

0.1 Stock Options

A stock price can sometimes follow a risk-neutral distribution at time t as

$$S_t \sim N(f(S_0, t), v^2(S_0, t)t), \quad (1)$$

where S_0 is the current stock price, S_t is the stock price at t ; f and v are calibrated functions. In this problem,

$$f(S_0, t) = S_0(\alpha T + \tan(\beta T)) + \theta \cos(\alpha T + \beta T), \quad (2)$$

$$v(S_0, t) = \frac{1}{2} \sigma (1 + \alpha T) (S_0 + \theta)^\gamma, \quad (3)$$

where we have parameters $\alpha=0.02$, $\beta = 0.02$, $\theta = 70$, $\gamma = 1.03$, $T = 1$, the time at option maturity in years, $\sigma = 0.19$, the volatility.

It can be shown that a European put option then has analytic solution

$$P(S_0, t = 0) = \left[XN(z) + v(S_0, T) \sqrt{\frac{T}{2\pi}} e^{-z^2/2} - f(S_0, T)N(z) \right] e^{-rT}, \quad (4)$$

where X is the strike price and r is the risk-free interest rate. $N(z)$ is the standard cumulative normal distribution for variable

$$z = \frac{X - f(S_0, T)}{v(S_0, T)\sqrt{T}}. \quad (5)$$

To carry out a Monte Carlo valuation of this option we use samples from a standard normal distribution, ϕ , to write the i -th stock path as

$$S_T^i = f(S_0, T) + v(S_0, T)\sqrt{T}\phi_i. \quad (6)$$

We can then average over n approximations of the discounted payoff to get the value of the put option:

$$P(S_0, t = 0) \approx e^{-rT} \frac{1}{n} \sum_{i=1}^n \max(X - S_T^i, 0). \quad (7)$$

0.2 Asian Option

Stock prices can also be viewed as following a risk neutral stochastic process with SDE

$$dS = f'(S, t)dt + v'(S, t)dW, \quad (8)$$

where W is a Wiener process and f' and v' are calibrated functions. In this problem,

$$f'(S, t) = \alpha\theta - \beta S, \quad (9)$$

$$v'(S, t) = \sigma(|S|)^\gamma dW. \quad (10)$$

These options are path dependent and so the pricing depends on the path taken. We can approximate this by taking K steps to maturity time T with a respective time-step of Δt . To go from one point to the next we can use an Euler type scheme to write

$$S^i(t_k) = S^i(t_{k-1}) + f(S^i(t_{k-1}), t_{k-1})\Delta t + v(S^i(t_{k-1}), t_{k-1})\sqrt{\Delta t}\phi_{i,k-1} \quad (11)$$

for t_k the k -th time-step. In this assignment a value of 70 was used for K .

For a discrete floating-strike Asian put option, the payoff function is

$$G(S, A) = \max(A - S(t_K), 0), \quad (12)$$

$$A = \frac{1}{K} \sum_{k=1}^K S(t_k). \quad (13)$$

As before, to then carry out a Monte Carlo valuation we average over n approximations of the discounted payoff:

$$P(S_0, t = 0) \approx e^{-rT} \frac{1}{n} \sum_{i=1}^n G(S^i, A). \quad (14)$$

0.3 Error

The Central Limit Theorem states that if $V(S_i^T)$ is a sequence of independent and identically distributed random variables with mean $E_t^Q[V(S_T)]$ and variance η^2 then we can say that

$$\lim_{n \rightarrow \infty} \sqrt{n} \left(\frac{1}{n} \sum_{i=1}^n V(S_i^T) - E_t^Q[V(S_T)] \right) = N(0, \eta^2). \quad (15)$$

Therefore, each estimate of a solution V_N with a Monte-Carlo simulation is such that

$$V_N \sim N(V^*, \eta^2), \quad (16)$$

where V^* is the true solution. Then if we choose to take a sample mean from a distribution \bar{V} of M samples of V_N we have that

$$\bar{V} \sim N(V^*, \frac{\eta^2}{M}). \quad (17)$$

To then obtain an estimate of the error on our approximation we find the variance of our M samples, divide this by M and then square root. This was taken to be the error on all succeeding results unless otherwise specified. It is also equivalent to the standard deviation here.