
g2tools Documentation

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MUON G-2 FROM LATTICE QCD USING G2TOOLS

Module `g2tools` contains a small number of tools useful for analyzing contributions to the muon's magnetic moment from (lattice) QCD vacuum polarization. These tools were developed by G.P. Lepage to implement the analysis presented in Chakraborty *et al*, Phys.Rev. D89 (2014) no.11, 114501 (arXiv:1403.1778) and subsequent papers by the same authors.

A typical application, illustrating the most important tools, is provided by the following code:

```
import g2tools as g2
import gvar as gv

def main():
    # data
    Z = gv.gvar('0.9938(17)')          # current Z factor
    Q = 1. / 3.                        # charge of quark (units of proton charge)
    ainv = gv.gvar('1.6280(86)')       # inverse lattice spacing (in GeV)

    G = gv.gvar([                      # G(t) for t=0..63 (in lattice units)
        '0.0870904(11)', '0.0435138(14)', '0.00509859(48)', '0.00305614(43)',
        '0.00069516(19)', '0.00045466(15)', '0.000166972(80)', '0.000102219(58)',
        '0.000045284(34)', '0.000026213(22)', '0.000012630(14)', '7.0635(91)e-06',
        '3.5569(57)e-06', '1.9469(37)e-06', '1.0027(24)e-06', '5.421(16)e-07',
        '2.834(10)e-07', '1.5174(67)e-07', '7.943(43)e-08', '4.253(28)e-08',
        '2.221(19)e-08', '1.183(12)e-08', '6.132(81)e-09', '3.292(51)e-09',
        '1.727(34)e-09', '9.19(22)e-10', '4.81(14)e-10', '2.643(96)e-10',
        '1.385(64)e-10', '7.61(44)e-11', '3.92(31)e-11', '2.67(24)e-11',
        '2.07(21)e-11', '2.90(23)e-11', '4.12(31)e-11', '8.20(42)e-11',
        '1.380(65)e-10', '2.788(98)e-10', '5.01(15)e-10', '9.72(23)e-10',
        '1.782(34)e-09', '3.406(53)e-09', '6.333(78)e-09', '1.212(12)e-08',
        '2.249(18)e-08', '4.283(28)e-08', '8.016(44)e-08', '1.5263(67)e-07',
        '2.843(10)e-07', '5.420(16)e-07', '1.0062(25)e-06', '1.9453(39)e-06',
        '3.5611(58)e-06', '7.0675(93)e-06', '0.000012647(14)', '0.000026240(22)',
        '0.000045282(32)', '0.000102285(56)', '0.000166993(79)', '0.00045479(15)',
        '0.00069503(19)', '0.00305647(42)', '0.00509870(47)', '0.0435158(14)'
    ])

    # N.B.: In general would construct G so that correlations from one t
    # to the next are included. Don't bother here since this is meant
    # just to illustrate g2tools.

    # compute moments, converting to physical units from lattice units
    mom = g2.moments(G, ainv=ainv, Z=Z, periodic=True, nlist=[4, 6, 8, 10])
    print('Taylor coefficients:', g2.mom2taylor(mom))
    print()

    # construct subtracted vac pol function using [2,2] Pade
    vpol = g2.vacpol(mom, order=(2,2))
```

```

# integrate vpol to get a_mu and print result
amu = g2.a_mu(vpol, Q=Q)
print('a_mu contribution =', amu)
print()

# error budget for a_mu
print(gv.fmt_errorbudget(
    outputs=dict(a_mu=a_mu, mom4=mom[4]),
    inputs=dict(G=G, Z=Z, ainv=ainv),
))

if __name__ == '__main__':
    main()

```

In this code, we first read the simulation data for the jj correlator into array G , where $G[i]$ is the correlator for (Euclidean) time i/ainv where $i=0, 1 \dots 63$. We then use `g2tools.moments()` to calculate temporal moments of the correlator, while also converting from lattice units to physical units (using the inverse lattice spacing ainv) and renormalizing the current (Z).

`vpol(q2)` is the vacuum polarization function at Euclidean q^2 equal to $q2$. Object `vpol` has type `g2tools.vacpol`. It constructs a $[2,2]$ Padé approximant from the moments, and uses that approximant to approximate the exact function. The approximants converge to the exact result as the order increases provided the momentum is space-like ($q2$ non-negative). Using a $[1,1]$ Padé instead of $[2,2]$ gives almost identical results here, so the approximants have converged for the present application.

We calculate the contribution from vacuum polarization `vpol` to the muon's anomalous magnetic moment a_μ using `g2tools.a_mu()`. We also use `gv.fmt_errorbudget()` to produce an error budget for it and the 4th moment.

Running this code gives the following output:

```

Taylor coefficients: [0.06629(74) -0.0527(11) 0.0472(15) -0.0435(18)]

a_mu contribution = 5.412(57)e-09

Partial % Errors:

```

	a_mu	mom4

ainv:	1.00	1.06
Z:	0.34	0.34
G:	0.01	0.01

total:	1.06	1.11

The contribution to the muon's anomalous magnetic moment is $54.12(57) \times 10^{-10}$. The error budget shows that the final uncertainty is dominated by the uncertainty in the inverse lattice spacing ainv ; statistical errors from G are completely negligible in this example.

`g2tools` is designed to work with module `gvvar` which we use here to represent the statistical and systematic uncertainties in the correlator values, inverse lattice spacing, and Z factor. Each of these quantities is an object of type `gvvar.GVar`, which represents a Gaussian random variable. `gvvar.GVars` describe not only means and standard deviations, but also statistical correlations between different objects. These correlations are propagated through arbitrary arithmetic statements. Adding the following code to the end of `main()`, for example,

```
print(gv.evalcorr([mom[4], mom[6], mom[8], mom[10]]))
```

prints out the correlation matrix for the moments, showing that they are highly correlated (as expected):


```
[[ 1.          0.98833867  0.9787737  0.97262094]
 [ 0.98833867  1.          0.99853653  0.99646438]
 [ 0.9787737   0.99853653  1.          0.99949934]
 [ 0.97262094  0.99646438  0.99949934  1.          ]]
```

The moments are also highly correlated with the final results `a_mu`: for example, adding the following to the end of `main()`

```
print(gv.evalcorr([a_mu, mom[4]]))
```

gives:

```
[[ 1.          0.96864247]
 [ 0.96864247  1.          ]]
```

This kind of correlation information is used by `gvar.fmt_errorbudget(...)` to create the error budget. See `gvar`'s documentation for more information.

G2TOOLS MODULE

2.1 Moments

The main tools for creating and manipulating moments are:

`g2tools.moments` (*G*, *Z*=1.0, *ainv*=1.0, *periodic*=True, *nlist*=[4, 6, 8, 10, 12, 14, 16, 18, 20])

Compute t^{**n} moments of correlator *G*.

Compute $\sum_t t^{**n} G(t)$ for *n* in *nlist*, where both positive and negative *t* are included.

Parameters

- **G** – Array of correlator values $G[t]$ for $t=0, 1, \dots$
- **Z** – Renormalization factor for current (moments multiplied by Z^{**2}). Default is 1.
- **ainv** – Inverse lattice spacing used to convert moments to physical units (*n*-th moment multiplied by $1/ainv^{** (n-2)}$). Default is 1.
- **periodic** – `periodic=True` implies $G[-t] = G[t]$ (default); `periodic=False` implies no periodicity in array $G[t]$ (and results doubled to account for negative *t*).
- **nlist** – List of moments to calculate. Default is *nlist*=[4, 6, 8...20].

Returns Dictionary *Gmom* where *Gmom*[*n*] is the *n*-th moment.

`g2tools.mom2taylor` (*mom*)

Convert moments in dictionary *mom* into Taylor series coefficients.

`g2tools.taylor2mom` (*tayl*)

Convert Taylor coefficients in array *tayl* to moments.

2.2 Subtracted Vacuum Polarization

A subtracted vacuum polarization function ($\hat{\Pi}$) is represented by the following class:

class `g2tools.vacpol` (*g*, *order*=None, *scale*=None, *rtol*=None)

Subtracted vac. $\text{pol}'n$ ($\hat{\Pi}(q^2)$) from correlator moments *Gmon*[*n*].

The vacuum polarization function is a Pade approximant to the Taylor series corresponding to the moments *g*[*n*]. The code estimates the precision of the moments and sets the tolerance for the Pade determination accordingly. The order (*m*, *n*) of the Pade can be specified, but might be reduced by the code if the data are noisy.

`vacpol` objects are used primarily as functions (of q^2) but also have several attributes. Attribute `pseries` is a dictionary containing various powerseries (see `gvar.powerseries`) describing the function: the vacuum polarization function is q^2 times a Pade approximant with a numerator given by `pseries['num']` and a denominator given by `pseries['den']`. The Taylor series for this function is given by q^2 times `pseries['taylor']`.

`vacpol` objects also have a method `vacpol.badpoles()` that tests the poles in the denominator of the Pade. `badpoles(qth)` returns `False` if any of the poles is complex or if any are located above $-(qth ** 2)$. `qth` should be set equal to the threshold energy for the correlator. If it is unset, `qth=0` is used.

`vacpol` has several static methods for creating specialized examples of vacuum polarizations (e.g., for testing):

- `vacpol.fermion(m)` – 1-loop fermion (mass m) contribution;
- `vacpol.scalar(m)` – 1-loop scalar (mass m) contribution;
- `vacpol.vector(m, f)` – tree-level contribution from vector with mass m and decay constant f .

Parameters

- **g** – Dictionary containing moments where $g[n] = \sum_t t^{*n} G(t)$, or array containing Taylor coefficients where $\hat{P}_i(q^2) = q^2 * \sum_j q^{2*j} * g[j]$.
- **order** – Tuple (m, n) specifying the order of the Pade approximant used to approximate the function. The order may be reduced (automatically) if the data are too noisy. If the order is not specified, it is set automatically according to the number of entries in **G**.
- **scale** – Scale factor used to rescale q^2 so that the Taylor coefficients are more uniform in size. This is normally set automatically (from the first two moments), but the automatic value is overridden if `scale` is set.
- **rtol** – Relative tolerance assumed when determining the Pade approximant. This is normally set automatically (from the standard deviations of the moments), but the automatic value is overridden if `rtol` is specified.

2.3 Padé Approximants

The following two functions are used for calculating Padé approximants from the Taylor coefficients of an arbitrary function. The first (`g2tools.pade_svd()`) implements an algorithm that uses *svd* cuts to address instabilities caused by uncertainties in the Taylor coefficients. The second function (`g2tools.pade_gvar()`) is built on the first but allows Taylor coefficients to have uncertainties (`gvar.GVars`). The statistical uncertainties and correlations between different coefficients are propagated through the analysis.

`g2tools.pade_svd(f, m, n, rtol=1e-14)`
[m, n] Pade approximant to $\sum_i f[i] x^{*i}$.

The [m, n] Pade approximant to a series given by $\sum_i f[i] * x^{*i}$ is the ratio of polynomials of order m (numerator) and n (denominator) whose Taylor expansion agrees with that of the original series up to order $m+n$.

This code is adapted from P. Gonnet, S. Guttel, L. N. Trefethen, SIAM Review Vol 55, No. 1, 101 (2013). It uses an *svd* algorithm to deal with imprecision in the input data, here specified by the relative tolerance `rtol` for the input coefficients `f[i]`. It automatically reduces the order of the approximant if the extraction of Pade coefficients is too unstable given tolerance `rtol`.

Parameters

- **f** – Array `f[i]` of power series coefficients for $i=0 \dots n+m$.
- **m** – Maximum order of polynomial in numerator of Pade approximant ($m \geq 0$).

- **n** – Maximum order of polynomial in denominator of Pade approximant ($m \geq 0$).
- **rtol** – Relative accuracy of input coefficients.

Returns Tuple of power series coefficients (p , q) such that $\sum_i p[i] x^{*i}$ is the numerator of the approximant, and $\sum_i q[i] x^{*i}$ is the denominator. $q[0]$ is normalized to 1.

`g2tools.pade_gvar(f, m, n, rtol=None)`

$[m, n]$ Pade approximant to $\sum_i f[i] x^{*i}$ for GVars.

The $[m, n]$ Pade approximant to a series given by $\sum_i f[i] * x^{*i}$ is the ratio of polynomials of order m (numerator) and n (denominator) whose Taylor expansion agrees with that of the original series up to order $m+n$.

This code uses an *svd* algorithm (see `pade_svd()`) to deal with imprecision in the input data. It automatically reduces the order of the approximant if the extraction of Pade coefficients is too unstable given noise in the input data.

Parameters

- **f** – Array $f[i]$ of power series coefficients for $i=0 \dots n+m$.
- **m** – Maximum order of polynomial in numerator of Pade approximant ($m \geq 0$).
- **n** – Maximum order of polynomial in denominator of Pade approximant ($m \geq 0$).
- **rtol** – Relative accuracy of input coefficients. Overrides default estimate from the $f[i]$ unless set equal to `None`.

Returns Tuple of power series coefficients (p , q) such that $\sum_i p[i] x^{*i}$ is the numerator of the approximant, and $\sum_i q[i] x^{*i}$ is the denominator. $q[0]$ is normalized to 1.

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