## **Computational Intelligence**

# Course Project Spring 1399

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Below are the packages used in this project.

```
[1]: import numpy as np
  import keras
  import matplotlib.pyplot as plt
  from keras.models import Sequential
  from sklearn.preprocessing import StandardScaler
  from keras.layers import Dense, Dropout
  from keras.layers.normalization import BatchNormalization
  import pandas as pd
```

Using TensorFlow backend.

```
[2]: input_path = 'data/Elearning-Data-cut.xls' train_data_count = 640
```

The data is split into train and test sets. the data is read using pandas and tha NaN values are replaced with 0. also the student No. column is dropped since we don't need it.

```
[3]: def read_input():
    input = pd.read_excel(input_path, 0)
    input = input.drop(['STUDENTN'], axis=1)
    input = input.fillna(0)
    output_gpa = pd.read_excel(input_path, 1)
    output_fail = pd.read_excel(input_path, 2)[['OUT_IN']]
    return input, output_gpa, output_fail
```

Here the data is converted from pandas datafram to numpy array. For the second part of the project (Classfication), Because the data is not balanced and there are lots more passed classes than failed classes, we duplicate rows belonging to failed classes. And then the data is split into train and test sets. Finally all the data are normalized using StandardScaler().fit\_transform.

```
[4]: input, output_gpa, output_fail = read_input()
input = input.to_numpy()
output_gpa = output_gpa.to_numpy()
```

```
output_fail = output_fail.to_numpy()
input2 = []
output_fail2 = []
for i in range(len(input)):
    if output_fail[i] == 0:
        input2.append(input[i])
        input2.append(input[i])
        output_fail2.append([output_fail[i][0]])
        output_fail2.append([output_fail[i][0]])
    else:
        input2.append(input[i])
        output_fail2.append([output_fail[i][0]])
input2 = np.array(input2)
output_fail2 = np.array(output_fail2)
X_train = input[:train_data_count]
X_train2 = input2[:train_data_count]
Y_train = output_gpa[:train_data_count]
Y_train2 = output_fail2[:train_data_count]
X_test = input[train_data_count:]
X_test2 = input2[train_data_count:]
Y_test = output_gpa[train_data_count:]
Y_test2 = output_fail2[train_data_count:]
Y_test22 = output_fail[train_data_count:]
ss = StandardScaler()
X_train = ss.fit_transform(X_train)
X_test = ss.fit_transform(X_test)
X_test2 = ss.fit_transform(X_test2)
X_train2 = ss.fit_transform(X_train2)
```

### 1 Regression

For this part, we design a sequential model having 2 hidden layers with 250 and 150 neurons respectively. The output layer has 1 nueron. Also all the layers are using relu as their activation function. Nadam is used as the optimizer and mse is used as loss function and the metric. These are obtained by trial and error. Also there's a batch normalizer layer between each of the layers that normalizes the current batch which is being fed to the network. finally the data is shuffled and the network is trained for 800 epochs.

```
[5]: model = Sequential()
   model.add(BatchNormalization())
   model.add(Dense(250, activation='relu'))
   model.add(BatchNormalization())
   model.add(Dense(150, activation='relu'))
   model.add(BatchNormalization())
   model.add(Dense(1))
   model.compile(optimizer='Nadam', loss='mse', metrics=['mse'])
   result = model.fit(X_train, Y_train,batch_size=32, epochs=400)
  Epoch 1/400
  640/640 [================ ] - 1s 1ms/step - loss: 195.2816 - mse:
  195.2816
  Epoch 2/400
  171.0429
  Epoch 3/400
  130.3741
  Epoch 4/400
  640/640 [============] - Os 80us/step - loss: 80.4407 - mse:
  80.4407
  Epoch 5/400
  640/640 [================ ] - Os 81us/step - loss: 37.5905 - mse:
  37.5905
  Epoch 6/400
  640/640 [============== ] - Os 85us/step - loss: 12.9863 - mse:
  12.9863
  Epoch 7/400
  4.6627
  Epoch 8/400
  2.4559
  Epoch 9/400
  2.2705
  Epoch 10/400
  2.1992
  Epoch 11/400
  640/640 [============== ] - Os 78us/step - loss: 1.9381 - mse:
  1.9381
  Epoch 12/400
  1.5955
  Epoch 13/400
```

```
1.2960
Epoch 14/400
1.3618
Epoch 15/400
1.3501
Epoch 16/400
640/640 [============== ] - Os 80us/step - loss: 1.3853 - mse:
1.3853
Epoch 17/400
1.5006
Epoch 18/400
1.2418
Epoch 19/400
1.0689
Epoch 20/400
1.2547
Epoch 21/400
640/640 [============== ] - Os 67us/step - loss: 1.0406 - mse:
1.0406
Epoch 22/400
640/640 [=============] - Os 72us/step - loss: 1.1682 - mse:
1.1682
Epoch 23/400
1.0527
Epoch 24/400
640/640 [============= ] - Os 112us/step - loss: 0.8363 - mse:
0.8363
Epoch 25/400
640/640 [============== ] - Os 107us/step - loss: 0.8688 - mse:
0.8688
Epoch 26/400
640/640 [=============== ] - Os 104us/step - loss: 0.9281 - mse:
0.9281
Epoch 27/400
640/640 [================ ] - Os 112us/step - loss: 0.8867 - mse:
0.8867
Epoch 28/400
0.8568
Epoch 29/400
```

```
0.9928
Epoch 30/400
640/640 [============== ] - Os 82us/step - loss: 0.9319 - mse:
0.9319
Epoch 31/400
0.8107
Epoch 32/400
640/640 [============== ] - Os 71us/step - loss: 0.7831 - mse:
0.7831
Epoch 33/400
0.7366
Epoch 34/400
0.7335
Epoch 35/400
640/640 [============== ] - Os 82us/step - loss: 0.6919 - mse:
0.6919
Epoch 36/400
0.8376
Epoch 37/400
0.8762
Epoch 38/400
640/640 [============] - Os 118us/step - loss: 0.7700 - mse:
0.7700
Epoch 39/400
640/640 [=============== ] - Os 113us/step - loss: 0.7460 - mse:
0.7460
Epoch 40/400
640/640 [============== ] - Os 97us/step - loss: 0.8032 - mse:
0.8032
Epoch 41/400
1.0118
Epoch 42/400
0.7800
Epoch 43/400
0.8010
Epoch 44/400
0.6135
Epoch 45/400
```

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0.8734
Epoch 46/400
640/640 [============== ] - Os 78us/step - loss: 0.7039 - mse:
0.7039
Epoch 47/400
0.6767
Epoch 48/400
0.7645
Epoch 49/400
0.6361
Epoch 50/400
0.7036
Epoch 51/400
640/640 [============== ] - Os 97us/step - loss: 0.7228 - mse:
0.7228
Epoch 52/400
0.7744
Epoch 53/400
0.6930
Epoch 54/400
640/640 [=============] - Os 80us/step - loss: 0.5111 - mse:
0.5111
Epoch 55/400
0.8187
Epoch 56/400
0.6219
Epoch 57/400
0.6732
Epoch 58/400
0.5999
Epoch 59/400
0.5927
Epoch 60/400
0.6565
Epoch 61/400
```

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0.5583
Epoch 62/400
0.6749
Epoch 63/400
0.7847
Epoch 64/400
0.5961
Epoch 65/400
0.6109
Epoch 66/400
0.9075
Epoch 67/400
0.6595
Epoch 68/400
0.7680
Epoch 69/400
0.6373
Epoch 70/400
0.6932
Epoch 71/400
0.6996
Epoch 72/400
0.7339
Epoch 73/400
0.6508
Epoch 74/400
0.6871
Epoch 75/400
0.7960
Epoch 76/400
0.6158
Epoch 77/400
```

```
0.5133
Epoch 78/400
0.6267
Epoch 79/400
0.4714
Epoch 80/400
0.5836
Epoch 81/400
0.5959
Epoch 82/400
0.5602
Epoch 83/400
640/640 [============== ] - Os 71us/step - loss: 0.4210 - mse:
0.4210
Epoch 84/400
0.7371
Epoch 85/400
640/640 [============== ] - Os 72us/step - loss: 0.6839 - mse:
0.6839
Epoch 86/400
640/640 [=============] - Os 75us/step - loss: 0.8328 - mse:
0.8328
Epoch 87/400
0.6442
Epoch 88/400
0.5691
Epoch 89/400
0.6019
Epoch 90/400
0.5932
Epoch 91/400
0.5887
Epoch 92/400
0.6236
Epoch 93/400
```

```
0.6302
Epoch 94/400
640/640 [=============] - Os 73us/step - loss: 0.4722 - mse:
0.4722
Epoch 95/400
0.5833
Epoch 96/400
0.5144
Epoch 97/400
0.7952
Epoch 98/400
0.5349
Epoch 99/400
0.6835
Epoch 100/400
0.4561
Epoch 101/400
0.5265
Epoch 102/400
640/640 [=============] - Os 63us/step - loss: 0.5366 - mse:
0.5366
Epoch 103/400
0.5675
Epoch 104/400
0.3779
Epoch 105/400
0.5246
Epoch 106/400
0.3732
Epoch 107/400
0.5546
Epoch 108/400
0.4738
Epoch 109/400
```

```
640/640 [================ ] - Os 104us/step - loss: 0.6577 - mse:
0.6577
Epoch 110/400
0.6806
Epoch 111/400
0.5416
Epoch 112/400
640/640 [============== ] - Os 74us/step - loss: 0.4652 - mse:
0.4652
Epoch 113/400
0.4023
Epoch 114/400
0.5503
Epoch 115/400
640/640 [============== ] - Os 62us/step - loss: 0.5649 - mse:
0.5649
Epoch 116/400
0.4386
Epoch 117/400
640/640 [============== ] - Os 57us/step - loss: 0.5442 - mse:
0.5442
Epoch 118/400
640/640 [============] - Os 66us/step - loss: 0.4824 - mse:
0.4824
Epoch 119/400
0.5432
Epoch 120/400
0.5504
Epoch 121/400
0.4369
Epoch 122/400
0.4537
Epoch 123/400
0.4550
Epoch 124/400
0.6031
Epoch 125/400
```

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0.4876
Epoch 126/400
640/640 [============== ] - Os 62us/step - loss: 0.4250 - mse:
0.4250
Epoch 127/400
0.4614
Epoch 128/400
640/640 [============== ] - Os 63us/step - loss: 0.4644 - mse:
0.4644
Epoch 129/400
0.6168
Epoch 130/400
0.5291
Epoch 131/400
640/640 [============== ] - Os 75us/step - loss: 0.3988 - mse:
0.3988
Epoch 132/400
0.4402
Epoch 133/400
640/640 [============== ] - Os 74us/step - loss: 0.4999 - mse:
0.4999
Epoch 134/400
640/640 [============] - Os 70us/step - loss: 0.4584 - mse:
0.4584
Epoch 135/400
0.4197
Epoch 136/400
0.4693
Epoch 137/400
0.5158
Epoch 138/400
0.5386
Epoch 139/400
0.5591
Epoch 140/400
0.4721
Epoch 141/400
```

```
0.5138
Epoch 142/400
640/640 [============== ] - Os 72us/step - loss: 0.6017 - mse:
0.6017
Epoch 143/400
0.5052
Epoch 144/400
640/640 [============== ] - Os 55us/step - loss: 0.5714 - mse:
0.5714
Epoch 145/400
0.5793
Epoch 146/400
0.6520
Epoch 147/400
640/640 [============== ] - Os 63us/step - loss: 0.7044 - mse:
0.7044
Epoch 148/400
0.6093
Epoch 149/400
0.4154
Epoch 150/400
640/640 [=============] - Os 64us/step - loss: 0.4091 - mse:
0.4091
Epoch 151/400
0.5765
Epoch 152/400
0.4806
Epoch 153/400
0.4547
Epoch 154/400
0.5213
Epoch 155/400
0.6227
Epoch 156/400
0.6539
Epoch 157/400
```

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0.4391
Epoch 158/400
0.5262
Epoch 159/400
0.5114
Epoch 160/400
0.5156
Epoch 161/400
640/640 [=============] - Os 118us/step - loss: 0.5390 - mse:
0.5390
Epoch 162/400
640/640 [================ ] - Os 103us/step - loss: 0.5184 - mse:
0.5184
Epoch 163/400
640/640 [============= ] - Os 102us/step - loss: 0.6504 - mse:
0.6504
Epoch 164/400
0.4801
Epoch 165/400
640/640 [============== ] - Os 97us/step - loss: 0.4709 - mse:
0.4709
Epoch 166/400
640/640 [============] - Os 128us/step - loss: 0.4178 - mse:
0.4178
Epoch 167/400
640/640 [=============== ] - Os 152us/step - loss: 0.4509 - mse:
0.4509
Epoch 168/400
640/640 [============= ] - Os 213us/step - loss: 0.3910 - mse:
0.3910
Epoch 169/400
640/640 [============== ] - Os 166us/step - loss: 0.4272 - mse:
0.4272
Epoch 170/400
640/640 [=============== ] - Os 131us/step - loss: 0.5176 - mse:
0.5176
Epoch 171/400
640/640 [============] - Os 155us/step - loss: 0.6255 - mse:
0.6255
Epoch 172/400
640/640 [=============== ] - Os 308us/step - loss: 0.5471 - mse:
0.5471
Epoch 173/400
```

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640/640 [================ ] - Os 163us/step - loss: 0.6517 - mse:
0.6517
Epoch 174/400
640/640 [============= ] - Os 157us/step - loss: 0.4971 - mse:
0.4971
Epoch 175/400
640/640 [=============== ] - 0s 114us/step - loss: 0.4771 - mse:
0.4771
Epoch 176/400
640/640 [============== ] - Os 98us/step - loss: 0.3247 - mse:
0.3247
Epoch 177/400
0.4652
Epoch 178/400
0.4141
Epoch 179/400
640/640 [============== ] - Os 85us/step - loss: 0.4845 - mse:
0.4845
Epoch 180/400
0.5354
Epoch 181/400
0.5216
Epoch 182/400
640/640 [============] - Os 103us/step - loss: 0.6243 - mse:
0.6243
Epoch 183/400
0.4611
Epoch 184/400
640/640 [============== ] - Os 91us/step - loss: 0.3919 - mse:
0.3919
Epoch 185/400
0.6254
Epoch 186/400
0.4606
Epoch 187/400
0.3850
Epoch 188/400
0.3392
Epoch 189/400
```

```
0.4280
Epoch 190/400
640/640 [============== ] - Os 61us/step - loss: 0.4200 - mse:
0.4200
Epoch 191/400
0.5494
Epoch 192/400
640/640 [============== ] - Os 58us/step - loss: 0.4489 - mse:
0.4489
Epoch 193/400
0.4278
Epoch 194/400
0.5221
Epoch 195/400
0.3433
Epoch 196/400
0.4833
Epoch 197/400
640/640 [============== ] - Os 65us/step - loss: 0.4848 - mse:
0.4848
Epoch 198/400
640/640 [=============] - Os 62us/step - loss: 0.5027 - mse:
0.5027
Epoch 199/400
0.4216
Epoch 200/400
640/640 [============== ] - Os 73us/step - loss: 0.4525 - mse:
0.4525
Epoch 201/400
0.4119
Epoch 202/400
640/640 [=============== ] - Os 122us/step - loss: 0.4047 - mse:
0.4047
Epoch 203/400
640/640 [================ ] - Os 110us/step - loss: 0.6396 - mse:
0.6396
Epoch 204/400
640/640 [=============== ] - Os 111us/step - loss: 0.4083 - mse:
0.4083
Epoch 205/400
```

```
640/640 [================ ] - Os 120us/step - loss: 0.5504 - mse:
0.5504
Epoch 206/400
640/640 [============== ] - Os 140us/step - loss: 0.5262 - mse:
0.5262
Epoch 207/400
640/640 [=============== ] - 0s 142us/step - loss: 0.4403 - mse:
0.4403
Epoch 208/400
640/640 [============= ] - Os 114us/step - loss: 0.5963 - mse:
0.5963
Epoch 209/400
0.3513
Epoch 210/400
0.3906
Epoch 211/400
640/640 [============== ] - Os 64us/step - loss: 0.4415 - mse:
0.4415
Epoch 212/400
0.4088
Epoch 213/400
640/640 [============== ] - Os 71us/step - loss: 0.3692 - mse:
0.3692
Epoch 214/400
640/640 [============] - Os 78us/step - loss: 0.4959 - mse:
0.4959
Epoch 215/400
0.6028
Epoch 216/400
0.5457
Epoch 217/400
0.4896
Epoch 218/400
0.5013
Epoch 219/400
0.4245
Epoch 220/400
0.3599
Epoch 221/400
```

```
0.4199
Epoch 222/400
0.4627
Epoch 223/400
0.4877
Epoch 224/400
640/640 [============== ] - Os 69us/step - loss: 0.4949 - mse:
0.4949
Epoch 225/400
0.4465
Epoch 226/400
0.3610
Epoch 227/400
640/640 [============== ] - Os 68us/step - loss: 0.4683 - mse:
0.4683
Epoch 228/400
0.4128
Epoch 229/400
0.5949
Epoch 230/400
640/640 [=============] - Os 68us/step - loss: 0.3338 - mse:
0.3338
Epoch 231/400
0.3897
Epoch 232/400
640/640 [============= ] - Os 120us/step - loss: 0.4451 - mse:
0.4451
Epoch 233/400
640/640 [============== ] - Os 117us/step - loss: 0.4270 - mse:
0.4270
Epoch 234/400
640/640 [============== ] - Os 126us/step - loss: 0.3643 - mse:
0.3643
Epoch 235/400
640/640 [================ ] - Os 106us/step - loss: 0.4987 - mse:
0.4987
Epoch 236/400
0.3600
Epoch 237/400
```

```
640/640 [=============== ] - Os 126us/step - loss: 0.3880 - mse:
0.3880
Epoch 238/400
640/640 [============== ] - Os 150us/step - loss: 0.4138 - mse:
0.4138
Epoch 239/400
0.5214
Epoch 240/400
640/640 [============== ] - Os 150us/step - loss: 0.4058 - mse:
0.4058
Epoch 241/400
640/640 [=============] - Os 138us/step - loss: 0.3471 - mse:
0.3471
Epoch 242/400
640/640 [=============== ] - Os 129us/step - loss: 0.3800 - mse:
0.3800
Epoch 243/400
640/640 [============= ] - Os 117us/step - loss: 0.5907 - mse:
0.5907
Epoch 244/400
640/640 [============= ] - Os 116us/step - loss: 0.4395 - mse:
0.4395
Epoch 245/400
640/640 [============= ] - Os 184us/step - loss: 0.4722 - mse:
0.4722
Epoch 246/400
640/640 [============] - Os 129us/step - loss: 0.4882 - mse:
0.4882
Epoch 247/400
640/640 [=============== ] - Os 112us/step - loss: 0.3647 - mse:
0.3647
Epoch 248/400
640/640 [============= ] - Os 136us/step - loss: 0.4760 - mse:
0.4760
Epoch 249/400
640/640 [============== ] - Os 105us/step - loss: 0.4398 - mse:
0.4398
Epoch 250/400
640/640 [============== ] - Os 107us/step - loss: 0.3303 - mse:
0.3303
Epoch 251/400
640/640 [================ ] - Os 138us/step - loss: 0.3419 - mse:
0.3419
Epoch 252/400
640/640 [================= ] - Os 133us/step - loss: 0.4275 - mse:
0.4275
Epoch 253/400
```

```
640/640 [=============== ] - Os 122us/step - loss: 0.3788 - mse:
0.3788
Epoch 254/400
640/640 [============= ] - Os 116us/step - loss: 0.4566 - mse:
0.4566
Epoch 255/400
640/640 [=============== ] - 0s 112us/step - loss: 0.3935 - mse:
0.3935
Epoch 256/400
640/640 [============== ] - Os 77us/step - loss: 0.4693 - mse:
0.4693
Epoch 257/400
0.5260
Epoch 258/400
640/640 [================= ] - 0s 133us/step - loss: 0.5338 - mse:
0.5338
Epoch 259/400
640/640 [============= ] - Os 114us/step - loss: 0.5539 - mse:
0.5539
Epoch 260/400
640/640 [============= ] - Os 108us/step - loss: 0.4785 - mse:
0.4785
Epoch 261/400
640/640 [============= ] - Os 115us/step - loss: 0.4754 - mse:
0.4754
Epoch 262/400
640/640 [============] - Os 102us/step - loss: 0.3422 - mse:
0.3422
Epoch 263/400
0.4118
Epoch 264/400
0.4949
Epoch 265/400
0.5105
Epoch 266/400
640/640 [=============== ] - Os 105us/step - loss: 0.5355 - mse:
0.5355
Epoch 267/400
0.3388
Epoch 268/400
0.5284
Epoch 269/400
```

```
0.3962
Epoch 270/400
640/640 [=============] - Os 76us/step - loss: 0.4454 - mse:
0.4454
Epoch 271/400
0.5003
Epoch 272/400
640/640 [============== ] - Os 93us/step - loss: 0.3784 - mse:
0.3784
Epoch 273/400
0.3704
Epoch 274/400
0.3915
Epoch 275/400
640/640 [============= ] - Os 108us/step - loss: 0.4681 - mse:
0.4681
Epoch 276/400
640/640 [============= ] - Os 113us/step - loss: 0.5507 - mse:
0.5507
Epoch 277/400
0.3324
Epoch 278/400
640/640 [============] - Os 112us/step - loss: 0.4883 - mse:
0.4883
Epoch 279/400
0.5128
Epoch 280/400
640/640 [============== ] - Os 73us/step - loss: 0.5134 - mse:
0.5134
Epoch 281/400
0.4225
Epoch 282/400
0.5154
Epoch 283/400
0.5000
Epoch 284/400
0.5422
Epoch 285/400
```

```
0.4057
Epoch 286/400
0.4629
Epoch 287/400
0.3808
Epoch 288/400
0.4395
Epoch 289/400
0.3742
Epoch 290/400
0.4655
Epoch 291/400
0.5637
Epoch 292/400
0.3570
Epoch 293/400
640/640 [============= ] - Os 83us/step - loss: 0.3200 - mse:
0.3200
Epoch 294/400
640/640 [=============] - Os 90us/step - loss: 0.4146 - mse:
0.4146
Epoch 295/400
0.3897
Epoch 296/400
0.4116
Epoch 297/400
0.3387
Epoch 298/400
640/640 [============== ] - Os 102us/step - loss: 0.5658 - mse:
0.5658
Epoch 299/400
640/640 [================ ] - Os 102us/step - loss: 0.3643 - mse:
0.3643
Epoch 300/400
0.3901
Epoch 301/400
```

```
640/640 [=============== ] - Os 100us/step - loss: 0.3657 - mse:
0.3657
Epoch 302/400
0.5815
Epoch 303/400
0.4115
Epoch 304/400
640/640 [============== ] - Os 72us/step - loss: 0.3831 - mse:
0.3831
Epoch 305/400
0.3471
Epoch 306/400
0.3723
Epoch 307/400
640/640 [============== ] - Os 82us/step - loss: 0.4474 - mse:
0.4474
Epoch 308/400
640/640 [============== ] - Os 101us/step - loss: 0.3540 - mse:
0.3540
Epoch 309/400
640/640 [============== ] - Os 97us/step - loss: 0.4465 - mse:
0.4465
Epoch 310/400
640/640 [============] - Os 103us/step - loss: 0.3202 - mse:
0.3202
Epoch 311/400
640/640 [=============== ] - Os 100us/step - loss: 0.4505 - mse:
0.4505
Epoch 312/400
640/640 [============== ] - Os 87us/step - loss: 0.2303 - mse:
0.2303
Epoch 313/400
0.2836
Epoch 314/400
0.2936
Epoch 315/400
640/640 [================ ] - Os 104us/step - loss: 0.3823 - mse:
0.3823
Epoch 316/400
640/640 [================= ] - Os 107us/step - loss: 0.3597 - mse:
0.3597
Epoch 317/400
```

```
0.4189
Epoch 318/400
0.3856
Epoch 319/400
0.4468
Epoch 320/400
640/640 [============== ] - Os 70us/step - loss: 0.3048 - mse:
0.3048
Epoch 321/400
0.3777
Epoch 322/400
0.4469
Epoch 323/400
640/640 [============== ] - Os 69us/step - loss: 0.3103 - mse:
0.3103
Epoch 324/400
0.4350
Epoch 325/400
640/640 [============== ] - Os 83us/step - loss: 0.4397 - mse:
0.4397
Epoch 326/400
640/640 [=============] - Os 94us/step - loss: 0.4019 - mse:
0.4019
Epoch 327/400
0.5710
Epoch 328/400
640/640 [============== ] - Os 96us/step - loss: 0.3629 - mse:
0.3629
Epoch 329/400
0.3832
Epoch 330/400
0.3931
Epoch 331/400
0.5326
Epoch 332/400
0.3311
Epoch 333/400
```

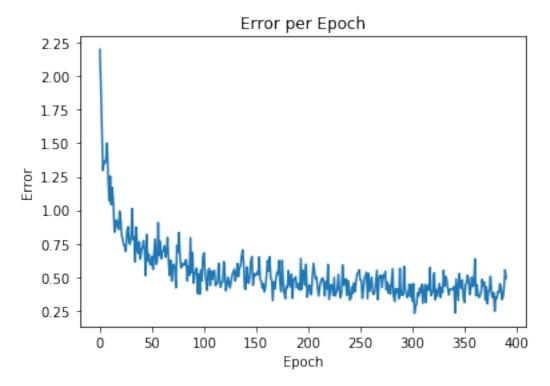
```
0.4074
Epoch 334/400
0.3581
Epoch 335/400
0.4030
Epoch 336/400
640/640 [============== ] - Os 72us/step - loss: 0.4532 - mse:
0.4532
Epoch 337/400
640/640 [=============] - Os 106us/step - loss: 0.3479 - mse:
0.3479
Epoch 338/400
640/640 [=============== ] - 0s 112us/step - loss: 0.4240 - mse:
0.4240
Epoch 339/400
640/640 [============== ] - Os 93us/step - loss: 0.3852 - mse:
0.3852
Epoch 340/400
0.5516
Epoch 341/400
640/640 [============== ] - Os 92us/step - loss: 0.4327 - mse:
0.4327
Epoch 342/400
640/640 [=============] - Os 87us/step - loss: 0.5037 - mse:
0.5037
Epoch 343/400
0.4786
Epoch 344/400
640/640 [============== ] - Os 79us/step - loss: 0.4136 - mse:
0.4136
Epoch 345/400
0.3667
Epoch 346/400
0.4105
Epoch 347/400
0.4132
Epoch 348/400
640/640 [=============== ] - Os 111us/step - loss: 0.4141 - mse:
0.4141
Epoch 349/400
```

```
0.4086
Epoch 350/400
640/640 [=============] - Os 137us/step - loss: 0.4303 - mse:
0.4303
Epoch 351/400
640/640 [=============== ] - Os 154us/step - loss: 0.2346 - mse:
0.2346
Epoch 352/400
640/640 [============= ] - Os 176us/step - loss: 0.4871 - mse:
0.4871
Epoch 353/400
640/640 [============] - Os 116us/step - loss: 0.4023 - mse:
0.4023
Epoch 354/400
0.3254
Epoch 355/400
640/640 [============= ] - Os 131us/step - loss: 0.5272 - mse:
0.5272
Epoch 356/400
640/640 [============= ] - Os 117us/step - loss: 0.4390 - mse:
0.4390
Epoch 357/400
0.4176
Epoch 358/400
640/640 [============] - Os 92us/step - loss: 0.4788 - mse:
0.4788
Epoch 359/400
0.3402
Epoch 360/400
0.3108
Epoch 361/400
0.4429
Epoch 362/400
0.4598
Epoch 363/400
0.5152
Epoch 364/400
640/640 [=============== ] - Os 127us/step - loss: 0.4818 - mse:
0.4818
Epoch 365/400
```

```
640/640 [================== ] - Os 147us/step - loss: 0.3695 - mse:
0.3695
Epoch 366/400
640/640 [============= ] - Os 132us/step - loss: 0.4116 - mse:
0.4116
Epoch 367/400
640/640 [=============== ] - 0s 136us/step - loss: 0.4998 - mse:
0.4998
Epoch 368/400
640/640 [============= ] - Os 135us/step - loss: 0.4310 - mse:
0.4310
Epoch 369/400
640/640 [=============] - Os 115us/step - loss: 0.4412 - mse:
0.4412
Epoch 370/400
640/640 [=============== ] - Os 110us/step - loss: 0.6379 - mse:
0.6379
Epoch 371/400
640/640 [============= ] - Os 147us/step - loss: 0.3628 - mse:
0.3628
Epoch 372/400
640/640 [============] - Os 110us/step - loss: 0.4431 - mse:
0.4431
Epoch 373/400
640/640 [============= ] - Os 121us/step - loss: 0.3628 - mse:
0.3628
Epoch 374/400
640/640 [============] - Os 132us/step - loss: 0.3484 - mse:
0.3484
Epoch 375/400
640/640 [================= ] - Os 125us/step - loss: 0.3617 - mse:
0.3617
Epoch 376/400
640/640 [============= ] - Os 124us/step - loss: 0.3901 - mse:
0.3901
Epoch 377/400
640/640 [============== ] - Os 163us/step - loss: 0.4841 - mse:
0.4841
Epoch 378/400
640/640 [=============== ] - Os 129us/step - loss: 0.4366 - mse:
0.4366
Epoch 379/400
640/640 [================= ] - Os 103us/step - loss: 0.4640 - mse:
0.4640
Epoch 380/400
0.3299
Epoch 381/400
```

```
0.4290
Epoch 382/400
0.3009
Epoch 383/400
0.3900
Epoch 384/400
640/640 [============== ] - Os 72us/step - loss: 0.4111 - mse:
0.4111
Epoch 385/400
0.5051
Epoch 386/400
0.3731
Epoch 387/400
640/640 [============== ] - Os 99us/step - loss: 0.3483 - mse:
0.3483
Epoch 388/400
640/640 [============= ] - Os 107us/step - loss: 0.3858 - mse:
0.3858
Epoch 389/400
640/640 [============= ] - Os 120us/step - loss: 0.2469 - mse:
0.2469
Epoch 390/400
640/640 [============] - Os 121us/step - loss: 0.3520 - mse:
0.3520
Epoch 391/400
640/640 [=============== ] - Os 111us/step - loss: 0.3424 - mse:
0.3424
Epoch 392/400
640/640 [============= ] - Os 118us/step - loss: 0.3921 - mse:
0.3921
Epoch 393/400
640/640 [============== ] - Os 112us/step - loss: 0.3836 - mse:
0.3836
Epoch 394/400
640/640 [=============== ] - Os 101us/step - loss: 0.4511 - mse:
0.4511
Epoch 395/400
0.4256
Epoch 396/400
0.3336
Epoch 397/400
```

```
640/640 [============= ] - Os 84us/step - loss: 0.3536 - mse:
    0.3536
    Epoch 398/400
    640/640 [=====
                                   ======] - Os 90us/step - loss: 0.4270 - mse:
    0.4270
    Epoch 399/400
                                         ==] - Os 85us/step - loss: 0.5540 - mse:
    640/640 [====
    0.5540
    Epoch 400/400
    640/640 [===
                                          =] - Os 84us/step - loss: 0.4918 - mse:
    0.4918
[6]: plt.plot(result.history['mse'][9:], label='Error')
    plt.title("Error per Epoch")
     plt.ylabel("Error")
     plt.xlabel("Epoch")
     plt.show()
```



Below are the evaluation and prediction of the test data.

```
[7]: [2.239789342880249, 2.2397892475128174]
[8]: model.predict(X_test)
[8]: array([[16.066086],
            [15.98393],
            [11.870772],
            [12.667431],
            [13.573432],
            [ 9.863728 ],
            [11.944125],
            [12.73495],
            [10.266866],
            [10.733826],
            [11.250207],
            [11.349945],
            [12.134901],
            [10.847218],
            [15.681121],
            [10.969329],
            [11.385669],
            [14.39683],
            [17.971432],
            [12.48136],
            [11.440838],
            [16.00816],
            [14.516333],
            [16.691385],
            [13.32311],
            [11.829491],
            [13.055396],
            [14.968102],
            [10.799687],
            [12.107593],
            [17.106564],
            [14.672067],
            [14.521335],
            [15.66048],
            [16.555681],
            [16.140974],
            [13.721002],
            [14.027346],
            [16.178965],
            [14.742581],
            [17.239153],
            [16.202415],
```

[15.156428],

```
[12.744971],
```

- [12.250231],
- [12.929467],
- [12.152541],
- [13.43007],
- [13.23071],
- [11.864444],
- [12.834146],
- [13.166191],
- [12.937608],
- [17.727953],
- [11.552488],
- [11.1039 ],
- [11.717118],
- [12.875093],
- [10.829951],
- [12.348716],
- [11.54097],
- [13.630561],
- [13.920239],
- [17.285248],
- [16.258028],
- [15.760074],
- [16.725489],
- [11.452701],
- [14.409041],
- [17.252888],
- [12.58702],
- [14.398694],
- [16.390593],
- [16.250315],
- [12.951282],
- [14.656769],
- [16.84159],
- [16.614618],
- [16.754156],
- [15.919413],
- [16.406979],
- [16.297445],
- [12.895689],
- [15.435889],
- [15.8416815],
- [17.175055],
- [14.358504],
- [16.414505],
- [15.16162],
- [14.263532],

```
[14.263532],

[16.734726],

[17.18942],

[16.884193],

[13.684597],

[17.30677],

[17.564352],

[13.62445],

[13.782347],

[18.745096]], dtype=float32)
```

#### 2 Classification

For the second part we had to classify the data into 2 categories. Same architecture is used as the previous model with the exception that the output layer's activation function is sigmoid instead of relu. binary\_crossentropy is used as loss function and accuracy is used as metric. The final result of the network's training is and accuracy between 70% and 80% on the train data.

```
Epoch 6/150
640/640 [============== ] - 0s 42us/step - loss: 0.5615 -
accuracy: 0.6844 - val_loss: 0.5242 - val_accuracy: 0.7658
Epoch 7/150
accuracy: 0.7156 - val_loss: 0.5113 - val_accuracy: 0.7793
accuracy: 0.7219 - val_loss: 0.5007 - val_accuracy: 0.7748
Epoch 9/150
640/640 [============== ] - 0s 39us/step - loss: 0.5241 -
accuracy: 0.7234 - val_loss: 0.4928 - val_accuracy: 0.7883
Epoch 10/150
640/640 [============= - - 0s 39us/step - loss: 0.5023 -
accuracy: 0.7234 - val_loss: 0.4862 - val_accuracy: 0.7928
Epoch 11/150
640/640 [============== ] - 0s 39us/step - loss: 0.4908 -
accuracy: 0.7422 - val_loss: 0.4797 - val_accuracy: 0.7883
Epoch 12/150
640/640 [============== ] - 0s 45us/step - loss: 0.4840 -
accuracy: 0.7437 - val_loss: 0.4748 - val_accuracy: 0.7883
Epoch 13/150
accuracy: 0.7578 - val_loss: 0.4703 - val_accuracy: 0.7793
Epoch 14/150
accuracy: 0.7547 - val_loss: 0.4662 - val_accuracy: 0.7793
Epoch 15/150
640/640 [============= - - os 80us/step - loss: 0.4749 -
accuracy: 0.7688 - val_loss: 0.4627 - val_accuracy: 0.7793
Epoch 16/150
accuracy: 0.7484 - val_loss: 0.4593 - val_accuracy: 0.7748
Epoch 17/150
accuracy: 0.7797 - val_loss: 0.4563 - val_accuracy: 0.7838
Epoch 18/150
accuracy: 0.7797 - val_loss: 0.4542 - val_accuracy: 0.7793
Epoch 19/150
accuracy: 0.7781 - val_loss: 0.4514 - val_accuracy: 0.7883
640/640 [============ ] - 0s 43us/step - loss: 0.4432 -
accuracy: 0.7750 - val_loss: 0.4505 - val_accuracy: 0.7883
Epoch 21/150
accuracy: 0.7891 - val_loss: 0.4483 - val_accuracy: 0.7883
```

```
Epoch 22/150
640/640 [============== ] - 0s 46us/step - loss: 0.4337 -
accuracy: 0.7828 - val_loss: 0.4469 - val_accuracy: 0.7883
Epoch 23/150
accuracy: 0.7953 - val_loss: 0.4443 - val_accuracy: 0.7883
Epoch 24/150
accuracy: 0.7844 - val_loss: 0.4434 - val_accuracy: 0.7883
Epoch 25/150
640/640 [============= ] - 0s 47us/step - loss: 0.4145 -
accuracy: 0.7922 - val_loss: 0.4430 - val_accuracy: 0.7793
Epoch 26/150
640/640 [============= ] - 0s 42us/step - loss: 0.4206 -
accuracy: 0.7859 - val_loss: 0.4430 - val_accuracy: 0.7838
Epoch 27/150
640/640 [============== ] - 0s 44us/step - loss: 0.4004 -
accuracy: 0.8078 - val_loss: 0.4414 - val_accuracy: 0.7838
Epoch 28/150
accuracy: 0.8031 - val_loss: 0.4413 - val_accuracy: 0.7838
Epoch 29/150
accuracy: 0.8016 - val_loss: 0.4406 - val_accuracy: 0.7883
Epoch 30/150
accuracy: 0.8078 - val_loss: 0.4400 - val_accuracy: 0.7883
Epoch 31/150
640/640 [============= - - 0s 38us/step - loss: 0.4047 -
accuracy: 0.8047 - val_loss: 0.4406 - val_accuracy: 0.7883
Epoch 32/150
640/640 [============== ] - 0s 38us/step - loss: 0.4025 -
accuracy: 0.7875 - val_loss: 0.4399 - val_accuracy: 0.7883
Epoch 33/150
accuracy: 0.8078 - val_loss: 0.4391 - val_accuracy: 0.7928
Epoch 34/150
accuracy: 0.7969 - val_loss: 0.4384 - val_accuracy: 0.8063
Epoch 35/150
640/640 [============= ] - Os 41us/step - loss: 0.3986 -
accuracy: 0.7922 - val_loss: 0.4374 - val_accuracy: 0.8063
Epoch 36/150
640/640 [============= - - os 48us/step - loss: 0.3976 -
accuracy: 0.8094 - val_loss: 0.4374 - val_accuracy: 0.8018
Epoch 37/150
accuracy: 0.7984 - val_loss: 0.4371 - val_accuracy: 0.7973
```

```
Epoch 38/150
640/640 [============== ] - 0s 43us/step - loss: 0.3856 -
accuracy: 0.8062 - val_loss: 0.4380 - val_accuracy: 0.8153
Epoch 39/150
accuracy: 0.8000 - val_loss: 0.4369 - val_accuracy: 0.8198
Epoch 40/150
accuracy: 0.8000 - val_loss: 0.4377 - val_accuracy: 0.8153
Epoch 41/150
640/640 [============== ] - 0s 43us/step - loss: 0.3929 -
accuracy: 0.8094 - val_loss: 0.4370 - val_accuracy: 0.8153
Epoch 42/150
640/640 [============= - - os 40us/step - loss: 0.4003 -
accuracy: 0.8000 - val_loss: 0.4362 - val_accuracy: 0.8153
Epoch 43/150
640/640 [============= ] - 0s 43us/step - loss: 0.3914 -
accuracy: 0.8109 - val_loss: 0.4362 - val_accuracy: 0.8198
Epoch 44/150
accuracy: 0.7922 - val_loss: 0.4367 - val_accuracy: 0.8198
Epoch 45/150
accuracy: 0.7969 - val_loss: 0.4362 - val_accuracy: 0.8198
Epoch 46/150
accuracy: 0.8109 - val_loss: 0.4368 - val_accuracy: 0.8153
Epoch 47/150
640/640 [============= - - os 41us/step - loss: 0.3861 -
accuracy: 0.8047 - val_loss: 0.4377 - val_accuracy: 0.8153
Epoch 48/150
accuracy: 0.8031 - val_loss: 0.4371 - val_accuracy: 0.8153
Epoch 49/150
accuracy: 0.8078 - val_loss: 0.4358 - val_accuracy: 0.8153
Epoch 50/150
accuracy: 0.8172 - val_loss: 0.4346 - val_accuracy: 0.8153
Epoch 51/150
accuracy: 0.8078 - val_loss: 0.4338 - val_accuracy: 0.8153
640/640 [============= - - os 40us/step - loss: 0.3762 -
accuracy: 0.8125 - val_loss: 0.4330 - val_accuracy: 0.8153
Epoch 53/150
accuracy: 0.8141 - val_loss: 0.4321 - val_accuracy: 0.8153
```

```
Epoch 54/150
640/640 [============= ] - 0s 43us/step - loss: 0.3841 -
accuracy: 0.8109 - val_loss: 0.4318 - val_accuracy: 0.8108
Epoch 55/150
accuracy: 0.8172 - val_loss: 0.4317 - val_accuracy: 0.8153
accuracy: 0.8141 - val_loss: 0.4322 - val_accuracy: 0.8108
Epoch 57/150
accuracy: 0.8062 - val_loss: 0.4315 - val_accuracy: 0.8153
Epoch 58/150
640/640 [============ - - os 40us/step - loss: 0.3961 -
accuracy: 0.7937 - val_loss: 0.4324 - val_accuracy: 0.8153
Epoch 59/150
640/640 [============== ] - 0s 37us/step - loss: 0.3831 -
accuracy: 0.8094 - val_loss: 0.4328 - val_accuracy: 0.8153
Epoch 60/150
accuracy: 0.8250 - val_loss: 0.4323 - val_accuracy: 0.8153
Epoch 61/150
accuracy: 0.8156 - val_loss: 0.4319 - val_accuracy: 0.8153
Epoch 62/150
accuracy: 0.8219 - val_loss: 0.4325 - val_accuracy: 0.8153
Epoch 63/150
640/640 [============ - - os 41us/step - loss: 0.3782 -
accuracy: 0.8172 - val_loss: 0.4332 - val_accuracy: 0.8243
Epoch 64/150
640/640 [============== ] - 0s 42us/step - loss: 0.3882 -
accuracy: 0.8047 - val_loss: 0.4326 - val_accuracy: 0.8198
Epoch 65/150
accuracy: 0.8125 - val_loss: 0.4324 - val_accuracy: 0.8198
Epoch 66/150
accuracy: 0.8125 - val_loss: 0.4321 - val_accuracy: 0.8198
Epoch 67/150
accuracy: 0.8125 - val_loss: 0.4323 - val_accuracy: 0.8198
640/640 [============ - - os 63us/step - loss: 0.3732 -
accuracy: 0.8062 - val_loss: 0.4316 - val_accuracy: 0.8198
Epoch 69/150
accuracy: 0.8141 - val_loss: 0.4330 - val_accuracy: 0.8198
```

```
Epoch 70/150
640/640 [============== ] - 0s 81us/step - loss: 0.3753 -
accuracy: 0.8203 - val_loss: 0.4337 - val_accuracy: 0.8198
Epoch 71/150
accuracy: 0.8078 - val_loss: 0.4344 - val_accuracy: 0.8153
accuracy: 0.8156 - val_loss: 0.4338 - val_accuracy: 0.8108
Epoch 73/150
accuracy: 0.8094 - val_loss: 0.4335 - val_accuracy: 0.8108
Epoch 74/150
640/640 [============= - - os 78us/step - loss: 0.3719 -
accuracy: 0.8125 - val_loss: 0.4335 - val_accuracy: 0.8108
Epoch 75/150
640/640 [============== ] - 0s 79us/step - loss: 0.3684 -
accuracy: 0.8250 - val_loss: 0.4335 - val_accuracy: 0.8153
Epoch 76/150
accuracy: 0.8203 - val_loss: 0.4337 - val_accuracy: 0.8153
Epoch 77/150
accuracy: 0.8172 - val_loss: 0.4327 - val_accuracy: 0.8108
Epoch 78/150
accuracy: 0.8141 - val_loss: 0.4328 - val_accuracy: 0.8108
Epoch 79/150
640/640 [============= - - 0s 62us/step - loss: 0.3864 -
accuracy: 0.8016 - val_loss: 0.4315 - val_accuracy: 0.8153
Epoch 80/150
accuracy: 0.8062 - val_loss: 0.4317 - val_accuracy: 0.8153
Epoch 81/150
accuracy: 0.8094 - val_loss: 0.4315 - val_accuracy: 0.8153
Epoch 82/150
accuracy: 0.8078 - val_loss: 0.4311 - val_accuracy: 0.8153
Epoch 83/150
accuracy: 0.8250 - val_loss: 0.4320 - val_accuracy: 0.8198
640/640 [============= - - os 47us/step - loss: 0.3848 -
accuracy: 0.8047 - val_loss: 0.4322 - val_accuracy: 0.8198
Epoch 85/150
accuracy: 0.8062 - val_loss: 0.4324 - val_accuracy: 0.8108
```

```
Epoch 86/150
accuracy: 0.8297 - val_loss: 0.4327 - val_accuracy: 0.8198
Epoch 87/150
accuracy: 0.8281 - val_loss: 0.4321 - val_accuracy: 0.8153
accuracy: 0.8359 - val_loss: 0.4318 - val_accuracy: 0.8198
Epoch 89/150
640/640 [============= ] - 0s 40us/step - loss: 0.3726 -
accuracy: 0.8266 - val_loss: 0.4338 - val_accuracy: 0.8198
Epoch 90/150
640/640 [============= - - os 43us/step - loss: 0.3671 -
accuracy: 0.8094 - val_loss: 0.4327 - val_accuracy: 0.8108
Epoch 91/150
640/640 [============== ] - 0s 36us/step - loss: 0.3706 -
accuracy: 0.8297 - val_loss: 0.4332 - val_accuracy: 0.8108
Epoch 92/150
640/640 [============== ] - 0s 46us/step - loss: 0.3680 -
accuracy: 0.8203 - val_loss: 0.4335 - val_accuracy: 0.8063
Epoch 93/150
accuracy: 0.8188 - val_loss: 0.4311 - val_accuracy: 0.8198
Epoch 94/150
accuracy: 0.8078 - val_loss: 0.4297 - val_accuracy: 0.8198
Epoch 95/150
640/640 [============= - - 0s 51us/step - loss: 0.3772 -
accuracy: 0.8094 - val_loss: 0.4326 - val_accuracy: 0.8153
Epoch 96/150
accuracy: 0.8188 - val_loss: 0.4327 - val_accuracy: 0.8198
Epoch 97/150
accuracy: 0.8078 - val_loss: 0.4322 - val_accuracy: 0.8153
Epoch 98/150
accuracy: 0.8125 - val_loss: 0.4322 - val_accuracy: 0.8153
Epoch 99/150
accuracy: 0.8062 - val_loss: 0.4326 - val_accuracy: 0.8108
640/640 [============ - - os 41us/step - loss: 0.3704 -
accuracy: 0.8219 - val_loss: 0.4320 - val_accuracy: 0.8063
Epoch 101/150
640/640 [=============== ] - 0s 40us/step - loss: 0.3610 -
accuracy: 0.8219 - val_loss: 0.4323 - val_accuracy: 0.8108
```

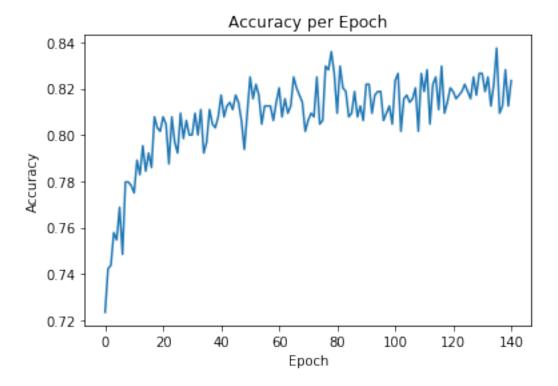
```
Epoch 102/150
640/640 [============== ] - 0s 39us/step - loss: 0.3692 -
accuracy: 0.8094 - val_loss: 0.4331 - val_accuracy: 0.8063
Epoch 103/150
accuracy: 0.8172 - val_loss: 0.4333 - val_accuracy: 0.8108
Epoch 104/150
accuracy: 0.8188 - val_loss: 0.4343 - val_accuracy: 0.8108
Epoch 105/150
640/640 [============== ] - 0s 50us/step - loss: 0.3661 -
accuracy: 0.8188 - val_loss: 0.4343 - val_accuracy: 0.8108
Epoch 106/150
640/640 [============ - - os 47us/step - loss: 0.3573 -
accuracy: 0.8062 - val_loss: 0.4344 - val_accuracy: 0.8108
Epoch 107/150
640/640 [============== ] - 0s 43us/step - loss: 0.3617 -
accuracy: 0.8094 - val_loss: 0.4342 - val_accuracy: 0.8108
Epoch 108/150
accuracy: 0.8125 - val_loss: 0.4347 - val_accuracy: 0.8153
Epoch 109/150
accuracy: 0.8047 - val_loss: 0.4342 - val_accuracy: 0.8108
Epoch 110/150
accuracy: 0.8234 - val_loss: 0.4347 - val_accuracy: 0.8198
Epoch 111/150
640/640 [============= - - os 39us/step - loss: 0.3610 -
accuracy: 0.8266 - val_loss: 0.4339 - val_accuracy: 0.8108
Epoch 112/150
accuracy: 0.8016 - val_loss: 0.4345 - val_accuracy: 0.8108
Epoch 113/150
accuracy: 0.8156 - val_loss: 0.4348 - val_accuracy: 0.8063
Epoch 114/150
accuracy: 0.8172 - val_loss: 0.4347 - val_accuracy: 0.8063
Epoch 115/150
accuracy: 0.8141 - val_loss: 0.4343 - val_accuracy: 0.8108
Epoch 116/150
640/640 [============= - - os 58us/step - loss: 0.3657 -
accuracy: 0.8156 - val_loss: 0.4336 - val_accuracy: 0.8063
Epoch 117/150
accuracy: 0.8203 - val_loss: 0.4348 - val_accuracy: 0.8063
```

```
Epoch 118/150
640/640 [============= ] - 0s 47us/step - loss: 0.3690 -
accuracy: 0.8016 - val_loss: 0.4341 - val_accuracy: 0.8108
Epoch 119/150
accuracy: 0.8266 - val_loss: 0.4371 - val_accuracy: 0.8108
Epoch 120/150
accuracy: 0.8188 - val_loss: 0.4346 - val_accuracy: 0.8108
Epoch 121/150
640/640 [============== ] - 0s 56us/step - loss: 0.3671 -
accuracy: 0.8281 - val_loss: 0.4340 - val_accuracy: 0.8018
Epoch 122/150
640/640 [============ ] - 0s 47us/step - loss: 0.3739 -
accuracy: 0.8047 - val_loss: 0.4355 - val_accuracy: 0.8108
Epoch 123/150
640/640 [============= ] - 0s 44us/step - loss: 0.3565 -
accuracy: 0.8219 - val_loss: 0.4374 - val_accuracy: 0.8108
Epoch 124/150
640/640 [============= ] - 0s 49us/step - loss: 0.3548 -
accuracy: 0.8250 - val_loss: 0.4366 - val_accuracy: 0.8108
Epoch 125/150
accuracy: 0.8109 - val_loss: 0.4366 - val_accuracy: 0.8108
Epoch 126/150
accuracy: 0.8297 - val_loss: 0.4352 - val_accuracy: 0.8108
Epoch 127/150
640/640 [============= - - os 49us/step - loss: 0.3717 -
accuracy: 0.8094 - val_loss: 0.4348 - val_accuracy: 0.8108
Epoch 128/150
640/640 [=============== ] - 0s 46us/step - loss: 0.3687 -
accuracy: 0.8141 - val_loss: 0.4352 - val_accuracy: 0.8108
Epoch 129/150
accuracy: 0.8203 - val_loss: 0.4341 - val_accuracy: 0.8063
Epoch 130/150
accuracy: 0.8188 - val_loss: 0.4340 - val_accuracy: 0.8108
Epoch 131/150
accuracy: 0.8156 - val_loss: 0.4343 - val_accuracy: 0.8018
Epoch 132/150
640/640 [============= - - os 48us/step - loss: 0.3678 -
accuracy: 0.8172 - val_loss: 0.4351 - val_accuracy: 0.8018
Epoch 133/150
accuracy: 0.8188 - val_loss: 0.4349 - val_accuracy: 0.7973
```

```
Epoch 134/150
640/640 [============== ] - 0s 48us/step - loss: 0.3641 -
accuracy: 0.8219 - val_loss: 0.4333 - val_accuracy: 0.8018
Epoch 135/150
accuracy: 0.8188 - val_loss: 0.4320 - val_accuracy: 0.8108
Epoch 136/150
accuracy: 0.8156 - val_loss: 0.4316 - val_accuracy: 0.8063
Epoch 137/150
accuracy: 0.8250 - val_loss: 0.4336 - val_accuracy: 0.7973
Epoch 138/150
640/640 [============= - - 0s 51us/step - loss: 0.3774 -
accuracy: 0.8172 - val_loss: 0.4327 - val_accuracy: 0.7973
Epoch 139/150
640/640 [============= ] - 0s 72us/step - loss: 0.3582 -
accuracy: 0.8266 - val_loss: 0.4320 - val_accuracy: 0.8018
Epoch 140/150
640/640 [============== ] - 0s 61us/step - loss: 0.3691 -
accuracy: 0.8266 - val_loss: 0.4318 - val_accuracy: 0.8018
Epoch 141/150
accuracy: 0.8188 - val_loss: 0.4319 - val_accuracy: 0.8018
Epoch 142/150
accuracy: 0.8250 - val_loss: 0.4318 - val_accuracy: 0.8018
Epoch 143/150
640/640 [============= - - 0s 52us/step - loss: 0.3607 -
accuracy: 0.8125 - val_loss: 0.4328 - val_accuracy: 0.7928
Epoch 144/150
640/640 [=============== ] - 0s 62us/step - loss: 0.3582 -
accuracy: 0.8219 - val_loss: 0.4340 - val_accuracy: 0.7973
Epoch 145/150
accuracy: 0.8375 - val_loss: 0.4346 - val_accuracy: 0.7973
Epoch 146/150
accuracy: 0.8094 - val_loss: 0.4350 - val_accuracy: 0.8018
Epoch 147/150
accuracy: 0.8125 - val_loss: 0.4354 - val_accuracy: 0.7973
Epoch 148/150
640/640 [============= - - os 49us/step - loss: 0.3586 -
accuracy: 0.8281 - val_loss: 0.4367 - val_accuracy: 0.7973
Epoch 149/150
accuracy: 0.8125 - val_loss: 0.4358 - val_accuracy: 0.7928
```

The accuracy of the network on the original test data (without duplicating the failed class) is also between 80% and 90%.

```
[10]: plt.plot(result2.history['accuracy'][9:], label='Accuracy')
    plt.title("Accuracy per Epoch")
    plt.ylabel("Accuracy")
    plt.xlabel("Epoch")
    plt.show()
```



```
[0.60959524],
```

- [0.07123813],
- [0.53515863],
- [0.5419489],
- [0.13775331],
- \_\_\_\_\_
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- [0.1.220000]
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- [0.9734995],
- [0.17926729],
- [0.8913601],
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- [0.3155466],
- [0.9448713],
- [0.97725964],
- [0.9732246],
- [0.8292054],
- [0.8036944],
- [0.9858477],
- [0.9871623],
- [0.9342852],
- [0.9541438],
- [0.9868249],
- [0.9216848],
- [0.4562843],
- [0.9852083],
- [0.8863766],
- [0.9937504],
- [0.75724834],
- [0.9761401],
- [0.90395194],
- [0.92769206],
- [0.92769206],
- [0.9934228],
- [0.9957795],
- [0.9292013],
- [0.69706684],
- [0.99145657],
- [0.9942192],
- [0.6315681],

```
[0.99908173]], dtype=float32)
[13]: [1 if n > 0.5 else 0 for n in model2.predict(X_test)]
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