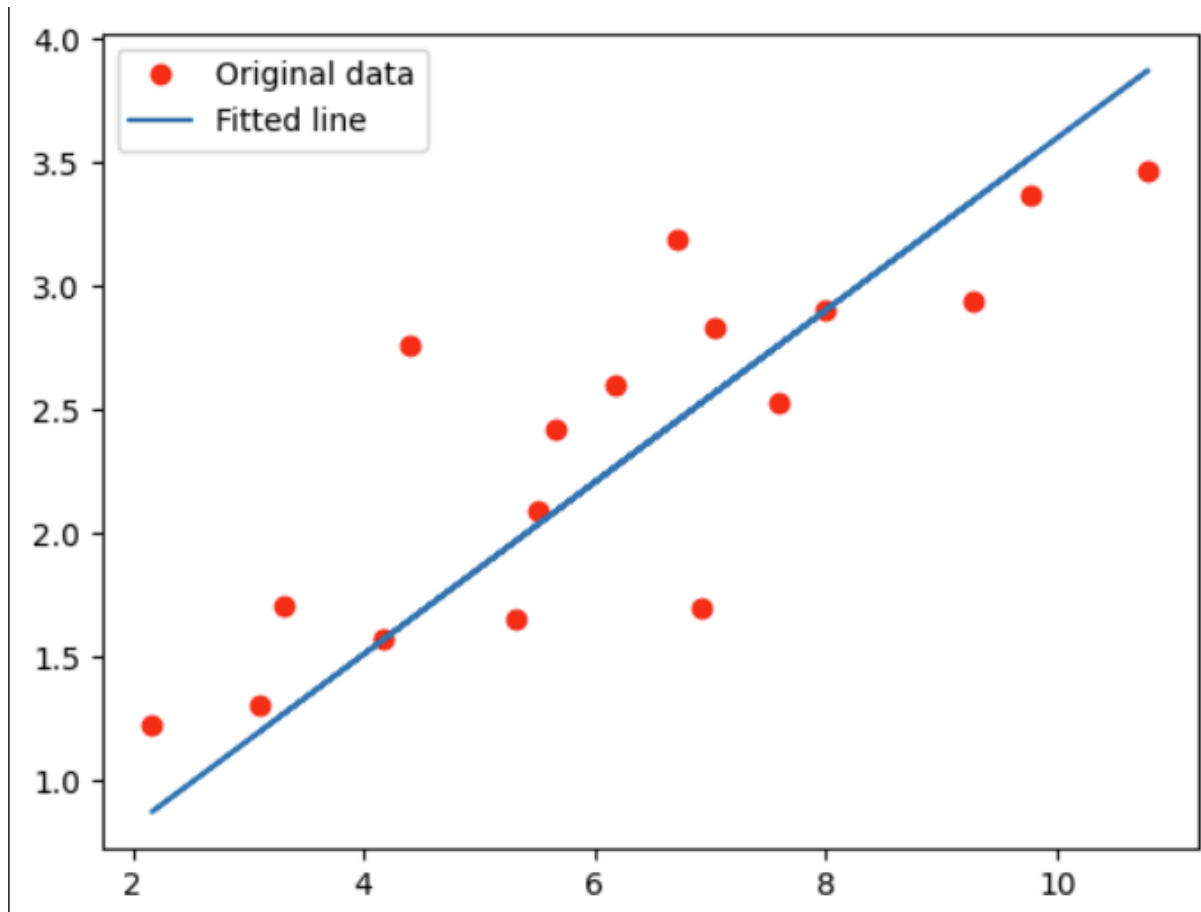


# LAB logbook

## Week - 1



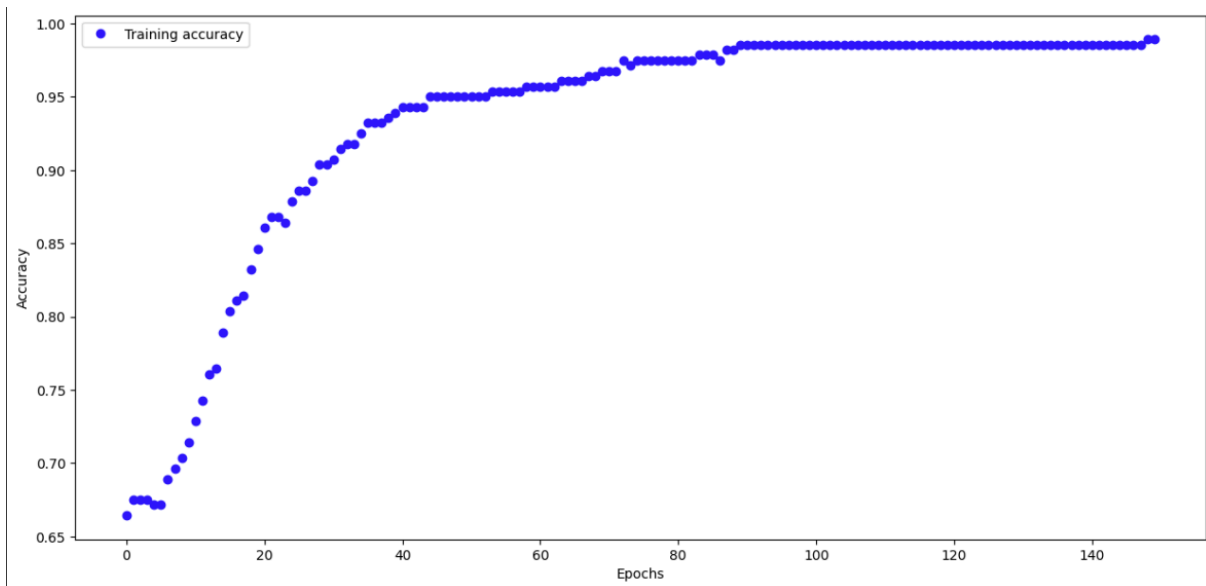
Dot product of  $A \cdot B^T$  results in a new Tensor:

```
[[8634 2719 8750]
 [2939 1329 2975]
 [7573 5341 7545]]
```

Matrix multiplication of  $A \cdot B^T$  results in a new Tensor:

```
[[8634 2719 8750]
 [2939 1329 2975]
 [7573 5341 7545]]
```

# Week-2



```
In [41]: # make a prediction - Copy a row of your choice!

row = [1,0,0.99539,-0.05889,0.85243,0.02306,
0.83398,-0.37708,1,0.03760,0.85243,-0.17755,
0.59755,-0.44945,0.60536,-0.38223,0.84356,
-0.38542,0.58212,-0.32192,0.56971,-0.29674,0.36946,
-0.47357,0.56811,-0.51171,0.41078,-0.46168,0.21266,
-0.34090,0.42267,-0.54487,0.18641,-0.45300]

yhat = model_3.predict([row])
print('Predicted: %.3f' % yhat)

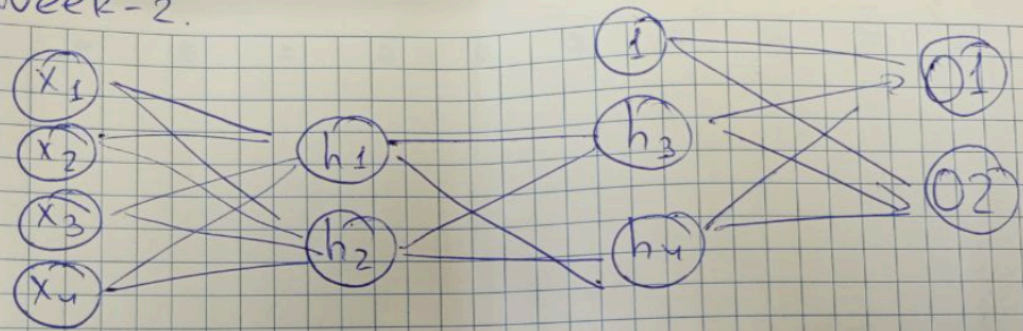
1/1 [=====] - 0s 56ms/step
Predicted: 0.985

/tmp/ipykernel_563982/4213957503.py:11: DeprecationWarning: Conversion of an array with ndim > 0 to a scalar is deprecated, and will error in future. Ensure you extract a single element from your array before performing this operation. (Deprecated NumPy 1.25.)
  print('Predicted: %.3f' % yhat)

In [ ]:
```

Task 1: accuracy = 0.998

Week-2.



$$SSID = 2288841$$

$$x_1 = (2288841 / 8964879) \times 23 = 5,8221$$

$$x_2 = (2288841 / 8964879) \times 32 = 8,1699$$

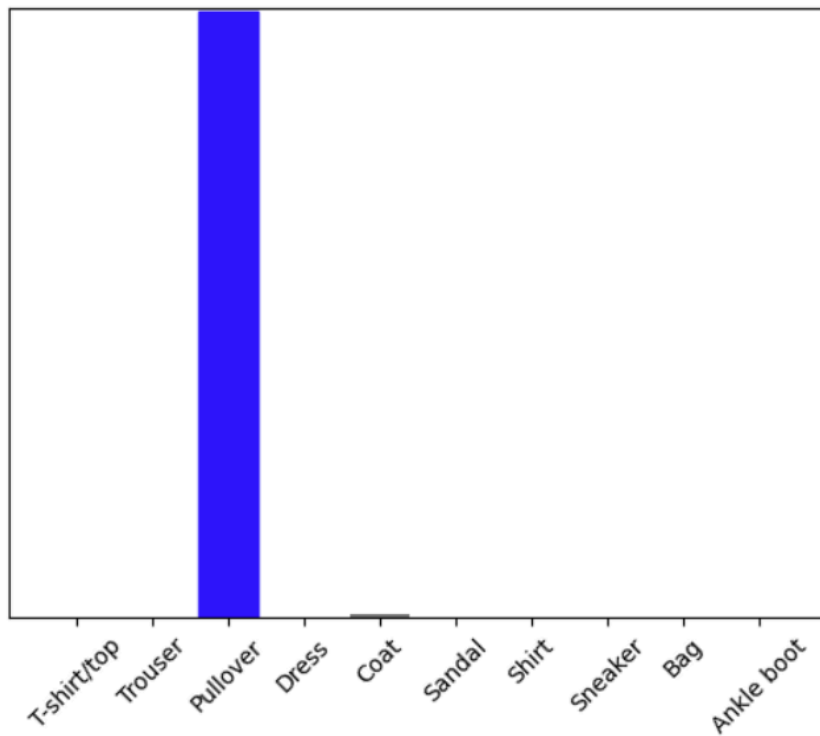
$$x_3 = (2288841 / 8964879) \times 56 = 14,2974$$

$$x_4 = (2288841 / 8964879) \times 48 = 12,2549$$

## week-3

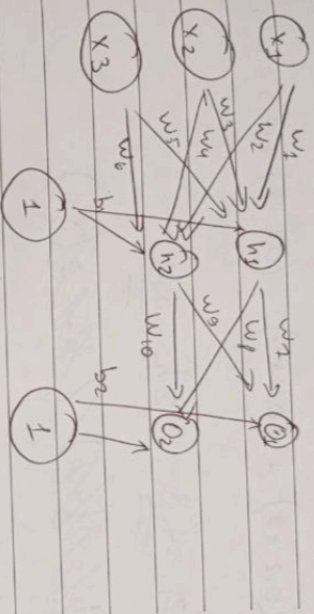


```
plot_value_array(1, predictions_single[0], test_labels)  
_ = plt.xticks(range(10), class_names, rotation=45)
```



Week-3

## Back propagation by hand



$$w_1 = 0.1 \quad w_3 = 0.3 \quad w_5 = 0.5 \quad w_7 = 0.2$$

$$w_2 = 0.2 \quad w_4 = 0.4 \quad w_6 = 0.6 \quad w_8 = 0.8$$

$$w_9 = 0.5 \quad w_{10} = 0.1 \quad b_1 = 0.5 \quad b_2 = 0.7$$

$$x_1 = 1, \quad x_2 = 4, \quad x_3 = 5 \quad t_1 = 0.1$$

$$t_2 = 0.05$$

Forward

Propagate

I → Hidden Layer

$$(0.1 \times 1) + (0.3 \times 4) + (0.5 \times 5) + 0.5 = 2h_1$$

$$(0.2 \times 1) + (0.4 \times 4) + (0.6 \times 5) + 0.15 = 2h_2$$

$$z_{h1} = 4.13 \quad z_{h2} = 5.13$$

$$h_1 = 0.98 \quad h_2 = 0.98$$

~~At the~~

$$h_1 = \frac{1}{1 + e^{-z_{h1}}} = 0.98$$

$$h_2 = \frac{1}{1 + e^{-z_{h2}}} = 0.98$$

(Pre-activation)

Output Layer

Memo No. / /  
Date

Mo	Tu	We	Th	Fr	Sa	Su
----	----	----	----	----	----	----

$$z_{o1} = 0.7 \cdot 0.38 + 0.8 \cdot 0.33 + 0.5 = 1.578$$

$$z_{o2} = 0.9 \cdot 0.38 + 0.1 \cdot 0.53 + 0.5 = 1.491$$

1

$$b_1 = \frac{1}{1 + e^{-1.328}} = 0.87$$

$$b_2 = \frac{1}{1 + e^{-1.491}} = 0.81$$

Error Calculation

$$E = \frac{1}{2} \left( (0.1 - 0.87)^2 + (0.05 - 0.81)^2 \right) = 0.58525$$

Backpropagation

$$\textcircled{1} \delta_{o1} = (0.87 - 0.1) \cdot 0.87 \cdot (1 - 0.87) = 0.08$$

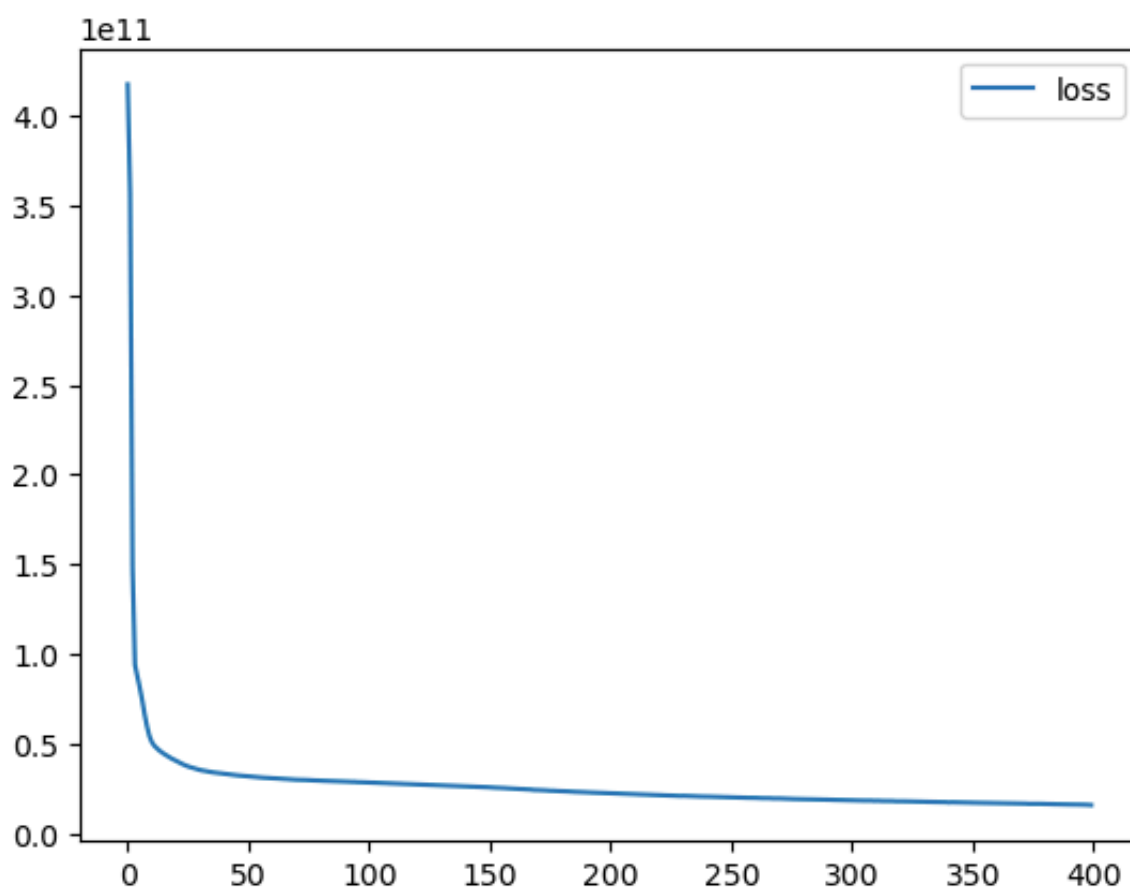
$$\delta_{o2} = (0.81 - 0.05) \cdot 0.81 \cdot (1 - 0.81) = 0.03$$

$$\textcircled{2} \delta_{h1} = (0.08 \cdot 0.7 + 0.03 \cdot 0.9) \cdot 0.38 \cdot (1 - 0.38) =$$

$$\delta_{h2} = (0.08 \cdot 0.8 + 0.03 \cdot 0.1) \cdot 0.33 \cdot (1 - 0.33) =$$

$$\delta_{h1} = 0.001 \quad \delta_{h2} = 0.0006$$

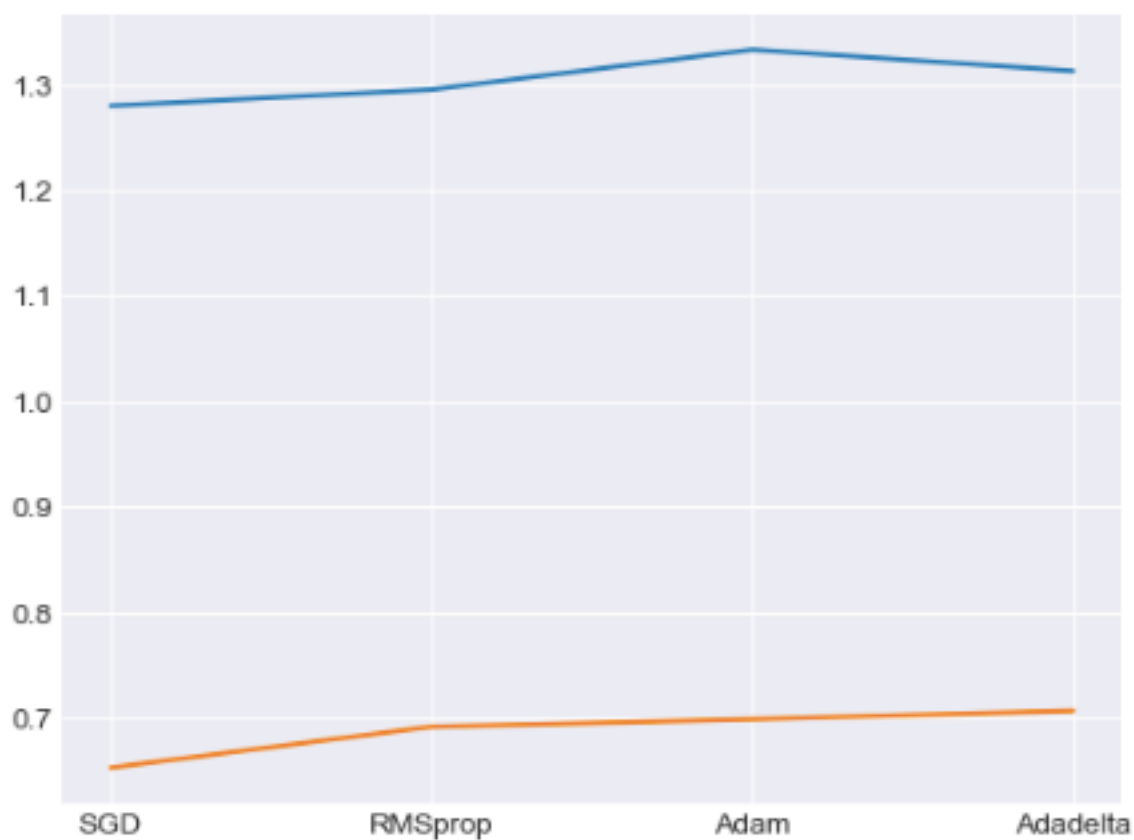
## Week-4



Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 128)	2560
dense_1 (Dense)	(None, 64)	8256
dense_2 (Dense)	(None, 32)	2080
dense_3 (Dense)	(None, 16)	528
dense_4 (Dense)	(None, 1)	17
Total params: 13441 (52.50 KB)		
Trainable params: 13441 (52.50 KB)		
Non-trainable params: 0 (0.00 Byte)		
None		

## Week - 5

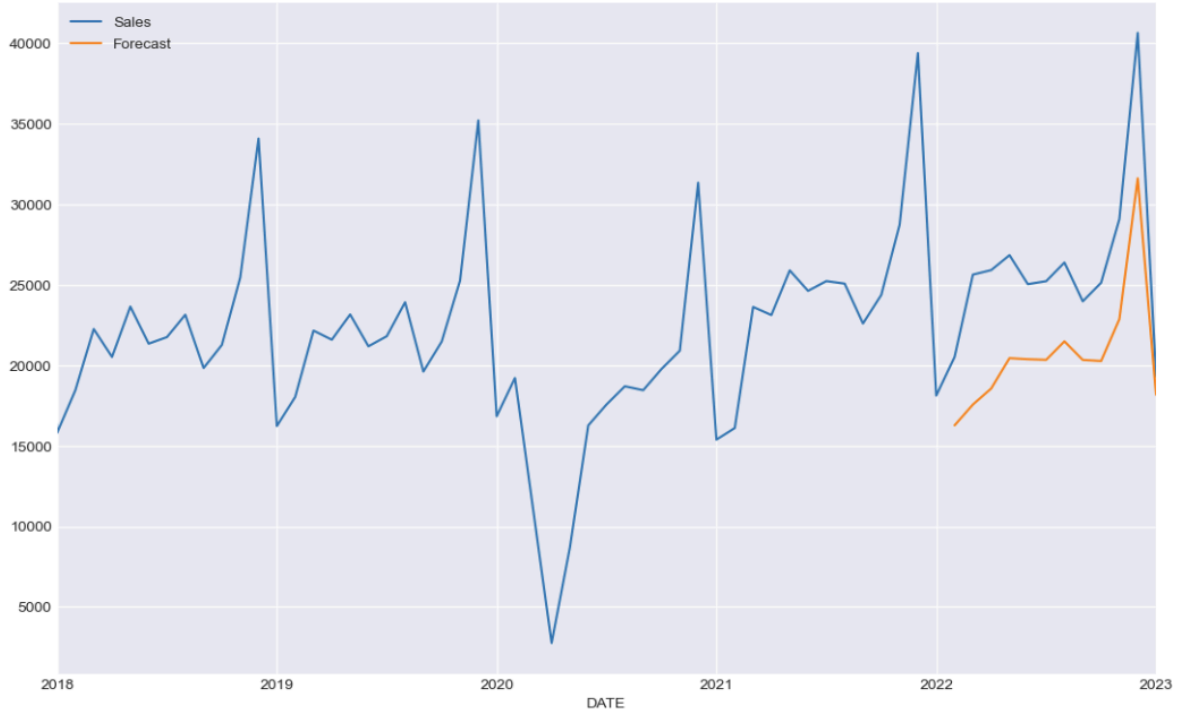




```
Epoch 1/3
1563/1563 [=====] - 12s 5ms/step - loss: 0.3863 - accuracy: 0.8716 - val_loss:
1.5317 - val_accuracy: 0.6758
Epoch 2/3
1563/1563 [=====] - 8s 5ms/step - loss: 0.2598 - accuracy: 0.9138 - val_loss:
1.2955 - val_accuracy: 0.6844
Epoch 3/3
1563/1563 [=====] - 8s 5ms/step - loss: 0.2298 - accuracy: 0.9219 - val_loss:
1.3430 - val_accuracy: 0.7076
Epoch 1/3
1563/1563 [=====] - 10s 6ms/step - loss: 0.4163 - accuracy: 0.8656 - val_loss:
1.4528 - val_accuracy: 0.6790
Epoch 2/3
1563/1563 [=====] - 9s 6ms/step - loss: 0.3921 - accuracy: 0.8724 - val_loss:
1.5664 - val_accuracy: 0.6802
Epoch 3/3
1563/1563 [=====] - 7s 5ms/step - loss: 0.3704 - accuracy: 0.8813 - val_loss:
1.3944 - val_accuracy: 0.6819
Epoch 1/3
1563/1563 [=====] - 10s 5ms/step - loss: 0.3244 - accuracy: 0.8934 - val_loss:
1.5107 - val_accuracy: 0.6815
Epoch 2/3
1563/1563 [=====] - 8s 5ms/step - loss: 0.2515 - accuracy: 0.9159 - val_loss:
1.5154 - val_accuracy: 0.6895
Epoch 3/3
1563/1563 [=====] - 8s 5ms/step - loss: 0.2054 - accuracy: 0.9315 - val_loss:
1.6015 - val_accuracy: 0.6602
Epoch 1/3
1563/1563 [=====] - 10s 5ms/step - loss: 0.2139 - accuracy: 0.9281 - val_loss:
1.5573 - val_accuracy: 0.6763
Epoch 2/3
1563/1563 [=====] - 8s 5ms/step - loss: 0.1658 - accuracy: 0.9455 - val_loss:
1.5491 - val_accuracy: 0.6843
Epoch 3/3
1563/1563 [=====] - 9s 5ms/step - loss: 0.1436 - accuracy: 0.9533 - val_loss:
1.5479 - val_accuracy: 0.6879
```

## Week - 6

.26... (576.0, 636.0)



[ 1 ]:

DATE

```
In [126... ax = df.plot()  
forecast_df.plot(ax=ax)  
plt.xlim('2018-01-01','2021-01-01')
```

Out[126... (576.0, 612.0)



## Week - 7

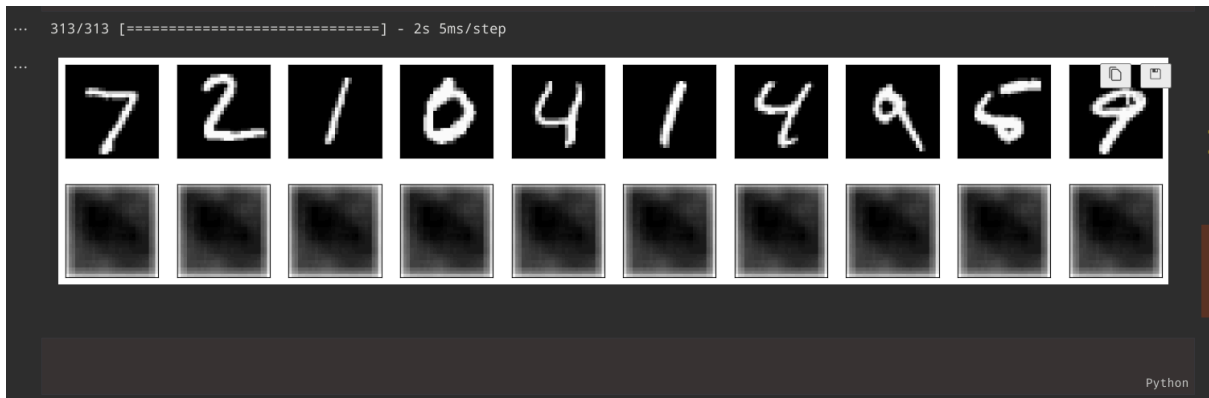
### task-1

Model: "sequential\_2"

Layer (type)	Output Shape	Param #
sequential (Sequential)	(None, 25)	420625
sequential_1 (Sequential)	(None, 28, 28)	421384

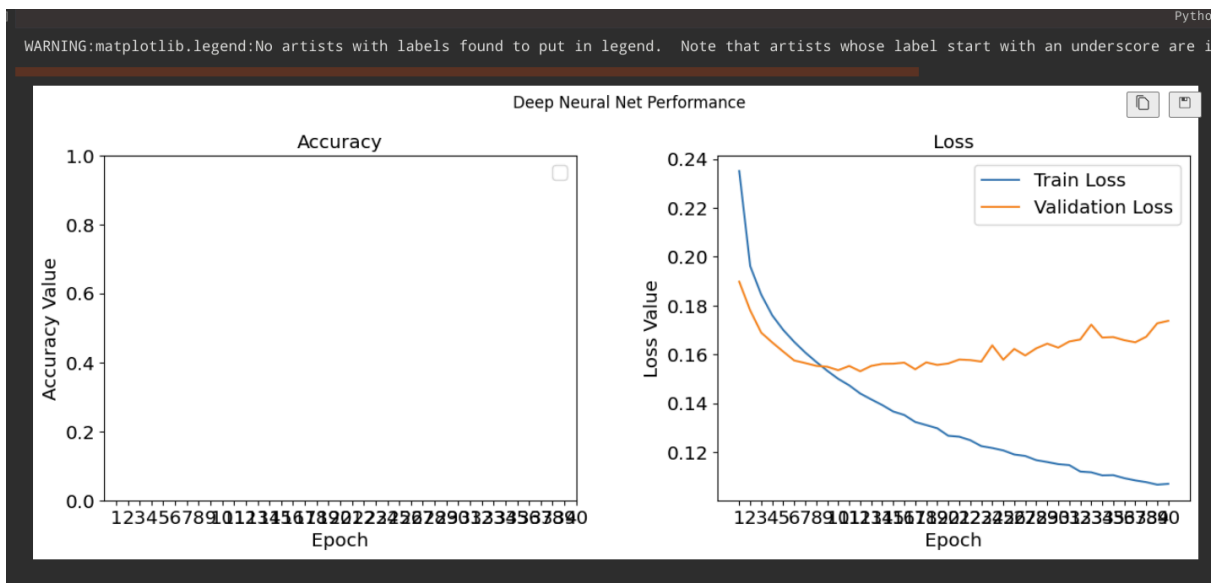
Total params: 842009 (3.21 MB)  
Trainable params: 842009 (3.21 MB)  
Non-trainable params: 0 (0.00 Byte)

### task - 2



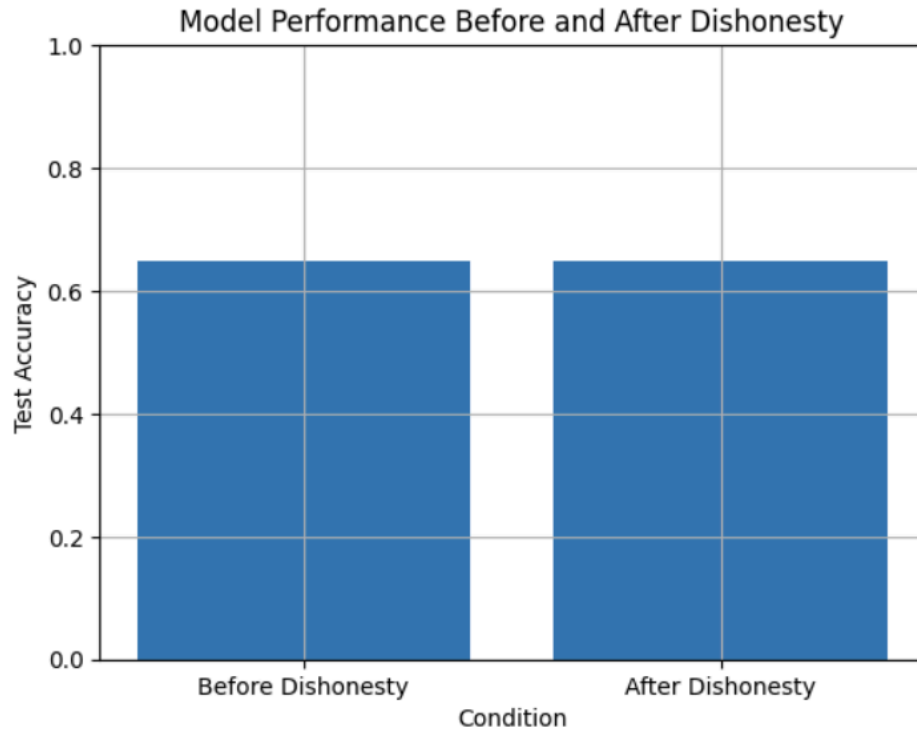
**Week - 8**

**Week - 9**



## Week - 10

Test Accuracy Before Dishonesty: 0.6486999988555908  
Test Accuracy After Dishonesty: 0.6486999988555908



In [ ]:

## Week - 11

