#### ΤΜΗΜΑ ΠΛΗΡΟΦΟΡΙΚΗΣ 🕆 ΤΗΛΕΠΙΚΟΙΝΩΝΙΩΝ



# NATIONAL AND KAPODISTRIAN UNIVERSITY OF ATHENS

DEPARTMENT OF INFORMATICS AND TELECOMMUNICATIONS

#### SOFTWARE DEVELOPMENT FOR ALGORITHMIC PROBLEMS

Assignment 3 -  $\Gamma$ - CNN Finetuning

Project implemented by team  $\mathbb{N}_{2}$  59:

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## 1. Fine-tuning

### 1.1 C

#### 1.1.1 BATCH SIZE

We experimented with batch sizes of 8, 32, 64, 128, 512, 1024 samples. The results showcased next, indicate that a batch size of **64** or **128** is the most preferable, since both Training and Validation loss converge at a very low value (less than 0.01) and require just a few epochs to train.

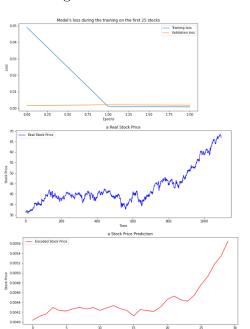
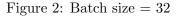
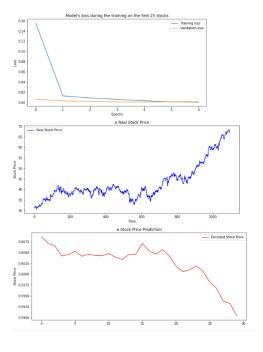


Figure 1: Batch size = 8





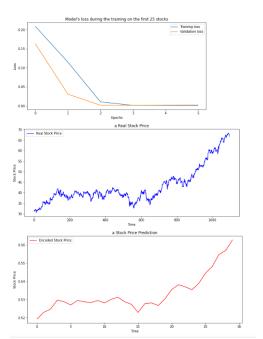
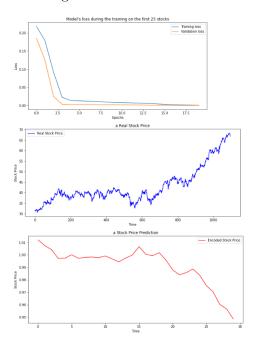


Figure 3: Batch size = 64

Figure 4: Batch size = 128



## 1.1.2 FILTER LAYERS AND FILTER SIZES

We experimented with  ${\bf 2}$  to  ${\bf 6}$  layers, and more specifically:

- 2 layers: [10, 5]
- 3 layers: [10, 10, 5], [20, 10, 5], [30, 10, 8]
- 4 layers: [40, 20, 10, 10]

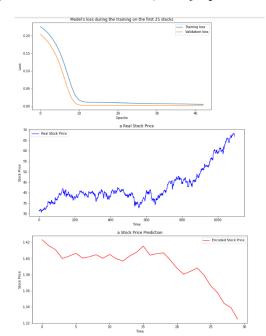
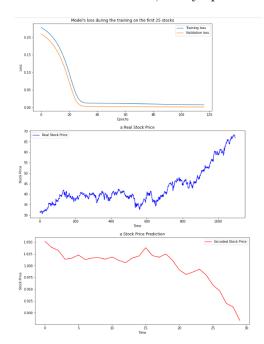


Figure 5: Batch size = 512, Many epochs to train

Figure 6: Batch size = 1024, Many epochs to train



- 5 layers: [50, 30, 10, 5, 5], [100, 75, 50, 25, 5]
- 6 layers: [50, 40, 20, 10, 5], [200, 150, 100, 50, 25, 5]

We present the results below. We conclude that **3 layers with 10, 10 and 5 nodes** give a representative compression while minimizing the Train and Validation losses.

#### 1.1.3 Latent dimensions

We experimented with the following Latent Dimensions: 2, 3, 5, 7, 10, 13, 15.

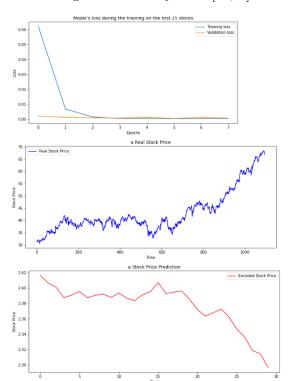


Figure 7: NN Layers = [10, 5]

### 1.1.4 Window size

We experimented with the following Window Sizes: 1, 15, 30, 60, 90. We conclude the the optimal Window Size, providing enough compression of our data while preserving information about major sudden changes is about 30.

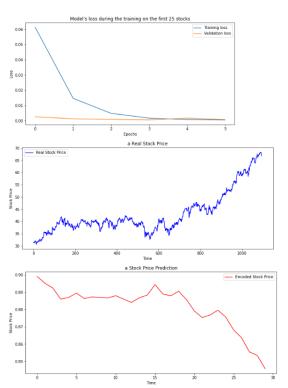
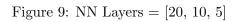


Figure 8: NN Layers = [10, 10, 5]



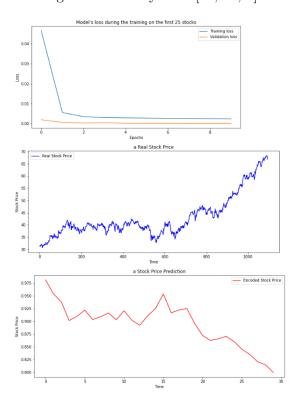


Figure 10: NN Layers = [30, 10, 8]

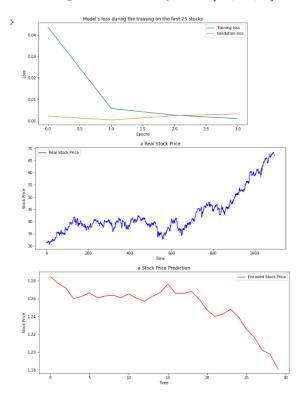
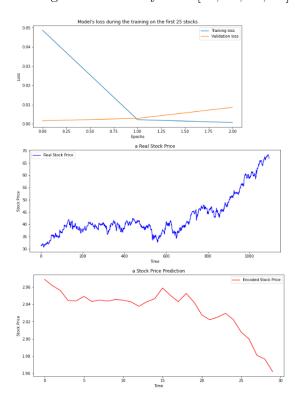


Figure 11: NN Layers = [40, 20, 10, 10]



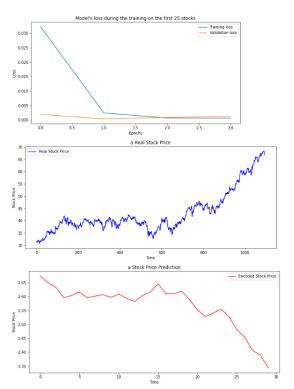
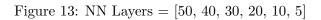


Figure 12: NN Layers = [50, 30, 10, 5, 5]



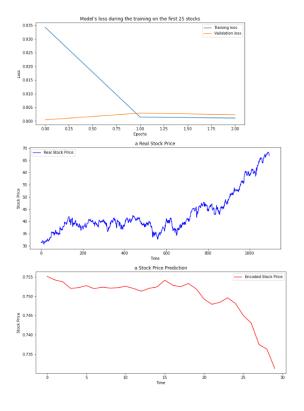


Figure 14: NN Layers = [100, 75, 50, 25, 5]

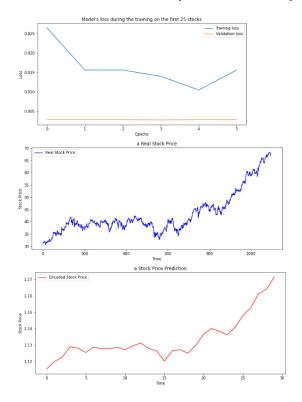
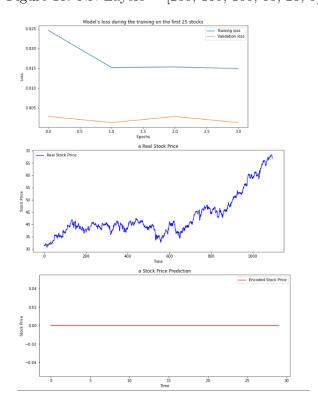


Figure 15: NN Layers = [200, 150, 100, 50, 25, 5]



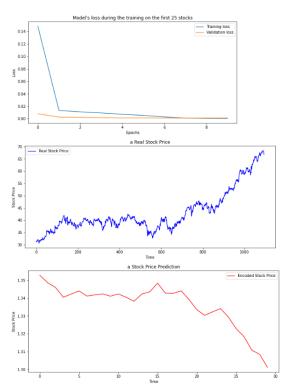
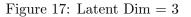


Figure 16: Latent Dim = 2



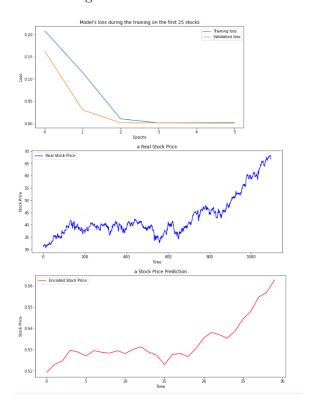


Figure 18: Latent Dim = 5

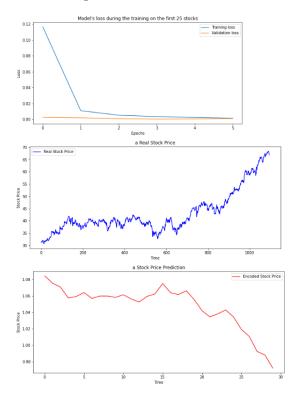
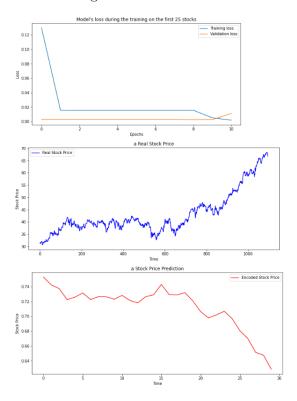


Figure 19: Latent Dim = 7



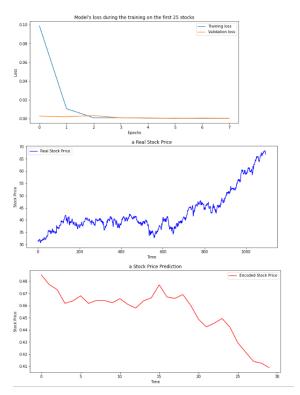


Figure 20: Latent Dim = 10

Figure 21: Latent Dim = 13

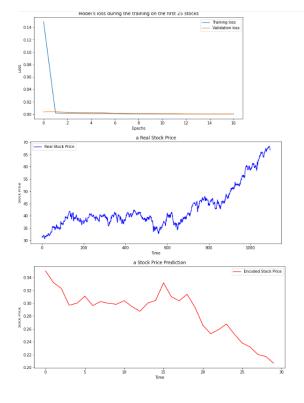


Figure 22: Latent Dim = 15

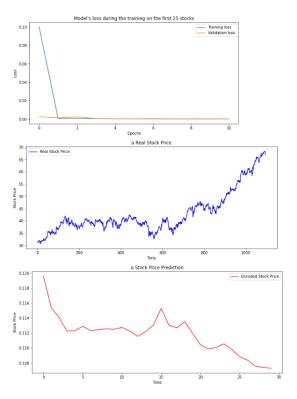
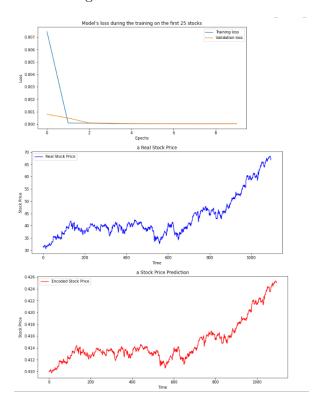


Figure 23: Window size = 1



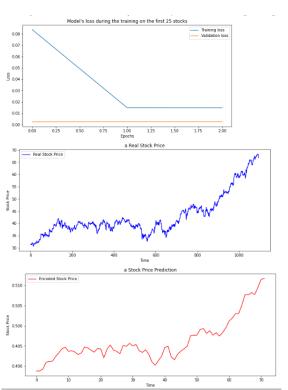
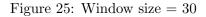


Figure 24: Window size = 15



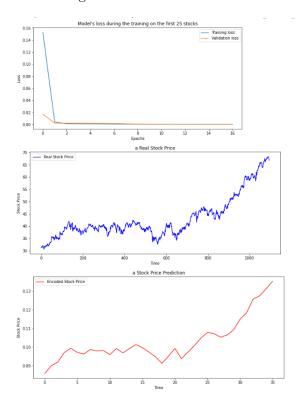


Figure 26: Window size = 60

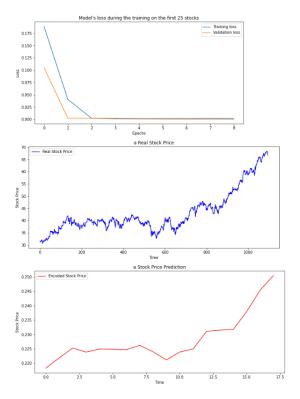


Figure 27: Window size = 90

