

HW5_Python

April 23, 2024

```
[ ]: # imports
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from scipy.stats import norm, multivariate_normal
from IPython.display import display, Markdown

# set up the environment
%matplotlib inline
plt.rc("figure", figsize=(16, 6)) # set default size of plots
sns.set_style("whitegrid") # set default seaborn style
rng = np.random.default_rng() # random generator

[ ]: def Black_Scholes_Call(S, K, r, sigma, q, T):
    """Black-Scholes Call option price.

    Args:
        S (float): initial stock price
        K (float): strike price
        r (float): risk-free rate
        sigma (float): volatility
        q (float): dividend yield
        T (float): time to maturity

    Returns:
        float: call option price
    """
    if sigma == 0:
        return max(S * np.exp(-q * T) - K * np.exp(-r * T), 0)
    else:
        d1 = (np.log(S / K) + (r - q + sigma**2 / 2) * T) / (sigma * np.sqrt(T))
        d2 = d1 - sigma * np.sqrt(T)
        return S * np.exp(-q * T) * norm.cdf(d1) - K * np.exp(-r * T) * norm.
        ↪cdf(d2)

def Black_Scholes_Put(S, K, r, sigma, q, T):
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"""Black-Scholes Put option price.

Args:
    S (float): initial stock price
    K (float): strike price
    r (float): risk-free rate
    sigma (float): volatility
    q (float): dividend yield
    T (float): time to maturity

Returns:
    float: put option price
    """

if sigma == 0:
    return max(K * np.exp(-r * T) - S * np.exp(-q * T), 0)
else:
    d1 = (np.log(S / K) + (r - q + sigma**2 / 2) * T) / (sigma * np.sqrt(T))
    d2 = d1 - sigma * np.sqrt(T)
    return K * np.exp(-r * T) * norm.cdf(-d2) - S * np.exp(-q * T) * norm.
↪cdf(-d1)

```

0.1 Q 8.5 Compare Chooser Option with Straddle Option

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[ ]: def Chooser(S, Kc, Kp, r, sigma, q, T, Tc, Tp, tol=1e-6):
    # Find the S* such that C(S*, Kc, r, sigma, q, Tc - T) = P(S*, Kp, r,
↪sigma, q, Tp - T)
    if (Tc == Tp) and (Kc == Kp):
        # simple chooser, through put-call parity
        S_star = np.exp((q - r) * (Tc - T)) * Kc
    else:
        # Find the S* by bisection
        def diff(S):
            """Difference between the call and put prices"""
            return Black_Scholes_Call(S, Kc, r, sigma, q, Tc - T) -
↪Black_Scholes_Put(
                S, Kp, r, sigma, q, Tp - T
            )

        lower = 0
        upper = np.exp(q * Tc) * (Kc + Kp)
        assert diff(lower) < 0 & diff(upper) > 0, "Invalid initial bounds for
↪bisection"
        guess = (Kc + Kp) / 2
        while upper - lower > tol:
            if diff(guess) < 0:
                lower = guess
            else:

```

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        upper = guess
        guess = (lower + upper) / 2
    S_star = guess

    # Calculate the option price
    d1 = (np.log(S / S_star) + (r - q + sigma**2 / 2) * T) / (sigma * np.
↪sqrt(T))
    d2 = d1 - sigma * np.sqrt(T)
    d1c = (np.log(S / Kc) + (r - q + sigma**2 / 2) * Tc) / (sigma * np.sqrt(Tc))
    d2c = d1c - sigma * np.sqrt(Tc)
    d1p = (np.log(S / Kp) + (r - q + sigma**2 / 2) * Tp) / (sigma * np.sqrt(Tp))
    d2p = d1p - sigma * np.sqrt(Tp)
    rhoc = np.sqrt(T / Tc)
    rhop = np.sqrt(T / Tp)
    M1c = multivariate_normal.cdf([d1, d1c], cov=[[1, rhoc], [rhoc, 1]])
    M2c = multivariate_normal.cdf([d2, d2c], cov=[[1, rhoc], [rhoc, 1]])
    M1p = multivariate_normal.cdf([-d1, -d1p], cov=[[1, rhop], [rhop, 1]])
    M2p = multivariate_normal.cdf([-d2, -d2p], cov=[[1, rhop], [rhop, 1]])
    return (
        S * np.exp(-q * Tc) * M1c
        - Kc * np.exp(-r * Tc) * M2c
        + Kp * np.exp(-r * Tp) * M2p
        - S * np.exp(-q * Tp) * M1p
    )

```

```

[ ]: # Parameters
S = 50
r = 0.05
sigma = 0.3
q = 0.02
Tc = Tp = 1
T = 0.5
Kc = Kp = S

# Calculate the option price
chooser_price = Chooser(S, Kc, Kp, r, sigma, q, T, Tc, Tp)
straddle_price = Black_Scholes_Call(
    S, Kc, r, sigma, q, Tc) + Black_Scholes_Put(S, Kp, r, sigma, q, Tp)
display(Markdown(f"Chooser Option Price: {chooser_price:.4f}"))
display(Markdown(f"Straddle Option Price: {straddle_price:.4f}"))

```

Chooser Option Price: 9.9052

Straddle Option Price: 11.5718

As in the Section 8.4: “A simple chooser must be cheaper than a straddle with the same exercise price and maturity $T = T_c = T_p$, because a straddle is always in the money at maturity, whereas a simple chooser has the same value as the straddle if it is in the money but is only in the money at T when the choice made at T turns out to have been the best one.”

0.2 Q 8.6 Floating-Strike Lookback Call

Floating-Strike Lookback Call Pricing Formula (Eq. 8.30):

$$e^{-qT}S(0)N(d_1) - e^{-rT}S_{min}N(d_2) + \frac{\sigma^2}{2(r-q)} \left(\frac{S_{min}}{S(0)} \right)^{2(r-q)/\sigma^2} e^{-rT}S(0)N(d'_2) - \frac{\sigma^2}{2(r-q)} e^{-qT}S(0)N(-d_1)$$

where

$$d_1 = \frac{\ln(S(0)/S_{min}) + (r - q + \sigma^2/2)T}{\sigma\sqrt{T}}, \quad d_2 = d_1 - \sigma\sqrt{T}$$

$$d'_1 = \frac{\ln(S_{min}/S(0)) + (r - q + \sigma^2/2)T}{\sigma\sqrt{T}}, \quad d'_2 = d'_1 - \sigma\sqrt{T}$$

```
[ ]: def Floating_Strike_Lookback_Call(S0, Smin, r, sigma, q, T):
    d1 = (np.log(S0 / Smin) + (r - q + sigma**2 / 2) * T) / (sigma * np.sqrt(T))
    d2 = d1 - sigma * np.sqrt(T)
    d1p = (np.log(Smin / S0) + (r - q + sigma**2 / 2) * T) / \
        (sigma * np.sqrt(T))
    d2p = d1p - sigma * np.sqrt(T)
    x = 2 * (r - q) / sigma**2
    call = (
        S0 * np.exp(-q * T) * norm.cdf(d1)
        - Smin * np.exp(-r * T) * norm.cdf(d2)
        + S0 * np.exp(-r * T) * norm.cdf(d2p) * (Smin / S0) ** x / x
        - S0 * np.exp(-q * T) * norm.cdf(-d1) / x
    )
    return call
```

```
[ ]: # Parameters
S0 = 100 # initial stock price
r = 0.05 # risk-free rate
sigma = 0.3 # volatility
q = 0 # non-dividend paying
T = 2 / 12 # time to maturity (two months)
Smin = S0 # minimum stock price at beginning

# Minimum fee
fee = Floating_Strike_Lookback_Call(S0, Smin, r, sigma, q, T) * np.exp(r * T)
# value at time T
display(Markdown(f"Minimum Fee: {fee:.4f}"))
```

Minimum Fee: 9.8684

0.3 Q 8.7 Down-and-Out Call

```
[ ]: def Hedged_Cost(S0, K, r, sigma, q, T, L, N, N_con):
    dt = T / N # time step
    t = np.linspace(0, T, N + 1) # time grid
    B = np.zeros(N + 1)
```

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B[1:] = np.cumsum(np.sqrt(dt) * rng.standard_normal(N)) # Brownian motion
S = S0 * np.exp((r - q - sigma**2 / 2) * t + sigma * B) # Underlying price
# whether the option is knocked out
if any(S < L):
    out = 1
else:
    out = 0
# end-of-year costs, net of the option values at maturity
cost_euro = S[-1] * N_con - np.maximum(S[-1] - K, 0) * N_con
cost_knockout = S[-1] * N_con - \
    np.maximum(S[-1] - K, 0) * N_con * (1 - out)
return cost_euro, cost_knockout

def Simulate_Hedge_Cost(S0, K, r, sigma, q, T, L, N, N_con, M):
    costs_euro = np.zeros(M)
    costs_knockout = np.zeros(M)
    for i in range(M):
        costs_euro[i], costs_knockout[i] = Hedged_Cost(
            S0, K, r, sigma, q, T, L, N, N_con
        )
    return costs_euro, costs_knockout

```

```

[ ]: # Parameters
S0 = 50 # initial stock price
K = S0 # strike price (ATM at the beginning)
L = (1 - 0.1) * S0 # barrier (10% below the initial price)
r = 0.05 # risk-free rate
sigma = 0.2 # volatility
q = 0.02 # dividend yield
T = 1 # time to maturity in years
N = 252 # number of periods in a year
N_con = 100 # number of contracts
M = 500 # number of simulations

```

```

# Simulation
costs_euro, costs_knockout = Simulate_Hedge_Cost(
    S0, K, r, sigma, q, T, L, N, N_con, M)

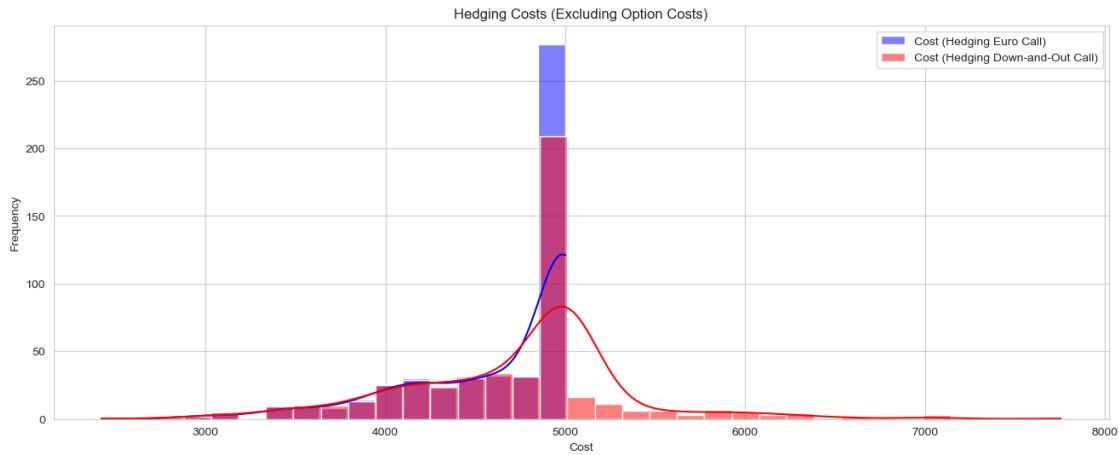
```

```

[ ]: # Histogram
fig, ax = plt.subplots(figsize=(16, 6))
sns.histplot(costs_euro, kde=True, color="blue", label="Cost (Hedging Euro_
↳Call)")
sns.histplot(
    costs_knockout, kde=True, color="red", label="Cost (Hedging Down-and-Out_
↳Call)"
)

```

```
plt.xlabel("Cost")
plt.ylabel("Frequency")
plt.title("Hedging Costs (Excluding Option Costs)")
plt.legend()
plt.show()
```



As shown in the histogram above, the costs using a European call hedging are truncated at K multiplying the number of contracts, while the costs using down-and-out call hedging might exceed this level.

0.3.1 Followup: Q 8.8 end-of-year costs including the costs of the options

```
[ ]: def Down_And_Out_Call(S, K, r, sigma, q, T, L):
    if K > L:
        d1 = (np.log(S / K) + (r - q + sigma**2 / 2) * T) / (sigma * np.sqrt(T))
        d1p = (np.log(L**2 / (K * S)) + (r - q + sigma**2 / 2) * T) / (
            sigma * np.sqrt(T)
        )
    else:
        d1 = (np.log(S / L) + (r - q + sigma**2 / 2) * T) / (sigma * np.sqrt(T))
        d1p = (np.log(L / S) + (r - q + sigma**2 / 2) * T) / \
            (sigma * np.sqrt(T))

    d2 = d1 - sigma * np.sqrt(T)
    d2p = d1p - sigma * np.sqrt(T)
    price = np.exp(-q * T) * S * (
        norm.cdf(d1)
        - (L / S) ** (2 * (r - q + sigma**2 / 2) / sigma**2) * norm.cdf(d1p)
    ) - np.exp(-r * T) * K * (
        norm.cdf(d2)
        - (L / S) ** (2 * (r - q - sigma**2 / 2) / sigma**2) * norm.cdf(d2p)
```

```
)
return price
```

```
[ ]: CO_euro = Black_Scholes_Call(S0, K, r, sigma, q, T)
display(Markdown(f"**Black-Scholes Call Option Price**: {CO_euro:.4f}"))

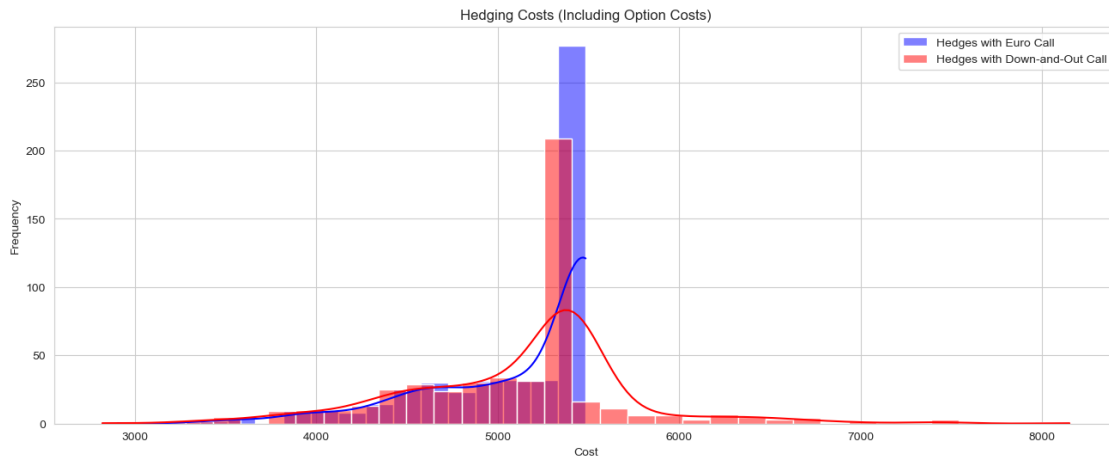
CO_knockout = Down_And_Out_Call(S0, K, r, sigma, q, T, L)
display(Markdown(f"**Down-and-Out Call Option Price**: {CO_knockout:.4f}"))
```

Black-Scholes Call Option Price: 4.6135

Down-and-Out Call Option Price: 3.7935

```
[ ]: # Modified costs
costs_euro = costs_euro + CO_euro * np.exp(r * T) * N_con
costs_knockout = costs_knockout + CO_knockout * np.exp(r * T) * N_con

# Histogram
fig, ax = plt.subplots(figsize=(16, 6))
sns.histplot(costs_euro, kde=True, color="blue", label="Hedges with Euro Call")
sns.histplot(costs_knockout, kde=True, color="red",
              label="Hedges with Down-and-Out Call")
plt.xlabel("Cost")
plt.ylabel("Frequency")
plt.title("Hedging Costs (Including Option Costs)")
plt.legend()
plt.show()
```



Similarly, the costs using a European call hedging are truncated at K multiplying the number of contracts, plus the costs of the options. The costs using down-and-out call hedging might exceed this level, but have a lower costs for the options.