$$E_0 = 0.53836 \pm 0.00450 \text{ from } E_{eff}(t) \text{ vs } t$$

In **sequential fit** procedure, the parameters ($|f_{10}|^2$ and ΔE) are substituted into the ratio ansatz —

$$\text{Re}[\text{Ratio}(P_3, t, z)] = \frac{1}{D(t_s)} \text{Re}(A_{00}) + \left[\frac{e^{-\Delta E(t_s - \tau)} + e^{-\Delta E \tau}}{D(t_s)} \right] \text{Re}(B) + \frac{|f_{10}|^2 e^{-\Delta E t_s}}{D(t_s)} \text{Re}(A_{11}) \text{ With } D(t_s) = (1 + e^{-\Delta t_s})$$

My calculations involved $t_s = 10a, 12a, 14a$ and $16a$ in fm, for $a = 0.057$ fm.

The data that takes contributions from nucleon ground state and the first excited state lies in the range: $t_s \in [3, t_s - 3]$

More on *Two State Fits* of the Ratio:

The model becomes linear in terms of parameters $\theta \in [\text{Re}(A_{00}), \text{Re}(B), \text{Re}(A_{11})]$:

$$\text{Generalized } \chi^2(\theta) = [Y - X_{\text{coeff}} \cdot \theta]^T [\text{Cov}(Y)]^{-1} [Y - X_{\text{coeff}} \cdot \theta]$$

Generalized χ^2 can be minimized as:

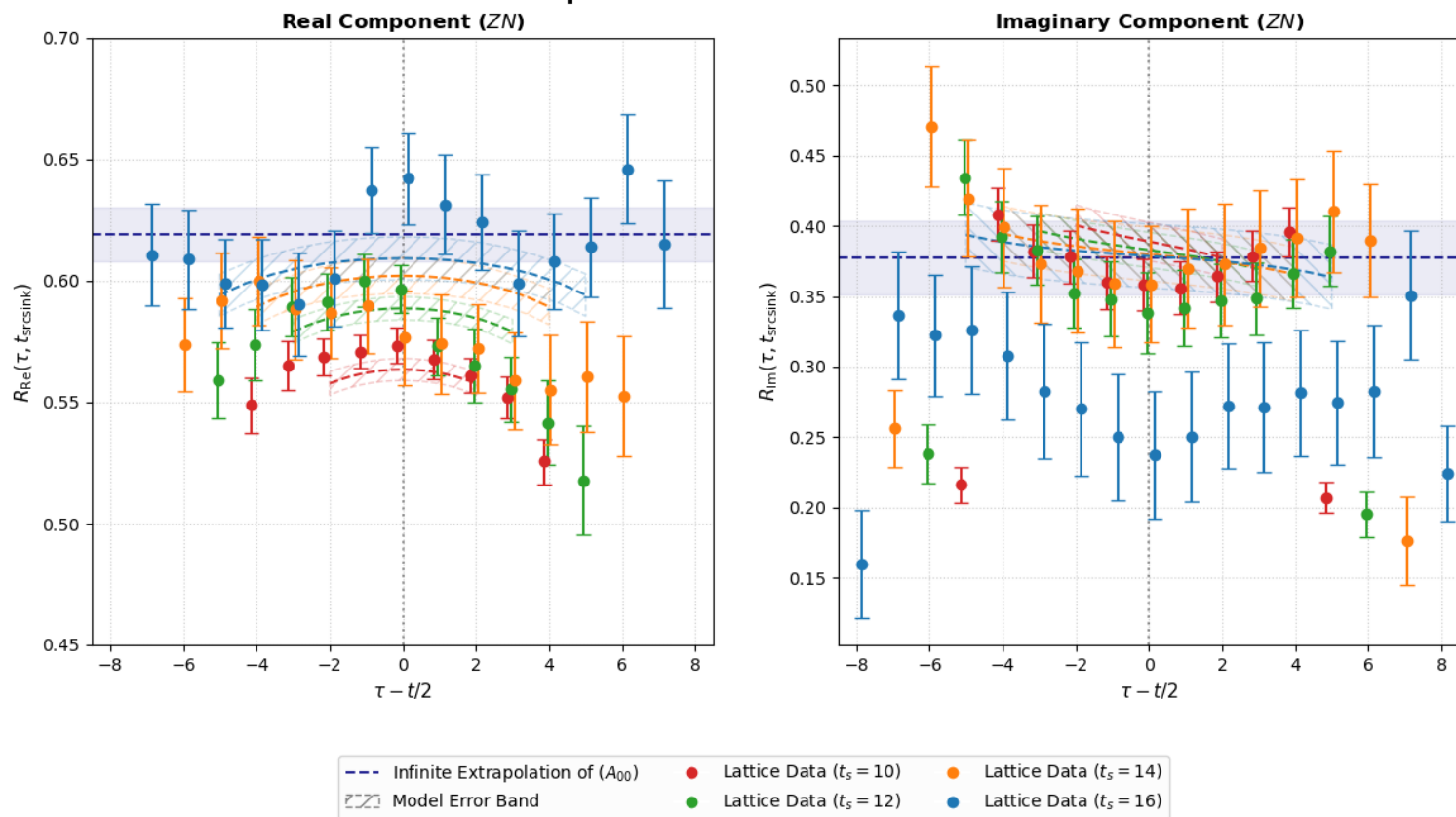
$$\nabla_{\theta^T} \chi^2(\theta) = 0 \longrightarrow \theta_{\text{sol}} = \left[X_{\text{coeff}}^T \text{Cov}(Y) X_{\text{coeff}} \right]^{-1} X_{\text{coeff}}^T W Y$$

Sequential Fit —

1. Solves the problem exactly and **uniquely** determines the parameters.
2. The analysis is exactly the same for real and imaginary parts.

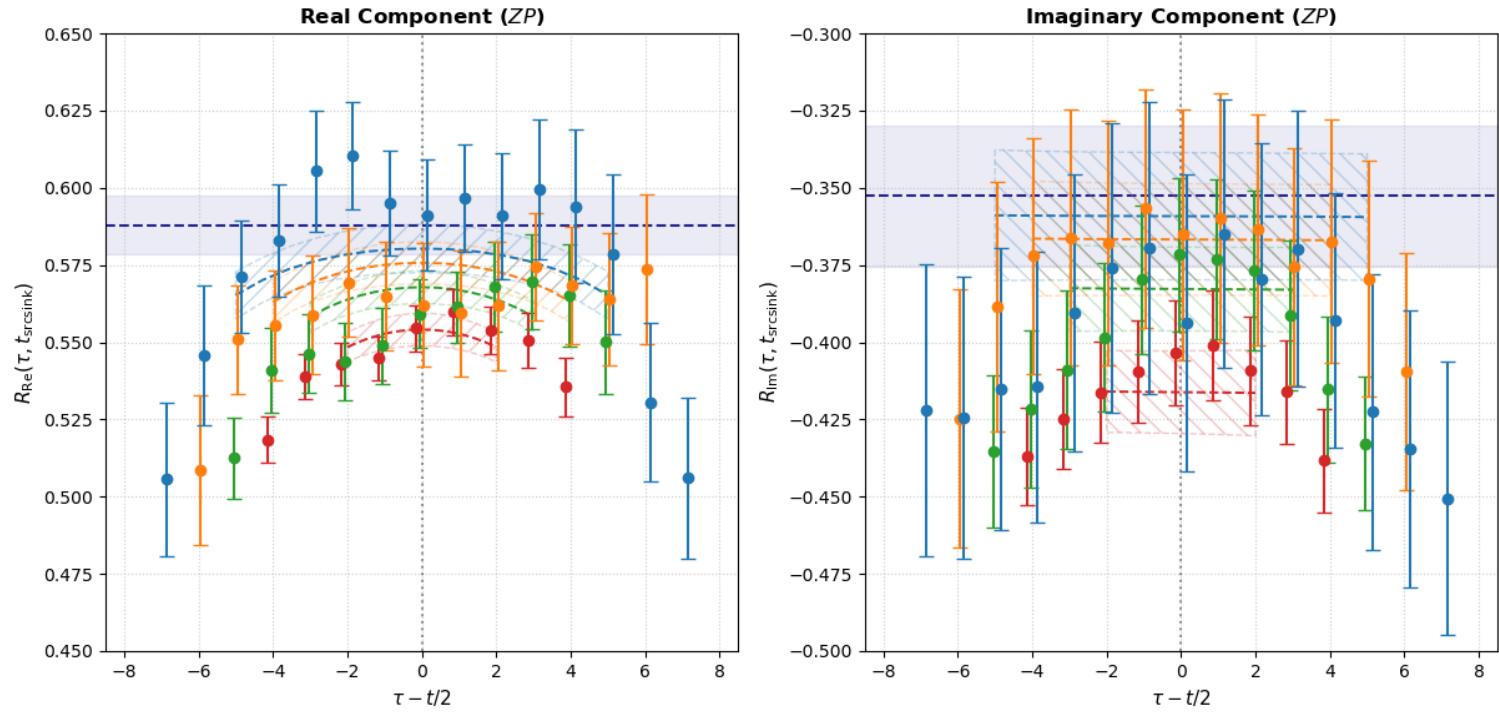
Results from *Two State Sequential Fits*:

Sequential Fit of ratio at $z = -5$



Results from *Two State Sequential Fits*:

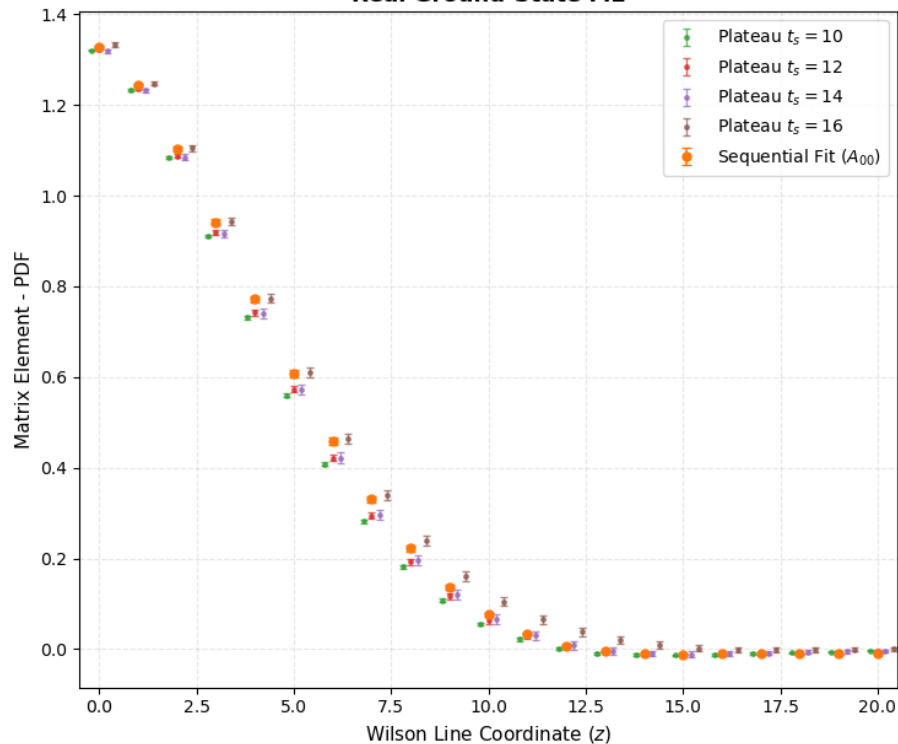
Sequential Fit of ratio at $z = +5$



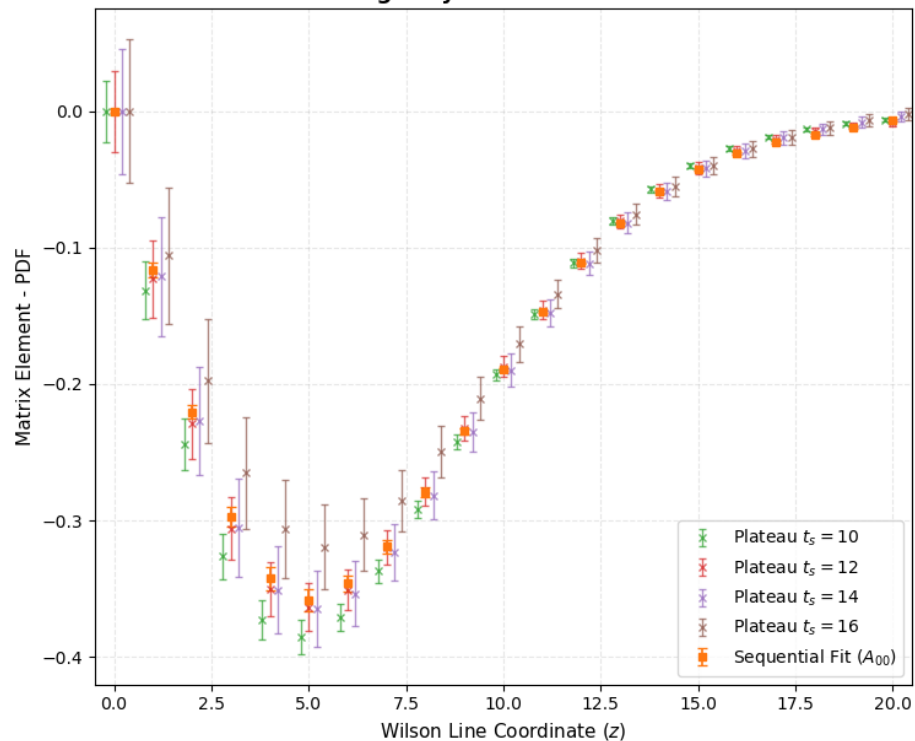
Ground State Matrix Elements from Sequential and Plateau fit:



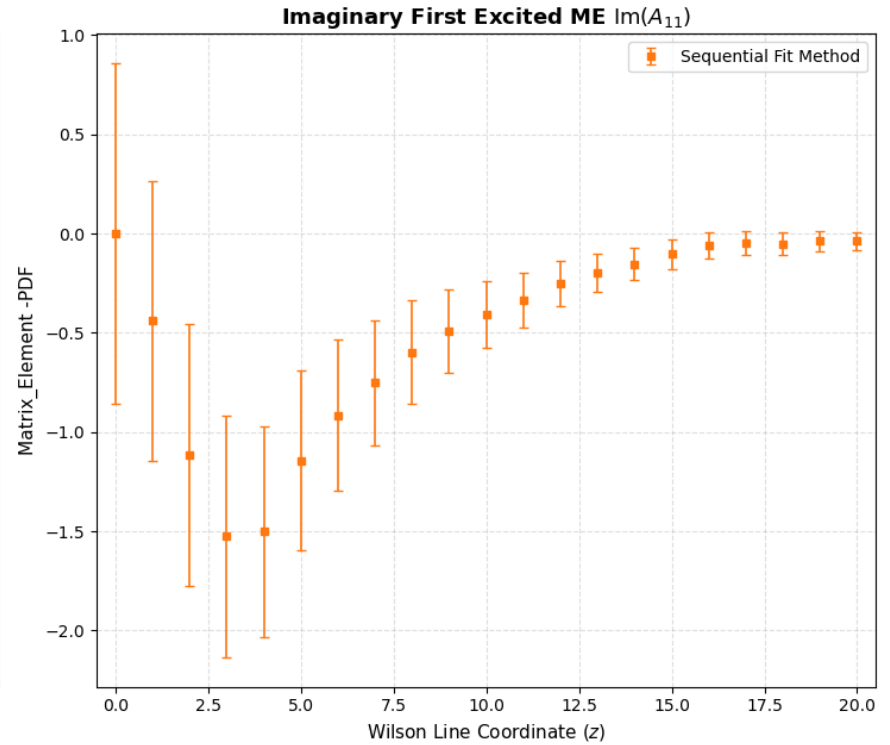
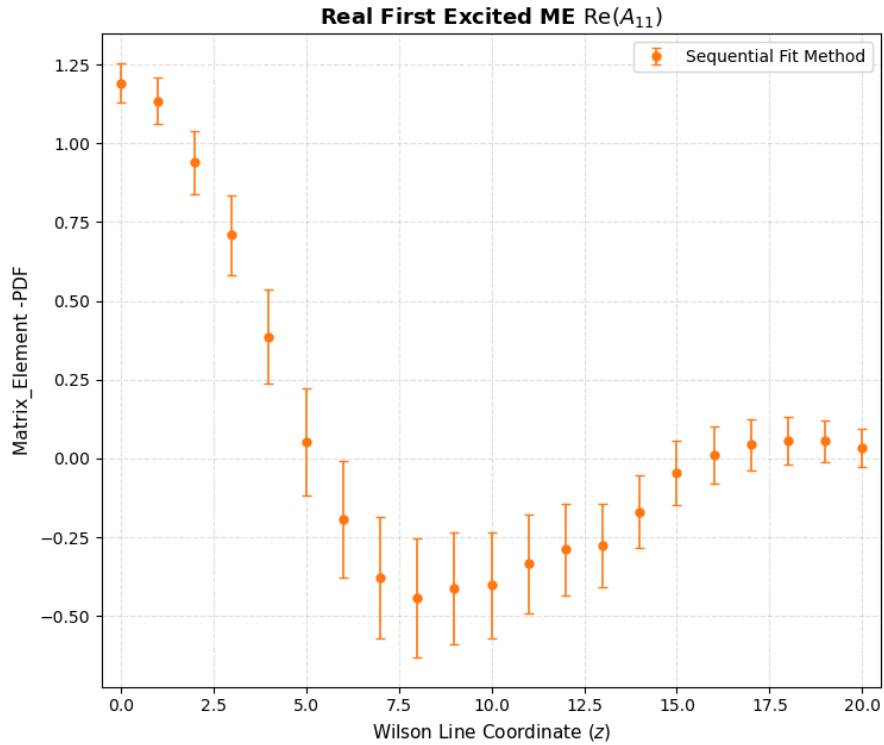
Real Ground-State ME



Imaginary Ground-State ME



First Excited State Matrix Elements from Sequential fit:



Why do we need Two-State Fits?

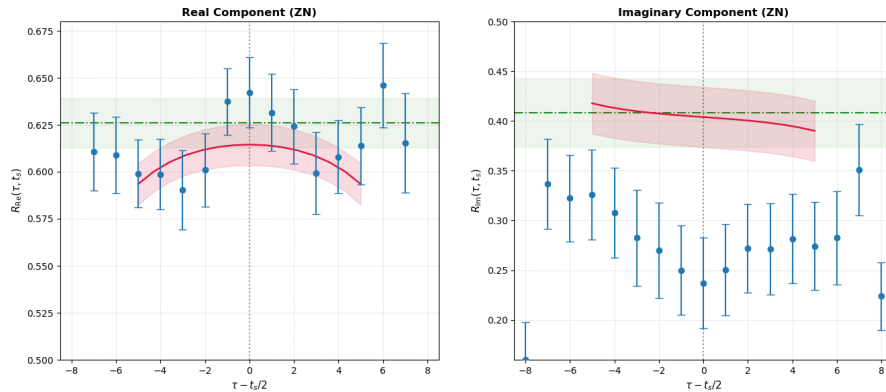


The signal-to-noise:
$$\frac{R(t)}{\sigma(t)} \simeq \sqrt{N_{cfg}} \exp\left(-\left[E_n(P) - \frac{3}{2}m_\pi\right]t\right)$$

Ref: T. DeGrand and C. E. Detar, Lattice methods for quantum chromodynamics (World Scientific, 2006).

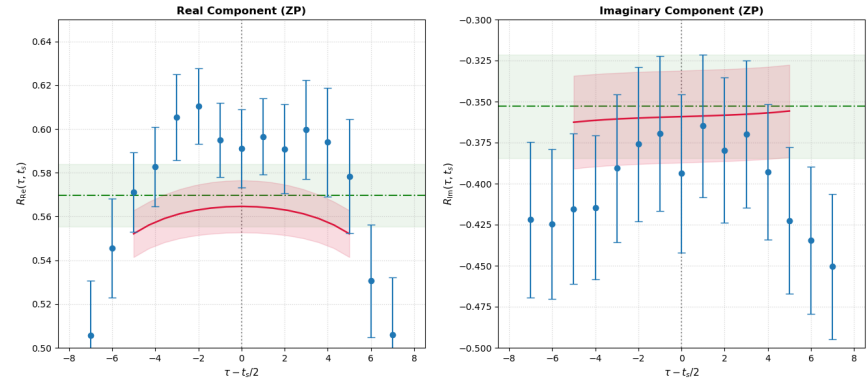
1. restricts from extracting the Matrix Elements at higher momentum boots (say $\sim 2\text{GeV}$), for sufficiently high value of t_s .
2. Let us see if lower values of $t_s \in [10a, 12a, 14a]$ can reproduce the results for $t_s = 16a$

Projection of $t_s = 16$ from $t_s = 10, 12, 14$ using seq. fit, at $z = -5$



■ Lattice Data ($t_s = 16$)
 — prediction for $t_s = 16$ from $t_s \in [10, 12, 14]$
- - - Sequential Fit: Infinite Extrapolation (A_0)

Projection of $t_s = 16$ from $t_s = 10, 12, 14$ using seq. fit, at $z = +5$

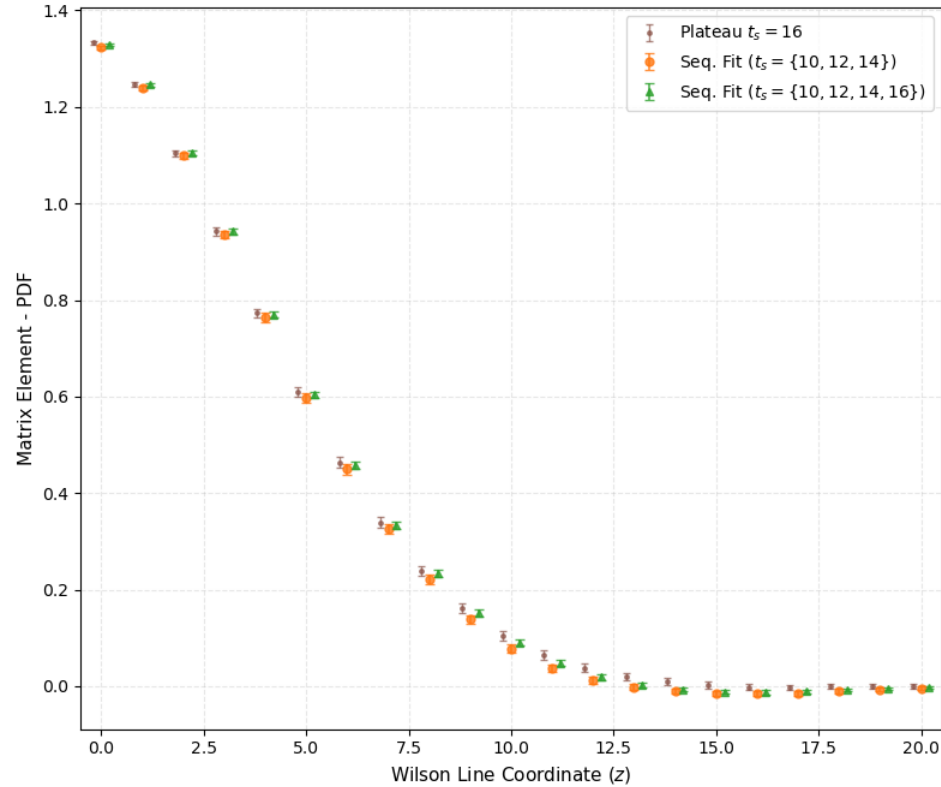


■ Lattice Data ($t_s = 16$)
 — prediction for $t_s = 16$ from $t_s \in [10, 12, 14]$
- - - Sequential Fit: Infinite Extrapolation (A_0)

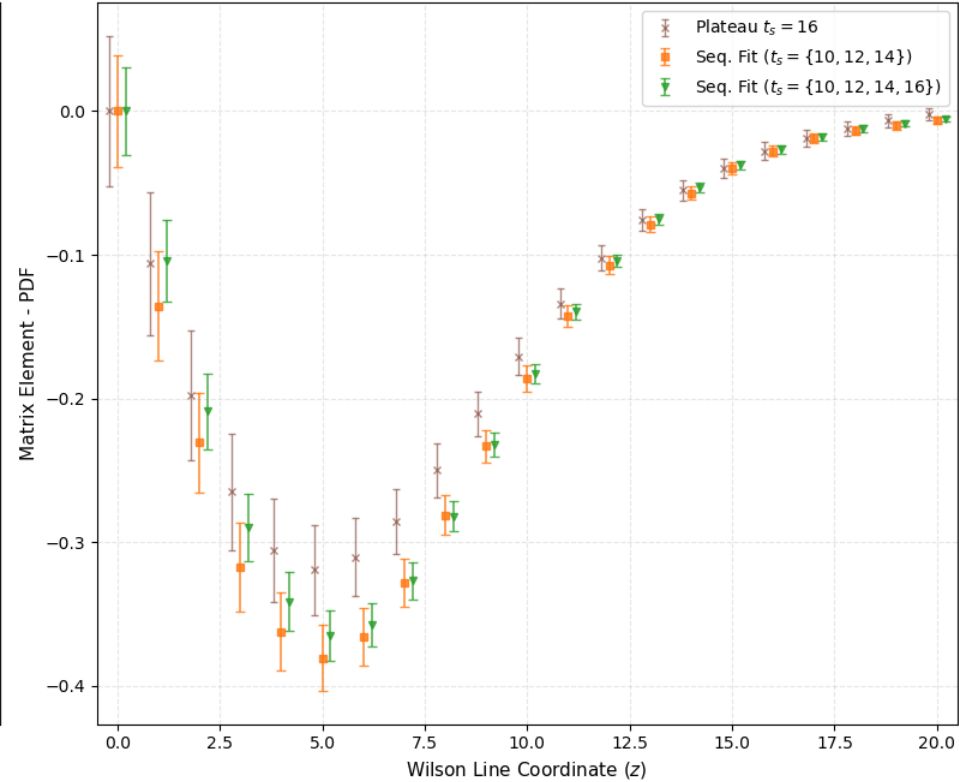
The comparison that $t_s = 10a, 12a, 14a$ can reproduce $16a$.



Real Ground-State ME



Imaginary Ground-State ME



Conclusions and What's Next?



- ANN captures full t -dependence of Mellin moments model-independently. It serves as the baseline for the continuum extrapolations in future.
- Quantifying ANN-related systematic errors is important, specifically on overfitting and underfitting via regularization techniques.
- We'll perform the excited-state analysis using the Direct Fit method and extend the entire framework for GPDs.

**THANK YOU
FOR YOUR
ATTENTION**

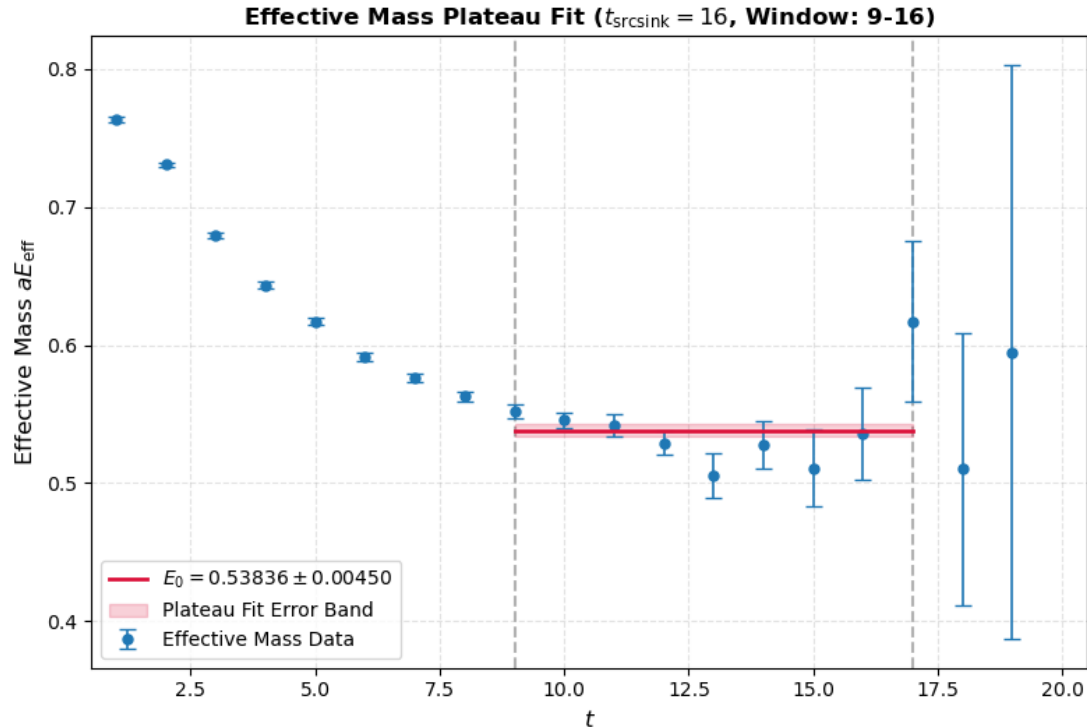


Questions, please?

Back-up slide [1] —



In sequential fit procedure, fitting C2pt also evaluates E_0 - which needs to be consistent with the value predicted from E_{eff} .



Back-up slide [2] —



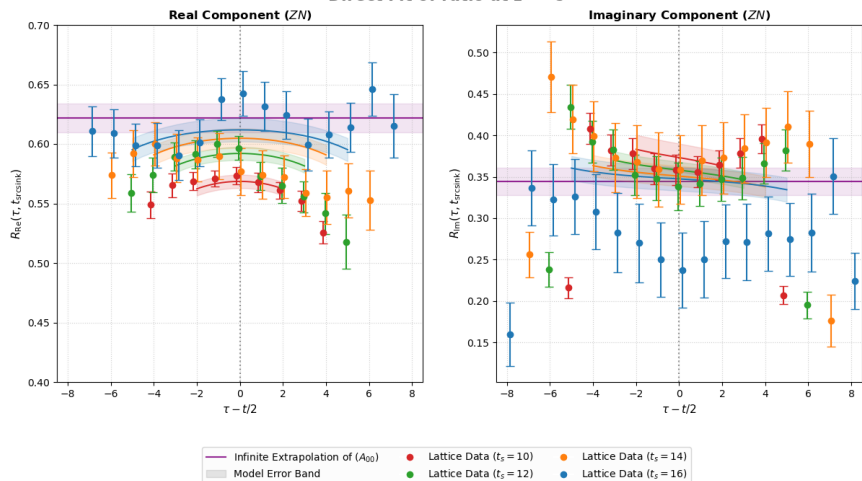
A few words about the direct fits:

- A. In this procedure, the $\text{Re}(\text{Ratio})$ ansatz is fitted against the data and the parameters: $\text{Re}(A_{00})$, $\text{Re}(B)$, $|f_{10}|^2$, ΔE and $\text{Re}(A_{11})$ are determined. Similarly for the imaginary part.
- B. Correlated fitting is performed such that, we can appropriately *capture the fitting window*, For example the reduced χ^2 turns out ~ 0.9 and ~ 1.2 for training sample 0 data, for window $\tau \in [3, t_s - 3]$, otherwise χ^2 doesn't get normalized.
- C. Procedure is stable, yet dependent on guess values. χ^2 is minimized in a controlled/constrained way such that that some parameters converges to an appropriate physical values.

Back-up slide [3] —

Some results from direct fits:

Direct Fit of ratio at $z = -5$



Direct Fit of ratio at $z = +5$

