

# Pion Distribution Amplitudes in the Continuum Limit

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# Meson Distribution Amplitudes

Meson DA: probability of finding the meson in  $q_1 - \bar{q}_2$  fock state.

$$\phi_M(x, \mu) = \frac{i}{f_M} \int \frac{d\mathcal{E}}{2\pi} e^{i(x-1)P \cdot \mathcal{E}n} \langle M(P) | \bar{\psi}(0) n \cdot \gamma \gamma_5 U(0, \mathcal{E}n) \psi(\mathcal{E}n) | 0 \rangle$$

Meson DAs important for understanding how light-quark hadron masses emerge from QCD

Like parton distribution functions:

- Universal quantity
- Non-perturbative

Unlike parton distribution functions:

- Less constrained by experiments
- Model-dependent calculations
- No global-fitting result to compare with

# Lattice Setup

- 2+1+1 highly improved staggered quarks (HISQ) in the sea
- Clover fermion action for the valence quarks
- $M_\pi \approx 310$  MeV, 220 MeV

Ensemble	a (fm)	$M_\pi$ (MeV)	$M_\pi L$	$L^3 \times T$	$P_z$ (GeV)	$N_{\text{src}}$	$N_{\text{conf}}$	$N_{\text{meas}}$
a15m310	0.1510	320(5)	3.93	$16^3 \times 48$	{1.02, 1.54, 2.05}	24	452	10,848
a12m310	0.1207	310(3)	4.55	$24^3 \times 64$	{1.28, 1.71, 2.14}	192	1013	194,496
a12m220	0.1184	228(2)	4.38	$32^3 \times 64$	{1.31, 1.63, 1.96}	384	959	368,256
a09m310	0.0888	313(3)	4.51	$32^3 \times 96$	{1.31, 1.74, 2.18}	48	889	39,648
a06m310	0.0582	320(2)	3.90	$48^3 \times 144$	{1.33, 1.77, 2.22}	4	593	2,372

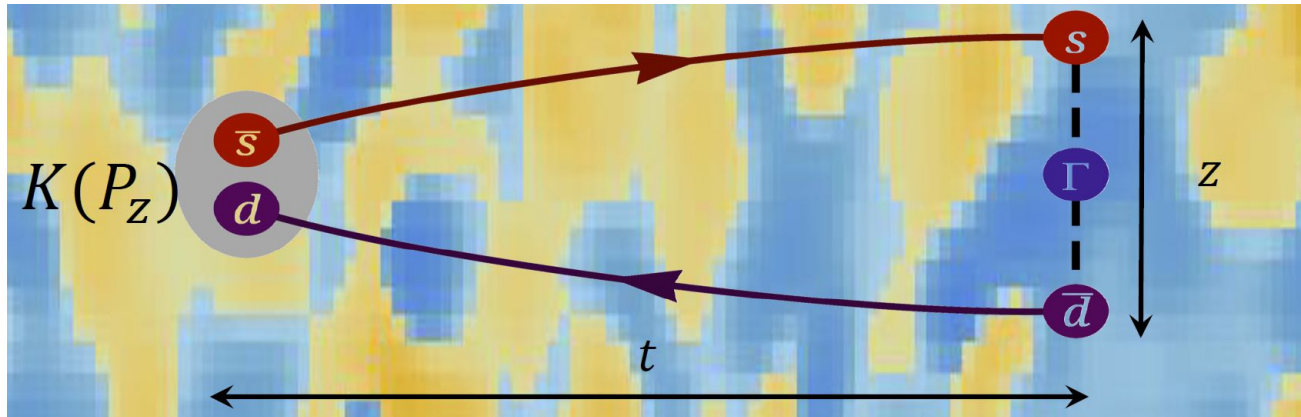
Thanks to MILC collaboration for sharing their 2+1+1 HISQ lattices

# Matrix Elements

DA two-point correlator:

$$C_M^{\text{DA}}(z, P, t) = \left\langle 0 \left| \int d^3y e^{i\vec{P}\cdot\vec{y}} \bar{\psi}_1(\vec{y}, t) \gamma_z \gamma_5 U(\vec{y}, \vec{y} + z \hat{z}) \psi_2(\vec{y} + z \hat{z}, t) \bar{\psi}_2(0, 0) \gamma_5 \psi_1(0, 0) \right| 0 \right\rangle$$

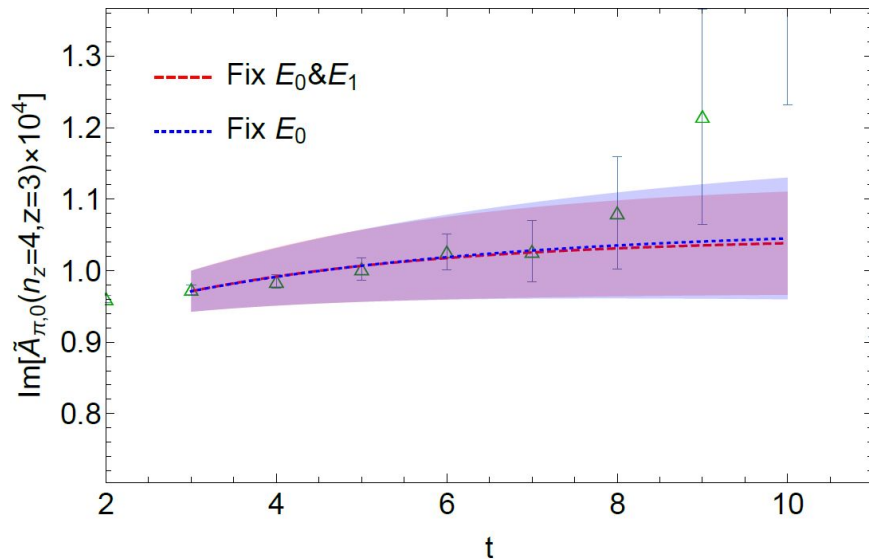
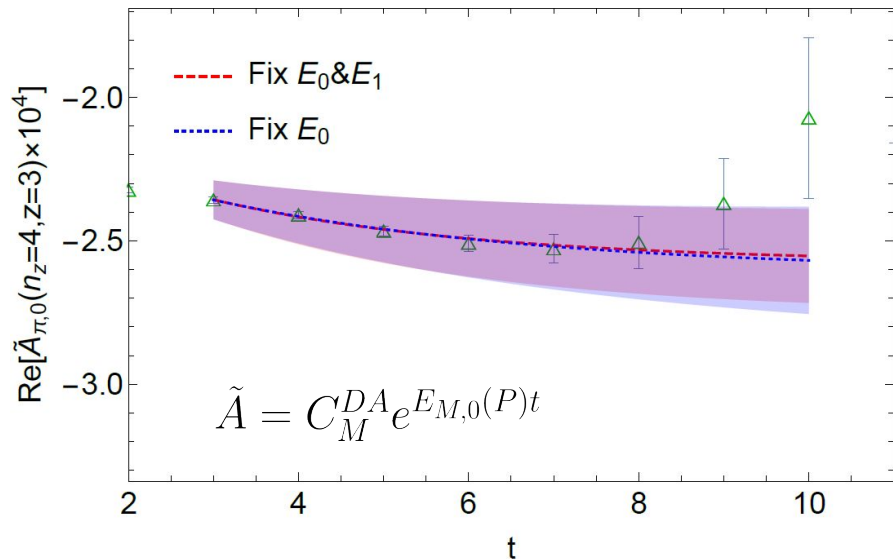
Matrix elements in coordinate space:  $\tilde{h}_M(z, P_z) = \langle M(P) | \bar{\psi}(0) \gamma^z \gamma_5 U(0, z) \psi(z) | 0 \rangle$



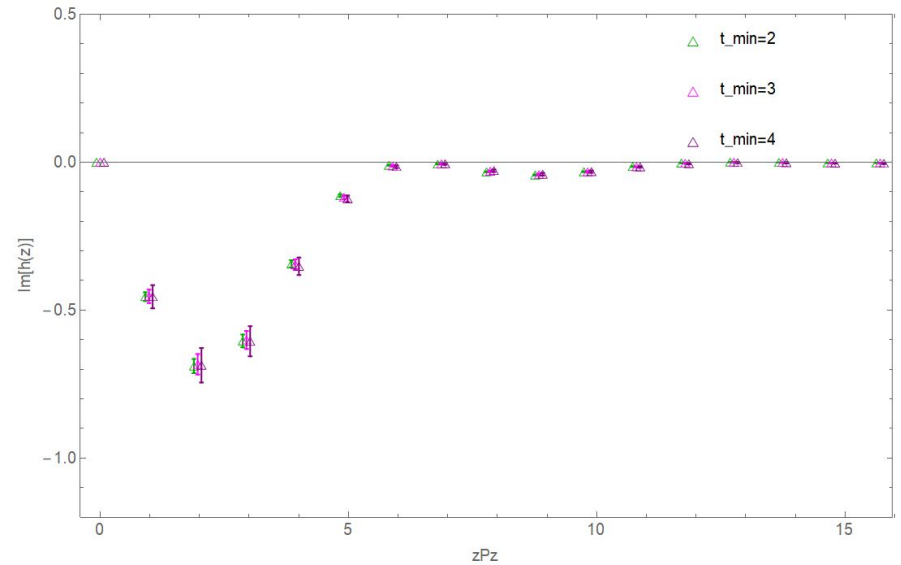
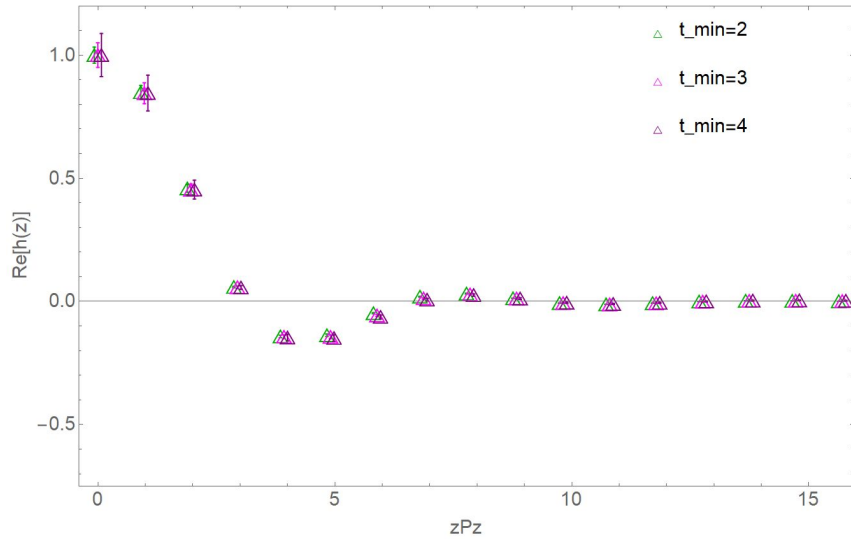
# Matrix Elements

The DA matrix elements can be extracted by a two-point correlator fit to the form:

$$C_M^{DA}(z, P, t) = A_{M,0}^{DA}(P, z)e^{-E_{M,0}(P)t} + A_{M,1}^{DA}(P, z)e^{-E_{M,1}(P)t} + \dots$$



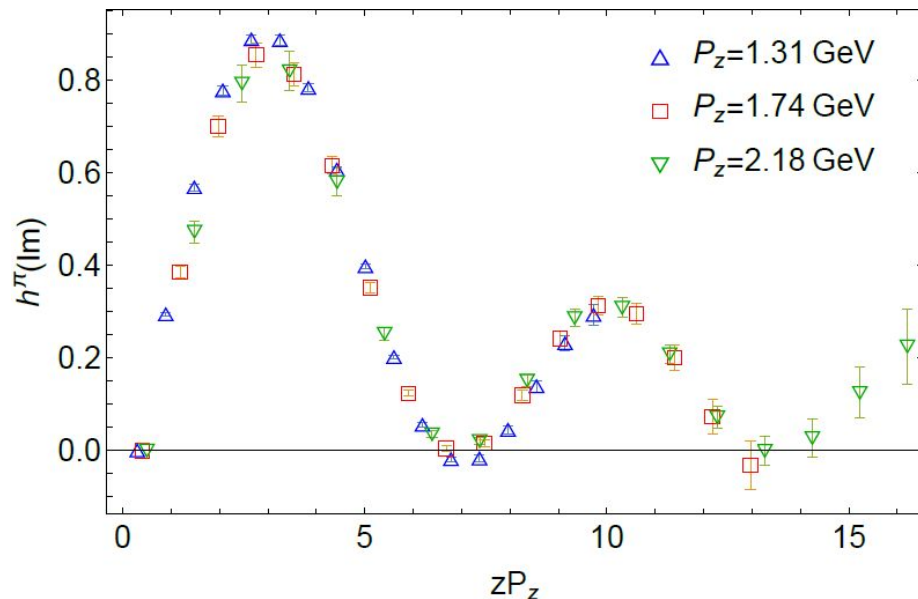
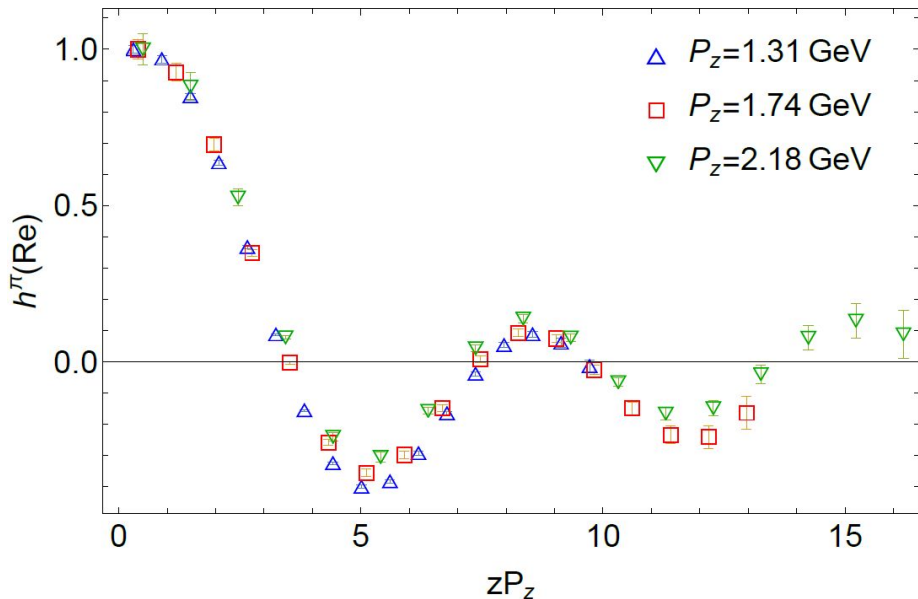
# Bare Matrix Elements: $t_{min}$ Dependence



- a09m310,  $P_z = 2.18$  GeV
- $t_{min} = \{2, 3, 4\} \rightarrow t$  dependence

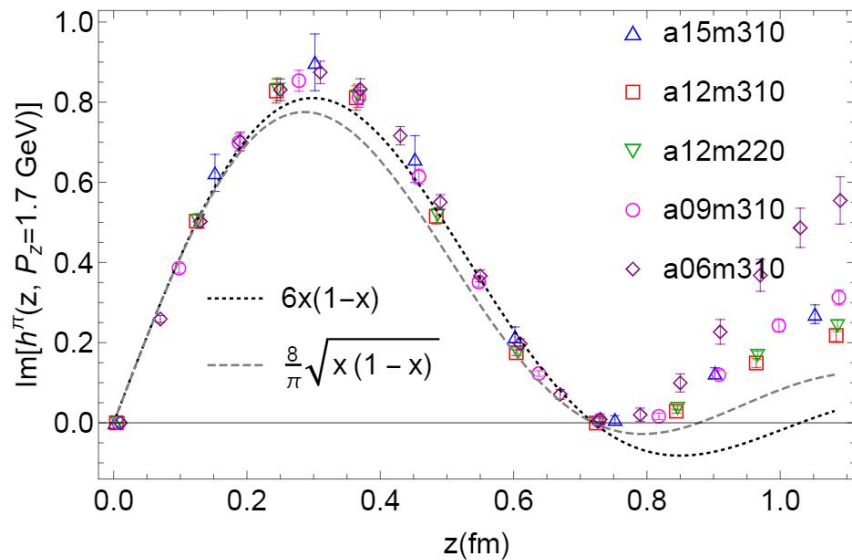
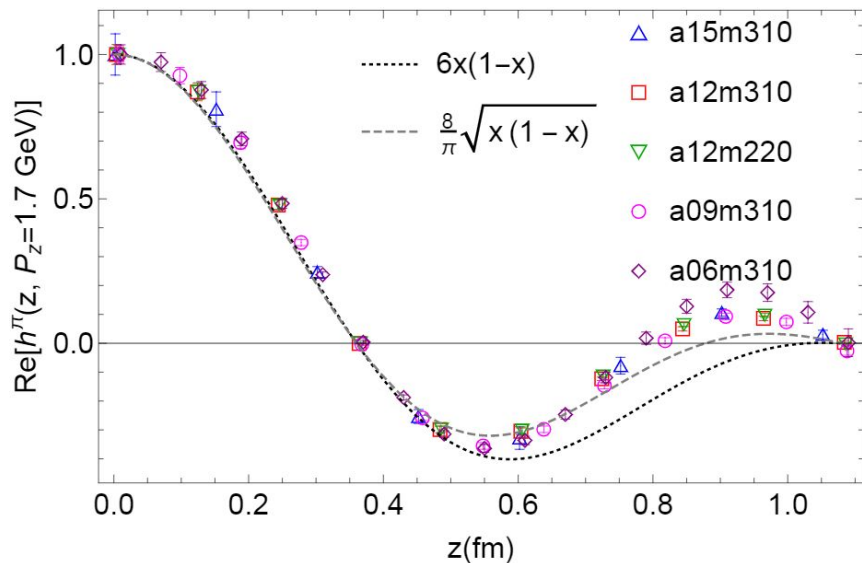
# $P_z$ Dependence of Renormalized Matrix Elements

- RI/MOM [see R. Zhang, C. Honkala, H.-W. Lin, and J.-W. Chen (2020)]
  - $\mu^R = 3.8 \text{ GeV}$ ,  $p_z^R = 0$
- For example: a09m310



# $M_\pi$ and a Dependence of Renormalized Matrix Elements

- RI/MOM
  - $\mu^R = 3.8 \text{ GeV}$ ,  $p_z^R = 0$
- $P_z \approx 1.7 \text{ GeV}$

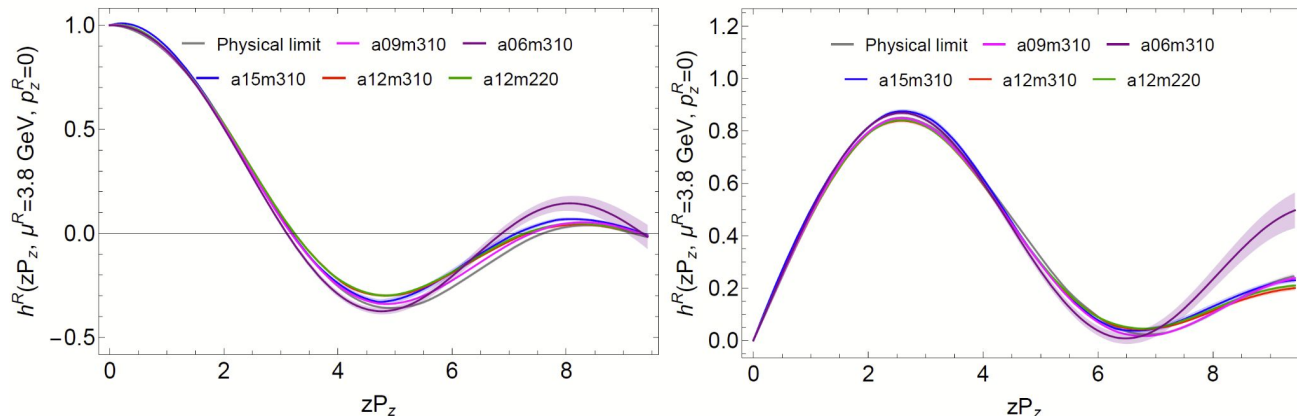




# Continuum Matrix Elements and Extracting Pion DA

$$h = h_0(1 + c_2 a^2 + d_2 M_\pi^2)$$

$$P_z = 1.7 \text{ GeV}$$



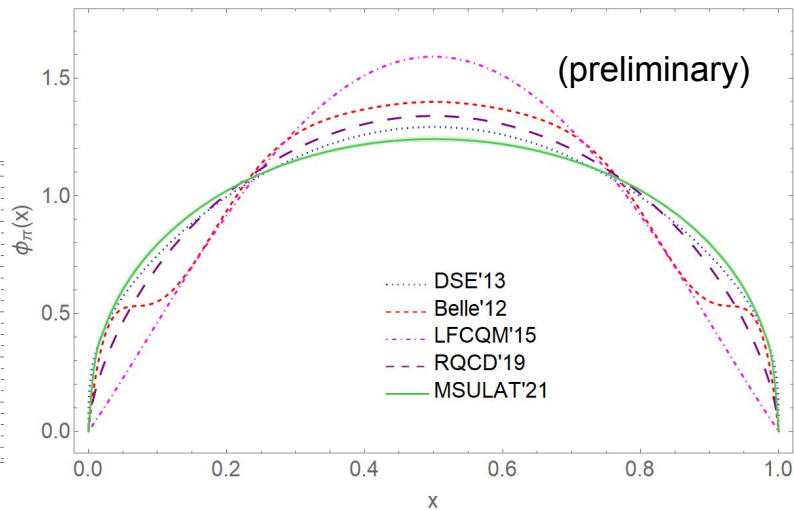
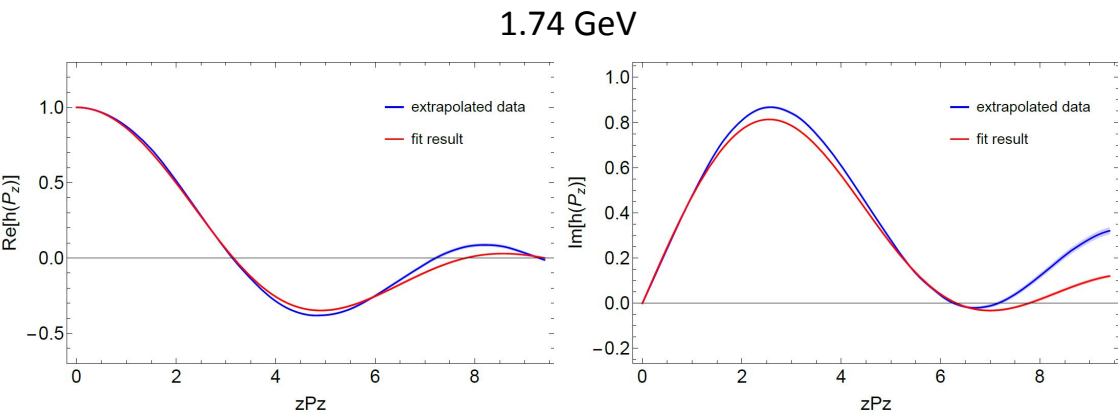
$$h(z, \mu^R, p_z^R, P_z) = \int_{-\infty}^{\infty} dx \int_0^1 dy C \left( x, y, \left( \frac{\mu^R}{p_z^R} \right)^2, \frac{P_z}{\mu^R}, \frac{P_z}{p_z^R} \right) f_{m,n}(y) e^{i(1-x)zP_z}$$

- C is the matching kernel (calculated perturbatively) [Y.-S. Liu et al. (2019)]
- We seek the lightcone DA,  $f(y)$

# Fitting to Analytical Form

- Fit  $h(z)$  to continuum-physical matrix elements with the functional form:

$$f_{m,n} = x^m (1-x)^n / B(m+1, n+1)$$

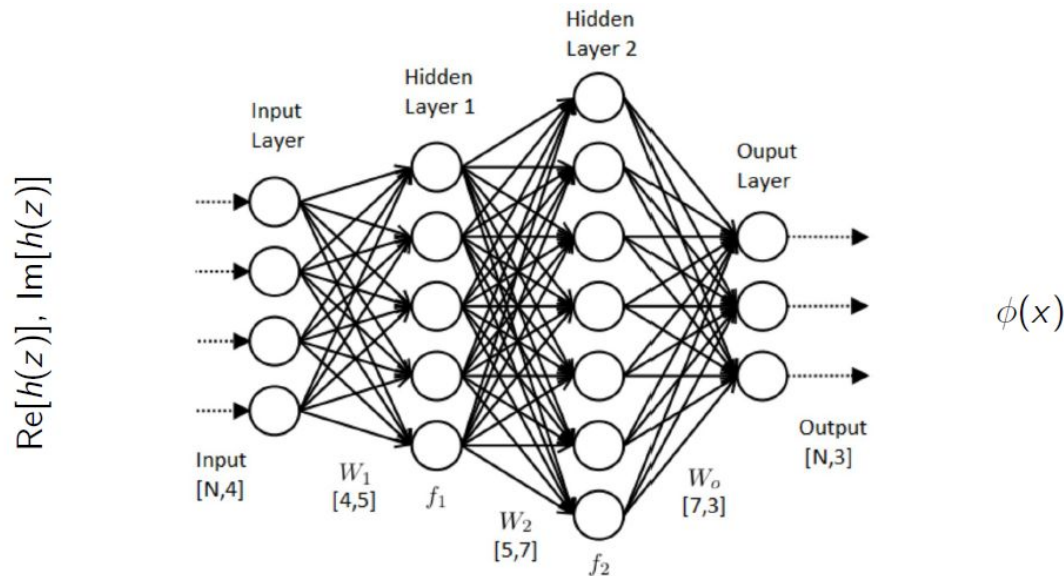


# Machine Learning Approach?

Alternatively, we use ML to predict meson DA

- MLP regressor (scikit-learn)
- 3 hidden layers, 20 nodes each
- Default loss function (MSE)
- activation function  
 $g(x) = \max(x, 0)$

[cf. *R. Zhang et al. (2020)*]

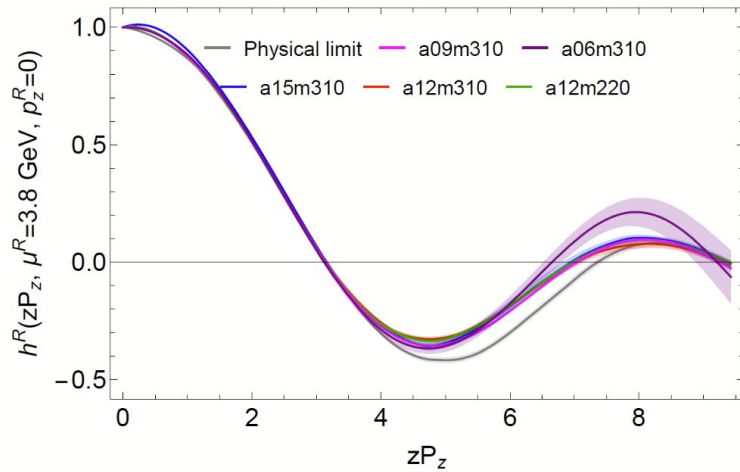
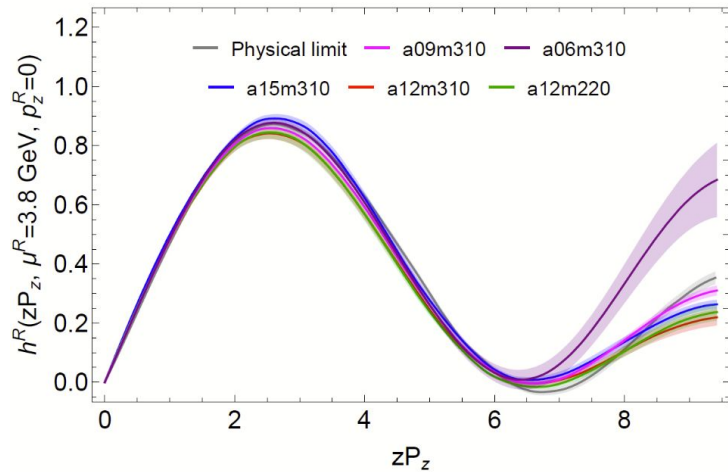


Credit: VIASAT

# Additional Slides

# Continuum Matrix Elements

$$h = h_0(1 + c_1 a^1 + d_2 M_\pi^2)$$



# Machine Learning Approach

Testing model predictions on known sine function

