

The background features a low-angle, black and white photograph of a modern building's interior, showing a series of white structural beams and glass panels that create a sense of depth and architectural complexity. This image is partially obscured by large, overlapping geometric shapes in shades of light gray, white, and a solid dark blue. A teal-colored triangle is visible in the top right corner.

BVA 529E FINAL

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STOCK TRADER

1. STOCK DATA ENVIRONMENT

2. SGD STOCK TRADER

3. DQN STOCK

4. COMPARISON AND CONCLUSION





STOCK DATA ENVIRONMENT

THE STOCK DATA

- . We have 3 different stocks from APPLE, MSI and STBUX
- . We provide the agent trade the stocks and maximize the rewards

	AAPL	MSI	SBUX
0	67.8542	60.30	28.185
1	68.5614	60.90	28.070
2	66.8428	60.83	28.130
3	66.7156	60.81	27.915
4	66.6556	61.12	27.775
...

```
env = MultiStockEnv(train_data,initial_investment=20000)
state=env.reset() #initializes the state 0 from each stock and $20000 at hand
while not done:
    action=get_action(state) #This will be got from agent
    #perform the trade and go to next state, info will give the current value,
    #reward is incremental value gain
    next_state,reward,done,info=env.step(action)
    state=next_state
```

- We have initial investment, time series of stocks.
- 3 different action for each stock 0,1,2 [sell,hold,buy]. (possible 27 action)
- S, state = [share1,share2,share3,price 1,price2,price3,portfolio] (2n+1 element)

MultiStockEnv the environment of our problem. You can see the .ipynb notebooks on class or my .ipynb notebooks.

SGD STOCK TRADER

SGD WITH MOMENTUM FUNC

$$\begin{aligned} V_t &= \beta V_{t-1} + \alpha \nabla L(W, X, y) \\ W &= W - V_t \end{aligned}$$

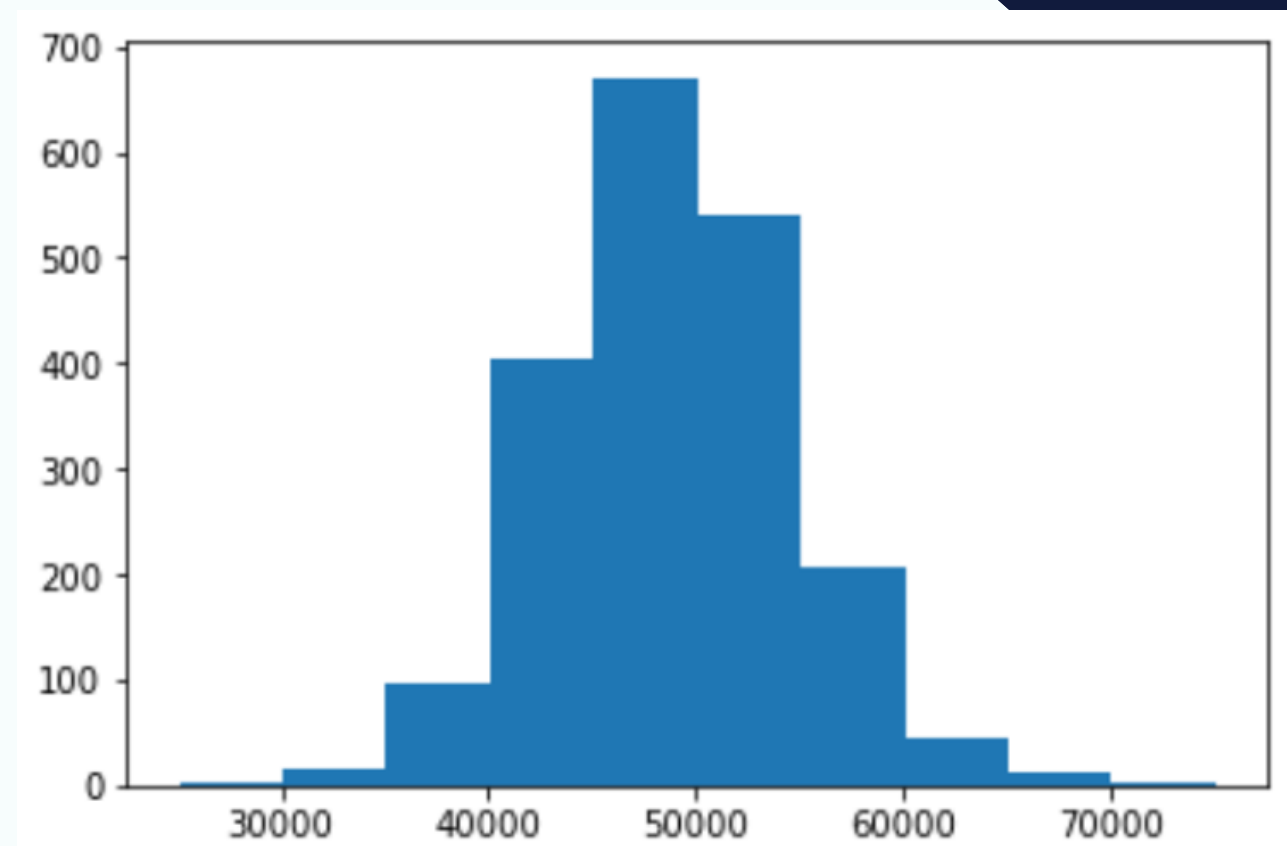
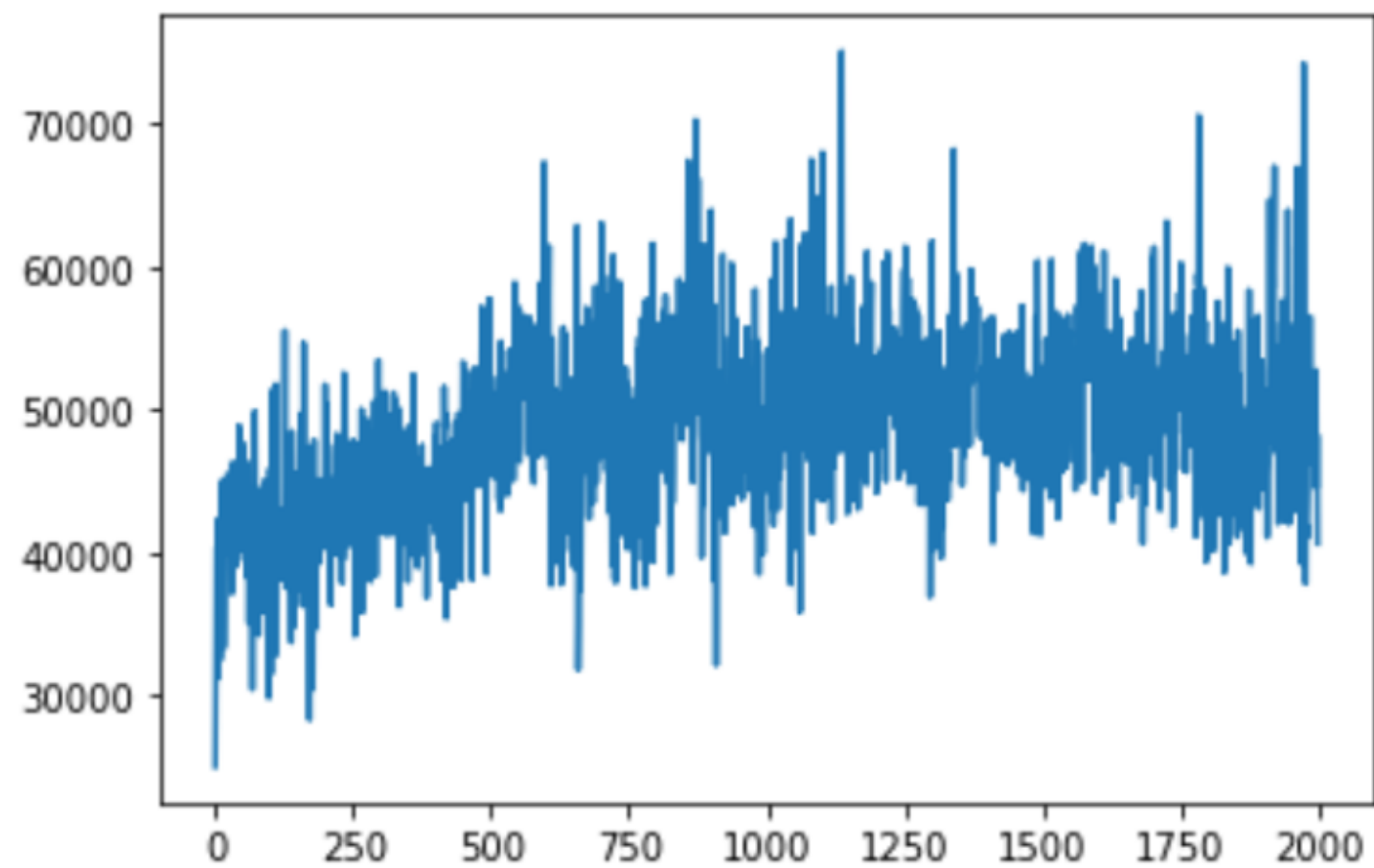
$$\nabla L(W, X, y) = \frac{1}{NK} 2X^T (\hat{y} - y)$$

- Where β is the momentum term, α is the learning rate, and W , is the weight vector (or matrix), and

is the gradient of the loss function
(e.g., MSE)
 K is the number of actions. Why do
we divide to K as well?

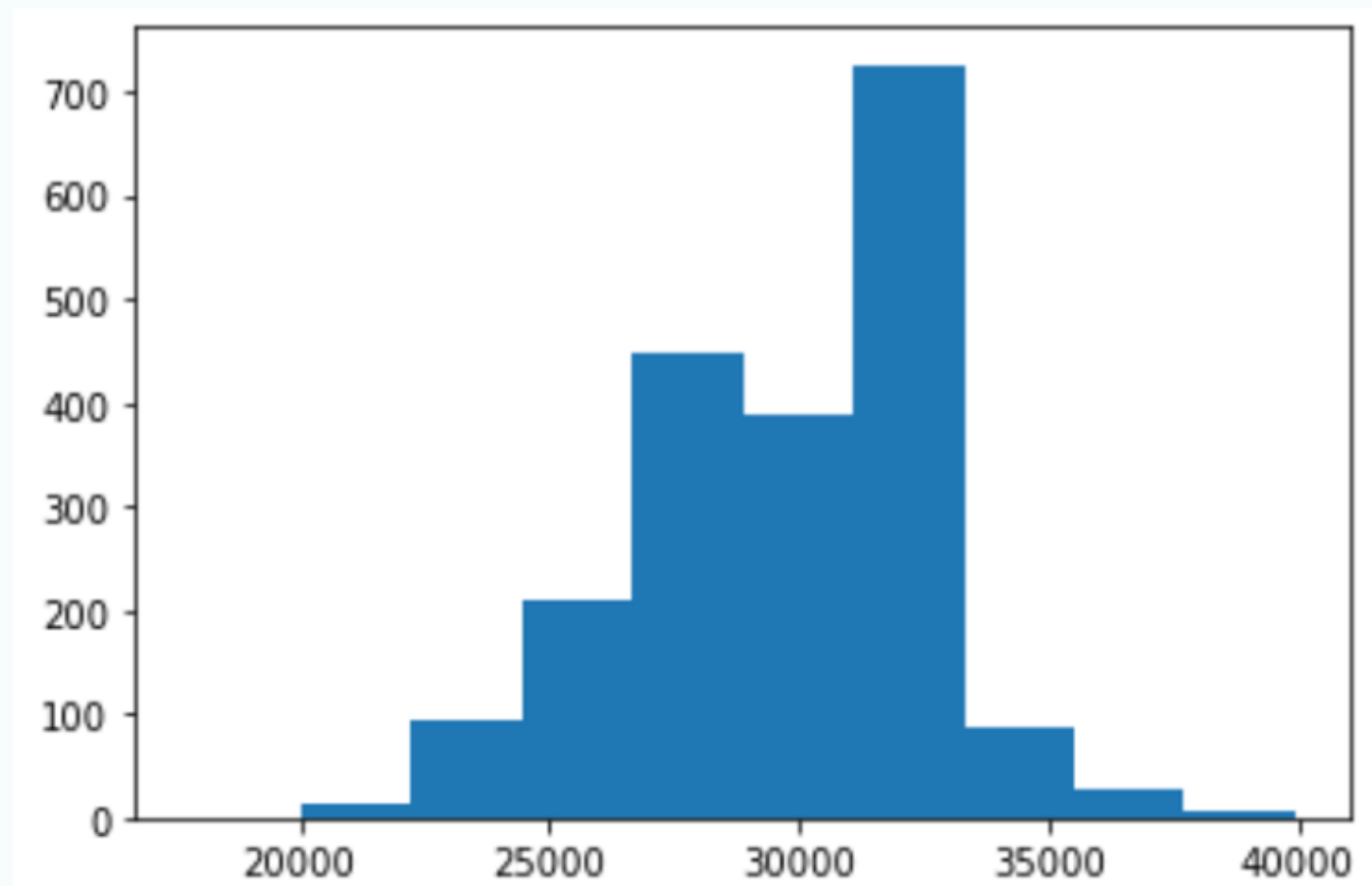
SGD Training Results

- . Dividing first %50 term of training
- . Other %50 term of test
- . Train Results;
- . Our # of episodes: 2000

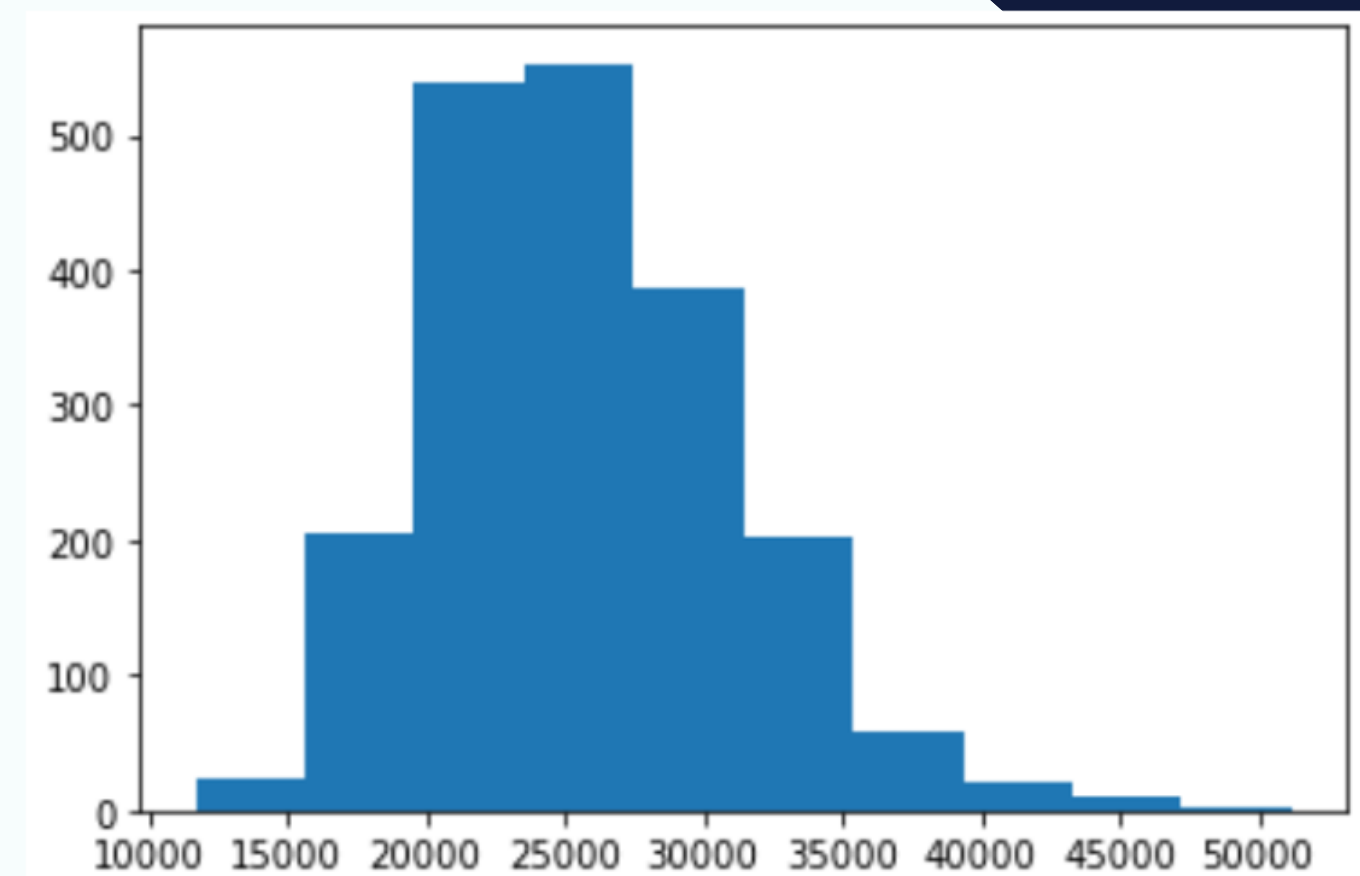


SGD Test and Random Results

- . Test and Random Policy Results;
- . Our # of episodes: 2000



.Test Policy



.Random Policy

DQN STOCK TRADER

● Code in .ipynb
notebook

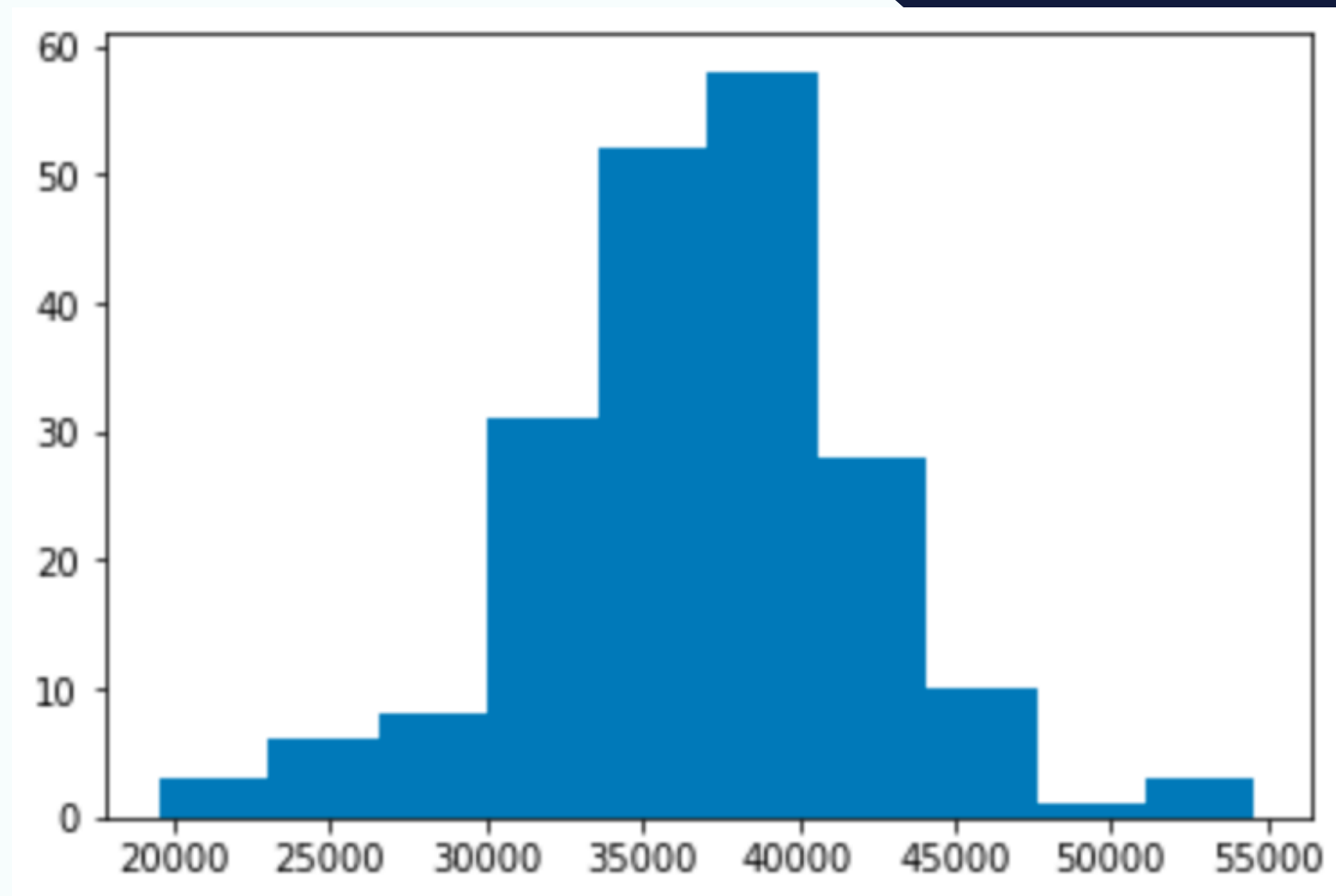
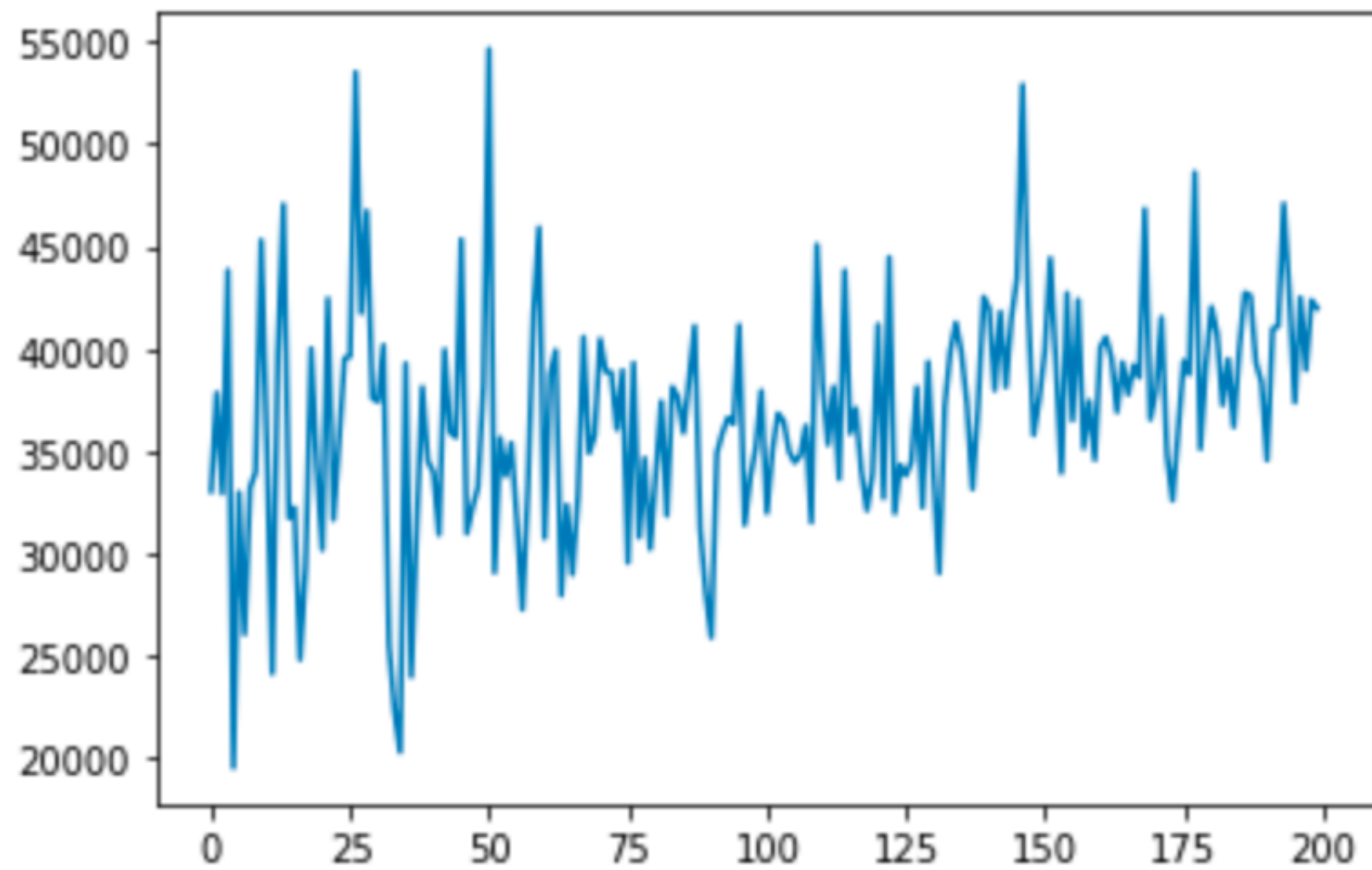
● Single DQN Approach

● NN Model Using For
Function of Algorithm

● Used same
environment

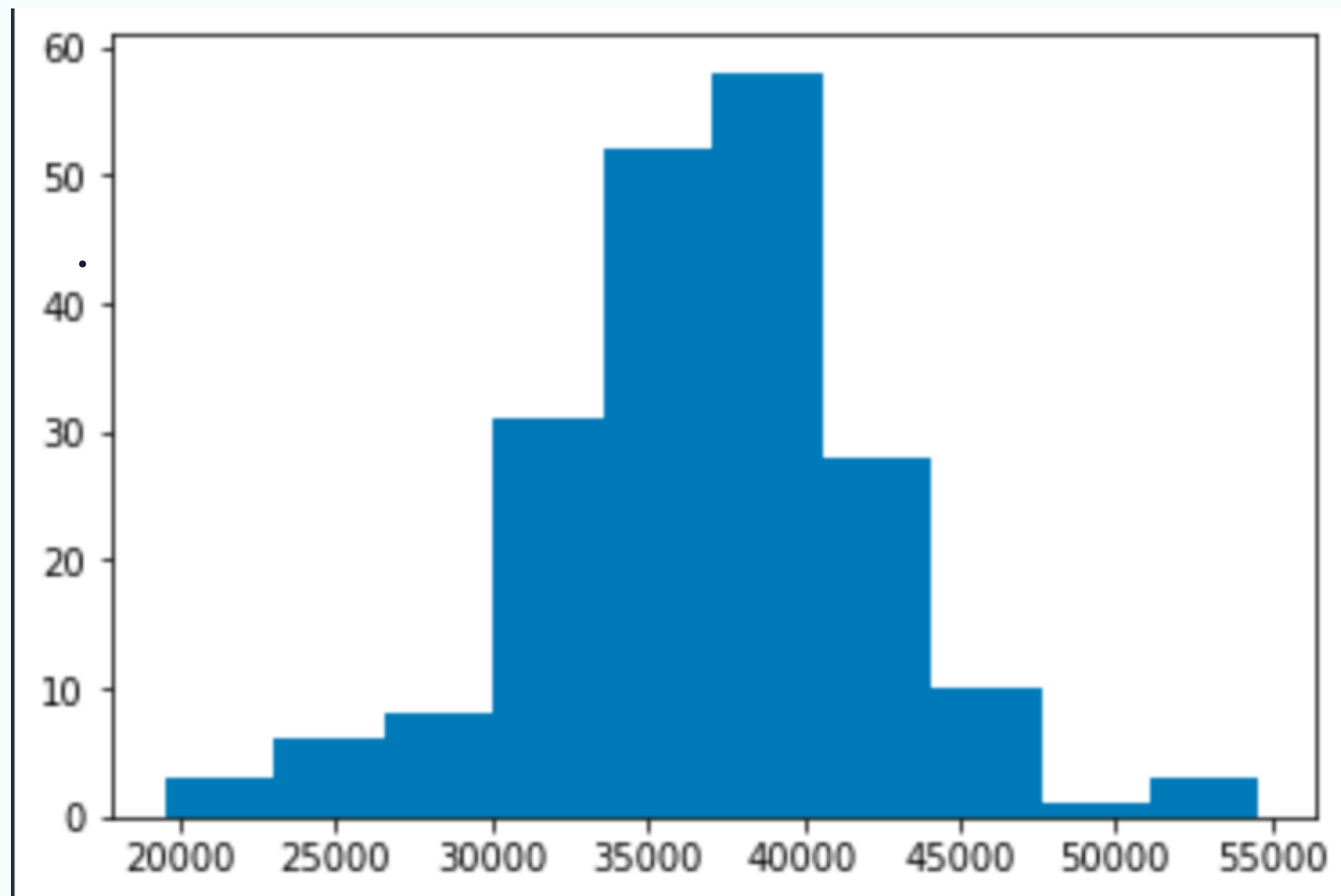
DQN Training Results

- . Used same training and test data
- # Of samples 200
- .. Training sample linearly increase
- . Mean 36781.782438000005

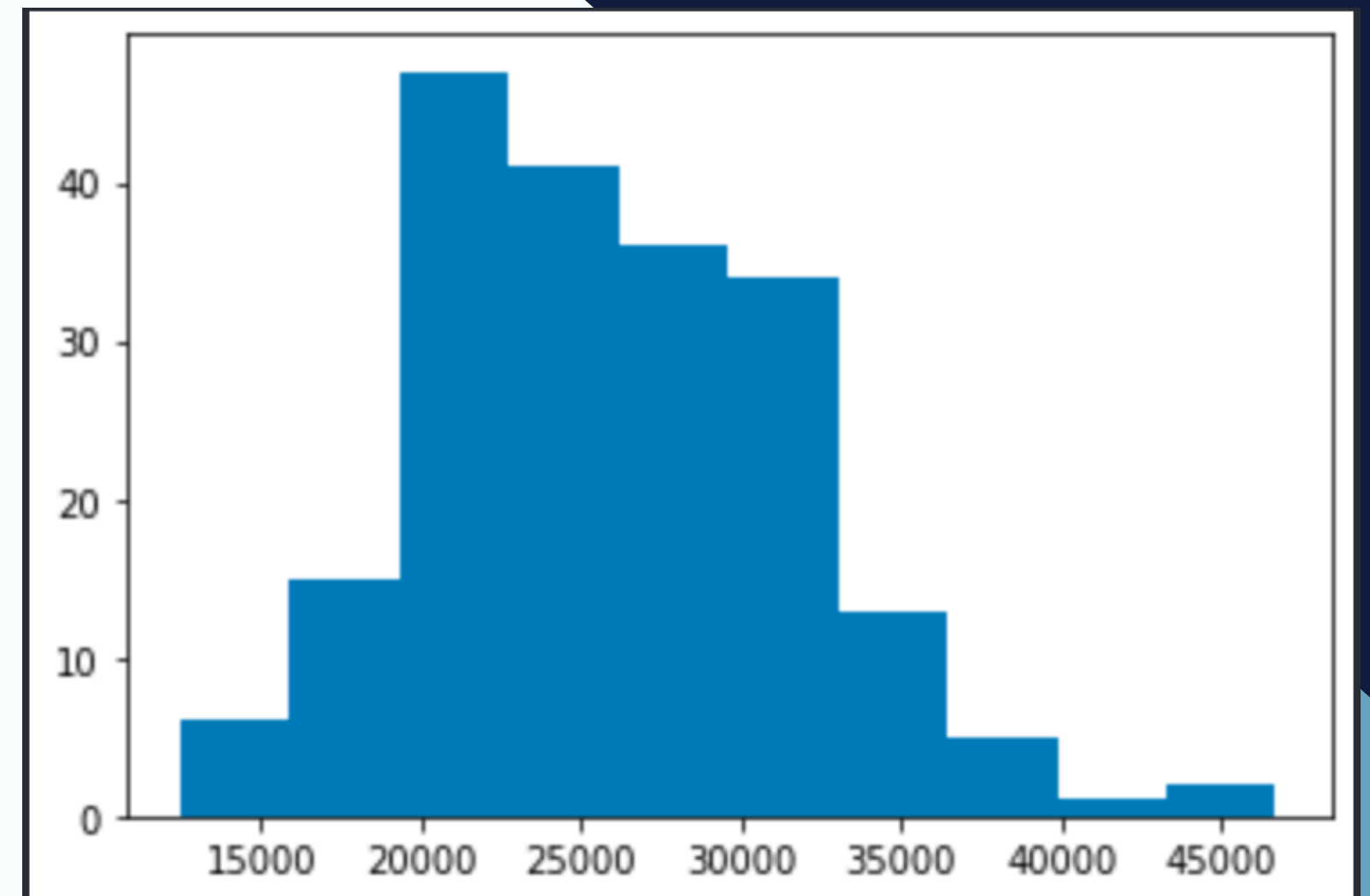


DQN Test Results

- . Used same training and test data.
- . Test portfolio on left
- . Random portfolio on right.



Test portfolio mean: 25045.305849999997



Random portfolio mean:
26051.4880999999682

Conculusion

- SGD function with # of 2000 better portfolio results than DQN function with # 200 samples.

- DQN function with # of 200 random results mean better than random portfolio but the histogram shows the median of test set better than random.

But we can see the test results little bit learn environment

- DQN function with # of 200 Results can be perform well when increase the # of training samples.

Training time is too long.