

Path Integral Monte Carlo

Lattice QCD for novices

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Discretizing Path Integrals for 1-dim Quantum Mechanics to evaluate the ground state

Propagator from t_i to t_f : $\langle x_f | e^{-\hat{H}(t_f - t_i)} | x_i \rangle = \int Dx(t) e^{-S[x]}$

w/ **Classical Action**: $S[x] \equiv \int_{t_i}^{t_f} L(x, \dot{x}) dt \equiv \int_{t_i}^{t_f} dt \left(\frac{m\dot{x}(t)^2}{2} + V(x(t)) \right)$

Setting $x_i = x_f = x$ and $t_f - t_i = T$ we have

$$\langle x | e^{-\hat{H}T} | x \rangle = \sum_k \langle x | E_k \rangle e^{-E_k T} \langle E_k | x \rangle.$$

At $T \rightarrow \infty$ only the **groundstate** contributes so we can extract it by integrating over x : $\int dx \langle x | e^{-\hat{H}T} | x \rangle \rightarrow_{T \rightarrow \infty} e^{-E_0 T}.$

We can also determine the **ground state wavefunction**: $\Psi_{E_0}(x) = \langle x | E_0 \rangle.$

Discretizing Path Integrals for 1-dim Quantum Mechanics to evaluate the ground state

How to develop a numerical procedure for evaluating the propagator using path integral

- We approximate $x(t)$ only at the nodes on a **discretized** t axis:
 $\Rightarrow t_j = t_i + ja$ with $j = 0, \dots, N$ and Lattice spacing $a = \frac{t_f - t_i}{N}$
 $\Rightarrow x(t) = \{x(t_0), x(t_1), \dots, x(t_N)\}$
 $\Rightarrow \int Dx(t) \rightarrow A \int_{-\infty}^{+\infty} dx_1 dx_2 \dots dx_{N-1}$ with $x_0 = x_N = x$
- The action for $t_j \leq t \leq t_{j+1}$ is

$$S_{lat}^j \approx a \left[\frac{m}{2} \left(\frac{x_{j+1} - x_j}{a} \right)^2 + \frac{1}{2} (V(x_{j+1}) + V(x_j)) \right]$$

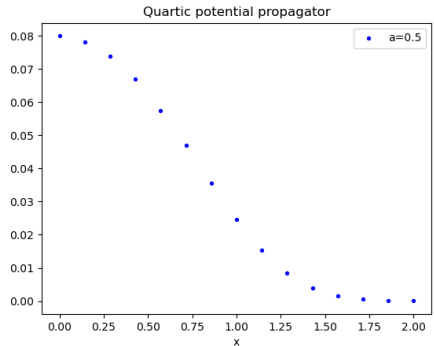
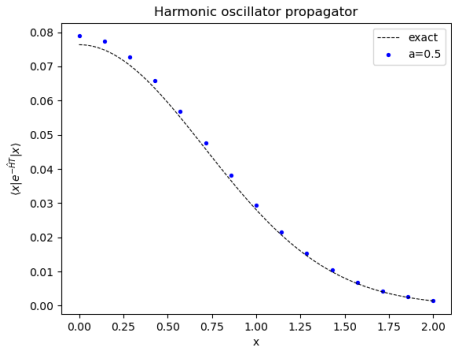
$$\langle x | e^{-\hat{H}(t_f - t_i)} | x \rangle \approx A \int_{-\infty}^{+\infty} dx_1 dx_2 \dots dx_{N-1} e^{-S_{lat}[x]}$$

Exercise 1: Evaluation of harmonic and quartic potential propagator for several values of $x_0 = x_N$

```
1 #Integrand function
2 def I(x):
3     S=0 #constructing the action
4     for i in range(N-1):
5         if i == 0:
6             #this is the action from x_i to x[0]
7             S += (1/(2*a))*(x[i]-x_i)**2 + (a/2)*(V(x_i) + V(x[i]))
8         elif i == N-2:
9             #this is the action from x[5] to x[6] and x[6] to x_f
10            S += (1/(2*a))*((x_f-x[i])**2 + (x[i]-x[i-1])**2) + (a/2)*(V(x[i])+
11            V(x[i-1])) + (a/2)*(V(x_f)+V(x[i]))
12        else:
13            #this is the action from x[i-1] to x[i]
14            S += (1/(2*a))*(x[i]-x[i-1])**2 + (a/2)*(V(x[i-1])+V(x[i]))
15    return ((1/(2*np.pi*a))*(N/2))*np.exp(-S)
16
17 #Specify integration details, including integration intervals for each of the
18 #independent variables.
19 integ = vegas.Integrator((N-1)*[[x_min, x_max]])
20
21 #Integration results of the integrand function I
22 result = integ(I, nitn=10, neval=100000)
23
```

Code extracted from github.com/MatildeGrassi/Lattice-QCD-for-novices—project

Exercise 1: Results for $a = 0.5$, $N = 8$, $x_0 = x_N = np.linspace(0, 2, 15)$



Monte Carlo evaluation of path integral to extract states beyond the ground state

Path integral averages $\langle\langle\Gamma(x)\rangle\rangle$ of arbitrary functionals $\Gamma(x)$

$$\langle\langle\Gamma(x)\rangle\rangle = \frac{\int D\mathbf{x}(t)\Gamma(\mathbf{x})e^{-S[\mathbf{x}]}}{\int D\mathbf{x}(t)e^{-S[\mathbf{x}]}}$$

are used to compute physical property of the excited states in quantum theory.

e.g. in exercise 2: $\Delta E = E_1 - E_0 = \frac{1}{a} \log \frac{G(t)}{G(t+a)}$

with $G(t) = \frac{1}{N} \sum_j \langle\langle x(t_{j+n \bmod N}) x(t_j) \rangle\rangle$

Metropolis algorithm

To evaluate the path integrals we need to generate a large number of configurations $x^\alpha = \{x_0^\alpha, x_1^\alpha, \dots, x_N^\alpha\}$ with $\alpha = 1, 2, \dots, N_{cf}$, in order to compute the **Monte Carlo estimator**: $\langle\langle \Gamma(x) \rangle\rangle \approx \bar{\Gamma} = \frac{1}{N_{cf}} \sum_{\alpha=1}^{N_{cf}} \Gamma[x^\alpha]$.

```
1  @njit
2  def update(x): #to update each site of path
3      for j in range(N):
4          #saving original values and compute the action
5          old_x = x[j]
6          old_Sj = S(j,x)
7          #update the x[j] using a uniformly distributed probability
8          x[j] += np.random.uniform(-eps,eps)
9          #compute the difference between the old and new action
10         dS = S(j,x) - old_Sj
11         #condition to restore the old value
12         if dS>0 and np.exp(-dS)<np.random.uniform(0,1):
13             x[j] = old_x
14
15  @njit
16  def S(j,x): #Action
17      jp = (j+1)%N #taking into account the periodicity xN=x0
18      jm = (j-1)%N
19      return a*v(x[j])+x[j]*(x[j]-x[jp]-x[jm])/a
```

Code extracted from github.com/MatildeGrassi/Lattice-QCD-for-novices—project

Metropolis algorithm

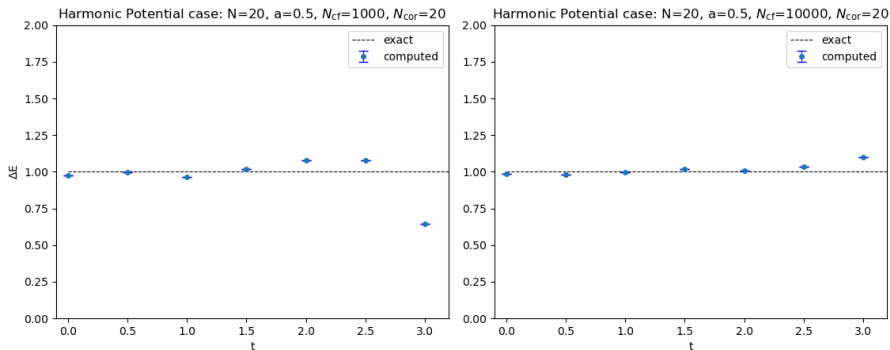
Successive paths generated by the Metropolis algorithm will be correlated. We need to **thermalize** them at the beginning and we need to keep only every N_{cor} updates.

```
1  @njit
2  def MCaverage(x,G): #erase correlations and thermalize the G's
3      for j in range(0,N):
4          x[j] = 0 #initialization of the path as all zeros
5      for j in range(10*N_cor): #thermalization
6          update(x)
7      for alpha in range(len(G)):
8          for j in range(N_cor): #erasing correlations
9              update(x)
10         for n in range(0,N):
11             G[alpha][n] = compute_G(x,n) #filling the G
12     return G
13
```

Code extracted from github.com/MatildeGrassi/Lattice-QCD-for-novices—project

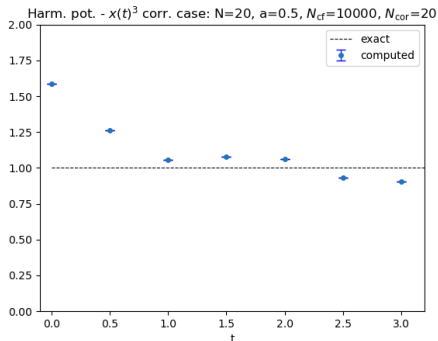
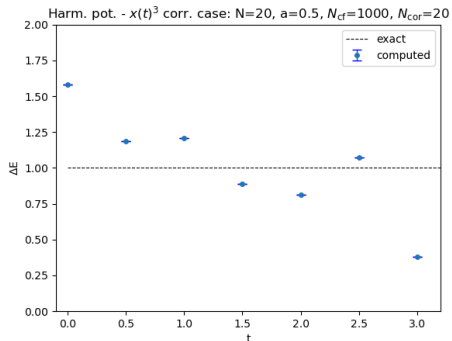
Exercise 2: Results for Harmonic potential and correlator

$$\frac{1}{N} \sum_j \langle \langle x(t_{(j+n) \bmod N}) x(t_j) \rangle \rangle$$



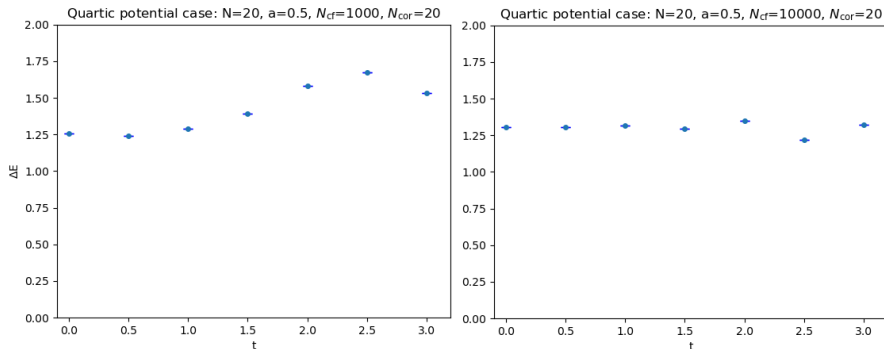
Exercise 3: Results for Harmonic potential and correlator

$$\frac{1}{N} \sum_j \langle \langle x(t_{(j+n) \bmod N})^3 x(t_j)^3 \rangle \rangle$$



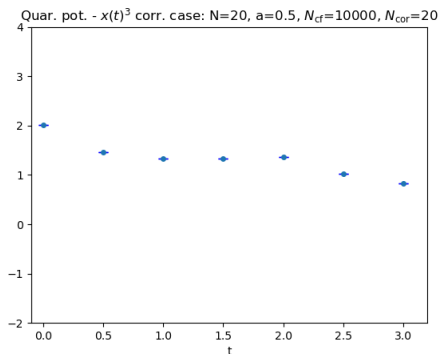
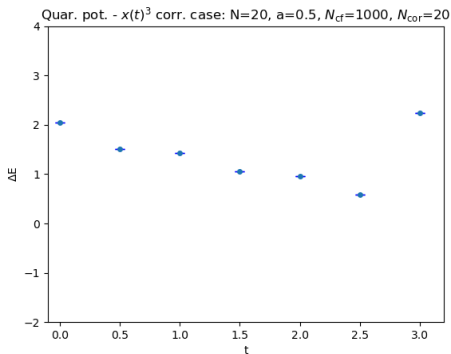
Exercise 2: Results for Quartic potential and correlator

$$\frac{1}{N} \sum_j \langle \langle x(t_{(j+n) \bmod N}) x(t_j) \rangle \rangle$$



Exercise 3: Results for Quartic potential and correlator

$$\frac{1}{N} \sum_j \langle \langle x(t_{(j+n) \bmod N})^3 x(t_j)^3 \rangle \rangle$$



Estimation of errors - Statistical bootstrap

Given an ensemble G^α with $\alpha = 1, \dots, N_{cf}$, we assemble a bootstrap copy of that ensemble by selecting $G^{\alpha'}$ s at random from the original ensemble, allowing duplications and omissions.

```
@njit
def bootstrap(G): #bootstrapping the G
    #new ensemble
    G_bootstrap = np.zeros((len(G), N))
    for i in range(len(G)):
        #choose random config
        k = int(np.random.uniform(0, len(G)))
        #keep G[k]
        G_bootstrap[i]=G[k]
    return G_bootstrap

1 @njit #DeltaE+errors
2 def bootstrap_deltaE(G,nbstrap=100):
3     bootstrap_E = np.zeros((nbstrap,N-1))
4     #bs copies of deltaE
5     for i in range(nbstrap):
6         g = bootstrap(G)
7         g_avg=average(g)
8         bootstrap_E[i]=deltaE(g_avg)
9     #spread of deltaEs
10    sdevE = stdDev(bootstrap_E)
11    #deltaEs
12    avgE=average(bootstrap_E)
13    return avgE,sdevE
14
```

Code extracted from github.com/MatildeGrassi/Lattice-QCD-for-novices—project

Estimation of errors - Binning

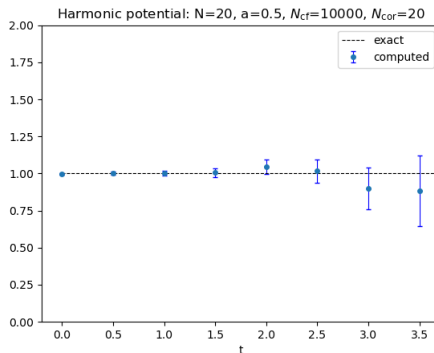
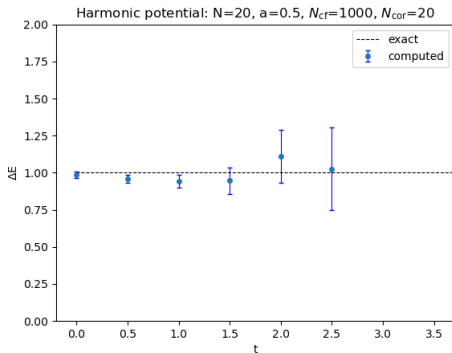
Binning allows us to save a lot of disk space, RAM and CPU time and it also reduces correlations.

```
1  def bin(G, binsize): #binning the G's
2      G_binned = np.zeros((int(len(G)/binsize), N))
3      k=0
4      for i in range(0, len(G), binsize):
5          G_avg = 0
6          for j in range(binsize): #summing for each bin
7              G_avg = G_avg + G[i+j]
8          G_binned[k]=(G_avg/binsize) #bin average
9          k+=1
10     return G_binned
11
```

Code extracted from github.com/MatildeGrassi/Lattice-QCD-for-novices—project

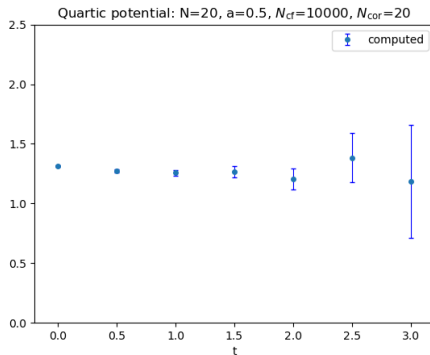
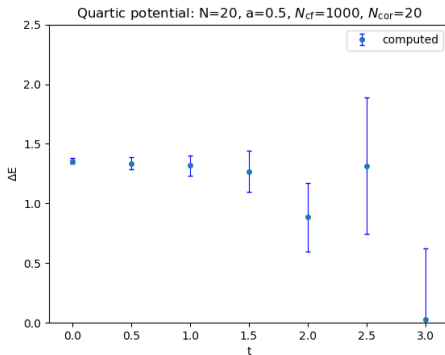
Exercise 4: Results for harmonic oscillator and correlator

$$\frac{1}{N} \sum_j \langle \langle x(t_{(j+n) \bmod N}) x(t_j) \rangle \rangle$$



Exercise 4: Results for quartic potential and correlator

$$\frac{1}{N} \sum_j \langle \langle x(t_{(j+n) \bmod N}) x(t_j) \rangle \rangle$$



Improved discretization of action

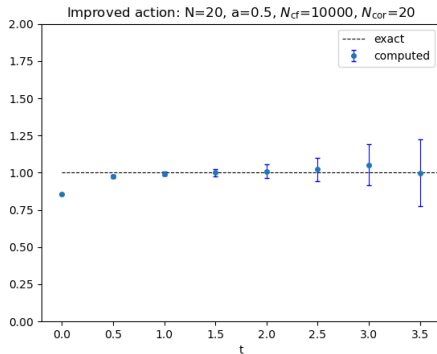
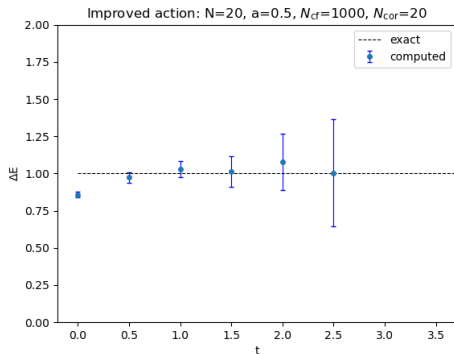
We use the **4th order central difference method** applied to the 2nd derivative:

$$\ddot{x}(t) \rightarrow \Delta^{(2)}x_j - \frac{a^2}{12}(\Delta^{(2)})^2x_j \quad w/ \quad \Delta^{(2)}x_j = \frac{x_{j+1} - 2x_j + x_{j-1}}{a^2}$$

Thus, we obtain

$$S_{imp}[x] = a \sum_{j=0}^{N-1} \left[-\frac{1}{2} m x_j \left(\frac{-x_{j+2} + 16x_{j+1} - 30x_j + 16x_{j-1} - x_{j-2}}{12a^2} \right) + V(x(t)) \right]$$

Exercise 6: Results with improved action for harmonic case



Avoiding ghost states

Let's derive the **discretized euclidean eq. of motion** from $\frac{\partial S[x]}{\partial x_j} = 0$.

Using the **unimproved S[x]** w/ harmonic pot. we get $\frac{x_{i+1} - 2x_i + x_{i-1}}{a^2} = \omega_0^2 x_i$.

Let's plug in the ansatz solution $x_i = e^{-\omega t_j}$ we obtain the frequency

$$\omega^2 = \omega_0^2 \left[1 - \frac{(a\omega_0)^2}{12} + \mathcal{O}((a\omega)^4) \right] \quad \text{error of the order of } a^2$$

Using the **improved S[x]** we obtain two solutions.

One is

$$\omega^2 = \omega_0^2 \left[1 + \frac{(a\omega_0)^4}{90} + \mathcal{O}((a\omega)^6) \right] \quad \text{error of the order of } a^4 \text{ (Improved)}$$

Avoiding ghost states - Field transformation

The other solutions comes from setting $\omega_0 = 0$:

$$\omega^2 \approx \left(\frac{2.6}{a}\right)^2$$

It corresponds to new oscillation modes which are an **artifact of the improved lattice theory**, called **ghost modes**.

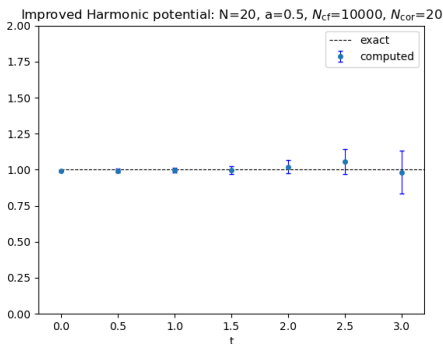
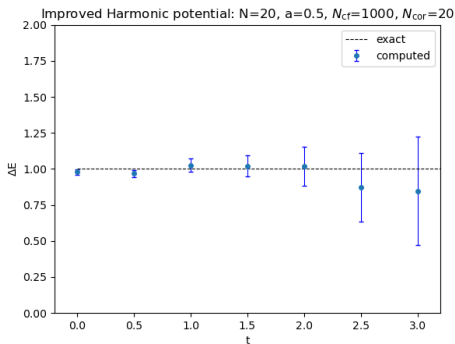
To avoid them we operate a change of variables:

$$x_j \rightarrow \tilde{x}_j \quad w/ \quad x_j = \tilde{x}_j + \delta\tilde{x}_j \quad \text{and} \quad \delta\tilde{x}_j = \xi_1 a^2 \Delta^{(2)} \tilde{x}_j + \xi_2 a^2 \omega_0^2 \tilde{x}_j$$

$$\Rightarrow \tilde{S}_{imp}(\tilde{x}) = \frac{1}{2} m \tilde{x}_j \Delta^{(2)} \tilde{x}_j + \tilde{V}_{imp}(\tilde{x}_j)$$

$$\text{with} \quad \tilde{V}_{imp}(\tilde{x}) = \frac{1}{2} m \omega_0^2 \tilde{x}_j^2 \left(1 + \frac{(a\omega_0)^2}{12}\right)$$

Exercise 7: Avoiding ghost states for harmonic case



Avoiding ghost states - Field transformation

Generalizing this trick to the anharmonic case:

$$V(x) = \frac{1}{2}m\omega_0^2 x^2(1 + cm\omega_0 x^2) \quad \text{with } c \text{ dimensionless parameter.}$$

We do the following variable change:

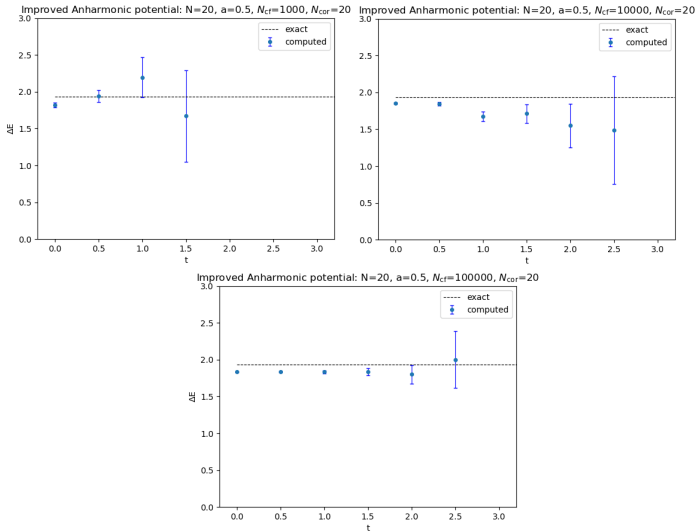
$$\delta\tilde{x}_j \equiv \xi_1 a^2 \Delta^{(2)} \tilde{x}_j + \xi_2 a^2 \omega_0^2 \tilde{x}_j + \xi_3 a^2 m \omega_0^3 \tilde{x}_j^3.$$

We obtain the new \tilde{V}_{imp} :

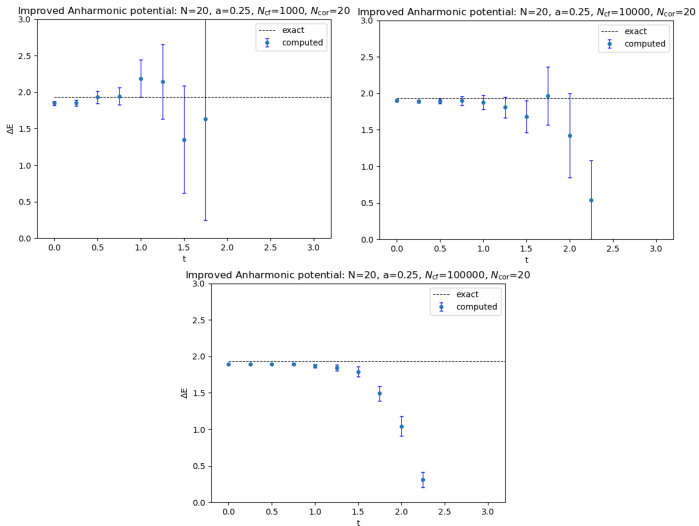
$$\tilde{V}_{imp}(\tilde{x}) = \frac{m\omega_0^2}{2} \tilde{x}^2(1 + cm\omega_0 \tilde{x}^2) + \frac{a^2 m \omega_0^4}{24} (\tilde{x} + 2cm\omega_0 \tilde{x}^3)^2 - a\delta v(\tilde{x}) + \frac{a^3}{2} \delta v(\tilde{x})^2$$

$$\text{with } \delta v(\tilde{x}) \equiv cm\omega_0^3 \tilde{x}^2/4.$$

Exercise 8: Avoiding ghost states for anharmonic case



Exercise 8: Avoiding ghost states for anharmonic case



Bibliography

- <https://github.com/MatildeGrassi/Lattice-QCD-for-novices—project>
- <https://arxiv.org/abs/hep-lat/0506036>
- https://github.com/paolofinelli/latticeQCD_novices