Hedging options with transaction costs

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Abstract

This paper investigates an optimal investment problem in which an investor hedges a short position in a European call option. Under a utility maximization framework, we first derive the optimal portfolio allocation without transaction costs by formulating and solving the corresponding Hamilton–Jacobi–Bellman equation. In the presence of a short call position, we extend the analysis to reveal how the indirect utility function adjusts and how the optimal strategy is consequently modified. To facilitate practical implementation, a numerical scheme based on a binomial tree is introduced, enabling the approximation of the value function and recovery of the optimal strategy in a discretized setting. We then incorporate proportional transaction costs into the model and extend the numerical method to account for the resulting variational inequality. Numerical results highlight the approach's effectiveness in capturing both the frictionless and cost-adjusted optimal strategies.

Key Words : Stochastic control, Option Hedging, Transaction costs

1 Optimal investment problem without transaction costs

1.1 Problem setting

We consider a portfolio with two assets : one risky and one non risky asset. The first one follows a geometric brownian motion :

$$dS = \alpha S dt + \sigma S dZ_t \tag{1}$$

The second is such that :

$$B_t = Be^{rt} \tag{2}$$

with B the initial amount invested in the asset. Lastly, the portfolio has a short position in an European call option on the risky asset, with strike K and maturity T, with payoff :

$$C_T = (S_T - K)^+ \tag{3}$$

The investor has a utility function u twice differentiable and concave, such that his objective is to maximise his expected utility of the terminal wealth W_T, with :

$$W_T = B_T + y_T - C_T \tag{4}$$

y_T being the cash value of his position on the risky asset at maturity (minus the transaction costs if there are any).

1.2 Optimal investment without options

There is no consumption before maturity so the portfolio is self-financing, therefore with π_t the proportion of wealth allocated in the risky asset at time t, we have :

$$\begin{aligned} dW_t &= \pi_t W_t dS_t + r(1-\pi_t)W_t dt && (5)\\ &= \pi_t W_t(\alpha S_t dt + \sigma S_t dZ_t) + (1-\pi_t) r W_t dt && (6)\\ &= (\alpha - r)\pi_t W_t dt + rW_t dt + \sigma \pi_t W_t dZ_t && (7) \end{aligned}$$

As we mentioned, the investor's goal is to maximize his expected utility at maturity, which translates as :

$$\max_\pi E(u(W_T)) \tag{8}$$

and we define the indirect function utility function at time t for a wealth W (the value function of our stochastic control problem) as :

$$U(t,W) = \max_\pi E_t(u(W_T)) \tag{9}$$

By Itô's formula, we have :

$$\begin{aligned} dU(t,W) &= \frac{\partial U}{\partial t}dt + \frac{\partial U}{\partial w}dW_t + \frac{1}{2}\frac{\partial^2 U}{\partial w^2}(dW_t)^2 && (10)\\ &= \frac{\partial U}{\partial t}dt + \frac{\partial U}{\partial w}((\alpha - r)\pi_t W_t dt + rW_t dt + \sigma\pi_t W_t dZ_t) && (11)\\ &+ \frac{1}{2}\frac{\partial^2 U}{\partial w^2}\sigma^2\pi^2 W_t^2 dt && (12) \end{aligned}$$

Therefore :

$$E(dU(t,W)) = \frac{\partial U}{\partial t}dt + \frac{\partial U}{\partial w}((\alpha - r)\pi_t W_t + rW_t)dt + \frac{1}{2}\frac{\partial^2 U}{\partial w^2}\sigma^2\pi^2 W_t^2 dt \tag{13}$$

And :

$$\max_\pi E(dU(t,W)) = \frac{\partial U}{\partial t}dt + \max_\pi\left(\frac{\partial U}{\partial w}((\alpha - r)\pi_t W_t + rW_t)dt + \frac{1}{2}\frac{\partial^2 U}{\partial w^2}\sigma^2\pi_t^2 W_t^2 dt\right) \tag{14}$$

By Bellman's principle of optimality ("If I'm already following the optimal strategy, then doing an optimal action in the next dt should give no advantage — I'm already on the best possible path."), if U is optimal for t onward, at time t the best decision we can make regarding the control π leads to $E(\frac{dU}{dt}) = 0$, which leads to the Hamilton-Jacobi-Bellman equation, dividing by dt and setting the derivative to 0 for optimality :

$$\boxed{\frac{\partial U}{\partial t} + \max_\pi L^\pi U = 0} \tag{15}$$

with the following differential operator :

$$L^\pi U = \frac{\partial U}{\partial w}((\alpha - r)\pi_t W_t + rW_t) + \frac{1}{2}\frac{\partial^2 U}{\partial w^2}\sigma^2\pi_t^2 W_t^2 \tag{16}$$

and the boundary condition :

$$U(T,W) = u(W), \forall W \geq 0 \tag{17}$$

Once we know the indirect utility function U, we can find the optimal allocation π^* using first order conditions to solve for $\max_\pi L^\pi U$:

$$\frac{\partial L^\pi U}{\partial \pi} = \pi_t W^2 \sigma^2 \frac{\partial^2 U}{\partial w^2} + (\alpha - r) W \frac{\partial U}{\partial w} \tag{18}$$

which leads to :

$$\pi^* = -\frac{(\alpha - r) U_w}{W \sigma^2 U_{ww}} \tag{19}$$

with $U_w = \frac{\partial U}{\partial w}$.
When there is no option position in the portfolio and for a utility function of the shape $u(W) = \frac{W^\gamma}{\gamma}, 0 < \gamma < 1$, the boundary condition gives $U(T, W) = \frac{W^\gamma}{\gamma}$ and then the indirect utility function can be found as :

$$U(t, W) = g(t) \frac{W^\gamma}{\gamma} \tag{20}$$

with :

$$g(t) = e^{vr(T-t)} \tag{21}$$

and

$$v = \frac{1}{2} \left(\frac{\alpha - r}{\sigma}\right)^2 \frac{1}{1-\gamma} + r \tag{22}$$

This leads to :

$$\boxed{\pi^* = -\frac{(\alpha - r) U_w}{W \sigma^2 U_{ww}} = \frac{\alpha - r}{\sigma^2 (1-\gamma)}} \tag{23}$$

1.3 When there is a short position in option

In the case where the investor has a short position in an European call option , the terminal wealth becomes :

$$\overline{W_T} = W_T - (S_T - K)^+ \tag{24}$$

With this new formulation, the indirect utility function (value function of the stochastic control problem) becomes :

$$U(t, S, W) = \max_\pi E_t(u(\overline{W_T})) \tag{25}$$

In this case, the stock price becomes a variable in the indirect utility function, which changes the computations leading to the HJB equation :

$$dU(t, S, W) = \frac{\partial U}{\partial t} dt + \frac{\partial U}{\partial w} dW_t + \frac{1}{2} \frac{\partial^2 U}{\partial w^2} (dW_t)^2 + \frac{1}{2} \frac{\partial^2 U}{\partial S^2} (dS_t)^2 + \frac{\partial^2 U}{\partial S \partial W} dS_t dW_t \tag{26}$$

We have $dS_t = \alpha S_t dt + \sigma S_t dZ_t$ and $dW_t = ((\alpha - r)\pi_t W_t + r W_t) dt + \sigma \pi_t W_t dZ_t$.
These two are correlated (same brownian motion), therefore $dS_t dW_t = \sigma^2 \pi_t S_t W_t dt$. This leads to :

$$E(dU(t, S, W)) = \left(\frac{\partial U}{\partial t} + \frac{\partial U}{\partial w}((\alpha - r)\pi_t W_t + r W_t) + \frac{\partial U}{\partial S} \alpha S_t + \frac{1}{2} \frac{\partial^2 U}{\partial w^2} \sigma^2 \pi_t^2 W_t^2 \right. \tag{27}$$

$$\left. + \frac{1}{2} \frac{\partial^2 U}{\partial S^2} \sigma^2 S_t^2 + \sigma^2 \pi_t S_t W_t \frac{\partial^2 U}{\partial S \partial W}\right) dt \tag{28}$$

This time, we define :

$$\overline{L}^\pi U = \frac{\partial U}{\partial w}((\alpha - r)\pi_t W_t + r W_t) + \frac{\partial U}{\partial S} \alpha S_t + \frac{1}{2} \frac{\partial^2 U}{\partial w^2} \sigma^2 \pi_t^2 W_t^2 + \frac{1}{2} \frac{\partial^2 U}{\partial S^2} \sigma^2 S_t^2 + \sigma^2 \pi_t S_t W_t \frac{\partial^2 U}{\partial S \partial W} \tag{29}$$

and by the same argument, this leads to the HJB equation :

$$\boxed{\frac{\partial U}{\partial t} + \max_{\pi} \overline{L}^{\pi} U = 0} \tag{30}$$

with the boundary condition :

$$U(T, S, W) = \frac{(W - (S-K)^+)^\gamma}{\gamma} \tag{31}$$

The same way as before, we can derive the optimal strategy which will depend on the form of the indirect utility function :

$$\boxed{\pi^* = -\frac{\alpha - r}{\sigma^2 W}\frac{U_w}{U_{ww}} - \frac{S}{W}\frac{U_{Sw}}{U_{ww}}} \tag{32}$$

We can prove that in this case, the indirect utility function is linked to the one when there is no position in the option, by the following relation :

$$\overline{U}(t, W) = U(t, W - f(t, S)) = g(t)\frac{(W - f(t,S))^\gamma}{\gamma} \tag{33}$$

where $f(t, S)$ is the price of the call option under the Black-Scholes framework.
This explicit form allows us to get :

$$\pi^* = \frac{\alpha - r}{\sigma^2(1-\gamma)}\frac{W - f}{W} + \frac{S}{W} f_S \tag{34}$$

$$\boxed{\pi^* W = \frac{\alpha - r}{\sigma^2(1-\gamma)}(W - f) + S f_S} \tag{35}$$

2 Numerical Scheme Without Transaction Costs Using a Binomial Tree

In practical applications, it is often difficult to derive an explicit form for the indirect utility function

$$U(t, S, W) = \max_{\pi} E_t\Big[u\Big(W_T - (S_T - K)^+\Big)\Big], \tag{36}$$

especially when additional complications such as transaction costs are present. In this section, we present a numerical scheme based on a binomial tree method that approximates the value function $U(t, S, W)$ and yields the optimal strategy by backward induction.

2.1 Overview of the Scheme

The aim is to compute, for each node on a discrete grid in time, stock price, and wealth, the optimal expected utility when starting from that node and acting optimally thereafter. In particular, the scheme will:

1. **Model the stock process:** We assume the stock follows a recombining binomial tree. Over a time step Δt, the stock price evolves as

$$S_{t+\Delta t} = \begin{cases} S_t\, u, & \text{with probability } p, \\ S_t\, d, & \text{with probability } 1 - p, \end{cases} \tag{37}$$

   where

$$u = \exp\Big(\sigma\sqrt{\Delta t}\Big), \quad d = \exp\Big(-\sigma\sqrt{\Delta t}\Big), \tag{38}$$

   and the probability p is given by

$$p = \frac{e^{\alpha\,\Delta t} - d}{u - d}. \tag{39}$$

2. **Model the wealth dynamics:** The investor allocates a fraction π_t of wealth W_t to the risky asset. For a small time step Δt, the wealth update is approximated by

$$W_{t+\Delta t} = W_t\Big(1 + r\,\Delta t + \pi_t\Big[(\alpha - r)\Delta t \pm \sigma\sqrt{\Delta t}\Big]\Big), \tag{40}$$

where the "+" corresponds to the up move (i.e., $S_{t+\Delta t} = S_t\,u$) and the "−" to the down move (i.e., $S_{t+\Delta t} = S_t\,d$).

3. **Set the terminal condition:** At maturity T, the investor's terminal wealth, is

$$\overline{W_T} = W_T - (S_T - K)^+. \tag{41}$$

Therefore, for the utility function $u(x) = \frac{x^\gamma}{\gamma}$ (with $0 < \gamma < 1$), the terminal condition is

$$U(T, S, W) = \frac{\Big(W - (S-K)^+\Big)^\gamma}{\gamma}. \tag{42}$$

4. **Dynamic Programming Recursion:** At an arbitrary node (t, S, W), the value function is computed by optimizing over the control π:

$$U(t, S, W) = \max_{\pi\in\Pi}\Big\{p\,U\Big(t+\Delta t, S\,u,\, W_{\mathrm{up}}(\pi)\Big) + (1-p)\,U\Big(t+\Delta t, S\,d,\, W_{\mathrm{down}}(\pi)\Big)\Big\}, \tag{43}$$

where the wealth updates for the up and down moves are:

$$W_{\mathrm{up}}(\pi) = W\Big(1 + r\,\Delta t + \pi\Big[(\alpha - r)\Delta t + \sigma\sqrt{\Delta t}\Big]\Big), \tag{44}$$

$$W_{\mathrm{down}}(\pi) = W\Big(1 + r\,\Delta t + \pi\Big[(\alpha - r)\Delta t - \sigma\sqrt{\Delta t}\Big]\Big). \tag{45}$$

5. **Backward Induction:** Starting from the terminal condition at $t = T$, we recursively compute the value function at each earlier time step. At every node, a discrete set of control values $\pi \in \Pi$ is tested. For each candidate π, the next-step wealth levels are computed and the corresponding U-values are interpolated (if necessary) on the wealth grid, since W_{up} and W_{down} might not belong in the wealth grid used to compute the U-values at the next step. The control that maximizes the expected utility is chosen and the value function is updated accordingly.

2.2 Detailed Algorithm

We now summarize the steps of the algorithm:

1. **Grid Setup:** Discretize the time interval $[0, T]$ into N_t steps with $\Delta t = T/N_t$, the stock price over a range $[S_{\min}, S_{\max}]$, and the wealth over a grid $[W_{\min}, W_{\max}]$.

2. **Terminal Condition:** For each terminal node at time T, where the stock price is

$$S_T = S_0\,u^i\,d^{N_t - i} \quad (i = 0, 1, \ldots, N_t), \tag{46}$$

and for each wealth value W, set

$$U(T, S_T, W) = \frac{\Big(W - (S_T - K)^+\Big)^\gamma}{\gamma}. \tag{47}$$

3. **Backward Induction:** For $n = N_t - 1, N_t - 2, \ldots, 0$ and for each node (characterized by the number of up moves i, with stock price $S = S_0\,u^i\,d^{n-i}$):

   (a) For each wealth grid point W, loop over the discrete set of candidate controls $\pi \in \Pi$.

(b) Compute the next-period wealth in both the up and down scenarios:

$$W_{\mathrm{up}}(\pi) = W\Big(1 + r\,\Delta t + \pi\Big[(\alpha - r)\Delta t + \sigma\sqrt{\Delta t}\Big]\Big), \tag{48}$$

$$W_{\mathrm{down}}(\pi) = W\Big(1 + r\,\Delta t + \pi\Big[(\alpha - r)\Delta t - \sigma\sqrt{\Delta t}\Big]\Big). \tag{49}$$

(c) For the up move, the next node is $(t + \Delta t, S\,u)$; for the down move, it is $(t + \Delta t, S\,d)$. Use linear interpolation on the wealth grid to approximate the value function at these nodes, i.e.,

$$U\Big(t + \Delta t, S\,u, W_{\mathrm{up}}(\pi)\Big) \quad \text{and} \quad U\Big(t + \Delta t, S\,d, W_{\mathrm{down}}(\pi)\Big). \tag{50}$$

(d) Compute the expected utility for the control π:

$$V(\pi) = p\,U\Big(t + \Delta t, S\,u, W_{\mathrm{up}}(\pi)\Big) + (1 - p)\,U\Big(t + \Delta t, S\,d, W_{\mathrm{down}}(\pi)\Big). \tag{51}$$

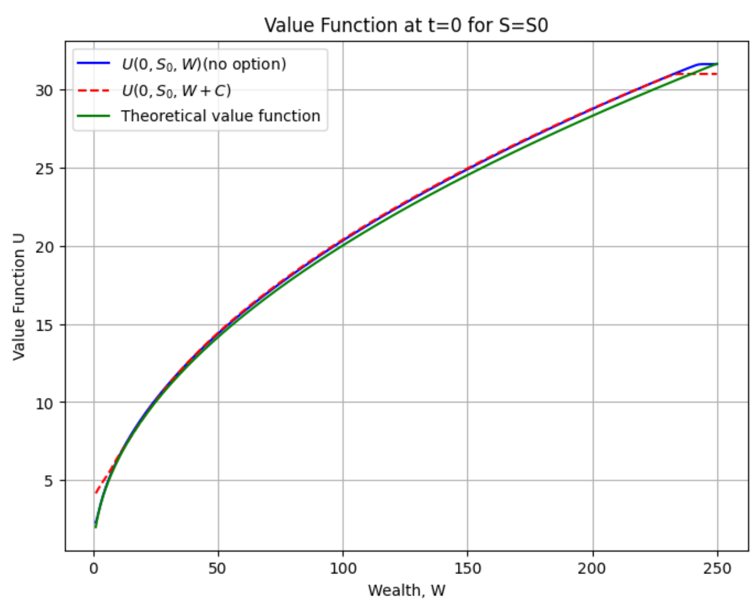
(e) Choose the control π^* that maximizes $V(\pi)$, and set

$$U(t, S, W) = \max_{\pi \in \Pi} V(\pi). \tag{52}$$

4. **Optimal Control Recovery:** At the end of the backward induction, the value function $U(0, S_0, W)$ at the root node (time 0, stock price S_0) is obtained. Also, by recording the control π^* that maximizes $V(\pi)$ at each node, we can recover the optimal investment strategy.

2.3 Results

For the following parameters : $T = 1, S_0 = 100, \sigma = 0.2, \alpha = 0.05, r = 0.03, K = 100, \gamma = 0.5$, we observe that when adding the Black-Scholes price (C) to the initial wealth for the portfolio with a position in option, the curve of the indirect utility function at $t = 0$, with regards to the intial wealth, corresponds with the one with no option position, and the theoretical one.



3 Optimal investment with transaction costs

We now extend the problem to account for proportional transaction costs. The investor can trade a risky asset S_t and a non-risky one B_t, and is charged a proportional transaction cost $\lambda > 0$ on both buying and selling. His holdings are represented by:

- y_t: quantity of risky asset held
- W_t: wealth held in the cash account

We model the system as follows:

$$dS_t = \alpha S_t dt + \sigma S_t dZ_t, \tag{53}$$
$$dy_t = l_t dt - m_t dt, \tag{54}$$
$$dW_t = rW_t dt - (1+\lambda) S_t l_t dt + (1-\lambda) S_t m_t dt, \tag{55}$$

where $l_t, m_t \geq 0$ are the purchases and sales of the risky asset, and λ is the transaction cost (assumed symmetric in our study).

Remark. When setting the transaction costs to 0, we observe that the portfolio has the following dynamics (total wealth being $X_t = W_t + y_t S_t$) :

$$dX_t = dW_t + dS_t y_t + dy_t S_t \tag{56}$$
$$= rW_t dt - S_t dy_t + y_t(\alpha S_t dt + \sigma S_t dZ_t) + S_t dy_t \tag{57}$$
$$= r(X_t - y_t S_t)dt + \alpha y_t S_t dt + \sigma y_t S_t dZ_t \tag{58}$$
$$= rX_t dt + y_t S_t(\alpha - r)dt + y_t S_t \sigma dZ_t. \tag{59}$$

By setting $\pi_t = \frac{y_t S_t}{X_t}$ (proportion of wealth in risky asset), we find the exact same expression as in Part 1, which confirms coherence between the formulations of both problems.

3.1 Optimal investment without options

The investor's objective is now to maximize expected utility of his terminal wealth including the liquidation value of his risky holdings:

$$\sup_{(l,m)} E\left[u\left(W_T + y_T S_T - \lambda S_T |y_T|\right)\right]. \tag{60}$$

This leads us to define the indirect utility function (value function):

$$U(t,S,y,W) = \sup_{(l,m)} E_{t,S,y,W}\left[u\left(W_T + y_T S_T - \lambda S_T |y_T|\right)\right]. \tag{61}$$

We apply Itô's lemma to the function $U(t, S_t, y_t, W_t)$, assuming regularity:

$$dU = \frac{\partial U}{\partial t}dt + \frac{\partial U}{\partial S}dS_t + \frac{\partial U}{\partial y}dy_t + \frac{\partial U}{\partial W}dW_t + \frac{1}{2}\frac{\partial^2 U}{\partial S^2}(dS_t)^2 \tag{62}$$

$$= \left[\frac{\partial U}{\partial t} + \alpha S\frac{\partial U}{\partial S} + \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 U}{\partial S^2} + rW\frac{\partial U}{\partial W} + l\left(\frac{\partial U}{\partial y} - (1+\lambda)S\frac{\partial U}{\partial W}\right) + m\left(-\frac{\partial U}{\partial y} + (1-\lambda)S\frac{\partial U}{\partial W}\right)\right]dt + \sigma S\frac{\partial U}{\partial S}dZ_t. \tag{63}$$

Then, we use the same reasoning as in Section 1.2, taking the expectation and using the dynamic programming principle (and optimality), we obtain the HJB equation:

$$\boxed{\frac{\partial U}{\partial t} + \alpha S \frac{\partial U}{\partial S} + \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 U}{\partial S^2} + rW \frac{\partial U}{\partial W} + \max_{l \geq 0, m \geq 0} \left\{ l \left(\frac{\partial U}{\partial y} - (1+\lambda) S \frac{\partial U}{\partial W} \right) + m \left(-\frac{\partial U}{\partial y} + (1-\lambda) S \frac{\partial U}{\partial W} \right) \right\} = 0.} \tag{64}$$

From this equation, we can isolate the terms that drive the control l and m:

$$F(U) := \max \left\{ \frac{\partial U}{\partial y} - (1+\lambda) S \frac{\partial U}{\partial W},\ -\frac{\partial U}{\partial y} + (1-\lambda) S \frac{\partial U}{\partial W} \right\}. \tag{65}$$

Then the problem can be reformulated as a variational inequality:

$$\max \left\{ F(U),\ -\frac{\partial U}{\partial t} - \alpha S \frac{\partial U}{\partial S} - \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 U}{\partial S^2} - rW \frac{\partial U}{\partial W} \right\} = 0. \tag{66}$$

This equation illustrates the existence of three regions : the No Transaction Region, the Buying Region and the Selling Region.

The terminal condition reflects the full liquidated value of the portfolio at maturity:

$$U(T, S, y, W) = u \left(W + yS - \lambda S |y| \right). \tag{67}$$

3.2 When there is a short position in otpion

We now consider the case where the investor holds a short position in a contingent claim (e.g., an option) with a payoff function $\phi(S_T)$. This means that, in addition to managing their liquid and risky positions, the investor must deliver a terminal payoff equal to $\phi(S_T)$ at time T.

Modified objective. The optimization problem becomes:

$$\sup_{(l,m)} E \left[u \left(W_T + y_T S_T - \lambda S_T |y_T| - \phi(S_T) \right) \right]. \tag{68}$$

This corresponds to an investor aiming to maximize expected utility after liquidating all risky holdings and delivering the payoff of the contingent liability.

Value function. The associated value function is:

$$U(t, S, y, W) = \sup_{(l,m)} E_{t,S,y,W} \left[u \left(W_T + y_T S_T - \lambda S_T |y_T| - \phi(S_T) \right) \right]. \tag{69}$$

HJB and variational inequality. As before, the HJB equation and the variational inequality remain unchanged in structure since the dynamics of the controls and state variables are unaffected by the presence of the claim. The only modification lies in the terminal condition.

Terminal condition. The new terminal condition becomes:

$$U(T, S, y, W) = u \left(W + yS - \lambda S |y| - \phi(S) \right). \tag{70}$$

Remark. If the investor receives a premium p at time t for selling the claim, one may define an indifference price by comparing the value functions with and without the claim, and solving:

$$U^{\text{with claim}}(t, S, y, W + p) = U^{\text{without claim}}(t, S, y, W). \tag{71}$$

4 Numerical Scheme for the Problem with Transaction Costs and Option Position

In this section, we present a numerical scheme to compute the indirect utility function $U(t,S,y,W)$ in the case where the investor is subject to proportional transaction costs and holds a short position in a European option. We recall that the effective terminal wealth accounts for the liquidation value of the remaining position in the risky asset:

$$\bar{W}_T = W_T + S_T y_T - \lambda S_T |y_T|.$$

As a result, the terminal condition becomes:

$$U(T,S,y,W) = u(W + Sy - \lambda S|y|).$$

4.1 Discretization of the State Space

We discretize the continuous variables t, S, y, and W as follows:

- The time interval $[0,T]$ is divided into N steps of size $\Delta t = T/N$.
- The stock price S_t is approximated by a binomial process:

$$S(i+1) = \begin{cases} S(i) \cdot e^{\alpha \Delta t + \sigma\sqrt{\Delta t}} & \text{with probability } 1/2, \\ S(i) \cdot e^{\alpha \Delta t - \sigma\sqrt{\Delta t}} & \text{with probability } 1/2. \end{cases}$$

- The number of shares held y is discretized in steps of Δy:

$$y_k = k\Delta y, \quad k \in \{-M, ..., M\}.$$

- The cash wealth W is discretized in steps of ΔW:

$$W_l = l\Delta W, \quad l \in \{0, ..., K\}.$$

We denote the discretized utility value by:

$$u(i,j,k,l) \approx U(i\Delta t, S_j, y_k, W_l).$$

4.2 Terminal Condition

At maturity $t = T$, we initialize:

$$u(N,j,k,l) = u\left(W_l + S_j y_k - \lambda S_j |y_k|\right).$$

4.3 Backward Recursion over Time Steps

For $i = N-1$ down to 0, we compute $u(i,j,k,l)$ at each grid point by considering the three possible actions: do nothing, buy the minimal allowed number of shares, or sell the minimal allowed number of shares.

1. **No Transaction (NT):**

   If no transaction occurs, wealth grows at the risk-free rate:

$$W' = W_l \cdot e^{r\Delta t}.$$

   Let l^+ be the index closest to W'. The continuation value is:

$$u^{\mathrm{NT}}(i,j,k,l) = \frac{1}{2}\left[u(i+1,j+1,k,l^+) + u(i+1,j-1,k,l^+)\right].$$

2. **Buy Action:**
   The investor buys Δy shares:
   $$y' = y_k + \Delta y, \quad W' = W_l - (1+\lambda) S_j \Delta y.$$
   Let l_b be the nearest index to W', then:
   $$u^{\text{Buy}}(i,j,k,l) = u(i,j,k+1,l_b).$$

3. **Sell Action:**
   The investor sells Δy shares:
   $$y' = y_k - \Delta y, \quad W' = W_l + (1-\lambda) S_j \Delta y.$$
   Let l_s be the nearest index to W', then:
   $$u^{\text{Sell}}(i,j,k,l) = u(i,j,k-1,l_s).$$

4. **Value and Optimal Action:**
   The optimal decision is to take the maximum value among the three possibilities:
   $$u(i,j,k,l) = \max\left\{u^{\text{NT}}(i,j,k,l), u^{\text{Buy}}(i,j,k,l), u^{\text{Sell}}(i,j,k,l)\right\}.$$
   The action that maximizes $u(i,j,k,l)$ determines whether the point lies in region $\mathcal{NT}$, $\mathcal{B}$ (buy), or $\mathcal{S}$ (sell).

4.4 Implementation Notes

- In practice, we use interpolation when W' does not exactly fall on a mesh point.
- The scheme tests only minimal trade increments ($\pm\Delta y$), rather than computing the exact trade required to re-enter $\mathcal{NT}$.
- Nevertheless, this recursive maximization allows us to capture the optimal control logic implied by the variational inequality.