

Model-independent parametrization of the hadronic vacuum polarization and $g-2$ for the muon on the lattice

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The leading hadronic contribution to the muon anomalous magnetic moment is given by a weighted integral over euclidean momentum of the hadronic vacuum polarization. This integral is dominated by momenta of order the muon mass. Since the finite volume in lattice QCD makes it difficult to compute the vacuum polarization at a large number of low momenta with high statistics (combined with the fact that one cannot compute it at zero momentum), a parametrization of the vacuum polarization is required to extrapolate the data. A much used functional form is based on vector meson dominance, which introduces model dependence into the lattice computation of the magnetic moment. Here we introduce a model-independent extrapolation method, and present a few first tests of this new method.

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I. INTRODUCTION

Recently, there has been an increased interest in the lattice computation of the hadronic contributions to the anomalous magnetic moment a_μ of the muon [1–6]. The aim is to provide a first-principles computation of the hadronic contributions to a_μ from lattice QCD assuming the Standard Model, with full control of the error. Since the experimental value of a_μ is known with great accuracy [7], this would provide an interesting test of the Standard Model if the theoretical computation can be carried out with a comparable precision [8].

The dominant hadronic contribution comes from the hadronic vacuum polarization, and can be written as [1,9]¹

$$a_\mu^{\text{HLO}} = 4\alpha^2 \int_0^\infty dQ^2 f(Q^2) (\Pi(0) - \Pi(Q^2)), \quad (1.1a)$$

$$f(Q^2) = \frac{m_\mu^2 Q^2 Z^3(Q^2) (1 - Q^2 Z(Q^2))}{1 + m_\mu^2 Q^2 Z^2(Q^2)}, \quad (1.1b)$$

$$Z(Q^2) = \frac{\sqrt{Q^4 + 4m_\mu^2 Q^2} - Q^2}{2m_\mu^2 Q^2}, \quad (1.1c)$$

$$\Pi(Q^2) = \sum_{i=u,d,s} q_i^2 \Pi_i(Q^2), \quad (1.1d)$$

where $\Pi_i(Q^2)$, defined by

$$\begin{aligned} \Pi_{\mu\nu;i}(q) &= (-q^2 \delta_{\mu\nu} + q_\mu q_\nu) \hat{\Pi}_i(q^2), \\ \Pi(Q^2) &= \hat{\Pi}(q^2 = -Q^2), \end{aligned} \quad (1.2)$$

in which

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¹Our sign convention for $\Pi(Q^2)$ is opposite to that of Ref. [3].

$$\Pi_{\mu\nu;i}(q) = \int d^4x e^{iqx} \langle J_\mu^i(x) J_\nu^i(0) \rangle, \quad (1.3)$$

with $J_{\mu i} = \bar{\psi}_i \gamma_\mu \psi_i$, is the (flavor-diagonal) vacuum polarization for quark flavor i . In Eq. (1.1d), q_i is the electric charge of quark i in units of the electron charge, α is the fine-structure constant, and m_μ is the muon mass. Disconnected contributions are expected to be small [10] and have been neglected in Eq. (1.1d).² Of course, Eq. (1.1a) also applies to the electron and tau magnetic moments, if one replaces the muon mass by the electron or tau mass.

In principle, the integral in Eq. (1.1a) could be approximated by a sum, if lattice values of $\Pi(Q^2)$ at sufficiently many low values of Q^2 (of order m_μ^2 , the region that dominates the integral) could be computed with high precision. At present, this is not yet possible. Instead, all lattice computations of a_μ^{HLO} rely on fitting the lattice data to a functional form for $\Pi(Q^2)$, which is then used to compute the integral. The most commonly used approach to date has been to choose a functional form inspired by vector-meson dominance (VMD), with either only a contribution from the ρ pole (possibly dressed up with pion loop contributions [3]), or with contributions from the ρ and the ρ' [5], where the ρ mass is fixed to its value as independently determined on the same ensemble of gauge-field configurations.

However, this introduces a model-dependent element in what is supposed to be a first-principles computation in lattice QCD. This results in a systematic error afflicting lattice computations of a_μ^{HLO} which is difficult to quantify.³

²The methods developed in this article apply if both connected and disconnected parts are included. To the extent that the disconnected part is numerically negligible, we can apply our method to the connected part only as well. For estimates of disconnected contributions, see Ref. [4].

³As already discussed in Ref. [3], chiral perturbation theory is of little help in this case.

While VMD fits to $\Pi(Q^2)$ look very good, and lead to quite small statistical errors, a_μ^{HLO} is extremely sensitive to the behavior of the fitted $\Pi(Q^2)$ at very small Q^2 ,⁴ and thus to any systematics affecting the small- Q^2 behavior of $\Pi(Q^2)$. Therefore, it would be very nice if a functional form of $\Pi(Q^2)$ could be constructed that is solely based on known mathematical properties of the vacuum polarization, and which can be systematically improved if data with higher precision become available.

It turns out that such a method exists. It is based on the well-known observation that the vacuum polarization can be expressed in terms of a positive spectral function through a (once-subtracted) dispersion relation. This makes it possible to express the vacuum polarization $\Pi(Q^2)$ in a form for which a *convergent* sequence of Padé approximants (PAs) is known to exist. Moreover, the convergence is uniform for any compact region in the complex plane excluding the cut along the negative real axis. This includes, in particular, any finite interval in euclidean Q^2 between 0 and ∞ . Since the contribution from say the region $Q^2 \geq 3 \text{ GeV}^2$ to a_μ is much smaller than currently attainable errors, this is sufficient to employ this observation for the computation of a_μ . Our goal is to present an exploration of this observation, using examples of available data for $\Pi(Q^2)$.

This article is organized as follows. In Sec. II we review elements of the necessary mathematical theory, beginning with the observation that $\Pi(Q^2)$ can be written in terms of a Stieltjes function, for which a converging sequence of PAs is known to exist. In Sec. III we explain how we will apply this construction in order to carry out fits to numerical data for $\Pi(Q^2)$. Section IV reports on two examples of such fits. We discuss future prospects of this approach in our concluding section.

II. STIELTJES FUNCTIONS AND PADÉ APPROXIMANTS

In this section, we review the necessary elements of the theory of multi-point Padé approximants (PAs) for functions that can be written as a Stieltjes integral with a finite radius of convergence. A good review about PAs in general is Ref. [11]; for multi-point PAs we refer to Refs. [12,13].

A. Stieltjes functions

Consider the function

$$\Phi(z) = \int_0^{1/R} \frac{d\nu(\tau)}{1 + \tau z}, \quad (2.1)$$

with $\nu(\tau)$ some real, bounded, nondecreasing function on the interval $[0, 1/R]$, taking infinitely many values on that interval. The function $\Phi(z)$ then is a Stieltjes function, and

⁴At values of order m_μ^2 , which dominate the integral in Eq. (1.1a).

it is analytic everywhere in the complex plane except on the negative real axis for $z \leq -R$. The function $\Phi(z = Q^2)$ decreases monotonically as a function of Q^2 for $Q^2 \in (-R, \infty)$.

The vacuum polarization $\Pi(Q^2)$ can be expressed in terms of Φ through a once-subtracted dispersion relation

$$\begin{aligned} \Pi(Q^2) &= \Pi(0) - Q^2 \Phi(Q^2), \\ \Phi(Q^2) &= \int_{4m_\pi^2}^{\infty} dt \frac{\rho(t)}{t(t + Q^2)}, \end{aligned} \quad (2.2)$$

where $\rho(t)$ is the spectral function, which, of course, satisfies the constraint $\rho(t) \geq 0$ for $4m_\pi^2 \leq t < \infty$.⁵ This can be seen by changing variables $\tau = 1/t$ in the integral, taking $R = 4m_\pi^2$, and choosing

$$d\nu(\tau) = \rho(1/\tau)d\tau, \quad \rho(t) = \lim_{\epsilon \rightarrow 0} \frac{1}{\pi} \text{Im} \hat{\Pi}(t + i\epsilon), \quad (2.3)$$

where $\Pi(Q^2) = \hat{\Pi}(t = -Q^2)$.

Let us consider an ordered sequence of positive values Q_i^2 of the variable Q^2 , with $i \in \{1, \dots, P\}$ and $0 \leq Q_1^2 < Q_2^2 < \dots < Q_P^2$, and assume that the function $\Phi(Q^2)$ is known at these points. We may now construct a sequence of Stieltjes functions as follows. We begin by defining a function $\Psi_1(Q^2)$ by writing $\Phi(Q^2)$ as

$$\Phi(Q^2) = \frac{\Phi(Q_1^2)}{1 + (Q^2 - Q_1^2)\Psi_1(Q^2)}. \quad (2.4)$$

Then $\Psi_1(Q^2)$ is also a Stieltjes function [12]. Moreover, $\Psi_1(Q^2)$ is positive on the interval $[-R, \infty)$, and, on that interval, has an upper bound⁶

$$\Psi_1(Q^2) \leq \Psi_1(-R) \leq \frac{1}{R + Q_1^2}, \quad Q^2 \in [-R, \infty). \quad (2.5)$$

This follows from the requirement that $\Phi(Q^2)$ not have a singularity on the real axis for $Q^2 > -R$, which implies that

$$\lim_{Q_1^2 \downarrow -R} (Q_1^2 - Q^2)\Psi_1(Q^2) \leq 1. \quad (2.6)$$

Clearly, a sequence of Stieltjes functions $\Psi_i(Q^2)$, $i \in \{1, \dots, P\}$, can be constructed by iteration:

$$\Psi_{i-1}(Q^2) = \frac{\Psi_{i-1}(Q_i^2)}{1 + (Q^2 - Q_i^2)\Psi_i(Q^2)}, \quad i \in \{2, \dots, P\}, \quad (2.7)$$

which on the interval $[-R, \infty)$ satisfy

⁵Positivity of $\rho(t)$ follows, for instance, from the fact that it is proportional to the $e^+e^- \rightarrow$ hadrons cross section.

⁶For the lower bound, recall that Ψ_1 monotonically decreasing.

$$0 \leq \Psi_i(Q^2) \leq \Psi_i(-R) = \frac{1}{R + Q_i^2} \left(1 - \frac{\Psi_{i-1}(Q_i^2)}{\Psi_{i-1}(-R)} \right),$$

$$i \in \{2, \dots, P\}, \quad (2.8)$$

where $\Psi_{i-1}(-R) = \lim_{Q^2 \rightarrow -R} \Psi_{i-1}(Q^2)$.

Equation (2.4) defines $\Psi_1(Q_2^2)$ in terms of $\Phi(Q_2^2)$ and $\Phi(Q_1^2)$. Likewise, in general, $\Psi_{i-1}(Q_i^2)$ can be expressed in terms of the values $\Phi(Q_j^2)$, $j \in \{1, \dots, i\}$ by using Eq. (2.7) recursively.

Applying Eq. (2.7), the original function $\Phi(Q^2)$ can be written as a continued fraction

$$\Phi(Q^2) = \frac{\Phi(Q_1^2)}{1 + \frac{\Phi(Q_1^2)}{(Q^2 - Q_1^2)\Psi_1(Q_2^2)}} \cdot \frac{1}{1 + \frac{(Q^2 - Q_{P-1}^2)\Psi_{P-1}(Q_P^2)}{1 + (Q^2 - Q_P^2)\Psi_P(Q^2)}}. \quad (2.9)$$

As already observed above, $\Psi_1(Q_2^2), \dots, \Psi_{P-1}(Q_P^2)$ can be expressed in terms of the values of the function $\Phi(Q^2)$ at the points Q_2^2, \dots, Q_P^2 .

B. Multi-point Padé's

A rational (or Padé) approximation to the function $\Phi(Q^2)$ can be constructed by setting $\Psi_P(Q^2)$ in Eq. (2.9) equal to its lower bound (*i.e.*, zero), or its upper bound, given by Eq. (2.8). A rational approximation $R_M^N(Q^2)$ is the ratio of two polynomials of degrees N and M

$$R_M^N(Q^2) = \frac{\sum_{n=0}^N a_n Q^{2n}}{\sum_{n=0}^{M-1} b_n Q^{2n} + Q^{2M}}. \quad (2.10)$$

We will refer to $R_M^N(Q^2)$ as an $[N, M]$ PA.⁷

If we choose $\Psi_i(Q^2) = 0$, the expression in Eq. (2.7) yields a $[0, 0]$ PA for $\Psi_{i-1}(Q^2)$. Working back to the original function, this choice leads to a PA for the function $\Phi(Q^2)$. If the number of points Q_i^2 , $i \in \{1, \dots, P\}$ is even, $P = 2k$, starting with $\Psi_P(Q^2) = 0$ yields a $[k-1, k]$ PA. Indeed, for a $[k-1, k]$ PA we need to solve for k coefficients a_n and k coefficients b_n in Eq. (2.10), for a total of

$$R_{[(P+1)/2]}^{[P/2]}(Q^2) = \frac{(R + Q_P^2)B_{P-1}(-R)A_P(Q^2) + (Q^2 - Q_P^2)B_P(-R)A_{P-1}(Q^2)}{(R + Q_P^2)B_{P-1}(-R)B_P(Q^2) + (Q^2 - Q_P^2)B_P(-R)B_{P-1}(Q^2)}. \quad (2.12)$$

As already mentioned, both these PAs are exact at the points Q_i^2 . Moreover, the complementary PAs have a pole at $Q^2 = -R$, as can be seen from Eq. (2.12). Between the points Q_i^2 and Q_{i+1}^2 , as well as between $-R$ and Q_1^2 and between Q_P^2 and ∞ , the standard and complementary PAs provide an upper and lower bound to the original function $\Phi(Q^2)$.⁸ Which is the lower bound and which the upper

$P = 2k$, determined by the values $\Phi(Q_i^2)$, $i \in \{1, \dots, P\}$. Likewise, for $P = 2k + 1$ odd, the procedure yields a $[k, k]$ PA. In short, from $\Psi_P(Q^2) = 0$ one obtains a $[(P-1)/2, [P/2]]$ PA, where $[x]$ denotes the integer part of x .

These “standard” multi-point PAs were studied in Refs. [12,13]. By construction, they are exact at the values $Q^2 = Q_i^2$, *i.e.*, the PA takes precisely the values $\Phi(Q_i^2)$ at these values of Q^2 . Moreover, these PAs converge to the function $\Phi(Q^2)$. More precisely, if we consider a sequence of standard multi-point PAs constructed from the values of $\Phi(Q^2)$ at a collection of points $-R < Q_1^2 < Q_2^2 < \dots < Q_P^2 < Q_*^2 < \infty$ with $\lim_{P \rightarrow \infty} Q_P^2 = Q_*^2$, the PAs converge uniformly on any closed and bounded domain in the complex Q^2 -plane excluding the cut $-\infty < Q^2 \leq -R$ [13], for $P \rightarrow \infty$.

If we choose $\Psi_P(Q^2)$ equal to the upper bound of Eq. (2.8), the expression in Eq. (2.7) yields a $[0, 1]$ PA for $\Psi_{P-1}(Q^2)$. Again working back to the original function, this choice also leads to a PA for the function $\Phi(Q^2)$. Now if the number of values Q_i^2 , $i \in \{1, \dots, P\}$ is even, $P = 2k$, this yields a $[k, k]$ PA. The counting argument is analogous to that above, but now this PA has, by construction, a pole at $Q^2 = -R$, which provides the extra information needed to find the $2k + 1$ coefficients in Eq. (2.10) for this case. Likewise, for $P = 2k + 1$ odd, the procedure yields a $[k, k + 1]$ PA. In short, in this case we obtain a $[P/2, [(P+1)/2]]$ PA. These “complementary” multi-point PAs are also exact at the values $Q^2 = Q_i^2$; they were introduced in Ref. [13].

If, given P values $\Phi(Q_i^2)$, $i \in \{1, \dots, P\}$, the standard PA is written as

$$R_{[P/2]}^{[(P-1)/2]}(Q^2) = \frac{A_P(Q^2)}{B_P(Q^2)}, \quad (2.11)$$

defining the polynomials $A_P(Q^2)$ and $B_P(Q^2)$, the complementary PA can be written as Ref. [13]

bound alternates as one progresses through these $P + 1$ intervals from $-R$ to ∞ [12,13].

C. Parametrization

All the poles of our standard PAs should have their poles on the negative real axis, at locations $Q^2 \leq -R$. Indeed, one can prove [11,13] that these PAs can be written in the form

$$R_{[P/2]}^{[(P-1)/2]}(Q^2) = a_0 + \sum_{n=1}^{[P/2]} \frac{a_n}{b_n + Q^2}, \quad (2.13)$$

⁷Redundancy between the coefficients a_n and b_n is removed by choosing one of them equal to 1. Here we choose $b_M = 1$.

⁸These bounds are optimal [12,13].

with $a_0 = 0$ for P even, and

$$\begin{aligned} a_n &> 0, & n \in \{1, \dots, \lfloor P/2 \rfloor\}, \\ b_{\lfloor P/2 \rfloor} &> b_{\lfloor P/2 \rfloor - 1} > \dots > b_1 \geq R. \end{aligned} \quad (2.14)$$

Once these parameters have been obtained for the $[\lfloor (P-2)/2 \rfloor, \lfloor (P-1)/2 \rfloor]$ and $[\lfloor (P-1)/2 \rfloor, \lfloor P/2 \rfloor]$ PAs, the complementary $[\lfloor P/2 \rfloor, \lfloor (P+1)/2 \rfloor]$ PA can be obtained from Eq. (2.12).

III. FIT STRATEGY

In the situation of an actual fit to values of $\Pi(Q^2)$ obtained from a numerical computation, these values are only known within some statistical errors. That implies that we do not know any points of the function exactly. Therefore, obviously, a multi-point sequence of PAs as described in Sec. II cannot be constructed. Instead, we will be fitting a fixed number of data points on a given interval, and the sequence of PAs we will find as a result of these fits is not a multi-point sequence in the sense of the theorem. However, the fact that $\Pi(Q^2)$, according to the theorem, can be described by a converging sequence implies that PAs of the form

$$\Pi(Q^2) = \Pi(0) - Q^2 \left(a_0 + \sum_{n=1}^N \frac{a_n}{b_n + Q^2} \right), \quad (3.1)$$

provide a valid functional form to which to fit the data. A number of PAs can be estimated by fitting this form to the data as a function of increasing N . For $a_0 = 0$ the parameters to be fitted are $\Pi(0)$ and the a_n and b_n for $n \in \{1, \dots, N\}$, and we obtain an $[N-1, N]$ PA. When also a_0 is fitted we obtain an $[N, N]$ PA.

In practice, errors on the data will limit how large we can take N in Eq. (3.1). For each value of N ,⁹ the value of χ^2 per degree of freedom will tell us how well Eq. (3.1) with given N fits the data. In particular, one would expect that with an increasing number of data points with decreasing errors, one would need to consider larger values of N to obtain a good fit.

With a given data set, adding more poles makes it harder to precisely determine the corresponding extra fit parameters as we increase N . However, as we will see, a_μ^{HLO} turns out to be quite stable as a function of N , because this quantity is rather insensitive to the higher poles.

As in Ref. [3], we will mostly explore fits in which we take the values of Q^2 for which we fit the PAs from the data for $\Pi(Q^2)$ in an interval between $Q^2 = 0$ and $Q^2 = 1 \text{ GeV}^2$. For each of our fits, we will compute the quantity

⁹Strictly speaking, there are two PAs for each value of N , one with $a_0 = 0$ and one with a_0 a fit parameter.

$$a_\mu^{\text{HLO}, Q^2 \leq 1} = 4\alpha^2 \int_0^{1 \text{ GeV}^2} dQ^2 f(Q^2) (\Pi(0) - \Pi(Q^2)), \quad (3.2)$$

with $f(Q^2)$ defined in Eq. (1.1b). This of course misses the part of the integral between 1 GeV^2 and ∞ , but this part is of order a percent of the low- Q^2 contribution. Since our goal here is to test the Padé approach to fitting $\Pi(Q^2)$ for $Q^2 \leq 1 \text{ GeV}^2$, we have restricted ourselves to the expression in Eq. (3.2) for comparisons between different fits.

We have not explored the complementary PAs defined in Sec. II B yet, but we anticipate that they may become useful in the future, when more precise data become available.

IV. TESTS

In this section, we explore fits to two different data sets. One set is the data for $\Pi(Q^2)$ with light quark mass equal to 0.0124 in lattice units (on a $24^3 \times 96$ lattice with lattice spacing $a \approx 0.09 \text{ fm}$, using “fine” configurations from the MILC collaboration [14]) that was also studied in Ref. [3] (see Table I of that paper). The other set is data obtained using the MILC “super-fine” gauge configurations with lattice spacing $a \approx 0.06 \text{ fm}$ on a $64^3 \times 144$ lattice with light quark mass equal to 0.0018 and a strange quark mass equal to 0.018, in lattice units. For both data sets the lattice strange quark mass is approximately equal to the physical strange quark mass. We will always assume that the theory of Sec. II applies to these data, i.e., that lattice artifacts are small enough to be ignored. In practice, rotational invariance is broken on the lattice. Since $\Pi(Q^2)$ is extracted from Eq. (1.2) using a lattice definition of the momentum components Q_μ [3], breaking of rotational invariance causes $\Pi(Q^2)$ to show small deviations from the monotonic decrease that follows from Eq. (2.2).

For each data set we carry out both correlated and uncorrelated fits, and compare those with each other. It turns out that this raises interesting questions about the behavior of the data and the fits at very low Q^2 .

A. $a = 0.09 \text{ fm}$ data at $m_{\text{light}}/m_{\text{strange}} = 0.4$

For our first example we consider the $am_{\text{light}} = 0.0124$ data that were also considered in Ref. [3]; this value of the light quark mass corresponds to about 2/5 times the physical strange quark mass. For these data, $m_\pi = 476 \text{ MeV}$ and $m_\rho = 962 \text{ MeV}$. In Tables I and II we show the result of a sequence of PA fits, with Table I showing correlated fits, and Table II showing uncorrelated fits. For the correlated fits we fitted data on the interval $0 < Q^2 \leq 0.6 \text{ GeV}^2$, because this interval yields the smallest values for the χ^2 per degree of freedom. For the uncorrelated case, we fitted the data for $\Pi(Q^2)$ on the interval $0 < Q^2 \leq 1 \text{ GeV}^2$, as was done in Ref. [3].

TABLE I. VMD and PA fits to the $a = 0.09$ fm, $am_{\text{light}} = 0.0124$ data for $\Pi(Q^2)$ of Ref. [3] with $Q^2 \leq 0.6$ GeV², except for the VMD fit, for which the fit interval is $Q^2 \leq 0.35$ GeV². Correlated fits; χ^2 errors.

	χ^2/dof	$10^{10} a_\mu^{\text{HLO}, Q^2 \leq 1}$	$\Pi(0)$	a_i	b_i	a_0
VMD	5.86/3	363(7)	0.0962(6)	0.0471(9)	0.9256 (fixed)	...
[0, 1]	11.4/8	338(6)	0.0960(5)	0.0600(7)	1.287(27)	...
[1, 1]	7.49/7	350(8)	0.0963(6)	0.049(4)	1.09(9)	0.0028(12)
[1, 2]	7.49/6	350(8)	0.0963(6)	0.049(4)	1.09(9)	...
				2(17)	$2(8) \times 10^3$	
[2, 2]	7.49/5	350(7)	0.0963(6)	0.049(4)	1.09(9)	0.0012(10)
				2.4(1.4)	$1.4(0.8) \times 10^3$	

TABLE II. VMD and PA fits to the $a = 0.09$ fm, $am_{\text{light}} = 0.0124$ data for $\Pi(Q^2)$ of Ref. [3] with $Q^2 \leq 1$ GeV². Uncorrelated fits; errors computed by a linear fluctuation analysis.

	χ^2/dof	$10^{10} a_\mu^{\text{HLO}, Q^2 \leq 1}$	$\Pi(0)$	a_i	b_i	a_0
VMD	4.37/18	413(8)	0.0980(7)	0.0536(10)	0.9256 (fixed)	...
[0, 1]	3.58/17	373(37)	0.0971(12)	0.0569(25)	1.10(16)	...
[1, 1]	3.36/16	424(116)	0.0979(22)	0.033(14)	0.6(4)	0.007(6)
[1, 2]	3.35/15	443(293)	0.098(4)	0.02(10)	0.4(1.7)	...
				0.058(12)	2(11)	
[2, 2]	3.35/14	445(432)	0.098(4)	0.02(29)	0.4(4.2)	0.0(4)
				0.1(3.8)	4(141)	

Table I shows that the value for $a_\mu^{\text{HLO}, Q^2 \leq 1}$ becomes very stable for PA fits starting at [1, 1]. For the [1, 1] PA $\chi^2/\text{dof} = 1.07$, indicating a good fit. For higher PAs, the value of χ^2 does not change, being very insensitive to the location of the second pole. It follows that the values of the parameters characterizing the second pole are not well determined, as can be seen in the table. However, the value of $a_\mu^{\text{HLO}, Q^2 \leq 1}$ is completely insensitive to the second pole. The explanation for this is that the integral for $a_\mu^{\text{HLO}, Q^2 \leq 1}$ is dominated by the Q^2 region around m_μ^2 , and thus very insensitive to the precise location of PA poles at large negative values of Q^2 .

The fit marked ‘‘VMD’’ is obtained by holding the parameter b_1 fixed at the square of the ρ mass in what would otherwise be a [0, 1] PA fit. It is thus *not* one of the sequence of PAs introduced in Sec. II. According to the theory there is no reason one should expect the parameter b_1 to be equal to the square of the ρ mass, as borne out by the values for b_1 found in the PA fits of Table I.¹⁰ For the correlated VMD fit a fitting interval $0 < Q^2 \leq 0.35$ GeV² leads to the lowest χ^2 per degree of freedom. With $\chi^2/\text{dof} \approx 2$, the VMD fit is not very good. It is already much better for the [0, 1] PA, in which the constraint on b_1 is relaxed, and it decreases further, to an acceptable value, for the [1, 1] PA.

Table II shows similar fits, but here all fits are uncorrelated. All errors have been estimated using a linear fluctuation analysis starting from the uncorrelated χ^2 , starting from the full data covariance matrix [15]. These errors agree with errors computed under a single-elimination jackknife. In these PA fits we have relaxed the constraint $b_1 \geq 4m_\pi^2 = 0.906$ GeV² (on this data set), but one notes that the values of b_1 are consistent with this bound within errors. Both correlated and uncorrelated [1, 1] PA fits are shown in Fig. 1.

The uncorrelated VMD fit reproduces ‘‘fit A’’ of Ref. [3], including the error.¹¹ One notes that the uncorrelated PA fits lead to results consistent with those of Table I, but with much larger errors. The uncorrelated VMD fit is not consistent with what we would expect to be the best fit,

$$a_\mu^{\text{HLO}, Q^2 \leq 1} = 350(8) \times 10^{-10}, \quad (4.1)$$

from the [1, 1] PA of Table I.

We may also compare the values in the tables with values obtained from a fit with a fourth order polynomial in Q^2 , which are

$$\begin{aligned} a_\mu^{\text{HLO}, Q^2 \leq 1} &= 410(91) \times 10^{-10}, & (\text{uncorrelated}), \\ a_\mu^{\text{HLO}, Q^2 \leq 1} &= 346(8) \times 10^{-10}, & (\text{correlated}). \end{aligned} \quad (4.2)$$

¹⁰See Sec. IV C for further discussion.

¹¹The parameters $\Pi(0)$ and a_1 are not the same as the parameters A and f_V of Ref. [3].

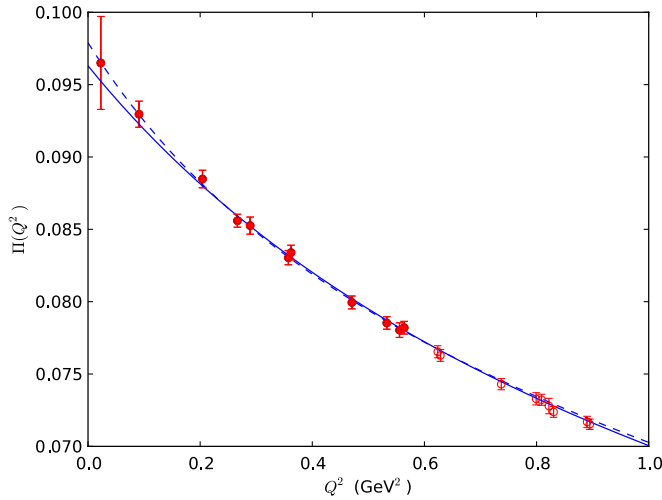


FIG. 1 (color online). [1, 1] fits of Tables I (correlated, solid curve) and II (uncorrelated, dashed curve) compared with data. Solid points have been included in the correlated fit while both solid and open points have been included in the uncorrelated fit.

The first line is in agreement with Ref. [3], and was fitted with $0 < Q^2 \leq 1 \text{ GeV}^2$, as in Table II, and the second is from a correlated fit on the interval $0 < Q^2 \leq 0.6 \text{ GeV}^2$, as in Table I. The latter fit has a χ^2/dof of 7.48/6, less good than the [1, 1] fit in Table I. Both are in good agreement with Eq. (4.1), given the size of the errors.

B. $a = 0.06 \text{ fm}$ data at $m_{\text{light}}/m_{\text{strange}} = 0.1$

For our second example, we consider the vacuum polarization computed on MILC configurations at $a = 0.06 \text{ fm}$ and $am_{\text{light}} = 0.0018$, which is about 1/10 times the physical strange quark mass. For these data, $m_\pi = 220 \text{ MeV}$ and $m_\rho \approx 800 \text{ MeV}$.¹² Correlated fits are shown in Table III, where we fitted the data for $0 < Q^2 \leq 0.53 \text{ GeV}^2$ (which corresponds to the 20 data points with the lowest values of Q^2). The χ^2 values per degree of freedom of the fits in Table III are slightly smaller than one, except for the VMD fit, for which χ^2/dof is about two. We find that the value of χ^2/dof increases if we fit over a larger range of Q^2 values, and we will therefore take the results of Table III as our optimal results (for more on this point, see the discussion in Sec. IV C). Uncorrelated fits are shown in Table IV, where, in line with Sec. IV A, fits were carried out on the interval $0 < Q^2 \leq 1 \text{ GeV}^2$.

It is again not surprising that the correlated fits become less good if one fits over a larger range in Q^2 . In order to improve the description of the data at larger Q^2 , most likely more poles would be needed. However, as before, it is clear from the tables that, given the quality of the data, it is very

¹²We thank Doug Toussaint for providing us with an unpublished rough estimate of the ρ mass for this data set.

hard to fit a second pole. The value of $a_\mu^{\text{HLO}, Q^2 \leq 1}$ is again completely insensitive to the location of the second pole.¹³

We show the [1, 1] fits of Tables III and IV in Fig. 2. As in Fig. 1 one notes the sensitivity of the fit near $Q^2 = 0$; this explains the different values for $a_\mu^{\text{HLO}, Q^2 \leq 1}$ shown in the tables.

From the [1, 1] PA fit of Table III we take what we would expect to be our best result for this data set:

$$a_\mu^{\text{HLO}, Q^2 \leq 1} = 572(41) \times 10^{-10}. \quad (4.3)$$

In Fig. 3 we show correlated and uncorrelated [1, 1] PA fits, now taking the range $0 < Q^2 \leq 0.53 \text{ GeV}^2$ as our fitting range also for the uncorrelated fit. We note that the uncorrelated fit appears to do better than the uncorrelated [1, 1] PA fit shown in Fig. 2 at the lowest Q^2 value, but much less well than the correlated fit for $Q^2 > 0.53 \text{ GeV}^2$. Accordingly, uncorrelated fits are quite sensitive to the fitting range. For instance, the central value of $a_\mu^{\text{HLO}, Q^2 \leq 1}$ from the uncorrelated fit shown in Fig. 3 is 42% larger than from a similar fit on the range $0 < Q^2 \leq 1 \text{ GeV}^2$ (shown in Table IV). A correlated fit on the latter range gives a central value which is only 3% larger than the value in Eq. (4.3), *i.e.*, it is within the error given in that equation.¹⁴

In Ref. [3] also polynomial fits with third- and fourth-order polynomials were considered, and it is thus interesting to compare PA fits with polynomial fits. For the data of this subsection, the radius of convergence, $4m_\pi^2 = 0.194 \text{ GeV}^2$.¹⁵ Therefore, fitted polynomials cannot be interpreted as estimates of the Taylor expansion of $\Pi(Q^2)$ around $Q^2 = 0$, as long as we use a fitting interval with upper bound larger than $4m_\pi^2$.

We show third- and fourth-order polynomial fits, as well as [1, 1] and [1, 2] PA fits in Table V, as a function of the number of data points in the fit (20 points corresponds to the fitting interval $0 < Q^2 \leq 0.53 \text{ GeV}^2$ used in Table III). All fits shown are correlated. Both ‘‘Poly 3’’ and ‘‘PA [1, 1]’’ are four-parameter fits, while ‘‘Poly 4’’ and ‘‘PA [1, 2]’’ are five-parameter fits. The χ^2/dof for all fits is good, except for fits with 26 data points, for which it shows a steep increase.

We observe that Poly 3, PA [1, 1] and PA [1, 2] fits all lead to values for $a_\mu^{\text{HLO}, Q^2 \leq 1}$ which are stable within the error given in Eq. (4.3). For the Poly 4 fit, however, this spread is much larger. Adding a fit parameter by going from Poly 3 to Poly 4 fits leads to significant changes in the central value for $a_\mu^{\text{HLO}, Q^2 \leq 1}$, whereas going from PA [1, 1] to PA [1, 2] fits the central values do not change much. In

¹³We even considered [2, 3] and [3, 3] fits, with the conclusion being the same.

¹⁴Despite the fact that for a correlated fit on the range $0 < Q^2 \leq 1 \text{ GeV}^2$ the value of χ^2 is about 2.5 per degree of freedom.

¹⁵The radius of convergence for the case of Sec. IV A is much larger, which is why we chose to make this comparison in this subsection.

TABLE III. PA fits to the $a = 0.06$ fm, $am_{\text{light}} = 0.0018$ data for $\Pi(Q^2)$ with $Q^2 \leq 0.53$ GeV². Correlated fits; χ^2 errors.

	χ^2/dof	$10^{10} a_\mu^{\text{HLO}, Q^2 \leq 1}$	$\Pi(0)$	a_i	b_i	a_0
VMD	38.6/18	646(8)	0.1222(6)	0.0595(8)	0.64 (fixed)	...
[0, 1]	14.3/17	550(20)	0.1203(7)	0.0646(16)	0.83(5)	...
[1, 1]	13.9/16	572(41)	0.1206(8)	0.052(16)	0.68(20)	0.005(7)
[1, 2]	13.9/15	572(37)	0.1206(8)	0.052(14)	0.68(19)	...
				1(6)	$0.3(1.0) \times 10^3$	
[2, 2]	13.9/14	572(38)	0.1206(8)	0.052(14)	0.68(18)	0.003(27)
				1(31)	$0.4(6.0) \times 10^3$	

TABLE IV. PA fits to the $a = 0.06$ fm, $am_{\text{light}} = 0.0018$ data for $\Pi(Q^2)$ with $Q^2 \leq 1$ GeV². Uncorrelated fits; errors from linear fluctuation analysis. For the [1, 2] and [2, 2] fits, b_1 is at the limit $4m_\pi^2 = 0.1936$ GeV² (for this ensemble), which was enforced in those fits.

	χ^2/dof	$10^{10} a_\mu^{\text{HLO}, Q^2 \leq 1}$	$\Pi(0)$	a_i	b_i	a_0
VMD	37.2/51	685.2(7.8)	0.1236(6)	0.0631(7)	0.64 (fixed)	...
[0, 1]	13.9/50	555(22)	0.1208(8)	0.0666(7)	0.85(4)	...
[1, 1]	12.0/49	645(66)	0.1221(13)	0.047(5)	0.54(11)	0.0071(21)
[1, 2]	11.4/48	788(482)	0.123(4)	0.015(20)	0.2(4)	...
				0.063(14)	1.4(9)	
[2, 2]	11.3/47	837(627)	0.124(5)	0.018(5)	0.2(5)	0.022(9)
				0.22(6)	3.9(6)	

other words, if we would do a correlated Poly 4 fit to 20 data points, for which we would find $a_\mu^{\text{HLO}, Q^2 \leq 1} = 535(45) \times 10^{-10}$, the error would be underestimated because of the spread of values for the Poly 4 fit shown in Table V, while the error shown in Eq. (4.3) encompasses the full range of $a_\mu^{\text{HLO}, Q^2 \leq 1}$ PA results shown in Table V. The Poly 4 and PA [1, 2] fits with 20 data points are shown in Fig. 4.

C. Discussion of fits

In this subsection we will discuss the fit results presented in Tables I through IV in more detail. We begin with the $a = 0.09$ fm results of Tables I and II.

It is important to emphasize again that the VMD fits are not part of the sequence of PAs introduced in Sec. II, because in the VMD fits the pole at $Q^2 = -b_1$ is held fixed at the estimated (squared) ρ mass on this ensemble. The actual QCD spectral function has a cut on the negative axis starting at $Q^2 = -4m_\pi^2$; any poles reside on the second Riemann sheet, away from the negative axis.¹⁶ While positivity of the spectral function implies that $\Pi(Q^2)$ can be expressed in terms of a Stieltjes function, with a convergent sequence of PAs given by Eq. (2.13), there is no reason that any of the poles of these Padé's should be equal to (the real part of) any pole representing

¹⁶Only in the limit of an infinite number of colors do the poles move toward the negative real axis, and the vacuum polarization becomes a meromorphic function.

a resonance in QCD. In particular, in the PA fits, the parameter b_1 should not be taken equal to the square of the ρ mass, but instead it should be left as a free parameter. We included the VMD fits in Tables I and II in order to compare them with the PA fits.

First, we note that the correlated VMD fit in Table I is a rather poor fit, with a high χ^2/dof , and there is no

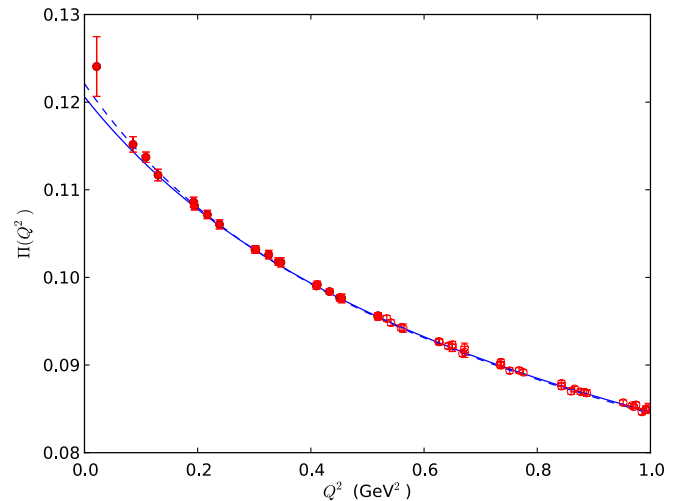


FIG. 2 (color online). [1, 1] PA fits of Tables III (correlated, solid curve) and IV (uncorrelated, dashed curve) compared with data. Solid points have been included in the correlated fit while both solid and open points have been included in the uncorrelated fit.

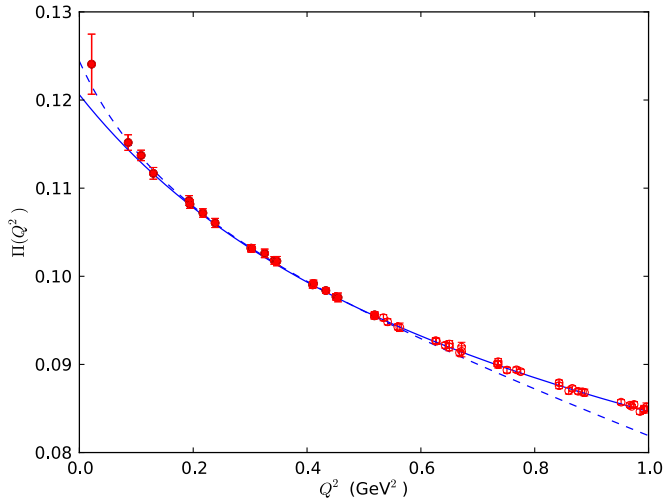


FIG. 3 (color online). [1, 1] correlated (solid curve) and uncorrelated (dashed curve) fits as in Table III, fitted on interval $0 < Q^2 \leq 0.53 \text{ GeV}^2$. Solid points have been included in the fits, open points have not been included.

agreement between the correlated and uncorrelated VMD fits. The quality of the correlated fits improves when we add more parameters, first by varying b_1 and then by adding in the parameter a_0 , by which time $\chi^2/\text{dof} \approx 1$.

The next observation is that the correlated fits do not get better by adding a second pole to the PA. The minimum value of χ^2 stays the same, and consequently, the parameters of the second pole are very poorly determined. We have checked that this does not depend on the fitting range employed. A possible explanation is that the lattice data for the vacuum polarization do not quite follow the behavior predicted by Eq. (2.2), because of the breaking of rotational invariance on the lattice (c.f., discussion at the beginning of this section). This can be seen from the fact that the data points in Fig. 1 show small deviations from a smooth, monotonically-decreasing behavior. However, we note that the value of $a_\mu^{\text{HLO}, Q^2 \leq 1}$ is completely insensitive to the parameters of the second (and higher) poles. Our best correlated fit value for $a_\mu^{\text{HLO}, Q^2 \leq 1}$ is given in Eq. (4.1).

Moving to the uncorrelated fits of Table II, we observe that the values for $a_\mu^{\text{HLO}, Q^2 \leq 1}$ for all fits in that table are consistent with each other, because of the rapid increase of errors with the order of the PA. Furthermore, all uncorrelated and correlated PA fits are consistent with each other as well, but clearly the correlated fits have much smaller errors.

It is instructive to compare the best correlated fit, the [1, 1] PA fit, with the uncorrelated VMD fit, because both have very small errors, and provide a good fit, as can be seen in Fig. 5. Both fits are good fits, but they lead to values for $a_\mu^{\text{HLO}, Q^2 \leq 1}$ which are not consistent with each other. The statistical error on the uncorrelated VMD fit is very small, but this fit has an unknown systematic error because of its model dependence. This may explain the discrepancy with the correlated [1, 1] PA fit. The latter may be expected to have a much smaller systematic error, since it is an estimate that should agree within errors with a member of a converging sequence of PAs, and clearly already provides a good fit also in the larger Q^2 region that was not included in the fit (the horizontal axis of Fig. 5 covers about twice the fitted region $0 < Q^2 \leq 0.53 \text{ GeV}^2$).

However, for the computation of $a_\mu^{\text{HLO}, Q^2 \leq 1}$ the region $Q^2 \sim m_\mu^2 = 0.011 \text{ GeV}^2$ dominates, and it is clear that the data do not distinguish between these two fits in that region. While the rapid increase of the goodness of fit seen in Table I from the VMD fit to the [1, 1] fit can be taken as an indication that correlated fits unbiased by model dependence are promising, we conclude that it is not possible to exclude either value of $a_\mu^{\text{HLO}, Q^2 \leq 1}$ on the basis of these data.

Similar remarks apply to the fits shown in Tables III and IV. Correlated PA fits all have $\chi^2/\text{dof} \approx 1$, unlike the correlated VMD fit for which $\chi^2/\text{dof} \approx 2$. In both cases, the correlated and uncorrelated VMD fits do not agree within errors (which, we recall, are purely statistical). The value we obtain for $a_\mu^{\text{HLO}, Q^2 \leq 1}$ is larger than that obtained in Sec. IV A; we believe that this is mostly due to a smaller pion mass, with $m_\pi \approx 220 \text{ MeV}$ for this data set, while $m_\pi \approx 480 \text{ MeV}$ for the $a = 0.09 \text{ fm}$ data set.

TABLE V. Correlated PA and polynomial fits to the data of Table III, as a function of the fitting interval. The first column shows the number of data points in the fit, with 20 points corresponding to the fitting interval $0 < Q^2 \leq 0.53 \text{ GeV}^2$ of Table III. The column ‘‘Poly n ’’ shows results from a fit to a polynomial of degree n ; $a_\mu^{(1)}$ stands for $10^{10} a_\mu^{\text{HLO}, Q^2 \leq 1}$.

# points	Poly 3		Poly 4		PA [1,1]		PA [1,2]	
	χ^2/dof	$a_\mu^{(1)}$	χ^2/dof	$a_\mu^{(1)}$	χ^2/dof	$a_\mu^{(1)}$	χ^2/dof	$a_\mu^{(1)}$
16	9.6/12	543(35)	9.5/11	483(244)	9.7/12	564(55)	9.7/11	565(41)
18	11.4/14	526(33)	10.5/13	596(79)	11.2/14	541(46)	11.5/13	561(21)
20	13.1/16	536(23)	13.1/15	535(45)	13.9/16	572(41)	13.9/15	572(37)
22	16.5/18	541(23)	15.9/17	513(44)	18.5/18	566(37)	18.5/17	566(33)
24	16.6/20	537(18)	16.4/19	521(41)	19.4/20	583(34)	19.4/19	583(33)
26	30.7/22	505(16)	23.6/21	580(32)	26.8/22	557(31)	26.7/21	560(27)

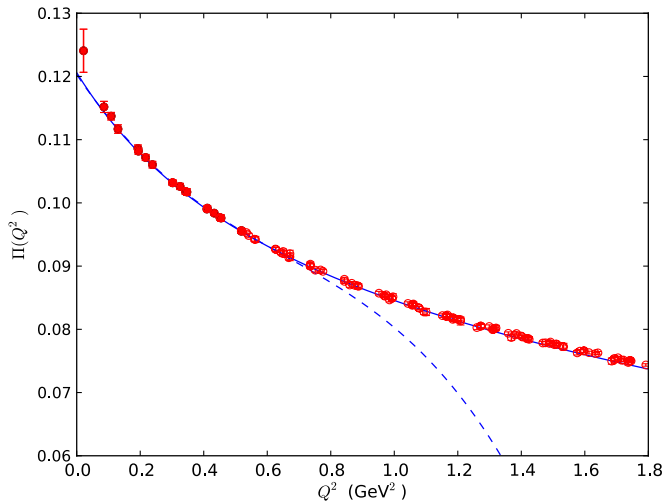


FIG. 4 (color online). Comparison of correlated [1, 2] PA (solid curve) and 4th-order polynomial (dashed curve) fits, both fitted on the interval $0 < Q^2 \leq 0.53 \text{ GeV}^2$. Solid points have been included in the fits, open points have not been included.

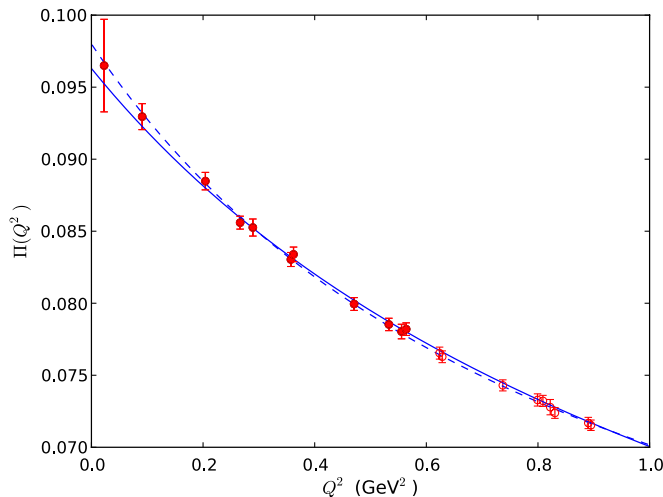


FIG. 5 (color online). [1, 1] fit of Table I (correlated, solid curve) and VMD fit of Table II (uncorrelated, dashed curve) compared with data. Solid points have been included in the correlated fit while both solid and open points have been included in the uncorrelated fit.

V. CONCLUSION

In this article, we presented a new, model-independent method for fitting the hadronic vacuum polarization $\Pi(Q^2)$ as a function of euclidean momentum Q^2 to data obtained from a lattice QCD computation. The method is based on the theory of PAs to a Stieltjes function, and yields, in principle, a converging sequence of PAs to the vacuum polarization.

These PAs can be used to obtain lattice estimates for the leading hadronic contribution to the anomalous magnetic moment of the muon from Eq. (1.1a). By comparing successive PAs in the sequence, one should be able to

check the convergence in practice. This would allow for a fully model-independent determination of the leading hadronic contribution a_μ^{HLO} , and thus help eliminate an unknown systematic error present in all lattice computations of a_μ^{HLO} to date.

In comparison with the VMD fits which have been employed in the literature, these PAs contain more parameters (the [0, 1] PA already contains three parameters, whereas the simplest VMD ansatz contains only two). One thus typically expects larger statistical errors given certain lattice data. However, the PA approach avoids model-dependent assumptions, and hence removes the unknown systematic error associated with the VMD approach.

We have explored this new framework on two state-of-the-art ensembles of gauge configurations, at different lattice spacings and pion masses. From these explorations, we conclude that this new method looks promising, but that better data at very low values of Q^2 will be needed in order to control the extrapolation necessary for a reliable computation of a_μ^{HLO} from the integral in Eq. (1.1a).

Our explorations show that given current lattice data for $\Pi(Q^2)$, there is a significant difference between our best PA fits (which are four-parameter [1, 1] PAs), and VMD fits.¹⁷ For instance, the difference between the values of $a_\mu^{\text{HLO}, Q^2 \leq 1}$ obtained from the correlated [1, 1] PA fit of Table I and the uncorrelated VMD fit of Table II and Ref. [3] is about 15–20%, much larger than the statistical fit error on each of these values. While it is tempting to view the value from the correlated [1, 1] PA fit as the more reliable one, it is clear from Fig. 5 that more data points with higher precision at low Q^2 are needed in order to reduce this uncertainty.

There are of course other systematic errors as well, including finite volume effects. In order to study those, simulations at larger volumes will be needed; at present it is not possible to assess what role they play in the results we obtained.

Our explorations also showed that with these data it is very difficult to fit the parameters characterizing the second and higher poles of the PAs. In order to test the convergence of the sequence of PAs fitted to $\Pi(Q^2)$, it would be desirable to investigate this issue, which is possibly related to breaking of rotational invariance at nonzero lattice spacing, in more detail in the future. This issue appears to have no direct effect on the value of $a_\mu^{\text{HLO}, Q^2 \leq 1}$, which we found to be very insensitive to the location and residues of the second and higher poles.

In conclusion, the new method presented here looks promising, but data for $\Pi(Q^2)$ with more values at

¹⁷We recall that VMD fits cannot be viewed as low-order PA fits, because there is no *a priori* relation between PA poles and QCD resonance parameters; consequently, the first pole should not be chosen equal to the square of the ρ mass, as is done in most VMD fits.

$Q^2 \sim m_\mu^2$ and with higher statistics will be necessary in order to attain the high precision determination of a_μ^{HLO} needed for a meaningful comparison with experiment. Work in this direction is in progress.

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