## **Bruno Morgado**

HW4

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Question 1

```
In []: import tensorflow as tf
    from tensorflow import keras
    from tensorflow.keras.datasets import cifar10
    from tensorflow.keras.models import Sequential
    from tensorflow.keras.layers import Dense
    from tensorflow.keras import Input
    from tensorflow.keras.optimizers import Adam
    from tensorflow.keras.utils import to_categorical

import numpy as np
    from skimage import io
    from matplotlib import pyplot as plt
    import copy

import ssl
    ssl._create_default_https_context = ssl._create_unverified_context
```

WARNING:tensorflow:From c:\Users\MorgadoBruno\AppData\Local\anaconda3\envs\ML\lib\si te-packages\keras\src\losses.py:2976: The name tf.losses.sparse\_softmax\_cross\_entrop y is deprecated. Please use tf.compat.v1.losses.sparse\_softmax\_cross\_entropy instea d.

Extracting data

```
In [ ]: (x_train, y_train), (x_test, y_test) = cifar10.load_data()
```

Plotting an image from each category

```
In []: """This plot finds an index for each of the types of images """
indeces = []

for label in range(10):
    iter = 0
    while y_train[iter] != label and iter<(y_train.size-1):
        iter += 1
    indeces.append(iter)

print(indeces)</pre>
```

[29, 4, 6, 9, 3, 27, 0, 7, 8, 1]

## Plotting images

```
In [ ]: plt.figure()
         label_names = ['airplane', 'automobile', 'bird', 'cat', 'deer', 'dog', 'frog', 'hor
         fig, ax = plt.subplots(1,10, figsize=(15,10))
         for label in range(10):
             ax[label].imshow(x_train[indeces[label]])
             ax[label].title.set_text(label_names[label])
         plt.show()
       <Figure size 640x480 with 0 Axes>
                  automobile
          airplane
                             bird
                                       cat
                                               deer
                                                         dog
                                                                          horse
                                                                                            truck
```

Reshaping the images for a fully connected NN

```
In [ ]: print(x_train.shape)
        print(x_train.shape[0])
        x_{train} = x_{train.reshape}(50000, -1)
        print(x_train.shape)
        print(x_train)
        # Convert the target data into one-hot encoding
        y_train = to_categorical(y_train, 10)
       (50000, 32, 32, 3)
       50000
       (50000, 3072)
       [[ 59 62 63 ... 123 92 72]
        [154 177 187 ... 143 133 144]
        [255 255 255 ... 80 86 84]
        [ 35 178 235 ... 12 31 50]
        [189 211 240 ... 195 190 171]
        [229 229 239 ... 163 163 161]]
In [ ]: def create_nn(n_layers, learn_rate):
            """ This function will be the function used for tuning the number of layers as
            Args:
                n_layers (_type_): _description_
            Returns:
                _type_: _description_
            model = Sequential()
            model.add(Input(shape=(3072,)))
            for n in range(n_layers):
```

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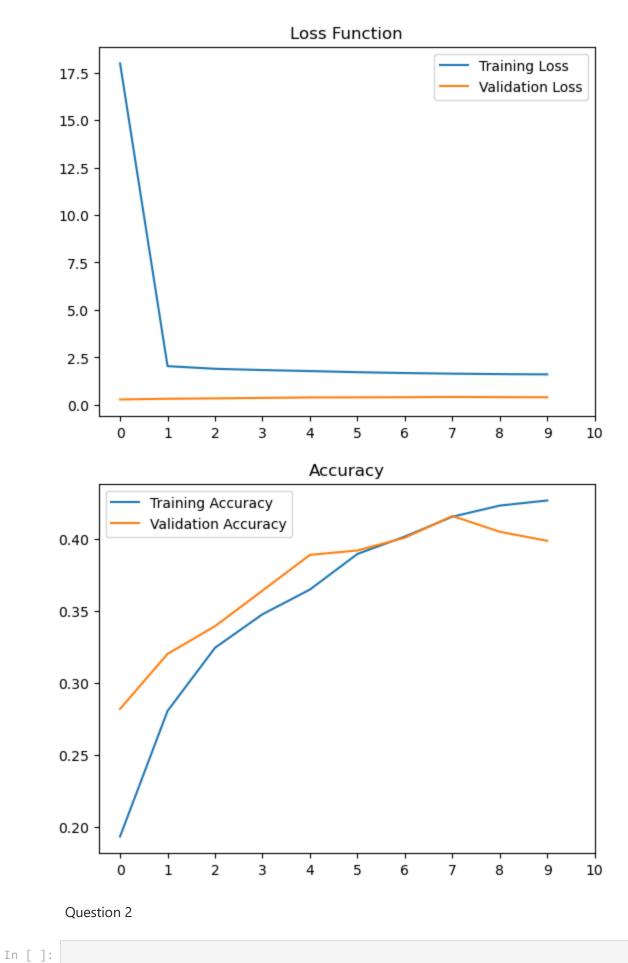
```
In [ ]: n_{ayers} = [4, 5, 6]
        learn_rates = [0.001, 0.0005]
        accuracy = 0
        opt_n = 0
        opt_rate = 0
        for n in n_layers:
            for rate in learn_rates:
                model = create_nn(n, rate)
                fit = model.fit(x_train, y_train, epochs=10,
                             batch_size=100, validation_split=0.3)
                 if fit.history['accuracy'][-1] > accuracy:
                    accuracy = fit.history['accuracy'][-1]
                    opt_n = n
                    opt_rate = rate
                    opt model = copy.deepcopy(model)
                    opt_fit = copy.deepcopy(fit)
                 del model
                 del fit
```

```
In [ ]: plt.figure()

plt.plot(opt_fit.history['loss'], label = "Training Loss")
plt.plot(opt_fit.history['val_accuracy'], label = "Validation Loss")
plt.legend()
plt.xticks(range(11))
plt.title("Loss Function")

plt.show()

plt.plot(opt_fit.history['accuracy'], label = "Training Accuracy")
plt.plot(opt_fit.history['val_accuracy'], label = "Validation Accuracy")
plt.legend()
plt.xticks(range(11))
plt.title("Accuracy")
```



```
from IPython.display import display
          import numpy as np
          import pandas as pd
          import seaborn as sns
          from sklearn.model_selection import train_test_split
          from tensorflow.keras.layers import Rescaling
          from tensorflow.keras.regularizers import L2
          from sklearn.preprocessing import MinMaxScaler
In [ ]: df = pd.read_csv('airfoil_self_noise.dat', sep="\t",header=None )
          sns.pairplot(df)
Out[]: <seaborn.axisgrid.PairGrid at 0x1a58e804f40>
         15000
       o 10000
         5000
           15
          0.30
          0.25
          0.20
        ° 0.15
          0.10
          0.05
           70
           60
         m 50
           40
          0.05
          0.04
        ₩ 0.03
          0.02
          0.01
          140
          130
          110
               5000 10000 15000 20000
                                  10
                                                    0.2
                                                        0.3
                                                                     60
                                                                        70
                                                                  50
                                                                          0.00
                                                                                   0.04
                                                                                        0.06
                                                                                             110 120 130
In [ ]: # Data Preparation
          df = pd.read_csv('airfoil_self_noise.dat', sep="\t",header=None )
          display(df)
         y = df.iloc[:, -1]
```

```
y = np.asmatrix(y)
        y = y.T
        print(y.shape)
        minmax = MinMaxScaler()
        X = df.iloc[:, :-1]
        X = np.asarray(X)
        X = minmax.fit_transform(X)
        print(X)
        print(X.shape)
                            2
                0
                     1
                                 3
                                          4
                                                  5
          0 800
                   0.0 0.3048 71.3 0.002663 126.201
          1 1000
                   0.0 0.3048 71.3 0.002663 125.201
          2 1250
                   0.0 0.3048 71.3 0.002663 125.951
          3 1600
                   0.0 0.3048 71.3 0.002663 127.591
          4 2000
                   0.0 0.3048 71.3 0.002663 127.461
       1498 2500 15.6 0.1016 39.6 0.052849 110.264
       1499 3150 15.6 0.1016 39.6 0.052849 109.254
       1500 4000 15.6 0.1016 39.6 0.052849 106.604
       1501 5000 15.6 0.1016 39.6 0.052849 106.224
       1502 6300 15.6 0.1016 39.6 0.052849 104.204
      1503 rows × 6 columns
       (1503, 1)
                                                     0.03900472]
       [[0.03030303 0.
                                          1.
        [0.04040404 0.
                               1.
                                          1.
                                                     0.03900472]
        [0.0530303 0.
                               1.
                                          1.
                                                     0.03900472]
        [0.19191919 0.7027027 0.27272727 0.19949495 0.90411066]
        [0.24242424 0.7027027 0.27272727 0.19949495 0.90411066]
        [0.30808081 0.7027027 0.27272727 0.19949495 0.90411066]]
       (1503, 5)
In [ ]: model = Sequential()
        model.add(Input(shape=(5,)))
        model.add(Dense(14, activation="linear"))
        model.add(Dense(1, activation="elu"))
        # Add output layer
        model.add(Dense(1, activation="linear"))
        optimizer = Adam(learning_rate=0.01)
        model.compile(optimizer=optimizer,
                        loss='mean_squared_error',
```

```
metrics=[keras.metrics.R2Score()])
model.summary()
```

Model: "sequential\_48"

Layer (type)	Output Shape	Param #
dense_222 (Dense)	(None, 14)	84
dense_223 (Dense)	(None, 1)	15
dense_224 (Dense)	(None, 1)	2

Layer (type)	Output Shape	Param #
dense_222 (Dense)	(None, 14)	84
dense_223 (Dense)	(None, 1)	15
dense_224 (Dense)	(None, 1)	2

\_\_\_\_\_\_

Total params: 101 (404.00 Byte)
Trainable params: 101 (404.00 Byte)
Non-trainable params: 0 (0.00 Byte)

```
In [ ]: print("train R2 Score: ", fit.history['r2_score'][-1])
print("test R2 Score: ", fit.history['val_r2_score'][-1])
```

train R2 Score: 0.5566079616546631 test R2 Score: 0.4110473394393921