

# Trinomial Model Construction

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## Hedging in a 1-Period Trinomial Model

We consider a 1-period model where the underlying has an initial value  $S_0 = S$  and can take 3 possible (future) values at time  $T$  that we note and with probability  $P(S_T = S_i) = p_i$  and where we assume that  $S_1 < S_2 < Se^{rT} < S_3$ ,  $r$  being the (continuous) interest rate from 0 to  $T$ .

## 1 Basic Model Considerations (10 pts)

- (a) Is the Model arbitrage free?
- (b) How many martingale measures (i.e. risk neutral measure) are there?
- (c) Is the market complete?

**Solution:**

$$\begin{cases} p_1 + p_2 + p_3 = 1 \\ p_1 S_1 + p_2 S_2 + p_3 S_3 = S \cdot e^{rT} \end{cases}$$

This is a case that 2 equations contain 3 variables  $(p_1, p_2, p_3)$ . Therefore, there are infinite sets of solutions  $(p_1, p_2, p_3)$ . So,

- (a) The model is arbitrage free.
- (b) There are infinite sets of martingale measures.
- (c) The market is not complete.

## 2 Mean-Variance Hedging (50 pts)

(a) Let's imagine you sold a derivatives with (generic) payoff  $g(S_T)$  that you will hedge with a self-financing portfolio with initial endowment  $p$  and quantity of underlying  $\Delta$ . We note  $P_T = P_T(p, \Delta)$  the value of the hedging portfolio at time  $T$ . To hedge our option, we want to minimize

$$\varepsilon(p, \Delta) = E \left[ (P_T - g(S_T))^2 \right]$$

that is, we want to find the initial portfolio value and delta-hedging strategy that minimizes our quadratic risk.

- i. What is the expression  $P_T$  of the self-financing portfolio at time  $T$  ?
- ii. What are the optimal  $(p, \Delta)$  that minimize  $\varepsilon(p, \Delta)$  ? (give explicit mathematical formulae)

iii. What is the expression of residual error, namely  $\varepsilon(p, \Delta)$  for the optimal solution?

**Solution:**

i. Given that the portfolio is self-financing,

$$\begin{aligned} P_0 &= p = \Delta S + B_0 \Rightarrow B_0 = p - \Delta S \\ P_T &= \Delta S_T + B_0 e^{rT} = \Delta S_T + (p - \Delta S) e^{rT} \end{aligned}$$

ii. The  $\varepsilon(p, \Delta)$  expression is:

$$\begin{aligned} \varepsilon(p, \Delta) &= E \left[ (p_T - g(S_T))^2 \right] = \sum_{i=1}^3 p_i (\Delta S_i + (p - \Delta S) e^{rT} - g(S_i))^2 \\ &= \sum_{i=1}^3 p_i ((S_i - S e^{rT}) \Delta + e^{rT} p - g(S_i))^2 \end{aligned}$$

The  $\frac{\partial \varepsilon(p, \Delta)}{\partial p}$  expression is:

$$\begin{aligned} \frac{\partial \varepsilon(p, \Delta)}{\partial p} &= 2e^{rT} \sum_{i=1}^3 p_i ((S_i - S e^{rT}) \Delta + e^{rT} p - g(S_i)) \\ &= 2e^{rT} \sum_{i=1}^3 p_i (S_i - S e^{rT}) \cdot \Delta + 2e^{2rT} \sum_{i=1}^3 p_i \cdot p - 2e^{rT} \sum_{i=1}^3 p_i g(S_i) \\ &= 2e^{rT} \left[ \sum_{i=1}^3 p_i (S_i - S e^{rT}) \cdot \Delta + e^{rT} \cdot p - \sum_{i=1}^3 p_i g(S_i) \right] \end{aligned}$$

The  $\frac{\partial \varepsilon(p, \Delta)}{\partial \Delta}$  expression is:

$$\begin{aligned} \frac{\partial \varepsilon(p, \Delta)}{\partial \Delta} &= 2 \sum_{i=1}^3 p_i (S_i - S e^{rT}) ((S_i - S e^{rT}) \Delta + e^{rT} p - g(S_i)) \\ &= 2 \sum_{i=1}^3 p_i (S_i - S e^{rT})^2 \Delta + 2e^{rT} \sum_{i=1}^3 p_i (S_i - S e^{rT}) \cdot p - 2 \sum_{i=1}^3 p_i (S_i - S e^{rT}) g(S_i) \end{aligned}$$

In order to find the minimum, the necessary condition is to Let  $\frac{\partial \varepsilon(p, \Delta)}{\partial p} = \frac{\partial \varepsilon(p, \Delta)}{\partial \Delta} = 0$ , and get

The  $\Delta$  is:

$$\begin{aligned} \Delta &= \frac{-2e^{rT} \sum p_i (S_i - S e^{rT}) g(S_i) + \sum p_i g(S_i) \cdot 2e^{rT} \sum p_i (S_i - S e^{rT})}{\sum p_i (S_i - S e^{rT}) \cdot 2e^{rT} \sum p_i (S_i - S e^{rT}) - 2e^{rT} \sum p_i (S_i - S e^{rT})^2} \\ &= \frac{E(g(S_T)) \cdot E(S_T) - E(g(S_T) \cdot S_T)}{E(S_T - S e^{rT})^2 - E^2(S_T - S e^{rT})} = \frac{Cov(g(S_T), S_T)}{Var(S_T - S e^{rT})} = \frac{Cov(g(S_T), S_T)}{Var(S_T)} \end{aligned}$$

The  $p$  is:

$$\begin{aligned} p &= e^{-rT} \left( \sum_{i=1}^3 P_i g(S_i) - \sum_{i=1}^3 P_i (S_i - Se^{rT}) \cdot \frac{\text{Cov}(g(S_T), S_T)}{\text{Var}(S_T - Se^{rT})} \right) \\ &= e^{-rT} \left( E(g(S_T)) - E(S_T - Se^{rT}) \cdot \frac{\text{Cov}(g(S_T), S_T)}{\text{Var}(S_T)} \right) \end{aligned}$$

iii. By plugging the above  $\Delta$  and  $p$  in the expression of  $\varepsilon(p, \Delta)$ , the  $\varepsilon(p, \Delta)$  is:

$$\begin{aligned} \varepsilon(p, \Delta) &= \sum_{i=1}^3 p_i ((S_i - Se^{rT}) \Delta + e^{rT} p - g(S_i))^2 \\ &= \sum P_i ((S_i - Se^{rT}) \cdot \Delta + E(g(S_T)) - E(S_T - Se^{rT}) \Delta - g(S_i))^2 \\ &= \text{Var}(g(T)) - 2\Delta \text{Cov}(g(S_T), S_T) + \Delta^2 \text{Var}(S_T - Se^{rT}) \\ &= \text{Var}(g(T)) - \frac{2\text{Cov}^2(g(S_T), S_T)}{\text{Var}(S_T)} + \frac{\text{Cov}^2(g(S_T), S_T)}{\text{Var}(S_T)} \\ &= \text{Var}(g(T)) - \frac{\text{Cov}^2(g(S_T), S_T)}{\text{Var}(S_T)} \end{aligned}$$

(b) Find a martingale measure  $Q = (q_1, q_2, q_3)$  which does not depend on  $g$  (at all) such that

$$p = E^Q(g(S_T) e^{-rT}) = \sum_i g(S_i) e^{-rT} q_i$$

**Solution:**

$$\begin{aligned} p &= e^{-rT} \left( E(g(S_T)) - E(S_T - Se^{rT}) \cdot \frac{\text{Cov}(g(S_T), S_T)}{\text{Var}(S_T)} \right) \\ &= e^{-rT} \left( E(g(S_T)) - E(S_T - Se^{rT}) \frac{\text{Cov}(g(S_T), S_T - e^{rT} S)}{\text{Var}(S_T)} \right) \\ &= e^{-rT} \left( E(g(S_T)) - E(S_T - Se^{rT}) \cdot \frac{E[g(S_T) \cdot (S_T - e^{rT} S)] - E(g(S_T)) \cdot E(S_T - e^{rT} S)}{\text{Var}(S_T)} \right) \\ &= e^{-rT} \left( \sum_{i=1}^3 p_i g(S_i) - E(S_T - Se^{rT}) \frac{\sum_{i=1}^3 p_i g(S_i) (S_i - e^{rT} S) - \sum_{i=1}^3 p_i g(S_i) \cdot E(S_T - e^{rT} S)}{\text{Var}(S_T)} \right) \\ &= e^{-rT} \left( \sum_{i=1}^3 p_i g(S_i) - E(S_T - Se^{rT}) \frac{\sum_{i=1}^3 p_i g(S_i) [S_i - e^{rT} S - E(S_T - e^{rT} S)]}{\text{Var}(S_T)} \right) \\ &= e^{-rT} \sum_{i=1}^3 p_i g(S_i) \left[ 1 - E(S_T - Se^{rT}) \cdot \frac{S_i - E(S_T)}{\text{Var}(S_T)} \right] \end{aligned}$$

Meanwhile,

$$p = \sum_{i=1}^3 g(S_i) e^{-rT} q_i$$

Therefore,

$$q_i = p_i \left[ 1 - E(S_T - Se^{rT}) \cdot \frac{S_i - E(S_T)}{\text{Var}(S_T)} \right]$$

(c) Show that  $\sum_i q_i = 1$  and that.

**Solution:**

$$\begin{aligned} \sum_{i=1}^3 q_i &= \sum_{i=1}^3 p_i \left[ 1 - E(S_T - Se^{rT}) \cdot \frac{S_i - E(S_T)}{\text{Var}(S_T)} \right] \\ &= \sum_{i=1}^3 P_i - E(S_T - Se^{rT}) \frac{\sum_{i=1}^3 P_i S_i - E(S_T) \cdot \sum_{i=1}^3 P_i}{\text{Var}(S_T)} \\ &= 1 - E(S_T - Se^{rT}) \frac{E(S_T) - E(S_T) \times 1}{\text{Var}(S_T)} = 1 \end{aligned}$$

(d) Compute  $E^Q(S_T)$ . What does it mean?

**Solution:**

$$\begin{aligned} E^Q(S_T) &= \sum_{i=1}^3 q_i S_i = \sum_{i=1}^3 p_i S_i \left[ 1 - E(S_T - Se^{rT}) \cdot \frac{S_i - E(S_T)}{\text{Var}(S_T)} \right] \\ &= \sum_{i=1}^3 p_i S_i - E(S_T - Se^{rT}) \frac{\sum_{i=1}^3 p_i S_i^2 - E(S_T) \cdot \sum_{i=1}^3 p_i S_i}{\text{Var}(S_T)} \\ &= \sum_{i=1}^3 p_i S_i - E(S_T - Se^{rT}) \frac{E(S_T^2) - E^2(S_T)}{\text{Var}(S_T)} \\ &= \sum_{i=1}^3 p_i S_i - E(S_T - Se^{rT}) \frac{\text{Var}(S_T)}{\text{Var}(S_T)} \\ &= E(S_T) - (E(S_T) - Se^{rT}) = Se^{rT} \end{aligned}$$

It indicates that  $Q = (q_1, q_2, q_3)$  is a martingale measure both for  $S$  (the price of the underlying asset) and  $g(S)$  (the payoff of derivatives).

(e) Show that  $0 < q_i < 1$  for all  $i$ .

**Solution:**

$$\forall i, \quad S_i - E(S_T) < \frac{Var(S_T - S_{e^{rT}})}{E(S_T - S_{e^{rT}})}$$

Therefore,

$$q_i = p_i \left[ 1 - E(S_T - S_e^{rT}) \cdot \frac{S_i - E(S_T)}{Var(S_T)} \right] > 0$$

Plus, in (d) I prove:

$$\sum_{i=1}^3 q_i = 1$$

Therefore,  $\forall i, 0 < q_i < 1$

### 3 Comparison with Binomial Tree (60 pts)

We set  $S = 100, S_1 = 92, S_2 = 98, S_3 = 105, r = 0$  and  $P(S_T = S_i) = p_i$  with  $(p_1, p_2, p_3) = (1/3, 1/4, 5/12)$  and assume that we sold a put option with strike  $K = 100$ .

(a) What are the values of  $p$  and  $\Delta$  in this specific trinomial model?

**Solution:**

$$\begin{aligned} S &= K = 100 \\ S_1 &= 92, \quad p_1 = \frac{1}{3}, g(S_1) = (K - S_1)^+ = 8 \\ S_2 &= 98, \quad p_2 = \frac{1}{4}, g(S_2) = (K - S_2)^+ = 2 \\ S_3 &= 105, \quad p_3 = \frac{5}{12}, g(S_3) = (K - S_3)^+ = 0 \end{aligned}$$

According to discussions and solutions in Task 2,

$$\begin{aligned} &\begin{cases} \frac{\partial \varepsilon(p, \Delta)}{\partial p} = 0 \\ \frac{\partial \varepsilon(p, \Delta)}{\partial \Delta} = 0 \end{cases} \\ \Rightarrow &\begin{cases} \frac{1}{3}(-8\Delta + p - 8) + \frac{1}{4}(-2\Delta + p - 2) + \frac{5}{12}(5\Delta + p - 0) = 0 \\ -\frac{8}{3}(-8\Delta + p - 8) - \frac{2}{4}(-2\Delta + p - 2) + \frac{25}{12}(5\Delta + p - 0) = 0 \end{cases} \\ \Rightarrow &\begin{cases} \Delta = -\frac{2722}{4547} \approx -0.5986 \\ p = \frac{11450}{4547} \approx 2.5181 \end{cases} \end{aligned}$$

(b) Because  $S_2 < S e^{rT} < S_3$ , we think of keeping  $S_3 = 105$  the same and look for a value  $\tilde{S}_1$  such that  $\Delta_{bin} = \Delta$  and the binomial model would have two

possible outcomes, namely  $\tilde{S}_1$  that needs to be computed and  $S_3 = 105$ .

- i. Explain why there is a solution  $\tilde{S}_1$  such that  $\Delta_{bin} = \Delta$  and how to compute it.
- ii. What is the value of  $p_{bin}$  ?

**Solution:**

- i. For the equal-delta binomial model, there is:

$$\begin{cases} g(\tilde{S}_1) = \Delta_{bin}\tilde{S}_1 + (P_{bin} - \Delta_{bin}S) \cdot e^{rT} \\ g(S_3) = \Delta_{bin}S_3 + (P_{bin} - \Delta_{bin} \cdot S) e^{rT} \end{cases}$$

There are 2 equations and 2 variables  $\tilde{S}_1$  and  $\Delta_{bin}$ , so there must be a solution  $(\tilde{S}_1, \Delta_{bin})$ , whose values are shown later.

$$\Rightarrow \Delta_{bin} = \frac{g(\tilde{S}_1) - g(S_3)}{\tilde{S}_1 - S_3}$$

Plus, given that  $\begin{cases} \Delta_{bin} = \Delta \\ g(\tilde{S}_1) = (K - \tilde{S}_1)^+ = K - \tilde{S}_1 \end{cases}$

where the second equation is because there must be  $\tilde{S}_1 < K < S_3$  in the binomial model.

Plugging these conditions in the equations above, we get:

$$\tilde{S}_1 = \frac{33778}{365} \approx 92.5425$$

- ii. It is aforementioned in i. that

$$g(S_3) = \Delta_{bin}S_3 + (P_{bin} - \Delta_{bin} \cdot S) e^{rT}$$

Therefore,

$$P_{bin} = e^{-0T} \left( 0 + \frac{2722}{4547} \cdot 105 \right) - \frac{2722}{4547} \cdot 100 = \frac{13610}{4547} \approx 2.9932$$

(c) We assume that in reality, the stock realized at 92, that is  $S_T = 92$ , which was a possible value in the trinomial model. The P&L of the hedge position is the difference between the value of the hedging portfolio and the derivatives payoff (we are supposed to pay to our client), that is  $P\&L = P_T - (K - S_T)^+$ .

- i. What if the P&L of the hedged position if we used the binomial model?
- ii. What if the P&L of the hedged position if we used the trinomial model?
- iii. What do you conclude?

**Solution:**

i.

$$\begin{aligned}
P\&L_{bin} &= \Delta_{bin}S_T + (P_{bin} - \Delta_{bin}S) \cdot e^{rT} - (K - S_T)^+ \\
&= -\frac{2722}{4547} \times 92 + \left( \frac{13610}{4547} + \frac{2722}{4547} \times 100 \right) \cdot e^0 - (100 - 92)^+ \\
&= -\frac{990}{4547} \approx -0.2177
\end{aligned}$$

ii.

$$\begin{aligned}
P\&L_{tri} &= \Delta S_T + (P - \Delta S) \cdot e^{rT} - (K - S_T)^+ \\
&= -\frac{2722}{4547} \times 92 + \left( \frac{11450}{4547} + \frac{2722}{4547} \times 100 \right) \cdot e^0 - (100 - 92)^+ \\
&= -\frac{3150}{4547} \approx -0.6928
\end{aligned}$$

iii. When  $S_T = 92$ , the equal-delta binomial model makes fewer loss than trinomial model.

(d) We now consider "out-of-sample" scenarii for the first model:

i. What if the underlying drops even more, namely  $S_T = 90$ : what is the PL from using each model?

ii. What if the underlying goes up to  $S_T = 105$ : what is the PL from using each model?

**Solution:**

i. Similar to (c) above, I can get:

$$\begin{aligned}
P\&L_{bin} &= \Delta_{bin}S_T + (P_{bin} - \Delta_{bin}S) \cdot e^{rT} - (K - S_T)^+ \\
&= -\frac{2722}{4547} \times 90 + \left( \frac{13610}{4547} + \frac{2722}{4547} \times 100 \right) \cdot e^0 - (100 - 90)^+ \\
&= -\frac{4640}{4547} \approx -1.0204
\end{aligned}$$

$$\begin{aligned}
P\&L_{tri} &= \Delta S_T + (P - \Delta S) \cdot e^{rT} - (K - S_T)^+ \\
&= -\frac{2722}{4547} \times 90 + \left( \frac{11450}{4547} + \frac{2722}{4547} \times 100 \right) \cdot e^0 - (100 - 90)^+ \\
&= -\frac{6800}{4547} \approx -1.4955
\end{aligned}$$

ii. Similarly:

$$P\&L_{bin} = \Delta_{bin}S_T + (P_{bin} - \Delta_{bin}S) \cdot e^{rT} - (K - S_T)^+$$

$$\begin{aligned}
&= -\frac{2722}{4547} \times 105 + \left( \frac{13610}{4547} + \frac{2722}{4547} \times 100 \right) \cdot e^0 - (100 - 105)^+ \\
&= 0
\end{aligned}$$

This is because I use the  $P\&L_{bin} = 0$  when  $S_T = 105$  when constructing the binomial model.

$$\begin{aligned}
P\&L_{tri} &= \Delta S_T + (P - \Delta S) \cdot e^{rT} - (K - S_T)^+ \\
&= -\frac{2722}{4547} \times 105 + \left( \frac{11450}{4547} + \frac{2722}{4547} \times 100 \right) \cdot e^0 - (100 - 105)^+ \\
&= -\frac{2160}{4547} \approx -0.4750
\end{aligned}$$

- (e) What if instead we choose  $\tilde{S}_1$  so that  $p_{bin} = p$ .
- How to compute the solution for such  $\tilde{S}_1$  in general and what is its numerical value in this specific example?
  - What is the corresponding  $\Delta_{bin}$ ?
  - Compare the P&L (in a table) of this model for each previous senarii against the trinomial model.

**Solution:**

- i. For the equal-p binomial model, there is:

$$\begin{cases} g(\tilde{S}_1) = \Delta_{bin} \tilde{S}_1 + (P_{bin} - \Delta_{bin} S) \cdot e^{rT} \\ g(S_3) = \Delta_{bin} S_3 + (P_{bin} - \Delta_{bin} \cdot S) e^{rT} \end{cases}$$

$$\Rightarrow \Delta_{bin} = \frac{g(\tilde{S}_1) - g(S_3)}{\tilde{S}_1 - S_3} = \frac{g(\tilde{S}_1) - P_{bin} \cdot e^{rT}}{\tilde{S}_1 - S e^{rT}}$$

Plus, given that  $\begin{cases} P_{bin} = P \\ g(\tilde{S}_1) = (K - \tilde{S}_1)^+ = K - \tilde{S}_1 \end{cases}$

where the second equation is because there must be  $\tilde{S}_1 < K < S_3$  in the binomial model.

Plugging these conditions in the equations above, we get:

$$\frac{100 - \tilde{S}_1 - \frac{11450}{4547} e^{0T}}{\tilde{S}_1 - 100 \cdot e^{0T}} = \frac{100 - \tilde{S}_1 - 0}{\tilde{S}_1 - 105}$$

Solving this function, we get:

$$\tilde{S}_1 = \frac{214250}{2257} \approx 94.9269$$



ii. According to the expression of  $\Delta_{bin}$  in i.,

$$\Delta_{bin} = \frac{g(\tilde{S}_1) - g(S_3)}{\tilde{S}_1 - S_3} = \frac{100 - \frac{214250}{2257} - 0}{\frac{214250}{2257} - 105} = -\frac{2290}{4547} \approx 0.5036$$

iii. Similarly to (c) and (d), I can get:

When  $S_T = 92$ ,

$$\begin{aligned} P\&L_{bin} &= \Delta_{bin}S_T + (P_{bin} - \Delta_{bin}S) \cdot e^{rT} - (K - S_T)^+ \\ &= -\frac{2290}{4547} \times 92 + \left( \frac{11450}{4547} + \frac{2290}{4547} \times 100 \right) \cdot e^0 - (100 - 92)^+ \\ &= -\frac{6606}{4547} \approx -1.4528 \end{aligned}$$

$P\&L_{tri}$  is calculated in (c):

$$\begin{aligned} P\&L_{tri} &= \Delta S_T + (P - \Delta S) \cdot e^{rT} - (K - S_T)^+ \\ &= -\frac{3150}{4547} \approx -0.6928 \end{aligned}$$

When  $S_T = 90$ ,

$$\begin{aligned} P\&L_{bin} &= \Delta_{bin}S_T + (P_{bin} - \Delta_{bin}S) \cdot e^{rT} - (K - S_T)^+ \\ &= -\frac{2290}{4547} \times 90 + \left( \frac{11450}{4547} + \frac{2290}{4547} \times 100 \right) \cdot e^0 - (100 - 90)^+ \\ &= -\frac{1112}{4547} \approx -2.4456 \end{aligned}$$

$P\&L_{tri}$  is calculated in (d):

$$\begin{aligned} P\&L_{tri} &= \Delta S_T + (P - \Delta S) \cdot e^{rT} - (K - S_T)^+ \\ &= -\frac{6800}{4547} \approx -1.4955 \end{aligned}$$

When  $S_T = 105$ ,

$$\begin{aligned} P\&L_{bin} &= \Delta_{bin}S_T + (P_{bin} - \Delta_{bin}S) \cdot e^{rT} - (K - S_T)^+ \\ &= -\frac{2290}{4547} \times 105 + \left( \frac{11450}{4547} + \frac{2290}{4547} \times 100 \right) \cdot e^0 - (100 - 105)^+ \\ &= 0 \end{aligned}$$

This is because I use the  $P\&L_{bin} = 0$  when  $S_T = 105$  when constructing the binomial model.

$P\&L_{tri}$  is calculated in (d):

$$\begin{aligned} P\&L_{tri} &= \Delta S_T + (P - \Delta S) \cdot e^{rT} - (K - S_T)^+ \\ &= -\frac{2160}{4547} \approx -0.4750 \end{aligned}$$

Table 1: P&L of 3 Models

P&L		$S_T$		
		90	92	105
model	equal-delta binomial	-1.0204	-0.2177	0
	equal-p binomial	-2.4456	-1.4528	0
	trinomial	-1.4955	-0.6928	-0.4750

(f) Summarize a table the P&L of each model for all the previous scenarii (and feel free to add some, best is to code it and show a graph).

**Solution:**

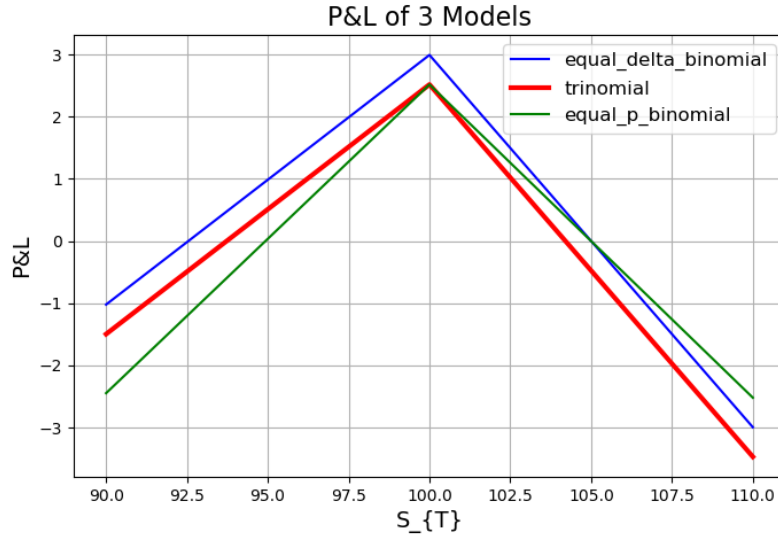


Figure 1: PNL of 3 Models

**Conclusions:**

1. The equal-delta binomial model, the equal-p binomial model, and the trinomial model reach zero P&L respectively at around (92.5, 105), (95.0, 105),

and (93.7, 104.2). These values of binomial models are exactly what I set when constructing the trees.

2. The equal-delta binomial model always have greater P&L than the trinomial model. And they have the same slope and trending. This is because in their expressions only the y-intercept containing the  $p_{bin}$  is different. The equal-delta binomial model has a higher  $p_{bin}$ .

3. When  $S_T < S = K$ , the equal-p binomial model has lower P&L than the trinomial model. However, when  $S_T > S = K$ , the equal-p binomial model has greater P&L than the trinomial model. When  $S_T = S = K$ , they reach the same P&L.

4. When  $S_T < S_3 = 105$ , the equal-p binomial model has lower P&L than the equal-delta binomial model. However, when  $S_T > S_3 = 105$ , the equal-p binomial model has greater P&L. When  $S_T = S_3 = 105$ , they reach the same P&L.