

PS5

May 5, 2022

1 Problem Set 5

```
[5]: import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import pandas as pd
from scipy.stats import norm
from sklearn.linear_model import LinearRegression
```

```
new_line = '\n'
pd.options.mode.chained_assignment = None
import warnings
warnings.filterwarnings('ignore')

sns.set_style('darkgrid')
sns.set(font_scale=1.5, rc={'text.usetex' : True,})
%config InlineBackend.figure_format='retina'
```

```
[15]: # simulate stock prices
def simulate_gbm(s_0, mu, sigma, n_sims, T, antithetic_var=False):

    """
    simulation of stock prices with antithetic variables to reduce variance
    """

    dt = 1/np.sqrt(n_sims)
    step = int(T/dt)

    # Antithetic variables
    if antithetic_var:
        dW_ant = np.random.normal(scale = np.sqrt(dt),
                                   size=(int(n_sims/2), step + 1))
        dW = np.concatenate((dW_ant, -dW_ant), axis=0)
    else:
        dW = np.random.normal(scale = np.sqrt(dt),
                               size=(n_sims, step + 1))
```

```

St = np.zeros((n_sims, step + 1))
St[:, 0] = s_0

for i in range (1, step + 1):
    St[:, i] = St[:, i - 1]*np.exp((mu - 1/2*np.power(sigma, 2))*dt +
↪sigma*dW[:, i - 1])

return St

```

```

[16]: # Polynomials function
def laguerre(x,k):
    x = np.array(x)
    n = len(x)

    s1 = np.exp(-x/2)
    s2 = np.exp(-x/2)*(1-x)
    s3 = np.exp(-x/2)*(1-2*x+x**2/2)
    s4 = np.exp(-x/2)*(1-3*x+3*x**2/2-x**3/6)

    if k==1: return s1
    elif k==2: return np.array([s1,s2]).reshape(2,n)
    elif k==3: return np.array([s1,s2,s3]).reshape(3,n)
    elif k==4: return np.array([s1,s2,s3,s4]).reshape(4,n)

def hermite(x,k):
    x = np.array(x)
    n = len(x)

    s1 = [1] * n
    s2 = 2*x
    s3 = 4*x**2-2
    s4 = 8*x**3-12*x

    if k==1: return s1
    elif k==2: return np.array([s1,s2]).reshape(2,n)
    elif k==3: return np.array([s1,s2,s3]).reshape(3,n)
    elif k==4: return np.array([s1,s2,s3,s4]).reshape(4,n)

def monomials(x,k):
    x = np.array(x)
    n = len(x)

    s1 = [1] * n
    s2 = x
    s3 = x**2
    s4 = x**3

```

```

    if k==1: return s1
    elif k==2: return np.array([s1,s2]).reshape(2,n)
    elif k==3: return np.array([s1,s2,s3]).reshape(3,n)
    elif k==4: return np.array([s1,s2,s3,s4]).reshape(4,n)

def basis_function(x,k,method):
    if method.lower() == 'laguerre':
        return laguerre(x,k)
    elif method.lower() == 'hermite':
        return hermite(x,k)
    elif method.lower() == 'monomials':
        return monomials(x,k)

```

```

[67]: def LCSM(s0,strike,T,r,sigma,path,k,method):
    """
    Least-Square Monte Carlo Function
    Method: function of choosing polynomials in
    → ['laguerre', 'hermite', 'monomials']
    """
    dt = 1/np.sqrt(path)
    step = int(T/dt)
    disc = np.exp(-r*dt)

    # scale stock price by strike price
    St =
    → simulate_gbm(s_0=s0,mu=r,sigma=sigma,n_sims=path,T=T,antithetic_var=True)/
    → strike

    # cashflow matrix
    payoff_matrix = np.maximum(1 - St, np.zeros_like(St))
    # continuation matrix
    value_matrix = np.zeros_like(payoff_matrix)
    value_matrix[:, -1] = payoff_matrix[:, -1]

    for i in range(step - 1, 0 , -1):

        X = St[:,i]
        y = value_matrix[:,i+1]*disc

        X = basis_function(X,k,method)
        A = np.dot(X,X.T)
        b = np.dot(X,y)

        # find coefficients
        coef = np.dot(np.linalg.inv(A), b)
        # calculate continuation value with coefficients
        continuation_value = np.dot(X.T,coef)

```

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        value_matrix[:, i] = np.where(
            payoff_matrix[:, i] > continuation_value,
            payoff_matrix[:, i],
            value_matrix[:, i + 1] * disc
        )

    option = np.mean(value_matrix[:, 1] * disc) * strike

    return option

# LCSM(s0=40, r=0.06, sigma=0.2, path=10, T=2, strike=40, k=4, method="monomials")

```

```

[37]: s0 = 40
      sigma = 0.2
      rf = 0.06
      N = 100000
      strike = 40
      T = [0.5, 1, 2]
      k = [2, 3, 4]

```

1.0.1 (a) Laguerre Polynomials

```

[38]: a_2 = []
      for t in T:
          price = _
          →LCSM(s0=s0, strike=strike, T=t, r=rf, sigma=sigma, path=N, k=2, method='laguerre')
          a_2.append(price)
      a_3 = []
      for t in T:
          price = _
          →LCSM(s0=s0, strike=strike, T=t, r=rf, sigma=sigma, path=N, k=3, method='laguerre')
          a_3.append(price)
      a_4 = []
      for t in T:
          price = _
          →LCSM(s0=s0, strike=strike, T=t, r=rf, sigma=sigma, path=N, k=4, method='laguerre')
          a_4.append(price)

```

```

[39]: answer_a = pd.DataFrame(list(zip(a_2, a_3, a_4)))
      answer_a.index = ['T=0.5', 'T=1', 'T=2']
      answer_a.columns = ['K=2', 'K=3', 'K=4']
      answer_a

```

```

[39]:           K=2      K=3      K=4
T=0.5  1.585069  1.702291  1.723197
T=1     2.018832  2.169624  2.216167

```

T=2 2.485110 2.643543 2.750408

1.0.2 (b) Hermite Polynomials

```
[40]: b_2 = []
      for t in T:
          price = _
          ↪LCSM(s0=s0,strike=strike,T=t,r=rf,sigma=sigma,path=N,k=2,method='hermite')
          b_2.append(price)
      b_3 = []
      for t in T:
          price = _
          ↪LCSM(s0=s0,strike=strike,T=t,r=rf,sigma=sigma,path=N,k=3,method='hermite')
          b_3.append(price)
      b_4 = []
      for t in T:
          price = _
          ↪LCSM(s0=s0,strike=strike,T=t,r=rf,sigma=sigma,path=N,k=4,method='hermite')
          b_4.append(price)
```

```
[41]: answer_b = pd.DataFrame(list(zip(b_2,b_3,b_4)))
      answer_b.index=['T=0.5','T=1','T=2']
      answer_b.columns = ['K=2','K=3','K=4']
      answer_b
```

```
[41]:
```

	K=2	K=3	K=4
T=0.5	1.572054	1.681312	1.726638
T=1	1.995157	2.120473	2.220375
T=2	2.448517	2.577618	2.729778

1.0.3 (c) Simple Monomials

```
[42]: c_2 = []
      for t in T:
          price = _
          ↪LCSM(s0=s0,strike=strike,T=t,r=rf,sigma=sigma,path=N,k=2,method='monomials')
          c_2.append(price)
      c_3 = []
      for t in T:
          price = _
          ↪LCSM(s0=s0,strike=strike,T=t,r=rf,sigma=sigma,path=N,k=3,method='monomials')
          c_3.append(price)
      c_4 = []
      for t in T:
          price = _
          ↪LCSM(s0=s0,strike=strike,T=t,r=rf,sigma=sigma,path=N,k=4,method='monomials')
          c_4.append(price)
```

```
[43]: answer_c = pd.DataFrame(list(zip(c_2,c_3,c_4)))
      answer_c.index=['T=0.5','T=1','T=2']
      answer_c.columns = ['K=2','K=3','K=4']
      answer_c
```

```
[43]:
```

	K=2	K=3	K=4
T=0.5	1.586788	1.677614	1.726501
T=1	1.998103	2.127337	2.225514
T=2	2.469076	2.567499	2.716316

1.0.4 (d)

```
[54]: def parameters(r,sigma,N,T,formula):
      dt = T/N
      if formula == 'a':
          c = 0.5 * (np.exp(-r*dt) + np.exp( (r+sigma**2)*dt) )
          d = c - np.sqrt(c**2 - 1)
          u = 1/d
          p = (np.exp(r*dt) - d) / (u-d)
      elif formula == 'b':
          u = np.exp(r*dt)*(1+np.sqrt(np.exp(sigma**2*dt)-1))
          d = np.exp(r*dt)*(1-np.sqrt(np.exp(sigma**2*dt)-1))
          p = 0.5
      elif formula == 'c':
          u = np.exp((r-sigma**2/2)*dt + sigma*np.sqrt(dt))
          d = np.exp((r-sigma**2/2)*dt - sigma*np.sqrt(dt))
          p=0.5
      elif formula == 'd':
          u = np.exp(sigma*np.sqrt(dt))
          d = np.exp(-sigma*np.sqrt(dt))
          p = 1/2 + 1/2*((r-sigma**2/2)*np.sqrt(dt))/sigma
```

```

    return u, d, p, N

def binoAmer_fast(s0,N,k,r,T,sigma,option_type,formula):
    dt = T/N
    params = parameters(r,sigma,N,T,formula)
    u = params[0]
    d = params[1]
    p = params[2]
    disc = np.exp(-r*dt)

    S = s0* u**np.arange(0,N+1,1)*d**np.arange(N,-1,-1)      # stock prices at
    ↪ last step

    if option_type.lower()=='c':
        C = np.maximum(0,S - k)
    else:
        C = np.maximum(0,k - S)

    for i in np.arange(N-1,-1,-1):
        S = s0* u**np.arange(0,i+1,1)*d**np.arange(i,-1,-1)

        C[:i+1] = disc * (p * C[1:i+2] + (1-p)* C[0:i+1])

        C = C[:-1]

        if option_type.lower()=='c':
            C = np.maximum(C,S - k)
        else:
            C = np.maximum(C,k - S)

    return C[0]

```

```

[64]: # compare to binomial prices
binoPrice = []
for t in T:
    price = ↪
    ↪ binoAmer_fast(s0=s0,N=158,k=strike,r=rf,T=t,sigma=sigma,option_type='put',formula='c')
    binoPrice.append(price)

```

```

[65]: binop = pd.DataFrame(binoPrice,index=['T=0.5','T=1','T=2'],columns=['Price'])
binop

```

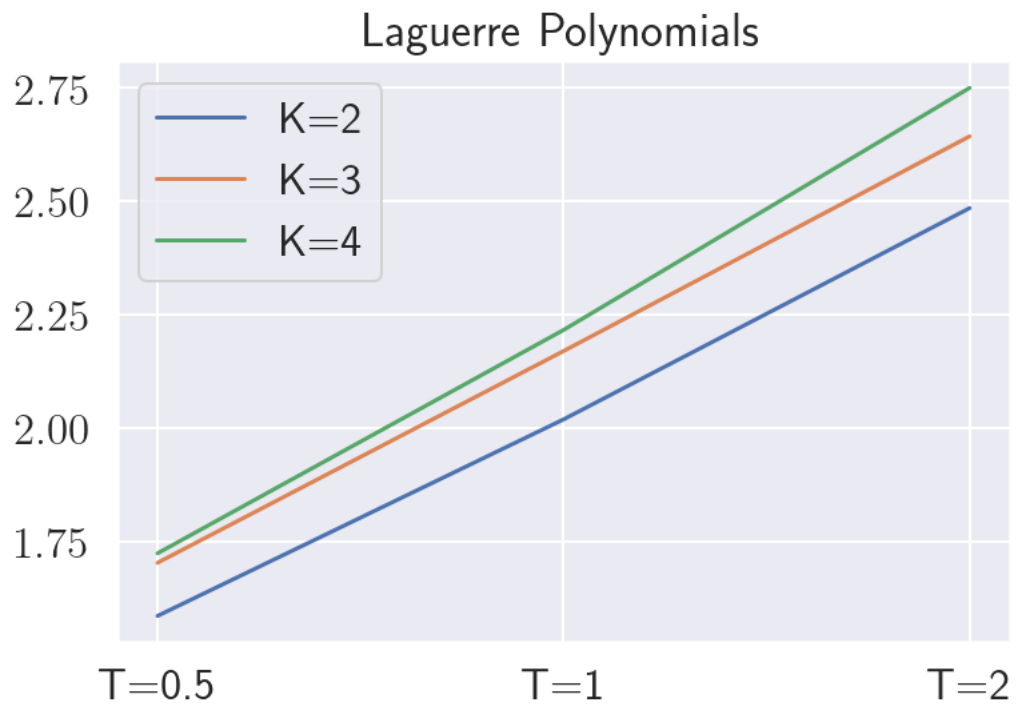
```

[65]:
      Price
T=0.5  1.797425
T=1    2.320763
T=2    2.890038

```

```
[45]: plt.figure()

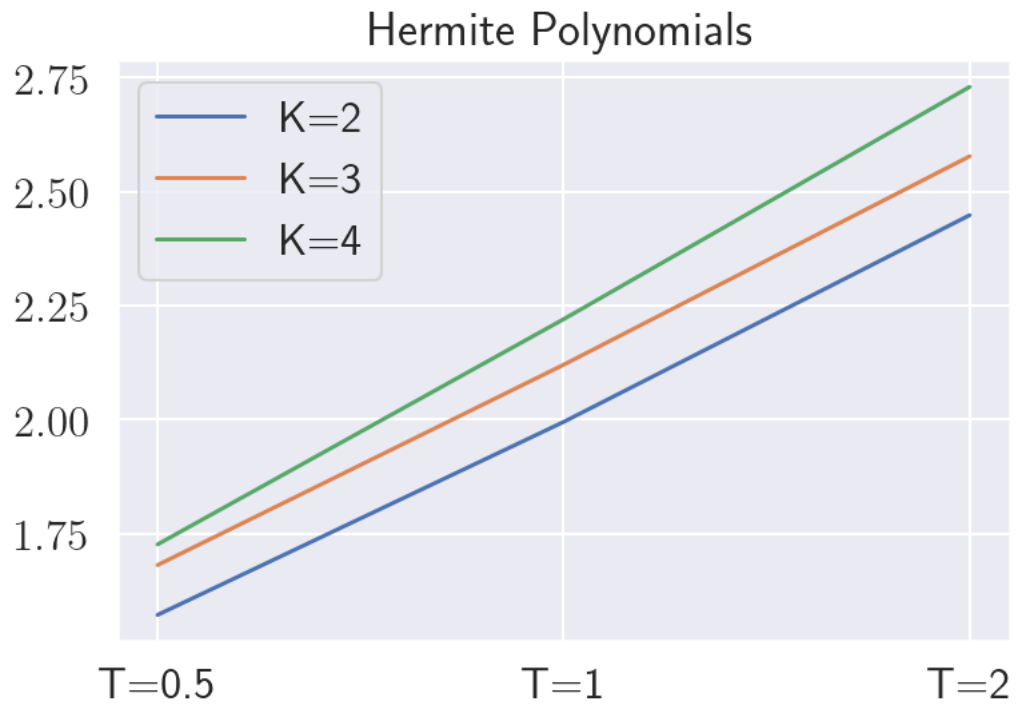
plt.plot(answer_a)
plt.legend(answer_a.columns)
plt.title('Laguerre Polynomials')
plt.show()
```



```
[46]: plt.figure()

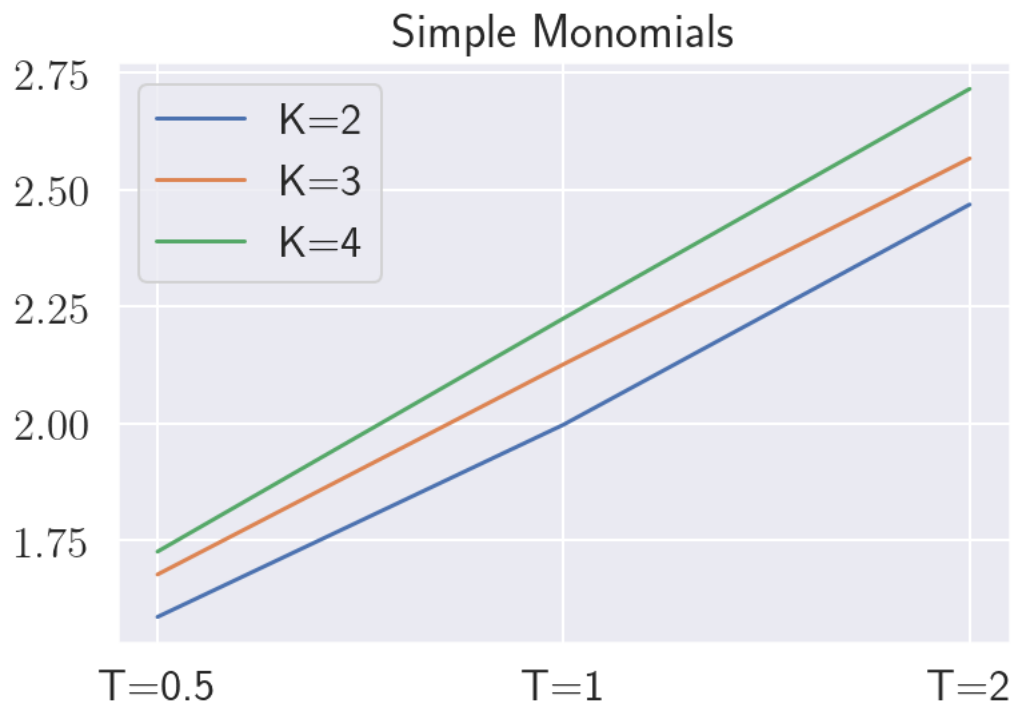
plt.plot(answer_b)
plt.legend(answer_b.columns)
plt.title('Hermite Polynomials')

plt.show()
```

```
[47]: plt.figure()

plt.plot(answer_c)
plt.legend(answer_c.columns)
plt.title('Simple Monomials')
plt.show()
```



```
[17]: # ## class material example
# k=97.5
# r =0.05
# dt = 3/3
# disc = np.exp(-r*dt)
# path = 10
# step = 3
# ss = np.array([[100,92.8,108.8,121.11],
#               [100,100.1,94.2,92.1],
#               [100,98.87,93.11,97.8],
#               [100,96.34,93.11,90.36],
#               [100,102.14,100.05,96.43],
#               [100,98.3,110.21,99.2],
#               [100,102.87,120.1,128.43],
#               [100,110.21,98.2,94.5],
#               [100,89.87,93.8,90],
#               [100,86.12,90.21,98.34]])
# # 0: do nothing, 1: exercise
# payoff = np.zeros_like(ss)
# payoff= np.maximum(k -ss,np.zeros_like(ss))
# index = np.zeros_like(ss)
# index[:,step] = np.where(k > ss[:,step],1,0)
# continue_matrix = np.zeros_like(payoff)
```

```

# continue_matrix[:,step] = payoff[:,step]

# for i in range(step-1,0,-1):

#     in_the_money_path = np.where(payoff[:,i]*disc > 0)[0]
#     X = ss[in_the_money_path,i]
#     y = continue_matrix[in_the_money_path,i+1]*disc

#     X = basis_function(X,3,method='hermite')
#     A = np.dot(X,X.T)
#     b = np.dot(X,y)
#     coef = np.dot(np.linalg.inv(A), b)

#     continue_matrix[in_the_money_path,i] = np.dot(X.T,coef)
#     exercise_value = payoff[in_the_money_path,i]

#     index[in_the_money_path,i] = np.
#     ↪where(exercise_value>continue_matrix[in_the_money_path,i],1,0)
#     continue_matrix[in_the_money_path,i] = np.
#     ↪maximum(exercise_value,continue_matrix[in_the_money_path,i])

# #     for j in range(path):
# #         if index[j,i] ==1:
# #             index[j,i+1:] = 0

#     continue_matrix[in_the_money_path,i+1] *= index[in_the_money_path,i+1]

#     continue_matrix[:,i] = np.maximum(disc* continue_matrix[
#     ↪,i+1],continue_matrix[:,i])

#     for j in range(path):
#         if index[j,i] ==1:
#             index[j,i+1:] = 0

# optimal_matrix = np.ones_like(payoff) * index * payoff
# dt_array = np.exp(-r*dt*np.arange(1,step+1,1))
# avg_payoff = np.sum(optimal_matrix,axis=0)[1:] # exclude the initial price
# option_price = np.dot(avg_payoff,dt_array)/path

# # print(f"option price: {option_price} {new_line} \
# #         index matrix : {new_line} {index} {new_line} \
# #         optimal exercise decision{new_line} {optimal_matrix}")

```

[25]: # def LCSM(s0,strike,T,r,sigma,path,k,method):

```

#     dt = 1/np.sqrt(path)

```

```

#     step = int(T/dt)
#     disc = np.exp(-r*dt)

#     St = □
#     ↳simulate_gbm(s_0=s0,mu=r,sigma=sigma,n_sims=path,T=T,antithetic_var=True)/
#     ↳strike

#     payoff_matrix = np.maximum(1 - St, np.zeros_like(St))

#     index_matrix = np.zeros_like(payoff_matrix)
#     index_matrix[:, -1] = np.where(payoff_matrix[:, -1]>0, 1, 0)

#     value_matrix = np.zeros_like(payoff_matrix)
#     value_matrix[:, -1] = payoff_matrix[:, -1]

#     for i in range(step - 1, 0, -1):
#         in_the_money_path = np.where(payoff_matrix[:, i]>0)[0]

#         X = St[in_the_money_path, i]
#         y = value_matrix[in_the_money_path, i+1]*disc

#         X = basis_function(X, k, method)
#         A = np.dot(X, X.T)
#         b = np.dot(X, y)
#         # find betas
#         coef = np.dot(np.linalg.inv(A), b)

#         continuation_value = np.dot(X.T, coef)

#         value_matrix[in_the_money_path, i] = np.where(
#             payoff_matrix[in_the_money_path, i] > continuation_value,
#             payoff_matrix[in_the_money_path, i],
#             value_matrix[in_the_money_path, i + 1] * disc
#         )

#     option_premium = np.mean(value_matrix[:, 1] * disc)* strike

#     return option_premium

# LCSM(s0=40, r=0.06, sigma=0.2, path=10, T=2, strike=40, k=4, method="monomials")

```

```

[27]: # def LCSM(s0, strike, T, r, sigma, path, k, method):
#     """
#     Least-square Monte Carlo Simulation
#     method in ['laguerre', 'hermite', 'monomials']
#     """
#     # define dt = 1/sqrt(N)

```

```

#     dt = 1/np.sqrt(path)
#     steps = int(T/dt)
#     disc = np.exp(-r*dt)
#     # simulate stock prices
#     St = simulate_gbm(s0=s0, mu=r, sigma=sigma, path=path, T=T)/strike #
    ↳ scale by strike price

#     # initialized payoff matrix
#     payoff = np.zeros_like(St)
#     payoff = np.maximum(1-St, np.zeros_like(St))

#     # index matrix
#     # 1: exercise , 0: doing nothing
#     index = np.zeros_like(St)
#     index[:,steps] = np.where(1>St[:,steps],1,0)

#     continue_matrix = np.zeros_like(payoff)
#     continue_matrix[:,step] = payoff[:,step]

#     for i in range(step-1,0,-1):
#         # find the in the money path
#         in_the_money_path = np.where(payoff[:,i]*disc >0)[0]
#         X = St[in_the_money_path,i]
#         y = continue_matrix[in_the_money_path,i+1]*disc

#         X = basis_function(X,k,method)
#         A = np.dot(X,X.T)
#         b = np.dot(X,y)
#         # find betas
#         coef = np.dot(np.linalg.inv(A), b)

#         continue_matrix[in_the_money_path,i] = np.dot(X.T,coef)

#         exercise_value = payoff[in_the_money_path,i]

#         index[in_the_money_path,i] = np.
    ↳ where(exercise_value>continue_matrix[in_the_money_path,i],1,0)

#         continue_matrix[in_the_money_path,i] = np.
    ↳ maximum(exercise_value,continue_matrix[in_the_money_path,i])

#         continue_matrix[in_the_money_path,i+1] *= index[in_the_money_path,i+1]
#         continue_matrix[:,i] = np.maximum(disc* continue_matrix[:,
    ↳ ,i+1],continue_matrix[:,i])

#         # make sure one row has only one 1
#         for j in range(path):
#             if index[j,i] ==1:

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```

#             index[j,i+1:] = 0

#     optimal_matrix = strike* np.ones_like(payload) * index * payoff
#     dt_array = np.exp(-r*dt*np.arange(1,steps+1,1))
#     avg_payoff = np.sum(optimal_matrix,axis=0)[1:] # exclude the initial price

#     # find the option price
#     option_price = np.dot(avg_payoff,dt_array)/path

#     return option_price

# # LCSM(s0=40,r=0.06, sigma=0.2,
# ↪ path=100000,T=2,strike=40,k=4,method="monomials")

```

[]: