

HW2_Q3_Python

March 1, 2024

```
[ ]: # imports
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from scipy.stats import norm
%matplotlib inline
plt.rcParams["figure.figsize"] = (10, 6) #set default figure size
sns.set_theme(style="whitegrid") #set default seaborn theme
```

```
[ ]: # load data
cboe_quotes = np.loadtxt('CBOEQuotes.txt')
print(cboe_quotes[:10])
```

```
[[650.  232.3 234.3]
 [675.  207.5 209.5]
 [700.  182.9 184.9]
 [725.  158.5 160.5]
 [750.  134.4 136.4]
 [760.  124.9 126.9]
 [765.  120.1 122.1]
 [775.  110.8 112.8]
 [800.   87.9  89.9]
 [810.   79.   81.  ]]
```

```
[ ]: # parameters
T = 30 / 365 # 30 days to maturity
S0 = 884.25 # current stock price
q = 0.0176 # dividend yield
r = 0.0125 # risk-free rate
K = cboe_quotes[:, 0] # strike prices
bid = cboe_quotes[:, 1] # bid prices
ask = cboe_quotes[:, 2] # ask prices
market = (bid + ask) / 2 # average of bid and ask prices
```

0.1 Bisection Method

```
[ ]: def Black_Scholes_Call(S, K, r, q, T, sigma):  
    """Black-Scholes call option price.  
  
    Args:  
        S (float): spot price  
        K (float): strike price  
        r (float): risk-free interest rate  
        q (float): dividend yield  
        T (float): time to maturity  
        sigma (float): volatility  
  
    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 implied_volatility_bisection(S, K, r, q, T, C, tol=1e-6, max_iter=1000):  
    """Implied volatility using bisection method.  
  
    Args:  
        S (float): spot price  
        K (float): strike price  
        r (float): risk-free interest rate  
        q (float): dividend yield  
        T (float): time to maturity  
        C (float): call option price  
        tol (float, optional): tolerance. Defaults to 1e-6.  
        max_iter (int, optional): maximum number of iterations. Defaults to  
↪1000.
```

```
    Returns:  
        float: approximate implied volatility  
    """  
    if C < max(S * np.exp(-q * T) - K * np.exp(-r * T), 0):  
        print("Option price $" + str(C) +  
              " violates the arbitrage bound (too low).")  
        return np.nan  
    elif C > S * np.exp(-q * T):  
        print("Option price $" + str(C) +
```

```

        " violates the arbitrage bound (too high).")
    return np.nan

lower = 0
upper = 1
while Black_Scholes_Call(S, K, r, q, T, upper) - C < 0:
    upper *= 2
guess = (lower + upper) / 2

while upper - lower > tol and max_iter > 0:
    diff = Black_Scholes_Call(S, K, r, q, T, guess) - C
    if diff < 0:
        lower = guess
    else:
        upper = guess
    guess = (lower + upper) / 2
    max_iter -= 1
return guess

```

```

[ ]: iv_bid_bisection = np.array([implied_volatility_bisection(
    S0, K[i], r, q, T, bid[i]) for i in range(len(K))])
iv_ask_bisection = np.array([implied_volatility_bisection(
    S0, K[i], r, q, T, ask[i]) for i in range(len(K))])
iv_market_bisection = np.array([implied_volatility_bisection(
    S0, K[i], r, q, T, market[i]) for i in range(len(K))])
iv_bisection = np.stack(
    (iv_bid_bisection, iv_ask_bisection, iv_market_bisection), axis=1)

print(" ")
print("Implied Volatility using Bisection Method:")
print(iv_bisection[:10])

```

Option price \$232.3 violates the arbitrage bound (too low).
 Option price \$207.5 violates the arbitrage bound (too low).
 Option price \$182.9 violates the arbitrage bound (too low).
 Option price \$158.5 violates the arbitrage bound (too low).
 Option price \$233.3 violates the arbitrage bound (too low).
 Option price \$208.5 violates the arbitrage bound (too low).

Implied Volatility using Bisection Method:

```

[[      nan 0.5046258      nan]
 [      nan 0.46623468     nan]
 [      nan 0.43761778 0.33935022]
 [      nan 0.41029119 0.35418081]
 [0.29337454 0.38361406 0.34674215]
 [0.30126905 0.37372255 0.34213114]
 [0.29936171 0.36651659 0.33675146]
 [0.30273294 0.35829592 0.33274221]

```

```
[0.29391813 0.33256769 0.31397581]
[0.28782797 0.32195616 0.30536699]]
```

0.2 Newton Method

```
[ ]: def vega(S, K, r, q, T, sigma):
    """Vega of a call option.

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

    Returns:
        float: vega
    """
    if sigma == 0:
        return 0
    else:
        d1 = (np.log(S / K) + (r - q + sigma**2 / 2) * T) / (sigma * np.sqrt(T))
        return S * np.exp(-q * T) * norm.pdf(d1) * np.sqrt(T)

[ ]: def implied_volatility_newton(S, K, r, q, T, C, sigma_guess=0.5, tol=1e-6,
    ↪max_iter=1000):
    """Implied volatility using Newton's method.

    Args:
        S (float): spot price
        K (float): strike price
        r (float): risk-free interest rate
        q (float): dividend yield
        T (float): time to maturity
        C (float): call option price
        sigma_guess (float): initial guess of volatility
        tol (float): tolerance. Defaults to 1e-6.
        max_iter (int): maximum number of iterations. Defaults to 1000.

    Returns:
        float: approximate implied volatility
    """
    if C < max(S * np.exp(-q * T) - K * np.exp(-r * T), 0):
        print("Option price $" + str(C) +
              " violates the arbitrage bound (too low).")
        return np.nan
```

```

elif C > S * np.exp(-q * T):
    print("Option price $" + str(C) +
          " violates the arbitrage bound (too high).")
    return np.nan

sigma = sigma_guess
while max_iter > 0:
    diff = Black_Scholes_Call(S, K, r, q, T, sigma) - C
    if abs(diff) < tol:
        break
    sigma -= diff / vega(S, K, r, q, T, sigma)
    max_iter -= 1
return sigma

```

```

[ ]: iv_bid_newton = np.array([implied_volatility_newton(
    S0, K[i], r, q, T, bid[i]) for i in range(len(K))])
iv_ask_newton = np.array([implied_volatility_newton(
    S0, K[i], r, q, T, ask[i]) for i in range(len(K))])
iv_market_newton = np.array([implied_volatility_newton(
    S0, K[i], r, q, T, market[i]) for i in range(len(K))])
iv_newton = np.stack((iv_bid_newton, iv_ask_newton, iv_market_newton), axis=1)

print(" ")
print("Implied Volatility using Newton's Method:")
print(iv_newton[:10])

```

Option price \$232.3 violates the arbitrage bound (too low).
 Option price \$207.5 violates the arbitrage bound (too low).
 Option price \$182.9 violates the arbitrage bound (too low).
 Option price \$158.5 violates the arbitrage bound (too low).
 Option price \$233.3 violates the arbitrage bound (too low).
 Option price \$208.5 violates the arbitrage bound (too low).

Implied Volatility using Newton's Method:

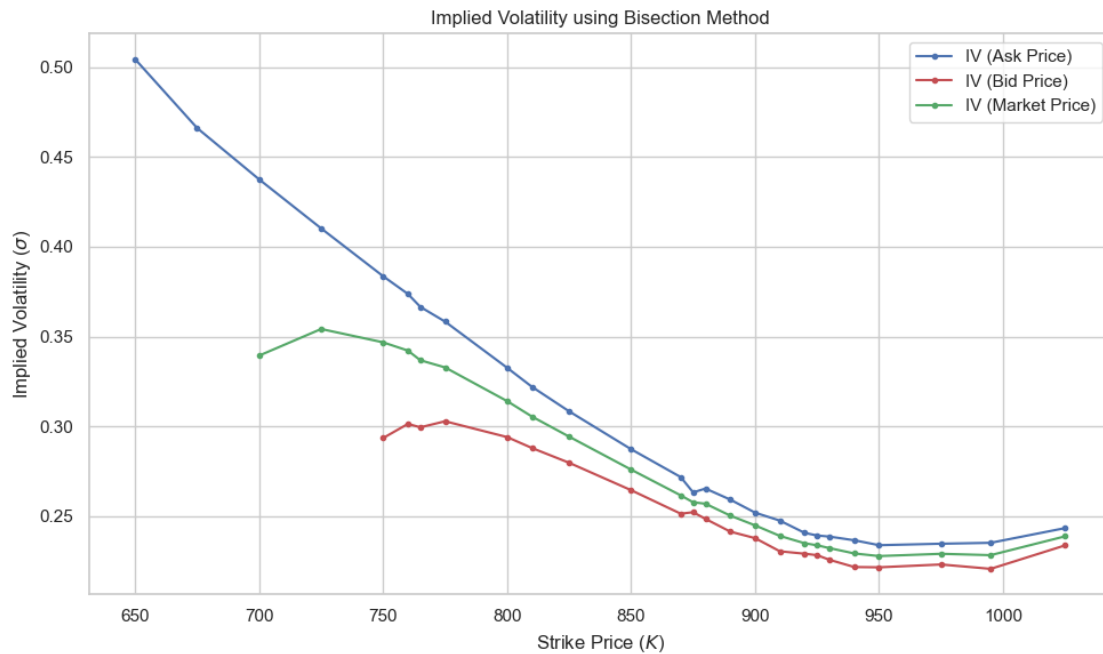
```

[[      nan 0.50462541      nan]
 [      nan 0.46623491      nan]
 [      nan 0.43761761 0.33935068]
 [      nan 0.41029135 0.35418082]
 [0.29337411 0.38361417 0.34674248]
 [0.30126863 0.37372226 0.34213076]
 [0.29936161 0.36651695 0.33675189]
 [0.3027326  0.3582963  0.33274255]
 [0.29391836 0.3325675  0.31397606]
 [0.28782754 0.32195602 0.30536718]]

```

0.3 Implied Volatility

```
[ ]: # Comparison of Different Prices Used
fig, ax = plt.subplots()
ax.plot(K, iv_ask_bisection, 'b.-', label='IV (Ask Price)')
ax.plot(K, iv_bid_bisection, 'r.-', label='IV (Bid Price)')
ax.plot(K, iv_market_bisection, 'g.-', label='IV (Market Price)')
ax.legend()
ax.set_title('Implied Volatility using Bisection Method')
ax.set_xlabel(r'Strike Price ($K$)')
ax.set_ylabel(r'Implied Volatility ($\sigma$)')
plt.tight_layout()
plt.show()
```

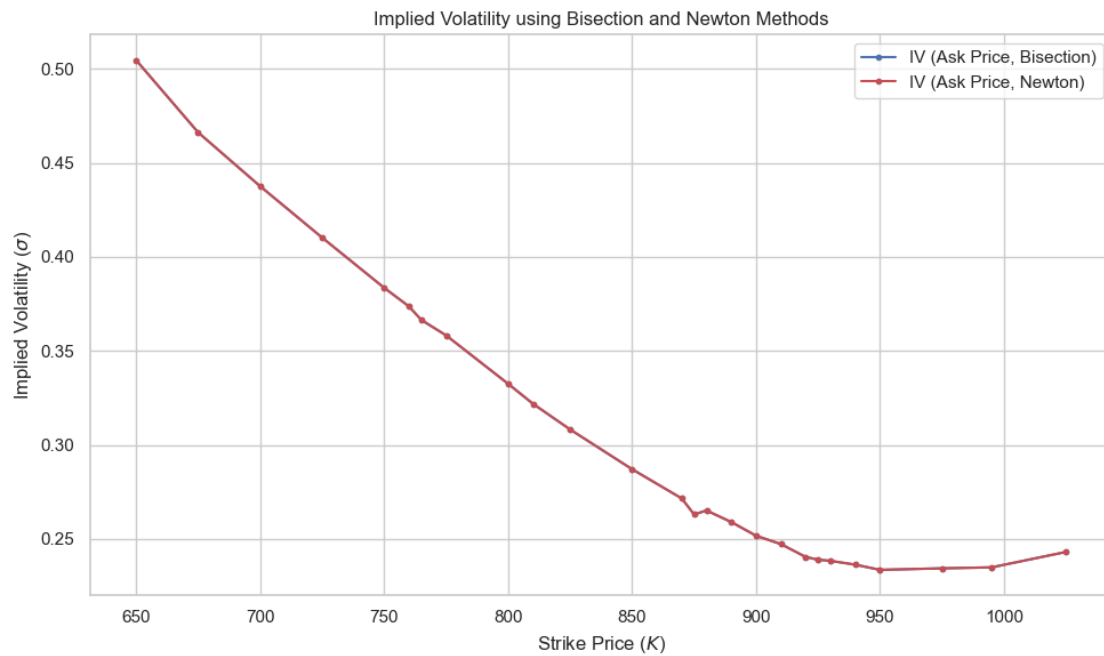


The missing values in the implied volatility that calculated using bid and average price are those that violate the no-arbitrage bounds:

$$C \geq Se^{-qT} - Ke^{-rT}$$

```
[ ]: # Comparison of Bisection and Newton Methods
fig, ax = plt.subplots()
ax.plot(K, iv_ask_bisection, 'b.-', label='IV (Ask Price, Bisection)')
ax.plot(K, iv_ask_newton, 'r.-', label='IV (Ask Price, Newton)')
ax.legend()
ax.set_title('Implied Volatility using Bisection and Newton Methods')
ax.set_xlabel(r'Strike Price ($K$)')
ax.set_ylabel(r'Implied Volatility ($\sigma$)')
```

```
plt.tight_layout()
plt.show()
```



As shown in the above figure, the difference between the implied volatility calculated using the two methods (Bisection and Newton) is very small.