In [104]:

```
import numpy as np
import math
import scipy as scp
import scipy.stats as ss
import matplotlib.pyplot as plt
```

In [105]:

```
class Option param():
    Option class wants the option parameters:
    S0 = current stock price
    K = Strike price
    T = time to maturity
    exercise = European or American
    def init (self, S0=100, K=100, T=1, payoff="call", exercise="Europ")
ean"):
        self.S0 = S0
        self.K = K
        self.T = T
        self.Delta = None
        self.Gamma = None
        self.Theta = None
        self.Vega = None
        self.Rho = None
        if (exercise=="European" or exercise=="American"):
            self.exercise = exercise
        else:
            raise ValueError("invalid type. Set 'European' or 'American'"
)
        if (payoff=="call" or payoff=="put"):
            self.payoff = payoff
        else:
            raise ValueError("invalid type. Set 'call' or 'put'")
```

In [106]:

```
class Diffusion process():
    Class for the diffusion process:
    r = continuously compounded risk-free interest rate (% p.a.)
    q = continuously compounded dividend yield (% p.a.)
    sig = volatility (% p.a.)
    def init (self, r=0.1, q=0.0, sig=0.2):
        self.r = r
        if (q<0):
            raise ValueError("Dividend must be positive")
        else:
            self.q = q
        if (sig<=0):
            raise ValueError("sig must be positive")
        else:
            self.sig = sig
    def exp RV(self, S0, T, N):
        W = ss.norm.rvs((self.r-0.5*self.sig**2)*T, np.sqrt(T)*self.sig
, N)
        S T = S0 * np.exp(W)
        return S_T
```

In [107]:

```
class Option pricer():
    def __init__(self, Option_info, Process_info ):
        Process info: of type Diffusion process. It contains (r, mu, sig)
i.e. interest rate, drift coefficient, diffusion coefficient
        Option info: of type Option param. It contains (SO,K,T) i.e. cur
rent price, strike, maturity in years
        self.r = Process info.r
                                         # interest rate
        self.q = Process info.q
                                        # diffusion coefficient
        self.sig = Process_info.sig
        self.S0 = Option_info.S0
                                          # current price
        self.K = Option_info.K
self.T = Option_info.T
                                          # strike
                                          # maturity in years
        self.exercise = Option info.exercise
        self.payoff = Option info.payoff
    def payoff f(self, S):
        if self.payoff == "call":
            Payoff = np.maximum( S - self.K, 0 )
        elif self.payoff == "put":
            Payoff = np.maximum( self.K - S, 0 )
        return Payoff
    def Binomial Tree(self, N=10000):
        dT = float(self.T) / N
                                                            # Delta t
                                                            # up factor
        u = np.exp(self.sig * np.sqrt(dT))
        d = 1.0 / u
                                                       # down factor
        V = np.zeros(N+1)
                                                       # initialize the pr
ice vector
        S T = np.array( [(self.S0 * u**j * d**(N - j)) for j in range(N + j)) for j in range(N + j)
1)] ) # price S T at time T
        a = np.exp((self.r - self.q ) * dT) # risk free compound retur
n
        p = (a - d)/(u - d) # risk neutral up probability
        q = 1.0 - p
                             # risk neutral down probability
        if self.payoff == "call":
            V[:] = np.maximum(S T-self.K, 0.0)
        elif self.payoff =="put":
            V[:] = np.maximum(self.K-S T, 0.0)
        else: raise ValueError("invalid type. Set 'call' or 'put'")
        if self.exercise == "American":
            for i in range(N-1, -1, -1):
                V[:-1] = np.exp(-self.r*dT) * (p * V[1:] + q * V[:-1])
# the price vector is overwritten at each step
```

```
S T = S T * u
                                                # it is a tricky way to
 obtain the price at the previous time step
                if self.payoff=="call":
                    V = np.maximum( V, S T-self.K )
                elif self.payoff=="put":
                    V = np.maximum( V, self.K-S T )
            return ("American", V[0])
        elif self.exercise == "European":
            for i in range (N-1, -1, -1):
                V[:-1] = np.exp(-self.r*dT) * (p * V[1:] + q * V[:-1])
# the price vector is overwritten at each step
            return ("European", V[0])
        else: raise ValueError("invalid type. Set 'American' or 'Europea
n'")
    def Monte Carlo Simulation(self, N=10000, paths=10000, order=2):
        Monte Carlo Simulation for pricing American options
        N = number of time steps
        paths = number of generated paths
        order = order of the polynomial for the regression
        11 11 11
                                  # time interval
        dt = self.T/(N-1)
        df = np.exp(-self.r * dt) # discount factor per time time interv
a 1
        X0 = np.zeros((paths, 1))
        increments = ss.norm.rvs(loc=(self.r - self.q - self.sig**2/2)*dt
, scale=np.sqrt(dt)*self.sig, size=(paths,N-1))
        X = np.concatenate((X0,increments), axis=1).cumsum(1)
        S = self.S0 * np.exp(X)
        if self.exercise == "American":
            if self.payoff == "put":
                H = np.maximum(self.K - S, 0) # intrinsic values for pu
t option
                                               # value matrix
                V = np.zeros like(H)
                V[:,-1] = H[:,-1]
                # Valuation by Monte Carlo Simulation
                for t in range(N-2, 0, -1):
                    good paths = H[:,t] > 0
                    rg = np.polyfit( S[good paths, t], V[good paths, t+1]
* df, order)
               # polynomial regression
                    C = np.polyval( rg, S[good_paths,t] )
# evaluation of regression
                    exercise = np.zeros( len(good paths), dtype=bool)
                    exercise[good_paths] = H[good_paths,t] > C
```

```
V[exercise,t] = H[exercise,t]
                    V[exercise, t+1:] = 0
                    discount path = (V[:,t] == 0)
                    V[discount path,t] = V[discount path,t+1] * df
                V0 = np.mean(V[:,1]) * df #
                return ("American" , V0)
            elif self.payoff == "call":
                H = np.maximum(S - self.K, 0) # intrinsic values for pu
t option
                V = np.zeros like(H)
                                               # value matrix
                V[:,-1] = H[:,-1]
                # Valuation by Monte Carlo Simulation
                for t in range (N-2, 0, -1):
                    good paths = H[:,t] > 0
                    if t==N-2: type(good paths)
                    rg = np.polyfit( S[good paths, t], V[good paths, t+1]
* df, order) # polynomial regression
                    C = np.polyval( rg, S[good paths,t] )
# evaluation of regression
                    exercise = np.zeros( len(good paths), dtype=bool)
                    exercise[good paths] = H[good paths,t] > C
                    V[exercise,t] = H[exercise,t]
                    V[exercise, t+1:] = 0
                    discount_path = (V[:,t] == 0)
                    V[discount path,t] = V[discount path,t+1] * df
                V0 = np.mean(V[:,1]) * df #
                return("American" , V0)
            else: raise ValueError("invalid type. Set 'call' or 'put'")
        elif self.exercise == "European":
            if self.payoff == "put":
                H = np.maximum(self.K - S, 0) # intrinsic values for pu
t option
                                               # value matrix
                V = np.zeros like(H)
                V[:,-1] = H[:,-1]
                # Valuation by Monte Carlo Simulation
                for t in range(N-2, 0, -1):
                    V[:,t] = V[:,t+1] * df
                V0 = np.mean(V[:,1]) * df #
                return("European" , V0)
            elif self.payoff == "call":
                H = np.maximum(S - self.K, 0) # intrinsic values for pu
t option
                V = np.zeros like(H)
                                               # value matrix
                V[:,-1] = H[:,-1]
```

```
# Valuation by Monte Carlo Simulation
                for t in range (N-2, 0, -1):
                    V[:,t] = V[:,t+1] * df
                V0 = np.mean(V[:,1]) * df #
                return("European" , V0)
            else: raise ValueError("invalid type. Set 'call' or 'put'")
        else: raise ValueError("invalid type. Set 'American' or 'Europea
n'")
   def Black Scholes(self):
        """ Black Scholes closed formula:
            payoff: call or put.
            S0: float.
                         initial stock/index level.
            K: float strike price.
            T: float maturity (in year fractions).
            r: float constant risk-free short rate.
            sigma: volatility factor in diffusion term. """
        d1 = (np.log(self.S0/self.K) + (self.r - self.g + self.sig**2 / 2
) * self.T) / (self.sig * np.sqrt(self.T))
        d2 = (np.log(self.S0/self.K) + (self.r - self.g - self.sig**2 / 2
) * self.T) / (self.sig * np.sgrt(self.T))
        if self.payoff=='call':
            return ("European", self.S0 * np.exp(-self.q * self.T) * ss.n
orm.cdf( d1 ) - self.K * np.exp(-self.r * self.T) * ss.norm.cdf( d2 ))
        elif self.payoff=='put':
            return ("European", self.K * np.exp(-self.r * self.T) * ss.no
rm.cdf( -d2 ) - self.S0 * np.exp(-self.q * self.T) * ss.norm.cdf( -d1 ))
            raise ValueError("invalid type. Set 'call' or 'put'")
   def plot(self):
        if self.exercise == "European":
            fig, axs = plt.subplots(2, figsize=(10,10))
            fig.tight layout(pad = 7)
            K = np.arange(self.K*0.1, self.K*1.9, self.K*0.1)
            if self.payoff == "call":
                payoff = np.array(np.maximum(self.S0-K, 0))
            if self.payoff == "put":
                payoff = np.array(np.maximum(K-self.S0, 0))
            axs[0].plot(K, payoff, label = "exercise price at S0")
            price = np.array([])
            for k in np.arange(self.K*0.1,self.K*1.9, self.K*0.1):
                opt param = Option param(S0=self.S0, K=k, T=self.T, exerc
```

```
ise=self.exercise, payoff=self.payoff)
                diff param = Diffusion process(r=self.r, sig=self.sig)
                option = Option pricer(opt param, diff param)
                price = np.append(price, option.Black Scholes()[1])
            axs[0].plot(K, price, label = "Black Scholes Pricing")
            axs[0].set title("Spot price = " + str(self.S0))
            axs[0].set(xlabel = "Strike price (K)", ylabel = "Option Valu
e")
            axs[0].legend()
            S0 = np.arange(self.S0*0.1, self.S0*1.9, self.S0*0.1)
            if self.payoff == "call":
                payoff = np.array(np.maximum(S0-self.K, 0))
            if self.payoff == "put":
                payoff = np.array(np.maximum(self.K-S0, 0))
            axs[1].plot(S0, payoff, label = "exercise price at S0")
            price = np.array([])
            for S in np.arange(self.S0*0.1,self.S0*1.9, self.S0*0.1):
                opt param = Option param(S0=S, K=self.K, T=self.T, exerci
se=self.exercise, payoff=self.payoff)
                diff param = Diffusion process(r=self.r, sig=self.sig)
                option = Option pricer(opt param, diff param)
                price = np.append(price, option.Black Scholes()[1])
            axs[1].plot(S0, price, label = "Black Scholes Pricing")
            axs[1].set title("Strike price = " + str(self.S0))
            axs[1].set(xlabel = "Spot price (S0)", ylabel = "Option Valu
e")
            axs[1].legend()
        elif self.exercise == "American":
            fig, axs = plt.subplots(2, figsize=(10,10))
            fig.tight layout(pad = 7)
            K = np.arange(self.K*0.1, self.K*1.9, self.K*0.1)
            if self.payoff == "call":
                payoff = np.array(np.maximum(self.S0-K, 0))
            if self.payoff == "put":
                payoff = np.array(np.maximum(K-self.S0, 0))
            axs[0].plot(K, payoff, label = "exercise price at S0")
            price = np.array([])
            for k in np.arange(self.K*0.1,self.K*1.9, self.K*0.1):
                opt_param = Option_param(S0=self.S0, K=k, T=self.T, exerc
ise=self.exercise, payoff=self.payoff)
                diff param = Diffusion process(r=self.r, sig=self.sig)
                option = Option_pricer(opt_param, diff_param)
                price = np.append(price, option.Binomial Tree()[1])
            axs[0].plot(K, price, label = "Binomial Tree Pricing")
```

```
axs[0].set title("Spot price = " + str(self.S0))
            axs[0].set(xlabel = "Strike price (K)", ylabel = "Option Valu
e")
            axs[0].legend()
            S0 = np.arange(self.S0*0.1, self.S0*1.9, self.S0*0.1)
            if self.payoff == "call":
                payoff = np.array(np.maximum(S0-self.K, 0))
            if self.payoff == "put":
                payoff = np.array(np.maximum(self.K-S0, 0))
            axs[1].plot(S0, payoff, label = "exercise price at S0")
            price = np.array([])
            for S in np.arange(self.S0*0.1, self.S0*1.9, self.S0*0.1):
                opt param = Option param(S0=S, K=self.K, T=self.T, exerci
se=self.exercise, payoff=self.payoff)
                diff param = Diffusion process(r=self.r, sig=self.sig)
                option = Option pricer(opt param, diff param)
                price = np.append(price, option.Binomial_Tree()[1])
            axs[1].plot(S0, price, label = "Binomial Tree Pricing")
            axs[1].set_title("Strike price = " + str(self.S0))
            axs[1].set(xlabel = "Spot price (S0)", ylabel = "Option Valu
e")
            axs[1].legend()
        else: raise ValueError("invalid type. Set 'American' or 'Europea
n'")
    def greeks(self):
        d1 = (np.log(self.S0/self.K) + (self.r - self.q + self.sig**2 / 2
) * self.T) / (self.sig * np.sqrt(self.T))
        d2 = (np.log(self.S0/self.K) + (self.r - self.q - self.sig**2 / 2
) * self.T) / (self.sig * np.sqrt(self.T))
        if self.payoff == "call":
            self.Delta = np.exp(-self.q * self.T) * ss.norm.cdf(d1)
            self.Gamma = (np.exp(-self.q * self.T - (d1**2)/2))/(self.S0)
* self.sig * np.sqrt(self.T) * np.sqrt(2 * math.pi))
            self.Theta = (1/(self.T*365)) * (-((self.S0 * self.sig * np.e))
xp(-self.q * self.T - (d1**2)/2))/(2 * np.sqrt(2 * self.T * math.pi))) -
self.r * self.K * np.exp(-self.r * self.T) * ss.norm.cdf(d2) + self.q * s
elf.S0 * np.exp(-self.q * self.T) * ss.norm.cdf(d1))
            self.Vega = (1/100) * self.S0 * np.exp(-self.q * self.T - (d)
1**2)/2) * np.sqrt(self.T) * (1/np.sqrt(2 * math.pi))
            self.Rho = (1/100) * self.K * self.T * np.exp(-self.r * sel
f.T) * ss.norm.cdf(d2)
        elif self.payoff == "put":
            self.Delta = np.exp(-self.q * self.T) * (ss.norm.cdf(d1)-1)
```

```
self.Gamma = (np.exp(-self.q * self.T - (d1**2)/2))/(self.S0
* self.sig * np.sqrt(self.T) * np.sqrt(2 * math.pi))
            self.Theta = (1/(self.T*365)) * (-((self.S0 * self.sig * np.e))
xp(-self.q * self.T - (d1**2)/2))/(2 * np.sqrt(2 * self.T * math.pi))) +
self.r * self.K * np.exp(-self.r * self.T) * ss.norm.cdf(-d2) - self.q *
self.S0 * np.exp(-self.q * self.T) * ss.norm.cdf(-d1))
            self.Vega = (1/100) * self.S0 * np.exp(-self.q * self.T - (d)
1**2)/2) * np.sqrt(self.T) * (1/np.sqrt(2 * math.pi))
            self.Rho = (-1/100) * self.K * self.T * np.exp(-self.r * se
lf.T) * ss.norm.cdf(-d2)
        else: raise ValueError("invalid type. Set 'call' or 'put'")
        print("Delta: ", self.Delta)
        print("Gamma: ", self.Gamma)
        print("Theta: ", self.Theta)
        print("Vega: " , self.Vega )
print("Rho: " , self.Rho )
        print("\n")
```

In [108]:

```
def implied vol(payoff, option price, S0, K, T, r, q):
    # apply bisection method to get the implied volatility by solving the
BSM function
    precision = 0.00001
    upper vol = 500.0
    lower vol = 0.0001
    iteration = 0
    opt param = Option param(S0=S0, K=K, T=T, exercise="European", payoff
=payoff)
    diff param = Diffusion process(r=r, q=q, sig=250.0)
    option = Option pricer(opt param, diff param)
    while 1:
        iteration +=1
        mid\ vol = (upper\ vol + lower\ vol)/2.0
        option.sig = mid vol
        price = option.Black Scholes()[1]
        if payoff == "call":
            option.sig = lower vol
            lower price = option.Black Scholes()[1]
            if (lower_price - option_price) * (price - option price) > 0:
                lower_vol = mid vol
            else:
                upper vol = mid vol
            if abs(price - option price) < precision: break</pre>
            if iteration > 100: raise ValueError("Computational error occ
ured. ")
        elif payoff == "put":
            option.sig = upper vol
            upper price = option.Black Scholes()[1]
            if (upper price - option price) * (price - option price) > 0:
                upper vol = mid vol
            else:
                lower vol = mid vol
            if abs(price - option price) < precision: break</pre>
            if iteration > 100: raise ValueError("Computational error occ
ured. ")
    return mid vol
```

In [109]:

```
def greeks hedge (*option):
    if Delta is being hedged: number of options needed = Number of h
edged Greeks
    if Delta is not being hedged: number of options needed = Number of h
edged Greeks + 1
   num option = 0
    for opt in option:
        num option += 1
        print("Option", num_option,":")
        opt.greeks()
    number = ['first', 'second', 'thirth', 'fourth', 'fifth']
    num greeks = int(input("How many greeks are going to hedge: "))
    greeks = [np.array([],dtype = str)]
    for n in range(num greeks):
        greeks = np.append(greeks, input("The " + number[n] + " hedging g
reek: "))
    stock Delta = np.array([1])
    array Delta = np.append(stock Delta, [option[index].Delta for index i
n range(num option-1)])
    stock Gamma = np.array([0])
    array Gamma = np.append(stock Gamma, [option[index].Gamma for index i
n range(num option-1)])
    stock Theta = np.array([0])
    array Theta = np.append(stock Theta, [option[index].Theta for index i
n range(num option-1)])
    stock Vega = np.array([0])
    array Vega = np.append(stock Vega , [option[index].Vega for index i
n range(num option-1)])
    stock_Rho = np.array([0])
    array Rho
              = np.append(stock Rho , [option[index].Rho for index i
n range(num option-1)])
    matrix_Left = np.array([])
    matrix Right = np.array([])
    for g in greeks:
        if q == "Delta":
            matrix Left = np.append(matrix Left, array Delta)
            matrix Right = np.append(matrix Right, -option[-1].Delta)
        if g == "Gamma":
            matrix Left = np.append(matrix Left, array Gamma)
            matrix Right = np.append(matrix Right, -option[-1].Gamma)
        if q == "Theta":
            matrix Left = np.append(matrix_Left, array_Theta)
            matrix Right = np.append(matrix Right, -option[-1].Theta)
        if g == "Vega":
            matrix Left = np.append(matrix Left, array Vega)
            matrix Right = np.append(matrix Right, -option[-1].Vega)
```

```
if q == "Rho":
            matrix Left = np.append(matrix Left, array Rho)
            matrix Right = np.append(matrix Right, -option[-1].Rho)
    matrix Left.shape = (num greeks, num option)
    x = np.linalg.lstsq(matrix Left, matrix Right)
    result = np.append(x[0],1)
    for i in range(num option+1):
        if i==0:
            print("\nPortfolio: ")
            print("Stock : ", result[i])
        else:
            print("Option ",i,": ", result[i])
    matrix greeks = np.array([np.append(array Delta, option[-1].Delta), n
p.append(array Gamma, option[-1].Gamma), np.append(array Theta, option[-1
]. Theta), np.append(array Vega, option[-1]. Vega), np.append(array Rho, op
tion[-1].Rho)
    portfolio greeks = matrix greeks.dot(result)
    print("\nThe Portfolio's Greeks: ")
    print("Delta: ", portfolio_greeks[0])
    print("Gamma: ", portfolio_greeks[1])
print("Theta: ", portfolio_greeks[2])
    print("Vega : ", portfolio_greeks[3])
    print("Rho : ", portfolio_greeks[4])
```

In [110]:

```
implied vol(payoff="call", option price=5.17174691, S0=100, K=110, T=1, r
=0.13, q=0.09)
```

Out[110]:

0.1999999687580392

In [111]:

```
implied vol(payoff="put", option price=0.10336389, S0=100, K=70, T=1, r=
0.13, q=0.02)
```

Out[111]:

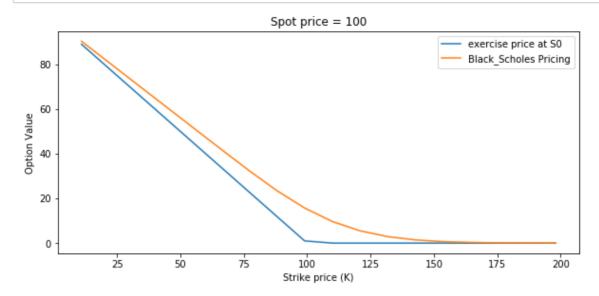
0.21991443538814784

```
In [112]:
implied vol(payoff="call", option price=6.0128267, S0=100, K=110, T=1, r=
0.13, q=0.07)
Out[112]:
0.20028961002081633
In [113]:
# Creates the option 1
opt param = Option param(S0=100, K=110, T=1, exercise="European", payoff=
"call")
diff param = Diffusion process(r=0.13, q=0.09, sig=0.2)
option 1 = Option pricer(opt param, diff param)
In [114]:
# Creates the option 2
opt param = Option param(S0=100, K=70, T=1, exercise="European", payoff=
diff param = Diffusion process(r=0.13, q=0.02, sig=0.22)
option 2 = Option pricer(opt param, diff param)
In [115]:
# Creates the option 3
opt param = Option param(S0=100, K=110, T=1, exercise="European", payoff=
"call")
diff param = Diffusion process(r=0.13, q=0.07, sig=0.2)
option 3 = Option pricer(opt param, diff param)
In [116]:
option 1.Binomial Tree()
Out[116]:
('European', 5.17173982431996)
In [117]:
option 1.Monte Carlo Simulation()
Out[117]:
('European', 5.296825291527466)
In [118]:
option 1.Black Scholes()
Out[118]:
('European', 5.171746910724259)
```

```
In [119]:
option 2.Binomial Tree()
Out[119]:
('European', 0.10361881912995327)
In [120]:
option 2.Monte Carlo Simulation()
Out[120]:
('European', 0.11649074315647964)
In [121]:
option 2.Black Scholes()
Out[121]:
('European', 0.10363892841819022)
In [122]:
option 3.Binomial Tree()
Out[122]:
('European', 6.002053141851009)
In [123]:
option 3.Monte Carlo Simulation()
Out[123]:
('European', 5.992604805106601)
In [124]:
option 3.Black Scholes()
Out[124]:
('European', 6.002075719878697)
```

In [125]:

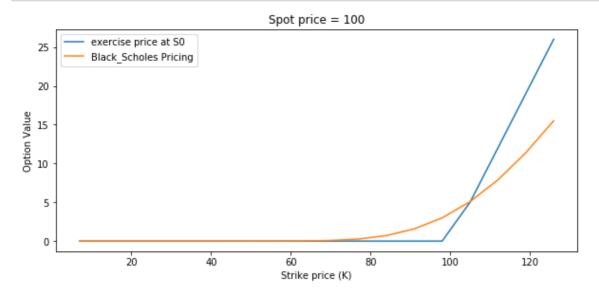
option_1.plot()

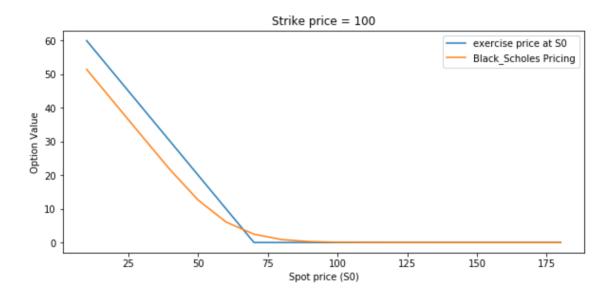




In [126]:

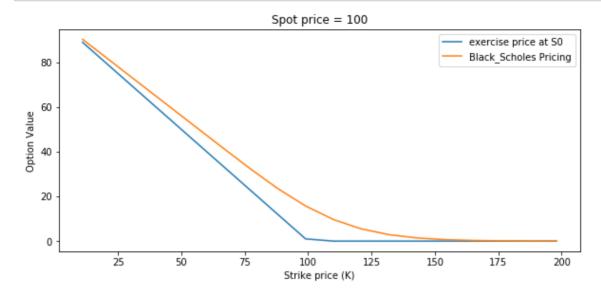
option_2.plot()





In [127]:

option_3.plot()





In [141]:

```
greeks hedge(option 1,option 2,option 3)
Option 1:
Delta: 0.39292696684251743
Gamma: 0.01794837106324773
Theta: -0.01229878077157661
Vega: 0.35896742126495457
Rho: 0.34120949773527487
Option 2:
Delta: -0.012578198789316201
Gamma: 0.0014748209310516587
Theta: -0.0005618449806511207
Vega: 0.03244606048313649
Rho: -0.013614588073498057
Option 3:
Delta: 0.43774995733591904
Gamma: 0.01854415142268068
Theta: -0.015219342972513475
Vega: 0.37088302845361354
Rho: 0.37772920013713207
How many greeks are going to hedge: 3
The first hedging greek: Delta
The second hedging greek: Theta
The thirth hedging greek: Rho
Portfolio:
Stock : 0.001956661742301191
Option 1: -1.1678427311703725
Option 2: -1.5240880948644884
Option 3: 1.0
The Portfolio's Greeks:
Delta: 0.0
Gamma: -0.004664480282954627
Theta: -4.336808689942018e-17
Vega: -0.09778511970523812
Rho : 5.551115123125783e-17
```

/Users/kenneth/anaconda3/lib/python3.7/site-packages/ipykerne l launcher.py:53: FutureWarning: `rcond` parameter will chang e to the default of machine precision times ``max(M, N)`` whe re M and N are the input matrix dimensions.

To use the future default and silence this warning we advise to pass `rcond=None`, to keep using the old, explicitly pass `rcond=-1`.

In []:			