

# Solving Merton's Portfolio Problem

---

Hyelin Choi

Department of Mathematics  
Sungkyunkwan University

# Merton's Portfolio Problem

---

# Problem Statement

- Balance constraint implies the following process for Wealth  $W_t$

$$dW_t = ((\pi_t \cdot (\mu - r) + r) \cdot W_t - c_t) \cdot dt + \pi_t \cdot \sigma \cdot W_t \cdot dz_t.$$

# Merton's Portfolio Problem

The goal of Merton's portfolio problem is to maximize the lifetime-aggregated utility of consumption by selecting the optimal allocation and consumption at each time point.

# Optimal Value Function

- At any time  $t$ , determine optimal  $[\pi(t, W_t), c(t, W_t)]$  to maximize:

$$\mathbb{E}_t \left[ \int_t^T \frac{e^{-\rho(s-t)} \cdot c_s^{1-\gamma}}{1-\gamma} ds + \frac{e^{-\rho(T-t)} \cdot \epsilon^\gamma \cdot W_T^{1-\gamma}}{1-\gamma} \right] \quad (1)$$

where  $\rho \geq 0$  is the utility discount rate.

# Continuous-Time Stochastic Control

- The State at time  $t$  is  $(t, W_t)$
- The Action at time  $t$  is  $[\pi_t, c_t]$
- The Reward per unit time at time  $t$  is  $U(c_t) = \frac{c_t^{1-\gamma}}{1-\gamma}$
- The Return at time  $t$  is the accumulated discounted Reward:

$$\int_t^T e^{-\rho(s-t)} \cdot \frac{c_s^{1-\gamma}}{1-\gamma} \cdot ds$$

- Find Policy:  $(t, W_t) \rightarrow [\pi_t, c_t]$  that maximizes the Expected Return

# Optimal Allocation and Consumption

Finally, we get:

$$\begin{aligned}\pi^*(t, W_t) &= \frac{\mu - r}{\sigma^2 \gamma} \\ c^*(t, W_t) &= \frac{W_t}{f(t)} = \begin{cases} \frac{\nu \cdot W_t}{1 + (\nu \epsilon - 1) \cdot e^{-\nu(T-t)}} & \text{for } \nu \neq 0 \\ \frac{W_t}{T-t+\epsilon} & \text{for } \nu = 0 \end{cases} \\ V^*(t, W_t) &= \begin{cases} \frac{(1 + (\nu \epsilon - 1) \cdot e^{-\nu(T-t)})^\gamma}{\nu^\gamma} \cdot \frac{W_t^{1-\gamma}}{1-\gamma} & \text{for } \nu \neq 0 \\ \frac{(T-t+\epsilon)^\gamma \cdot W_t^{1-\gamma}}{1-\gamma} & \text{for } \nu = 0 \end{cases}\end{aligned}$$

# Optimal Allocation and Consumption

- With Optimal Allocation Consumption, the Wealth process is:

$$\frac{dW_t}{W_t} = \left( r + \frac{(\mu - r)^2}{\sigma^2 \gamma} - \frac{1}{f(t)} \right) \cdot dt + \frac{\mu - r}{\sigma \gamma} \cdot dz_t$$



$$(x, t) \in [W_{min}, W_{max}] \times [0, T_{max}]$$

We use 3 neural networks to predict

- Fraction of wealth allocated to risky asset,  $\pi(t, W_t)$ ,
- Wealth consumption per unit time,  $c(t, W_t)$ ,
- Value function,  $V(t, W_t)$

# Structure of RiskNN

- Args: State  $(t, W_t)$
- 4 hidden layers with 100 perceptrons for each layers
- Softplus as activation functions
- Returns: Prediction of fraction of wealth allocated to risky asset,  $\pi(t, W_t)$

# Structure of ConsumpNN

- Args: State  $(t, W_t)$
- 4 hidden layers with 100 perceptrons for each layers
- LeakyReLU and Softplus as activation functions
- Returns: Prediction of wealth consumption per unit time,  $c(t, W_t)$

# Structure of ValueNN

- Args: State  $(t, W_t)$
- 4 hidden layers with 100 perceptrons for each layers
- LeakyReLU and Softplus as activation functions
- Returns: Prediction of Value function,  $V(t, W_t)$

Optimizer: Adam with learning rate  $1e-4$

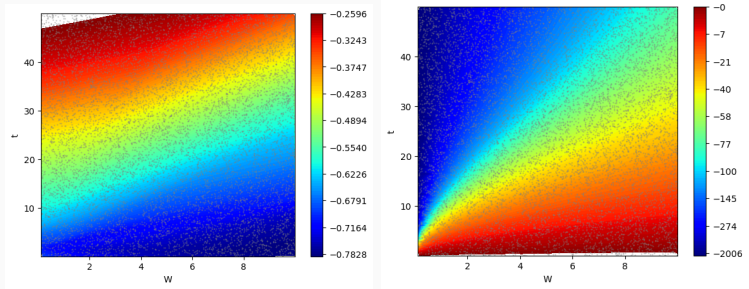
$$dW_t = ((\pi_t \cdot (\mu - r) + r) \cdot W_t - c_t) \cdot dt + \pi_t \cdot \sigma \cdot W_t \cdot dz_t \quad (2)$$

We rewrite (2) as follows:

$$dW_t = -c_t \cdot dt$$

$$dW_t = (\pi_t \cdot (\mu - r) + r) \cdot W_t \cdot dt + \pi_t \cdot \sigma \cdot W_t \cdot dz_t$$

# Results for Value



**Figure 1:** (Left) Value predicted by the ValueNN (Right) The exact Value



# Results for Value

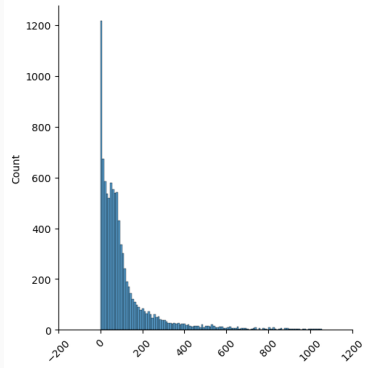
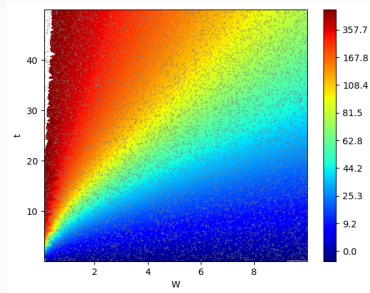
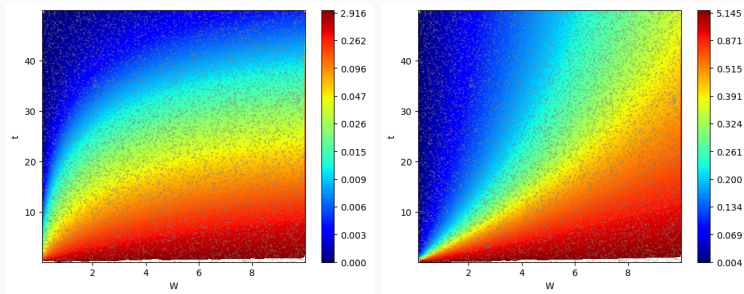


Figure 2: Error for Value

# Results for Consumption



**Figure 3:** (Left) Consumption predicted by  $\text{ConsumpNN} \times W_t / (1e - 6 + t)$   
(Right) The exact Consumption value

# Results for Consumption

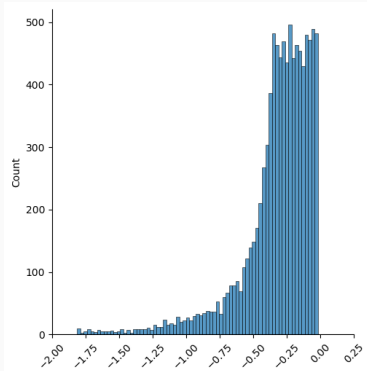
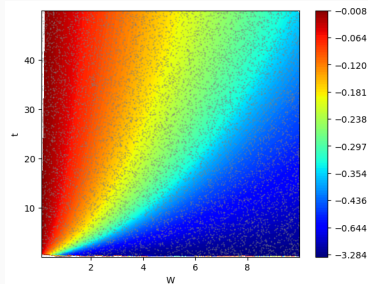
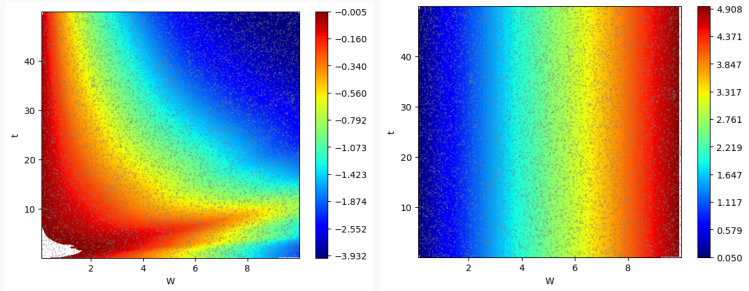


Figure 4: Error for Consumption

# Results for Risk



**Figure 5:** (Left) Risk predicted by RiskNN (Right) The exact Risk value

# Results for Risk

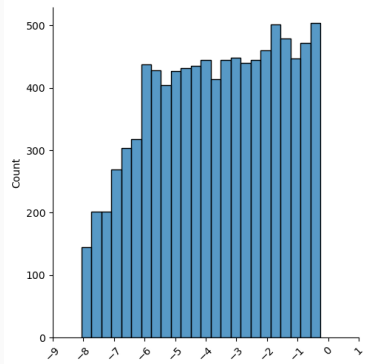
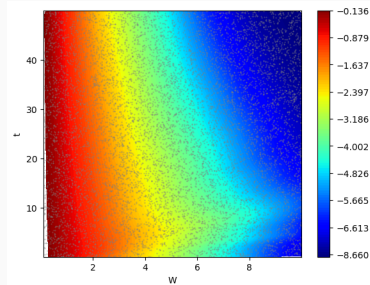


Figure 6: Error for Risk

