

The pseudoscalar poles contributions to $(g - 2)_\mu$
based on P. Masjuan, PS: Phys.Rev. D95 (2017)

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Outline

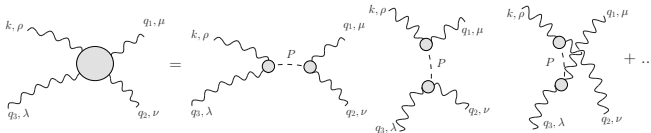
1. Pseudoscalar poles contribution to a_μ^{HLbL} : the essentials
2. Describing the pseudoscalar form factors: Padé approximants
3. Application to HLbL: results
4. Summary

Section 1

Pseudoscalar poles contribution to a_μ^{HLbL} : the essentials

— The pseudoscalar poles in brief —

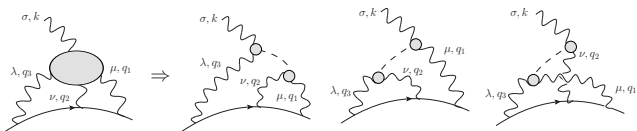
- We want the pseudoscalar (π^0, η, η') pole contributions to a_μ^{HLbL}



unambiguously identified (see G. Colangelo's talk)

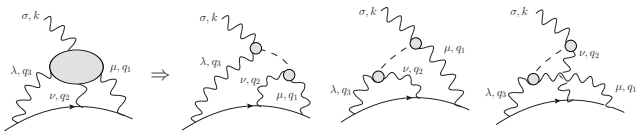
— The pseudoscalar poles in brief —

- Plugging it into the a_μ^{HLbL} contribution



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- Plugging it into the a_μ^{HLbL} contribution



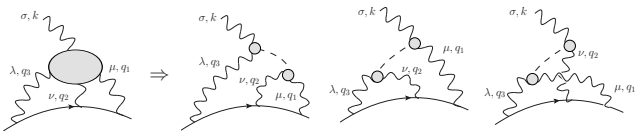
- Result as weighted integral over spacelike form factors [Phys.Rept., 477 (2009)]

$$a_\ell^{\text{HLbL};P} = \frac{-2\pi}{3} \left(\frac{\alpha}{\pi}\right)^3 \int_0^\infty dQ_1 dQ_2 \int_{-1}^{+1} dt \sqrt{1-t^2} Q_1^3 Q_2^3$$

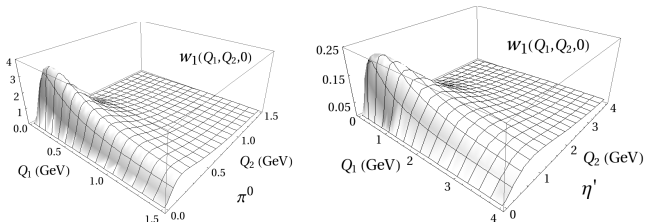
$$\times \left[\frac{F_{P\gamma^*\gamma^*}(Q_1^2, Q_3^2) F_{P\gamma^*\gamma}(Q_2^2, 0) h_1(Q_1, Q_2, t)}{Q_2^2 + m_p^2} + \frac{F_{P\gamma^*\gamma^*}(Q_1^2, Q_2^2) F_{P\gamma^*\gamma}(Q_3^2, 0) h_2(Q_1, Q_2, t)}{Q_3^2 + m_p^2} \right]$$

— The pseudoscalar poles in brief —

- Plugging it into the a_μ^{HLbL} contribution



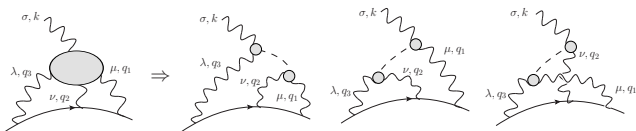
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notice the peaks at the relevant low energies

— The pseudoscalar poles in brief —

- Plugging it into the a_μ^{HLbL} contribution



- Solved if form factor known precisely at Q^2 and HE extrapolation appropriate

— Whitepaper criteria [Phys.Rept. 887 (2020) 1-166] —

- In addition to the normalization [...], also HE constraints must be fulfilled
- At least the SL data [...] must be reproduced
- Systematic uncertainties must be assessed [...]

⇒ Padé approximants, Dispersion relations, Lattice QCD

Here I will focus on Padé approximants

Section 2

Describing the pseudoscalar form factors: Padé approximants

— Padé approximants: the philosophy —

—The challenge—

Get a description for the SL TFF

—The philosophy—

Nature has solved QCD: make use of data

—The problem—

Inter(extra)polate the data in the SL region while (WP criteria):

- Incorporating QCD constraints
- Assessing systematics from procedure

—Our proposal—

Padé & Canterbury approximants: not a model but a mathematical toolkit

— Padé approximants: singly virtual _____

- How to approximate (not model) non-perturbative hadronic functions?

Taylor series: $F_{\pi\gamma\gamma^*}(q^2) = F_{\pi\gamma\gamma}(1 + b_P q^2 + \dots)$ **X** poles(cuts)

— Padé approximants: singly virtual _____

- How to approximate (not model) non-perturbative hadronic functions?

Laurent exp.: $F_{\pi\gamma\gamma^*}(q^2) = \sum_{n=-1} c_n (q^2 - M^2)^n$ \times next pole(cuts)

— Padé approximants: singly virtual —————

- How to approximate (not model) non-perturbative hadronic functions?

$$\text{PAs: } F_{\pi\gamma\gamma^*}(q^2) = P_M^N = \frac{Q_N(q^2)}{R_M(q^2)} = F_{\pi\gamma\gamma}(1 + b_P q^2 + \dots + \mathcal{O}(q^{N+M+1}))$$

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- Convergence to meromorphic and Stieltjes in the SL \Rightarrow beyond large- N_c
Convergence applies to sequences: $P_1^N, P_N^N, P_{N+1}^N, \dots$

Sequence convergence: model-independence and systematics

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Sequence convergence: model-independence and systematics

- Mixed $q^2 = 0$ and $q^2 = \infty$ expansions if required \rightarrow pQCD constraints

$$P_1^0(q^2) = \frac{F_{P\gamma\gamma^*}(0)}{1 - b_P Q^2} = F_{P\gamma\gamma^*}(0)(1 + b_P Q^2 + \mathcal{O}(q^4)), \quad b_P \neq M_V^{-2}!!!$$

Can incorporate relevant QCD constraints; $q^2 = 0$ relevant for a_μ^{HLbL}

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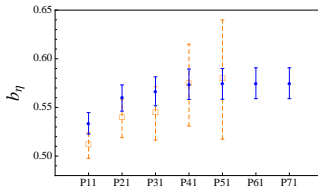
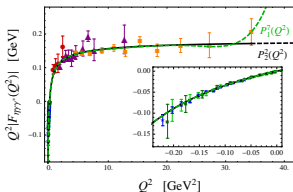
$$P_1^0(q^2) = \frac{F_{P\gamma\gamma^*}(0)}{1 - b_P Q^2} = F_{P\gamma\gamma^*}(0)(1 + b_P Q^2 + \mathcal{O}(Q^4)), \quad b_P \neq M_V^{-2}!!!$$

Can incorporate relevant QCD constraints; $q^2 = 0$ relevant for a_μ^{HLbL}

- Obtain the derivatives from data (nor spectroscopy nor direct fitting!!!)

Inputs from data, PAs from derivatives

- Inputs: sequences data fitting (not just fitting models)



Inputs from data, PAs from derivatives

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$$\begin{array}{l|l}
 \text{Reduced} & P_{N+1}^N : \quad P_1^{0;\text{fit}} \rightarrow b_P, \dots, P_2^{1;\text{fit}} \quad \times \\
 \text{Data Set} & P_1^N : \quad P_1^{0;\text{fit}} \rightarrow b_P, \dots, P_1^{4;\text{fit}} \rightarrow \frac{b_P}{c_P}, P_1^{5;\text{fit}} \quad \times
 \end{array}$$

- Let's see impact on $a_\mu^{\text{HLbL};\eta}$ assuming $F_{P\gamma^*\gamma^*}(q_1^2, q_2^2) = F_{P\gamma\gamma^*}(q_1^2)F_{P\gamma\gamma^*}(q_2^2)$

	Fit
P_1^0	14.5
P_2^1	—

Inputs from data, PAs from derivatives

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$$\begin{array}{l|l}
 \text{Reduced} & P_{N+1}^N : \quad P_1^{0;\text{der}} \rightarrow b_P, \dots, P_2^{1;\text{der}} \quad \checkmark \\
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	Fit	Der
P_1^0	14.5	13.2
P_2^1	—	13.3

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$$\begin{array}{l|l}
 \text{New larger} & P_{N+1}^N : \quad P_1^{0;\text{fit}} \rightarrow b_P, \dots, P_2^{1;\text{fit}}, P_3^{2;\text{fit}} \quad \times \\
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	Fit	Der	New Fit
P_1^0	14.5	13.2	14.0
P_2^1	—	13.3	13.4

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	Fit	Der	New Fit	New Der
P_1^0	14.5	13.2	14.0	13.1
P_2^1	—	13.3	13.4	13.3

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	Fit	Der	New Fit	New Der
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- Advantage of PAs: interrelation, acceleration of convergence, systematics
- Cannot overemphasize: not directly fitting functions, nor spectroscopy!!!

___ Padé approximants: doubly virtual _____

- Commonly referred to as Canterbury approximants

$$F_{\pi\gamma^*\gamma^*}(q_1^2, q_2^2) = C_M^N = \frac{Q_N(q_1^2, q_2^2)}{R_M(q_1^2, q_2^2)}; \quad Q_N(R_M) = \sum_{i,j}^{N(M)} c_{n,m}^{Q(R)} q_1^{2i} q_2^{2j}$$

Again, match derivatives to get $c_{ij}^{Q,R}$'s

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$$F_{P\gamma^*\gamma^*}^{\log}(Q_1^2, Q_2^2) = \frac{F_{P\gamma\gamma} M^2}{Q_1^2 - Q_2^2} \ln \left(\frac{1 + Q_1^2/M^2}{1 + Q_2^2/M^2} \right)$$

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- Simple example with appropriate power-like behavior

$$C_1^0 = \frac{1}{1 + c_{0,1}^R(q_1^2 + q_2^2) + c_{1,1}^R q_1^2 q_2^2} \rightarrow \frac{1}{1 + c_{0,1}^R(q_1^2 + q_2^2)}$$

Again, benefits of not doing spectroscopy!

- Unfortunately not available data yet (except for η' ; expected for π^0)

See to believe: toy models and systematics

— a_μ^π : Regge Model—

— a_μ^π : Logarithmic Model—

$$F_{\pi^0\gamma^*\gamma^*}^{\text{Regge}}(Q_1^2, Q_2^2) = \frac{aF_{P\gamma\gamma}}{Q_1^2 - Q_2^2} \frac{\left[\psi^{(0)}\left(\frac{M^2 + Q_1^2}{a}\right) - \psi^{(0)}\left(\frac{M^2 + Q_2^2}{a}\right) \right]}{\psi^{(1)}\left(\frac{M^2}{a}\right)}$$

$$F_{\pi^0\gamma^*\gamma^*}^{\text{log}}(Q_1^2, Q_2^2) = \frac{F_{P\gamma\gamma} M^2}{Q_1^2 - Q_2^2} \ln\left(\frac{1 + Q_1^2/M^2}{1 + Q_2^2/M^2}\right)$$

	C_1^0	C_2^1	C_3^2	C_4^3
LE	55.2	59.7	60.4	60.6
OPE	65.7	60.8	60.7	60.7
Fit ^{OPE}	66.3	62.7	61.1	60.8
Exact	60.7			

	C_1^0	C_2^1	C_3^2	C_4^3
LE	56.7	64.4	66.1	66.8
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Fit ^{OPE}	79.6	71.9	69.3	68.4
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- The convergence result is excellent!
- The OPE choice seems the best → high energy matters
- Still, low energies provide a good performance
- Error \sim difference among elements → Systematics!



Section 3

Application to HLbL and results

— Reconstructing $F_{P\gamma^*\gamma^*}(Q_1^2, Q_2^2)$ —

- Simplest approach: $C_1^0(Q_1^2, Q_2^2)$

$$C_1^0(Q_1^2, Q_2^2) = \frac{F_{P\gamma\gamma}(0, 0)}{1 + b_P(Q_1^2 + Q_2^2)}$$

Reconstructing $F_{P\gamma^*\gamma^*}(Q_1^2, Q_2^2)$

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- Next element: $C_2^1(Q_1^2, Q_2^2)$

$$C_2^1(Q_1^2, Q_2^2) = \frac{F_{P\gamma\gamma}(0, 0)(1 + \alpha_1(Q_1^2 + Q_2^2) + \alpha_{1,1}Q_1^2Q_2^2)}{1 + \beta_1(Q_1^2 + Q_2^2) + \beta_2(Q_1^4 + Q_2^4) + \beta_{1,1}Q_1^2Q_2^2 + \beta_{2,1}Q_1^2Q_2^2(Q_1^2 + Q_2^2) + \beta_{2,2}Q_1^4Q_2^4}$$

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Reconstruction

1. Reduce to Padé Approximants

$F_{P\gamma\gamma}(0, 0), \alpha_1, \beta_1, \beta_2 \rightarrow$ from PAs

Reconstructing $F_{P\gamma^*\gamma^*}(Q_1^2, Q_2^2)$

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Reconstruction

- Reduce to Padé Approximants
- Reproduce the OPE behavior (high energies)

$$F_{\pi\gamma^*\gamma^*} = \frac{1}{3Q^2}(2F_\pi) \left(1 - \frac{8}{9} \frac{\delta^2}{Q^2} + \mathcal{O}(\alpha_s(Q^2)) \right) \Rightarrow \beta_{2,2} = 0, \alpha_{1,1}, \beta_{2,1}$$

Reconstructing $F_{P\gamma^*\gamma^*}(Q_1^2, Q_2^2)$

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Reconstruction

1. Reduce to Padé Approximants
2. Reproduce the OPE behavior (high energies)
3. Reproduce the low energies ($a_{P;1,1}$)

Reconstructing $F_{P\gamma^*\gamma^*}(Q_1^2, Q_2^2)$

- Simplest approach: $C_1^0(Q_1^2, Q_2^2)$

$$C_1^0(Q_1^2, Q_2^2) = \frac{F_{P\gamma\gamma}(0, 0)}{1 + b_P(Q_1^2 + Q_2^2)}$$

- Next element: $C_2^1(Q_1^2, Q_2^2)$

$$C_2^1(Q_1^2, Q_2^2) = \frac{F_{P\gamma\gamma}(0, 0)(1 + \alpha_1(Q_1^2 + Q_2^2) + \alpha_{1,1}Q_1^2Q_2^2)}{1 + \beta_1(Q_1^2 + Q_2^2) + \beta_2(Q_1^4 + Q_2^4) + \beta_{1,1}Q_1^2Q_2^2 + \beta_{2,1}Q_1^2Q_2^2(Q_1^2 + Q_2^2)}$$

Reconstruction

1. Reduce to Padé Approximants
2. Reproduce the OPE behavior (high energies)
3. Reproduce the low energies ($a_{P;1,1}$)

Be generous: all configurations with no poles $\Rightarrow a_{P;1,1}^{\min} < a_{P;1,1} < a_{P;1,1}^{\max}$

Reconstructing $F_{P\gamma^*\gamma^*}(Q_1^2, Q_2^2)$

- Simplest approach: $C_1^0(Q_1^2, Q_2^2)$

$$C_1^0(Q_1^2, Q_2^2) = \frac{F_{P\gamma\gamma}(0, 0)}{1 + b_P(Q_1^2 + Q_2^2)}$$

- Next element: $C_2^1(Q_1^2, Q_2^2)$

$$C_2^1(Q_1^2, Q_2^2) = \frac{F_{P\gamma\gamma}(0, 0)(1 + \alpha_1(Q_1^2 + Q_2^2) + \alpha_{1,1}Q_1^2Q_2^2)}{1 + \beta_1(Q_1^2 + Q_2^2) + \beta_2(Q_1^4 + Q_2^4) + \beta_{1,1}Q_1^2Q_2^2 + \beta_{2,1}Q_1^2Q_2^2(Q_1^2 + Q_2^2)}$$

Reconstruction

1. Reduce to Padé Approximants
2. Reproduce the OPE behavior (high energies)
3. Reproduce the low energies ($a_{P;1,1}^{\min} < a_{P;1,1} < a_{P;1,1}^{\max}$)

— Pseudoscalar-pole contribution: Final results —

$$-C_1^0(Q_1^2, Q_2^2)-$$

$a_\mu^{\text{HLbL},P} \times 10^{11}$	OPE ($a_{P,1,1} = 2b_P^2$)	Fact ($a_{P,1,1} = b_P^2$)
π^0	65.3(1.4) _F (2.4) _{b_π} [2.8] _t	54.3(1.5) _F (2.2) _{b_π} [2.5] _t
η	17.1(0.6) _F (0.2) _{b_η} [0.6] _t	13.0(0.4) _F (0.2) _{b_η} [0.5] _t
η'	16.0(0.5) _F (0.3) _{b_{η'}} [0.6] _t	12.0(0.4) _F (0.3) _{b_{η'}} [0.5] _t
Total	98.4[2.9] _t	79.3[2.6] _t

— Pseudoscalar-pole contribution: Final results —

$$—C_1^0(Q_1^2, Q_2^2)—$$

$a_\mu^{\text{HLbL};P} \times 10^{11}$	OPE ($a_{P,1,1} = 2b_P^2$)	Fact ($a_{P,1,1} = b_P^2$)
π^0	65.3(1.4) _F (2.4) _{b_π} [2.8] _t	54.3(1.5) _F (2.2) _{b_π} [2.5] _t
η	17.1(0.6) _F (0.2) _{b_η} [0.6] _t	13.0(0.4) _F (0.2) _{b_η} [0.5] _t
η'	16.0(0.5) _F (0.3) _{b_{η'}} [0.6] _t	12.0(0.4) _F (0.3) _{b_{η'}} [0.5] _t
Total	98.4[2.9] _t	79.3[2.6] _t

$$—C_2^1(Q_1^2, Q_2^2)—$$

$a_\mu^{\text{HLbL};P} \times 10^{11}$	$a_{P,1,1}^{\text{min}}$	$a_{P,1,1}^{\text{max}}$
π^0	64.1(1.3) _L (0) _δ [1.3] _t	63.0(1.1) _L (0.5) _δ [1.2] _t
η	16.3(0.8) _L (0) _δ [0.8] _t	16.2(0.8) _L (0.6) _δ [1.0] _t
η'	14.7(0.7) _L (0) _δ [0.7] _t	14.3(0.5) _L (0.5) _δ [0.7] _t
Total	95.1[1.7] _t	93.5[1.7] _t

— Pseudoscalar-pole contribution: Final results —

$$—C_1^0(Q_1^2, Q_2^2)—$$

$a_\mu^{\text{HLbL};P} \times 10^{11}$	OPE ($a_{P,1,1} = 2b_P^2$)	Fact ($a_{P,1,1} = b_P^2$)
π^0	65.3(1.4) _F (2.4) _{b_π} [2.8] _t	54.3(1.5) _F (2.2) _{b_π} [2.5] _t
η	17.1(0.6) _F (0.2) _{b_η} [0.6] _t	13.0(0.4) _F (0.2) _{b_η} [0.5] _t
η'	16.0(0.5) _F (0.3) _{b_{η'}} [0.6] _t	12.0(0.4) _F (0.3) _{b_{η'}} [0.5] _t
Total	98.4[2.9] _t	79.3[2.6] _t

$$—C_2^1(Q_1^2, Q_2^2)—$$

$a_\mu^{\text{HLbL};P} \times 10^{11}$	$a_{P,1,1}^{\text{min}}$	$a_{P,1,1}^{\text{max}}$
π^0	64.1(1.3) _t (0) _δ [1.3] _t {1.2} _{sys}	63.0(1.1) _t (0.5) _δ [1.2] _t {2.3} _{sys}
η	16.3(0.8) _t (0) _δ [0.8] _t	16.2(0.8) _t (0.6) _δ [1.0] _t
η'	14.7(0.7) _t (0) _δ [0.7] _t	14.3(0.5) _t (0.5) _δ [0.7] _t
Total	95.1[1.7] _t	93.5[1.7] _t

— Pseudoscalar-pole contribution: Final results —

$$\text{—} C_1^0(Q_1^2, Q_2^2) \text{—}$$

$a_\mu^{\text{HLbL};P} \times 10^{11}$	OPE ($a_{P,1,1} = 2b_P^2$)	Fact ($a_{P,1,1} = b_P^2$)
π^0	65.3(1.4) _F (2.4) _{b_π} [2.8] _t	54.3(1.5) _F (2.2) _{b_π} [2.5] _t
η	17.1(0.6) _F (0.2) _{b_η} [0.6] _t	13.0(0.4) _F (0.2) _{b_η} [0.5] _t
η'	16.0(0.5) _F (0.3) _{b_{η'}} [0.6] _t	12.0(0.4) _F (0.3) _{b_{η'}} [0.5] _t
Total	98.4[2.9] _t	79.3[2.6] _t

$$\text{—} C_2^1(Q_1^2, Q_2^2) \text{—}$$

$a_\mu^{\text{HLbL};P} \times 10^{11}$	$a_{P,1,1}^{\text{min}}$	$a_{P,1,1}^{\text{max}}$
π^0	64.1(1.3) _L (0) _δ [1.3] _t {1.2} _{sys}	63.0(1.1) _L (0.5) _δ [1.2] _t {2.3} _{sys}
η	16.3(0.8) _L (0) _δ [0.8] _t {0.8} _{sys}	16.2(0.8) _L (0.6) _δ [1.0] _t {0.9} _{sys}
η'	14.7(0.7) _L (0) _δ [0.7] _t {1.3} _{sys}	14.3(0.5) _L (0.5) _δ [0.7] _t {1.7} _{sys}
Total	95.1[1.7] _t {3.3} _{sys}	93.5[1.7] _t {4.9} _{sys}

— Pseudoscalar-pole contribution: Final results —

— $C_1^0(Q_1^2, Q_2^2)$ —

$a_\mu^{\text{HLbL};P} \times 10^{11}$	OPE ($a_{P,1,1} = 2b_P^2$)	Fact ($a_{P,1,1} = b_P^2$)
π^0	65.3(1.4) _F (2.4) _{b_π} [2.8] _t	54.3(1.5) _F (2.2) _{b_π} [2.5] _t
η	17.1(0.6) _F (0.2) _{b_η} [0.6] _t	13.0(0.4) _F (0.2) _{b_η} [0.5] _t
η'	16.0(0.5) _F (0.3) _{b_{η'}} [0.6] _t	12.0(0.4) _F (0.3) _{b_{η'}} [0.5] _t
Total	98.4[2.9] _t	79.3[2.6] _t

— $C_2^1(Q_1^2, Q_2^2)$ —

$a_\mu^{\text{HLbL};P} \times 10^{11}$	$a_{P,1,1}^{\text{min}}$	$a_{P,1,1}^{\text{max}}$
π^0	64.1(1.3) _L (0) _δ [1.3] _t {1.2} _{sys}	63.0(1.1) _L (0.5) _δ [1.2] _t {2.3} _{sys}
η	16.3(0.8) _L (0) _δ [0.8] _t {0.8} _{sys}	16.2(0.8) _L (0.6) _δ [1.0] _t {0.9} _{sys}
η'	14.7(0.7) _L (0) _δ [0.7] _t {1.3} _{sys}	14.3(0.5) _L (0.5) _δ [0.7] _t {1.7} _{sys}
Total	95.1[1.7] _t {3.3} _{sys}	93.5[1.7] _t {4.9} _{sys}

—Final Result (combining errors just for clarity)

$$a_\mu^{\pi, \eta, \eta'} = (63.6(2.7) + 16.3(1.3) + 14.5(1.8)) \times 10^{-11} = 94.3(5.3) \times 10^{-11}$$

— Pseudoscalar-pole contribution: Final results —

— $C_1^0(Q_1^2, Q_2^2)$ —

$a_\mu^{\text{HLbL};P} \times 10^{11}$	OPE ($a_{P,1,1} = 2b_P^2$)	Fact ($a_{P,1,1} = b_P^2$)
π^0	65.3(1.4) _F (2.4) _{b_π} [2.8] _t	54.3(1.5) _F (2.2) _{b_π} [2.5] _t
η	17.1(0.6) _F (0.2) _{b_η} [0.6] _t	13.0(0.4) _F (0.2) _{b_η} [0.5] _t
η'	16.0(0.5) _F (0.3) _{b_{η'}} [0.6] _t	12.0(0.4) _F (0.3) _{b_{η'}} [0.5] _t
Total	98.4[2.9] _t	79.3[2.6] _t

— $C_2^1(Q_1^2, Q_2^2)$ —

$a_\mu^{\text{HLbL};P} \times 10^{11}$	$a_{P,1,1}^{\text{min}}$	$a_{P,1,1}^{\text{max}}$
π^0	64.1(1.3) _L (0) _δ [1.3] _t {1.2} _{sys}	63.0(1.1) _L (0.5) _δ [1.2] _t {2.3} _{sys}
η	16.3(0.8) _L (0) _δ [0.8] _t {0.8} _{sys}	16.2(0.8) _L (0.6) _δ [1.0] _t {0.9} _{sys}
η'	14.7(0.7) _L (0) _δ [0.7] _t {1.3} _{sys}	14.3(0.5) _L (0.5) _δ [0.7] _t {1.7} _{sys}
Total	95.1[1.7] _t {3.3} _{sys}	93.5[1.7] _t {4.9} _{sys}

—Final Result (combining errors just for clarity)

$$a_\mu^{\pi, \eta, \eta'} = (63.6(2.7) + 16.3(1.3) + 14.5(1.8)) \times 10^{-11} = 94.3(5.3) \times 10^{-11}$$

- Good agreement with $a_\mu^{\text{HLbL};\pi} = 63.0(2.7)/62.3(2.3)$ (DR/lattice)
- Current number in WP for η, η'

Summary

- Padé approximants to reconstruct form factors
- Full use of data and theory in a systematic approach; not pheno model
- OPE for all the pseudoscalars implemented
- Bypass $\eta - \eta'$ mixing (output): non-trivial if fully theory-driven approach
- Good agreement for π with DR/lattice (th. on safe grounds)
- State of the art for η, η'

Outlook

- π re-evaluation with forthcoming BES III and PrimEx data
- Besides pseudoscalars, interesting relations with axials and $\langle VVA \rangle$ constraints [Masjuan, Roig, PSP, 2005.11761]

Section 4

Backup

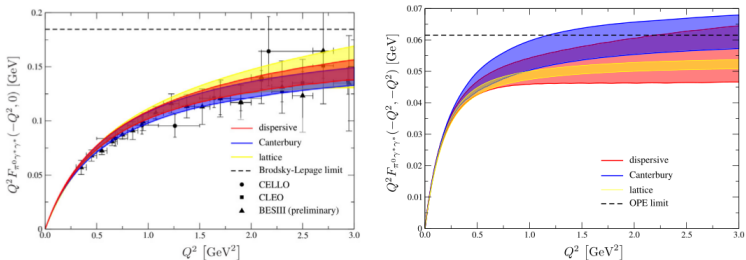


Fig. 60. Comparison of the π^0 TFF from dispersion theory [21,497] (red), CA [19] (blue), and lattice QCD [22] (yellow). We show both the singly- (left) and the doubly-virtual (right) form factors.

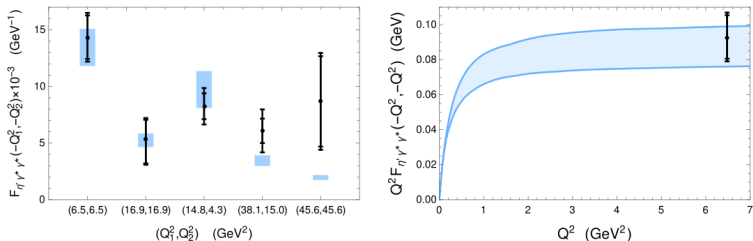
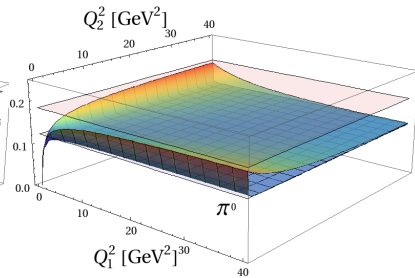
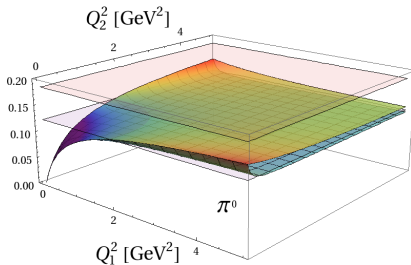
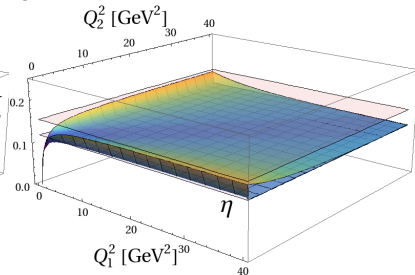
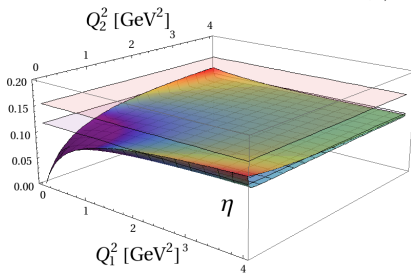


Fig. 59. Left: BABAR data points [108] with statistical errors (inner bars) and statistical and systematic combined (outer bars) in black, together with the CA prediction including errors (blue bands). Right: The analogous plot for the diagonal $Q^2 F_{\eta/\gamma^* \gamma^*}(-Q^2, -Q^2)$ TFF.

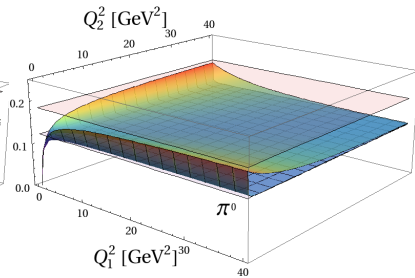
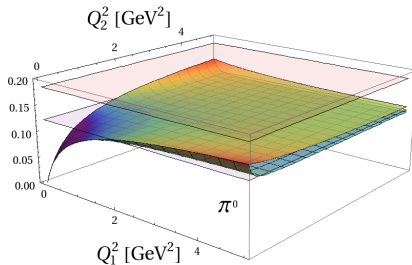
— $F_{P\gamma^*\gamma^*}(Q_1^2, Q_2^2)$ —



The two planes: boundaries for the $a_{P;1,1}$ region



$$F_{P\gamma^*\gamma^*}(Q_1^2, Q_2^2)$$



The two planes: boundaries for the $a_{P;1,1}$ region

