

HOMWORK 5

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To derive up-and-in-put option, $(K - S_T)^+ 1_{M_T^S \geq H}$, we need to find the joint distribution, $P(Z_T \leq x, M_T^S \geq y)$. Before we find the joint distribution of Z_T with drift, we find the joint distribution of W_T first. Then we use Girsanov's Theorem to find the Z_T with drift.

$$\begin{aligned} P(W_T \leq x, M_T^S \geq y) &= P(W_T \leq x | M_T^S \geq y) P(M_T^S \geq y) \\ &= P(W_T \geq 2y - x | M_T^S \geq y) P(M_T^S \geq y) \\ &= P(M_T^S \geq y | W_T \geq 2y - x) P(W_T \geq 2y - x) \\ &= P(W_T \geq 2y - x) \end{aligned}$$

Let event A be:

$$P(Z_T \leq x, M_T^S \geq y)$$

$$Z_t = \mu t + \sigma W_t$$

And let $Z_t = \sigma B_t$, $B_t = \mu t + W_t$, and $\mu = \frac{v}{\sigma}$

The probability A can be computed as:

$$\begin{aligned} P(A) &= E^P(1_A) \\ &= E^P(1_{\{Z_T \leq x, M_T^S \geq y\}}) \\ &= E^Q\left(1_{\{Z_T \leq x, M_T^S \geq y\}} \frac{dP}{dQ}\right) \text{ as } \frac{dP_t}{dQ_t} = e^{\mu B_t - \frac{1}{2}\mu^2 t} \\ &= E^Q\left(1_{\{Z_T \leq x, M_T^S \geq y\}} \frac{dP}{dQ}\right) \\ &= E^Q\left(1_{\{Z_T \leq x, M_T^S \geq y\}} e^{\mu B_t - \frac{1}{2}\mu^2 t}\right) \\ &= E^Q\left(1_{\{Z_T \leq x, M_T^S \geq y\}} e^{\frac{v}{\sigma} Z_t - \frac{1}{2}\mu^2 t}\right) \\ &= E^Q(1_{\{Z_T \geq 2y - x\}}) e^{\left(\frac{v}{\sigma}(2y - Z_t) - \frac{1}{2}\mu^2 t\right)} \\ &= e^{\left(\frac{2vy}{\sigma^2}\right)} E^Q(1_{\{Z_T \geq 2y - x\}}) e^{\left(-\mu B_t - \frac{1}{2}\mu^2 t\right)} \end{aligned}$$

We can regard the $e^{\left(-\mu B_t - \frac{1}{2}\mu^2 t\right)}$ as the Radon-Nikodym derivative which changes measure from Q to measure S .

$X_t = \mu t + B_t$ is a Brownian motion under S measure.

$$\frac{dS_t}{dQ_t} = e^{\left(-\mu B_t - \frac{1}{2}\mu^2 t\right)}$$

So, we now change measure from Q to S measure,

$$\begin{aligned} P(A) &= e^{\left(\frac{2vy}{\sigma^2}\right)} E^Q(1_{\{Z_T \geq 2y - x\}}) e^{\left(-\mu B_t - \frac{1}{2}\mu^2 t\right)} \\ &= e^{\left(\frac{2vy}{\sigma^2}\right)} E^Q(1_{\{Z_T \geq 2y - x\}}) \frac{dS_t}{dQ_t} \end{aligned}$$

$$\begin{aligned}
&= e^{\left(\frac{2vy}{\sigma^2}\right)} E^S(1_{\{Z_T \geq 2y-x\}}) \\
&= e^{\left(\frac{2vy}{\sigma^2}\right)} E^S(1_{\{\sigma B_t \geq 2y-x\}}) \\
&= e^{\left(\frac{2vy}{\sigma^2}\right)} E^S(1_{\{\sigma(X_t - \mu t) \geq 2y-x\}}) \\
&= e^{\left(\frac{2vy}{\sigma^2}\right)} E^S(1_{\{\sigma X_t \geq 2y-x+vt\}}) \\
&= e^{\left(\frac{2vy}{\sigma^2}\right)} E^S(1_{\{\sigma X_t \geq 2y-x+vt\}}) \\
&= e^{\left(\frac{2vy}{\sigma^2}\right)} \left(1 - \Phi\left(\frac{2y-x+vt}{\sigma\sqrt{t}}\right)\right) \\
&= e^{\left(\frac{2vy}{\sigma^2}\right)} \left(\Phi\left(\frac{-2y+x-vt}{\sigma\sqrt{t}}\right)\right)
\end{aligned}$$

Now, we can start to price our up-and-input option.

$$\begin{aligned}
UIP &= N_0 E_0 \left[\frac{(K - S_T)^+}{N_T} 1_{M_T^S \geq H} \right] \\
&= N_0 E_0 \left[\frac{(K - S_T)}{N_T} 1_{S_T \leq K, M_T^S \geq H} \right] \\
&= N_0 E_0 \left[\frac{(K - S_T)}{N_T} 1_{S_T \leq K, M_T^S \geq H} \right] \\
&= N_0 E_0 \left[\frac{K}{N_T} 1_{S_T \leq K, M_T^S \geq H} \right] - N_0 E_0 \left[\frac{S_T}{N_T} 1_{S_T \leq K, M_T^S \geq H} \right] \\
&= e^{-rT} K E_0^\beta \left[1_{S_T \leq K, M_T^S \geq H} \right] - S_0 e^{-qT} E_0^S \left[1_{S_T \leq K, M_T^S \geq H} \right]
\end{aligned}$$

we set $V_0^1 = e^{-rT} K E_0^\beta \left[1_{S_T \leq K, M_T^S \geq H} \right]$, we choose $N_t = \beta_t = e^{rt}$

we set $V_0^2 = S_0 e^{-qT} E_0^S \left[1_{S_T \leq K, M_T^S \geq H} \right]$, we choose $N_t = S_t e^{qt}$

We set $Z_t = \ln\left(\frac{S_t}{S_0}\right)$, $x = \ln\left(\frac{K}{S_0}\right)$, $y = \ln\left(\frac{H}{S_0}\right)$, $v = \mu - \frac{1}{2}\sigma^2$, hence, we have:

$$P(Z_T \leq x, M_T^S \geq y) = P(S_T \leq K, M_T^S \geq H)$$

$$\begin{aligned}
V_0^1 &= e^{-rT} K E_0^\beta \left[1_{S_T \leq K, M_T^S \geq H} \right] \\
&= e^{-rT} K P(Z_T \leq x, M_T^S \geq y) \\
&= K e^{-rT} \left(e^{\left(\frac{2vy}{\sigma^2}\right)} \left(\Phi\left(\frac{-2y+x-vt}{\sigma\sqrt{t}}\right) \right) \right) \\
&= K e^{-rT} \left(\frac{H}{S_0} \right)^{\frac{2\mu}{\sigma^2}-1} \Phi\left(\frac{\ln\left(\frac{S_0 K}{H^2}\right) - \left(\mu - \frac{1}{2}\sigma^2\right)T}{\sigma\sqrt{T}}\right)
\end{aligned}$$

for V_0^2 , under S measure, $v = \mu + \frac{1}{2}\sigma^2$, we have:

$$\begin{aligned}
 V_0^2 &= S_0 e^{-qT} E_0^S \left[1_{S_T \leq K, M_T^S \geq H} \right] \\
 &= S_0 e^{-qT} P(Z_T \leq x, M_T^S \geq y) \\
 &= S_0 e^{-qT} \left(e^{\left(\frac{2vy}{\sigma^2}\right)} \left(\Phi \left(\frac{-2y + x - vt}{\sigma\sqrt{t}} \right) \right) \right) \\
 &= S_0 e^{-qT} \left(\frac{H}{S_0} \right)^{\frac{2\mu}{\sigma^2}+1} \Phi \left(\frac{\ln \left(\frac{S_0 K}{H^2} \right) - \left(\mu + \frac{1}{2} \sigma^2 \right) T}{\sigma\sqrt{T}} \right)
 \end{aligned}$$

Put it all together, the price of up-and-in-put is:

$$\begin{aligned}
 V_0 &= V_0^1 + V_0^2 \\
 &= K e^{-rT} \left(\frac{H}{S_0} \right)^{\frac{2\mu}{\sigma^2}-1} \Phi \left(\frac{\ln \left(\frac{S_0 K}{H^2} \right) - \left(\mu - \frac{1}{2} \sigma^2 \right) T}{\sigma\sqrt{T}} \right) \\
 &\quad - S_0 e^{-qT} \left(\frac{H}{S_0} \right)^{\frac{2\mu}{\sigma^2}+1} \Phi \left(\frac{\ln \left(\frac{S_0 K}{H^2} \right) - \left(\mu + \frac{1}{2} \sigma^2 \right) T}{\sigma\sqrt{T}} \right)
 \end{aligned}$$