

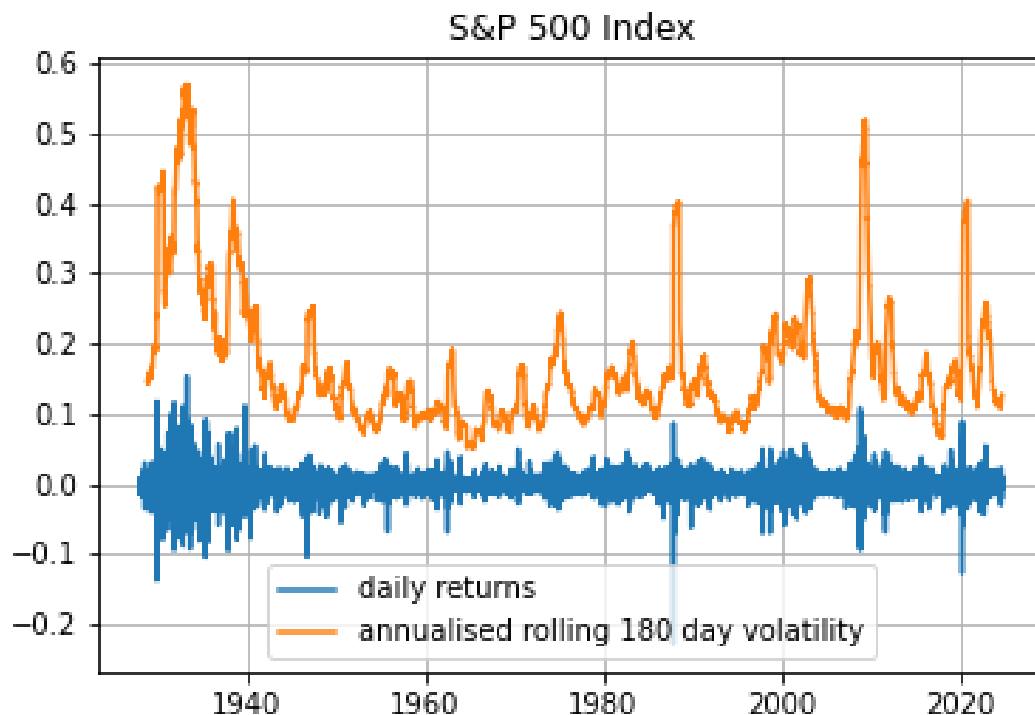
FINM3405 Derivatives and risk management

Tutorial Sheet 7: Options - Numerical methods (binomial and Monte Carlo)

Suggested solutions

September 22, 2024

Question 1. Download the historical daily returns of your favourite stock or share market index from yahoo!finance or some other data provider. Then using Excel or some other means, on the same graph plot the (i) time series of daily returns and (i) a rolling annualised 180 day historical volatility. What does the plot tell you about stock or share market index volatilities over time?



Volatility σ is not constant over time.

```
1 import numpy as np, yfinance as yf
2 S = yf.download("^GSPC")["Adj Close"]
3 ret = np.log(S).diff(1).dropna()
4 sigma = ret.rolling(window=180).std()*np.sqrt(252)
5 plt.figure()
6 plt.plot(ret, label='daily returns')
7 plt.plot(sigma, label='annualised rolling 180 day volatility')
8 plt.grid(); plt.title("S&P 500 Index"); plt.legend()
```

In the following questions let $S = 50$, $K = 50$, $r = 5\%$, $T = \frac{1}{2}$, $\sigma = 25\%$ and dividend yield $y = 0$ unless otherwise stated.

Binomial model

Question 2. Calculate by hand the 1-step binomial prices of ATM European call options and compare them to the Black-Scholes prices. Do the same for strikes of $K = 47.5$ and $K = 52.5$. Use the CRR or JR schemes. I did the ATM options in the lecture. For the others, and using the Jarrow-Rudd scheme (since it most resembles geometric Brownian motion), we still have that

$$u = e^{(r-\sigma^2/2)T+\sigma\sqrt{T}} = 1.2046 \quad \text{and} \quad d = e^{(r-\sigma^2/2)T-\sigma\sqrt{T}} = 0.84586,$$

as well as

$$q = \frac{u^{rT} - d}{u - d} = 0.500231.$$

We calculate that

$$S_u = Su = 60.23 \quad \text{and} \quad S_d = Sd = 42.293.$$

For the strike of $K = 47.5$, the call option payoffs are

$$C_u = \max\{0, S_u - K\} = 12.73 \quad \text{and} \quad S_d = \max\{0, S_d - K\} = 0,$$

and the put payoffs are

$$P_u = \max\{0, K - S_u\} = 0 \quad \text{and} \quad P_d = \max\{0, K - S_d\} = 5.207.$$

Call prices are given by

$$\begin{aligned} C &= e^{-rT} \mathbb{E}^q[C_T] \\ &= e^{-rT} [qC_u + (1-q)C_d] \\ &= 6.211 \end{aligned}$$

and the put prices are

$$\begin{aligned} P &= e^{-rT} \mathbb{E}^q[P_T] \\ &= e^{-rT} [qP_u + (1-q)P_d] \\ &= 2.538. \end{aligned}$$

When $K = 52.5$ I get $C = 3.77$ and $P = 4.9752$.

```

1 S = 50
2 K = 52.5
3 r = 0.05
4 T = 1/2
5 sigma = 0.25
6 u = np.exp((r - 0.5*sigma**2)*T + sigma*np.sqrt(T)) # JR
7 d = np.exp((r - 0.5*sigma**2)*T - sigma*np.sqrt(T)) # JR
8 q = (np.exp(r*T)-d)/(u-d)
9 Su = S*u
10 Sd = S*d
11 Cu = max(0, Su-K)
12 Cd = max(0, Sd-K)
13 Pu = max(0, K-Su)
14 Pd = max(0, K-Sd)
15 C = np.exp(-r*T)*(q*Cu + (1-q)*Cd)
16 P = np.exp(-r*T)*(q*Pu + (1-q)*Pd)

```

Question 3. Use excel to create a 7 layer binomial model asset price tree and calculate the binomial model prices of ATM European call and put options, and compare your prices to the Black-Scholes model prices. Do the same for strikes of $K = 47.5$ and $K = 52.5$. Use the CRR or JR schemes. [See the spreadsheet provided](#), which gives the following asset, call and put price trees for ATM options:

Asset price tree					Call price tree					Put price tree				
				80.568					30.568					0.000
			75.260						25.438				0	0.000
			70.302	70.490					20.490			0	0	0.000
		65.670		65.846					16.203			0	0	0.000
		61.343	61.508	61.673					12.22			0.172	0	0.000
		57.302	57.456	57.610					8.882			0.695	0.3451	0
	53.527	53.670	53.814	53.958					6.237			1.65	1.224	0.6928
	50.134	50.269	50.403	50.543					3.636			2.617	2.1109	1.3905
50	46.831	46.957	47.083	47.209	4.250				2.293			4.402	4.029	3.5441
	43.863	43.981	44.099	44.217					0.966			6.218	5.9761	5.7231
		41.083	41.193	41.304					0.244			8.452	8.4509	8.696
		38.479	38.582	38.685								10.988	11.239	11.491
		36.040	36.137	36.234								13.604	13.863	14.122
		33.756	33.853	33.950								16.065	16.324	16.583
			31.617	31.714								18.383	18.642	18.901

But it's not feasible to do numerical option pricing in Excel due to the manual work involved and inflexibility in choosing parameters values such as the number N of layers or dates characterising the binomial model trees. It's more efficient and flexible to use coding languages such as Python, Java, C++, etc. The below Python code calculates binomial option prices and produces the following trees (in which we ignore the NumPy NaNs):

	0	1	2	3	4	5	6	7
0	50	46.8311	43.863	41.083	38.4792	36.0405	33.7563	31.6169
1	nan	53.5265	50.1341	46.9567	43.9806	41.1932	38.5824	36.1372
2	nan	nan	57.3018	53.6701	50.2686	47.0826	44.0986	41.3037
3	nan	nan	nan	61.3434	57.4555	53.8141	50.4034	47.2089
4	nan	nan	nan	nan	65.6699	61.5079	57.6096	53.9584
5	nan	nan	nan	nan	nan	70.3017	65.8461	61.6729
6	nan	nan	nan	nan	nan	nan	75.2601	70.4903
7	nan	nan	nan	nan	nan	nan	nan	80.5683

	0	1	2	3	4	5	6	7
0	4.2496	2.29278	0.965963	0.243915	0	0	0	0
1	nan	6.23673	3.63594	1.69489	0.489563	0	0	0
2	nan	nan	8.88202	5.60291	2.91228	0.982605	0	0
3	nan	nan	nan	12.2245	8.3335	4.8627	1.97219	0
4	nan	nan	nan	nan	16.2028	11.8638	7.78786	3.9584
5	nan	nan	nan	nan	nan	20.6576	16.0243	11.6729
6	nan	nan	nan	nan	nan	nan	25.4384	20.4903
7	nan	nan	nan	nan	nan	nan	nan	30.5683

	0	1	2	3	4	5	6	7
0	3.0151	4.40168	6.21804	8.45169	10.9879	13.6036	16.0655	18.3831
1	nan	1.65016	2.6169	4.029	5.97607	8.45092	11.2393	13.8628
2	nan	nan	0.695273	1.2236	2.11085	3.54411	5.72314	8.6963
3	nan	nan	nan	0.171952	0.345142	0.692771	1.39053	2.79109
4	nan	nan	nan	nan	0	0	0	0
5	nan	nan	nan	nan	nan	0	0	0
6	nan	nan	nan	nan	nan	nan	0	0
7	nan	nan	nan	nan	nan	nan	nan	0

The differences here are that the Python code creates the trees “upside down” relative to how we may visualise them in our heads (the row number counts the number of up movements in the asset price), and computers like to work with rectangular matrices or data frames or arrays, hence the NaNs (not a number). But neither of these are major obstacles to conceptualise and work with.

```

1 import numpy as np
2 S = 50
3 K = 50
4 r = 0.05
5 T = 1/2
6 sigma = 0.25
7 N = 7; dt = T/N
8 u = np.exp((r - 0.5*sigma**2)*dt + sigma*np.sqrt(dt)) # JR
9 d = np.exp((r - 0.5*sigma**2)*dt - sigma*np.sqrt(dt)) # JR
10 q = (np.exp(r*dt)-d)/(u-d)
11 # asset price tree
12 St = np.zeros([N+1, N+1])*np.nan # initially create array of NaNs
13 for j in range(N+1):
14     for i in range(j+1):
15         St[i,j] = S*(u**i)*d**(j-i)
16 # option price trees
17 Ct = np.zeros([N+1, N+1])*np.nan # initially create array of NaNs
18 Pt = np.zeros([N+1, N+1])*np.nan # initially create array of NaNs
19 for i in range(N+1): # option payoffs at expiry
20     ST = St[i,N]
21     Ct[i,N] = max(0, ST-K)
22     Pt[i,N] = max(0, K-ST)
23 for j in reversed(range(N)): # step backwards through asset tree to create option trees
24     for i in range(j+1):
25         Ct[i,j] = np.exp(-r*dt)*(q*Ct[i+1, j+1] + (1-q)*Ct[i, j+1])
26         Pt[i,j] = np.exp(-r*dt)*(q*Pt[i+1, j+1] + (1-q)*Pt[i, j+1])
27 C = Ct[0,0]
28 P = Pt[0,0]

```

Question 4. In the derivation of the binomial model in the lecture notes, we suggestively defined and found an equation for the parameter Δ , namely

$$\Delta = \frac{C_u - C_d}{S(u - d)} = \frac{C_u - C_d}{S_u - S_d}.$$

The notation was not a coincidence since it does in fact define the delta of a call option as calculated by the binomial model. Calculate by hand the 1-period binomial model deltas of ATM calls and puts and compare them to the Black-Scholes model deltas. You can use the CRR or JR schemes. From the first question (or lecture notes), when $K = 50$ we had $S_u = 60.23$, $S_d = 42.293$, $C_u = 10.23$, $C_d = 0$, $P_u = 0$ and $P_d = 7.707$. This gives

$$\Delta_C = 0.57 \quad \text{and} \quad \Delta_P = -0.4297.$$

These compare to the Black-Scholes deltas of $\Delta_C = 0.59088$ and $\Delta_P = -0.40912$. We'd add the following lines to the code from Question 1 to calculate them:

```
1 deltaC = (Cu-Cd)/(Su-Sd)
2 deltaP = (Pu-Pd)/(Su-Sd)
```

Question 5. Following on from questions 3 and 4, calculate the ATM put and call deltas at each node of your 7 layer asset price tree (not including the final expiry date nodes). Then compare the deltas at the time $t_0 = 0$ node to the Black-Scholes model deltas. See the Excel spreadsheet provided. But it's much better to just add a few lines to the Python code. The below code calculates $\Delta_C = 0.589$ and $\Delta_P = -0.41095$, and creates the following delta trees:

	0	1	2	3	4	5	6		0	1	2	3	4	5	6
0	0.589048	0.425758	0.24703	0.08898...	0	0	0	0	-0.4109...	-0.5742...	-0.75297	-0.9110...	-1	-1	-1
1	nan	0.731905	0.582121	0.385296	0.166842	0	0	1	nan	-0.2680...	-0.4178...	-0.6147...	-0.8331...	-1	-1
2	nan	nan	0.862947	0.754317	0.576415	0.312808	0	2	nan	nan	-0.1370...	-0.2456...	-0.4235...	-0.6871...	-1
3	nan	nan	nan	0.957984	0.909958	0.807036	0.586474	3	nan	nan	nan	-0.0420...	-0.0900...	-0.1929...	-0.4135...
4	nan	nan	nan	nan	1	1	1	4	nan	nan	nan	nan	0	0	0
5	nan	nan	nan	nan	nan	1	1	5	nan	nan	nan	nan	nan	0	0
6	nan	nan	nan	nan	nan	nan	1	6	nan	nan	nan	nan	nan	nan	0
7	nan	nan	nan	nan	nan	nan	nan	7	nan	nan	nan	nan	nan	nan	nan

```
1 import numpy as np
2 S = 50
3 K = 50
4 r = 0.05
5 T = 1/2
6 sigma = 0.25
7 N = 7; dt = T/N
8 u = np.exp((r - 0.5*sigma**2)*dt + sigma*np.sqrt(dt)) # JR
9 d = np.exp((r - 0.5*sigma**2)*dt - sigma*np.sqrt(dt)) # JR
10 q = (np.exp(r*dt)-d)/(u-d)
11 # asset price tree
12 St = np.zeros([N+1, N+1])*np.nan
13 for j in range(N+1):
14     for i in range(j+1):
15         St[i,j] = S*(u**i)*d**(j-i)
16 # option premiums and deltas
17 Ct = np.zeros([N+1, N+1])*np.nan
18 Pt = np.zeros([N+1, N+1])*np.nan
19 deltaCt = np.zeros([N+1, N])*np.nan
20 deltaPt = np.zeros([N+1, N])*np.nan
```

```

21 for i in range(N+1): # option payoffs at expiry
22     ST = St[i,N]
23     Ct[i,N] = max(0, ST-K)
24     Pt[i,N] = max(0, K-ST)
25 for j in reversed(range(N)): # step backwards through the tree
26     for i in range(j+1):
27         Ct[i,j] = np.exp(-r*dt)*(q*Ct[i+1, j+1] + (1-q)*Ct[i, j+1])
28         Pt[i,j] = np.exp(-r*dt)*(q*Pt[i+1, j+1] + (1-q)*Pt[i, j+1])
29         deltaCt[i,j] = (Ct[i+1, j+1] - Ct[i, j+1])/(St[i+1, j+1] - St[i, j+1])
30         deltaPt[i,j] = (Pt[i+1, j+1] - Pt[i, j+1])/(St[i+1, j+1] - St[i, j+1])
31 C = Ct[0,0]
32 P = Pt[0,0]
33 deltaC = deltaCt[0,0]
34 deltaP = deltaPt[0,0]

```

Question 6. A benefit of numerical option pricing methods is they can handle a very wide variety of payoffs at expiry, and hence price a wide variety of exotic options. Use the 7 layer binomial model to price the cash-or-nothing and asset-or-nothing binary options presented in a previous tutorial sheet and compare the results to their Black-Scholes prices. Use the CRR or JR schemes. For pricing the binary options, all we need to do here in either the Excel spreadsheets or the Python code is simply change the terminal payoffs at expiry. For example, the below Python code does it for the cash-or-nothing options and gives $C = 0.4877$ and $P = 0.4876$. The Black-Scholes prices are $C = 0.5083$ and $P = 0.467$. When I set $N = 10,000$ in the code I get $C = 0.5071$ and $P = 0.4682$.

```

1 import numpy as np
2 S = 50
3 K = 50
4 r = 0.05
5 T = 1/2
6 sigma = 0.25
7 N = 10000; dt = T/N
8 u = np.exp((r - 0.5*sigma**2)*dt + sigma*np.sqrt(dt)) # JR
9 d = np.exp((r - 0.5*sigma**2)*dt - sigma*np.sqrt(dt)) # JR
10 q = (np.exp(r*dt)-d)/(u-d)
11 # asset price tree
12 St = np.zeros([N+1, N+1])*np.nan
13 for j in range(N+1):
14     for i in range(j+1):
15         St[i,j] = S*(u**i)*d**(j-i)
16 # option premiums and deltas
17 Ct = np.zeros([N+1, N+1])*np.nan
18 Pt = np.zeros([N+1, N+1])*np.nan
19 for i in range(N+1): # option payoffs at expiry
20     ST = St[i,N]
21     Ct[i,N] = 1 if ST>K else 0 # cash-or-nothing call payoff
22     Pt[i,N] = 1 if ST<K else 0 # cash-or-nothing put payoff
23 for j in reversed(range(N)): # step backwards through the tree
24     for i in range(j+1):
25         Ct[i,j] = np.exp(-r*dt)*(q*Ct[i+1, j+1] + (1-q)*Ct[i, j+1])
26         Pt[i,j] = np.exp(-r*dt)*(q*Pt[i+1, j+1] + (1-q)*Pt[i, j+1])
27 C = Ct[0,0]
28 P = Pt[0,0]

```

Monte Carlo method

Question 7. When pricing European options by the Monte Carlo method, we actually only need to calculate the N final asset price prices S_i at expiry for $i = 1, \dots, N$. We can use the formula in the lecture notes to simulate them via

$$S_i = S e^{(r - \frac{1}{2}\sigma^2)T + \sigma\sqrt{T}Z_i} \quad \text{for } i = 1, \dots, N,$$

where each Z_i is an independent, identically distributed standard normal random variable. The option prices are then simply given by

$$C = e^{-rT} \frac{1}{N} \sum_{i=1}^N \max\{0, S_i - K\} \quad \text{and} \quad P = e^{-rT} \frac{1}{N} \sum_{i=1}^N \max\{0, K - S_i\}.$$

Use Excel, Python, R or Matlab, etc (whatever is your software of choice - I use these and Java and C⁺⁺), to price ATM call and put options via this “simplified” Monte Carlo method that simulates only the final asset prices at expiry. *Remark:* It’s particularly easy in Excel by simulating 1 final asset price at expiry and then dragging this cell down for say another 499 cells to get 500 simulated final asset prices at expiry. Compare your prices to the Black-Scholes and binomial model prices. Do the same for strikes of $K = 47.5$ and $K = 52.5$. I posted the R code for this on Blackboard, upon request from some students. When $K = 50$, the Python code below gives $C = 4.131$ and $P = 2.8904$.

```
1 import numpy as np
2 from scipy.stats import norm
3 S = 50; K = 50; r = 0.05; T = 1/2; sigma = 0.25; N = 1000000
4 CT = np.zeros(N); PT = np.zeros(N)
5 for i in range(N):
6     ST = S*np.exp((r - 0.5*sigma**2)*T + sigma*np.sqrt(T)*norm.rvs())
7     CT[i] = max(0, ST-K)
8     PT[i] = max(0, K-ST)
9 C = np.exp(-r*T)*np.mean(CT)
10 P = np.exp(-r*T)*np.mean(PT)
```

Question 8. Now use this simplified Monte Carlo method to price the ATM cash-or-nothing and asset-or-nothing binary options and compare your prices to the Black-Scholes and binomial model prices. Again, for binary options we need only modify the option payoffs in the above code. The below code does this for the cash-or-nothing options and gives $C = 0.50885$ and $P = 0.4665$.

```
1 import numpy as np
2 from scipy.stats import norm
3 S = 50; K = 50; r = 0.05; T = 1/2; sigma = 0.25; N = 1000000
4 CT = np.zeros(N); PT = np.zeros(N)
5 for i in range(N):
6     ST = S*np.exp((r - 0.5*sigma**2)*T + sigma*np.sqrt(T)*norm.rvs())
7     CT[i] = 1 if ST>K else 0
8     PT[i] = 1 if ST<K else 0
9 C = np.exp(-r*T)*np.mean(CT)
10 P = np.exp(-r*T)*np.mean(PT)
```