

Question 6)

(this is my old Q6 code from last week. My new code will be submitted in HW8)

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1. # importing the necessary packages
2. import numpy as np
3. from numpy import sqrt, exp, log, mean
4. from scipy.stats import norm
5. from scipy import optimize
6. import matplotlib.pyplot as plt
7.
8. #####
9.
10. # increment step method for Brownian Motion
11. def SimBMStep(T, N):
12.     # initialize W brownian motion array
13.     W = np.zeros(int(N))
14.     W = np.append(W, 0)    # initial 0 value
15.
16.     # loop
17.     num = 0
18.     while num < N:
19.         W[num + 1] = W[num] + sqrt(T/N)*np.random.standard_normal()
20.         num += 1
21.
22.     return W
23.
24. # initial vars
25. s = 139.    # retrieved from APPL stock
26. #s = 100.
27. y = 0.08
28. lam = 3
29. kappa = 0.1
30. rho = -0.8
31. xi = 0.25
32.
33. # initial vars array
34. x0 = np.array([lam, kappa, xi, y, rho])
35.
36. # number retrieved from 30yr treasury bond yield curve rate
37. r = 0.0320
38.
39. # day form
40. T = 60/365.
41. runs = 1000
42.
43. # this is all XDATA
44. # these are the K strike CALL options that have trading volume of more than 10
45. CK = np.array([50, 130, 135, 140, 145, 150, 155, 160, 165, 170, 180])
46. # these are the K strike PUT options that have trading volume of more than 10
47. PK = np.array([70, 75, 80, 100, 105, 110, 115, 120, 125, 130, 135, 140, 145, 150])
48.
49. #CK_thetaPrices = np.zeros(0)
50. CK_actualPrices = np.array([mean([96.85,97.60]),
51.                             mean([10.55,10.60]),
52.                             mean([6.80,6.90]),
53.                             mean([3.90,4.00]),
54.                             mean([2.04,2.06]),
55.                             mean([0.95,0.97]),
56.                             mean([0.43,0.44]),
57.                             mean([0.20,0.21]),
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58.             mean([0.09,0.10]),
59.             mean([0.04,0.05]),
60.             mean([0.04,0.05]))]
61.
62. #PK_thetaPrices = np.zeros(0)
63. PK_actualPrices = np.array([mean([0, 0.03]),
64.                             mean([0.02,0.04]),
65.                             mean([0.01,0.05]),
66.                             mean([0.06,0.07]),
67.                             mean([0.11,0.12]),
68.                             mean([0.16,0.17]),
69.                             mean([0.27,0.28]),
70.                             mean([0.45,0.47]),
71.                             mean([0.81,0.84]),
72.                             mean([1.55,1.57]),
73.                             mean([2.89,2.92]),
74.                             mean([5.10,5.15]),
75.                             mean([8.20,8.30]),
76.                             mean([11.90,12.25])])
77.
78. print("K strike values for Call options (these have volume traded > 10):")
79. print(CK)
80. print("Call option actual prices (ordered with their respective K's): ")
81. print(CK_actualPrices)
82.
83. print("K strike values for Put options:")
84. print(PK)
85. print("Put option actual prices are: ")
86. print(PK_actualPrices)
87.
88. # payoff function for European Call option
89. def payoffEC(ST, K):
90.     return max(ST - K, 0)
91.
92. # payoff function for European Put option
93. def payoffEP(ST, K):
94.     return max(K - ST, 0)
95.
96. # big MAIN FUNCTION that has all parameters for optimization
97. # and outputs the estimated model prices
98. # xdata is the K value
99. def PricingUsingModelEC(K, Plam, Pkappa, Pxi, Py, Prho):
100.     #lam, kappa, xi, y, rho, = params
101.     #errVal = abs(ydata - CTheta)
102.
103.     # K loop?
104.     CK_estimatedPrices = np.zeros(0)
105.     i = 0
106.     while i < K.size:
107.         # array holding payoffs for price calculation
108.         payoffsArray = np.zeros(0)
109.         num = 0
110.         while num < runs:
111.             # get W1 and W~ Brownian Motions, in order to get the W2 BM
112.             # note: need to use rho correlation equation
113.             W1 = SimBMStep(T, 500.)
114.             Wtilda = SimBMStep(T, 500.)
115.             W2 = Prho*W1 + sqrt(1 - Prho**2)*Wtilda
116.
117.             # simulating Yt values and St values
118.             temp = 0
119.             Y = np.zeros(0)
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120.         # getting initial Y_0, which is little y = -1
121.         Y = np.append(Y, Py)
122.         St = np.zeros(0)
123.         St = np.append(St, s)
124.         # loop for Yt and St simulation
125.         # NOTE, Yt follows CIR process form
126.         while temp < 500:
127.             Y = np.append(Y, max(Y[temp]
128.                                   + Plam*(Pkappa - Y[temp])*(T/500.)
129.                                   + Pxi*sqrt(Y[temp])*(W2[temp+1] - W2[temp]), 0))
130.             St = np.append(St, St[temp]
131.                            + r*St[temp]*(T/500.)
132.                            + sqrt(Y[temp])*St[temp]*(W1[temp+1] - W1[temp])
133.                            + (0.5 * sqrt(Y[temp])*sqrt(Y[temp])*St[temp]*(W1[temp+1] - W1[tem
134.                                p])**2 - (T/500.)))
135.             temp += 1
136.         #
137.         ST = St[St.size - 1]
138.
139.         # calculating payoff
140.         #print("K size is: ")
141.         #print(K.size)
142.         payment = payoffEC(ST, K[i])
143.         payoffsArray = np.append(payoffsArray, payment)
144.
145.         num += 1
146.         # calculating price
147.         EstimatedPrice = exp(-r*T)*np.mean(payoffsArray)
148.         print("estimated price is: " + str(EstimatedPrice))
149.         CK_estimatedPrices = np.append(CK_estimatedPrices, EstimatedPrice)
150.         i += 1
151.         return CK_estimatedPrices
152.
153.
154.         # finally, we do curve_fit optimization
155.         popt, pcov = optimize.curve_fit(PricingUsingModelEC, CK, CK_actualPrices, x0)
156.         print(popt)
```