jjoy6_assignment_1

February 16, 2025

Assignent_1 Jacob Joy Loading the IMDB dataset

```
[1]: #Importing libraries that will be used
     import tensorflow as tf
     import numpy as np
     import matplotlib.pyplot as plt
     import os
     import random
     from tensorflow import keras
     from tensorflow.keras import layers
     from tensorflow.keras.regularizers import 12
     #importing IMBD Dataset and setting word count to top 10,000
     from tensorflow.keras.datasets import imdb
     (train_data, train_labels), (test_data, test_labels) = imdb.
      →load_data(num_words=10000)
     #Attempting to add some standardisation to RNG for better comparisons
     #Setting seed to 113 as this is the default seed used in the IMBD data set
     os.environ['PYTHONHASHSEED'] = str(113)
     random.seed(113)
     # Set the seed for NumPy
     np.random.seed(113)
     # Set the seed for TensorFlow
     tf.random.set_seed(113)
     # Enable deterministic operations
     tf.config.experimental.enable op determinism()
     \# Setting the seed may not make exactly reproducable results but helps to keep \sqcup
      →them as close as can be without more complex coding.
```

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/imdb.npz

17464789/17464789

0s

Ous/step

Multi-hot encoding of data and labels

```
# Using manual Multi-hot encoding
# Attempted to use keras to_categorical but would run out of memory
def vectorize_sequences(sequences, dimension=10000):
    results = np.zeros((len(sequences), dimension))
    for i, sequence in enumerate(sequences):
        for j in sequence:
            results[i, j] = 1.
    return results

x_train = vectorize_sequences(train_data)
x_test = vectorize_sequences(test_data)

y_train = np.asarray(train_labels).astype("float32")
y_test = np.asarray(test_labels).astype("float32")
```

Validation Set

```
[3]: #Creation of validation set from the training data
x_val = x_train[:10000]
partial_x_train = x_train[10000:]
y_val = y_train[:10000]
partial_y_train = y_train[10000:]
```

Building of different models for testing.

```
[4]: #Experiment 1
     model = keras.Sequential([
         layers.Dense(8, activation="relu"),
         layers.Dense(1, activation="sigmoid")
     ])
     11 11 11
     #Experiment 2
     model = keras.Sequential([
         layers.Dense(16, activation="relu"),
         layers.Dense(1, activation="sigmoid")
     7)
     ,,,,,,
     #Experiment 3
     HHHH
     model = keras.Sequential([
         layers.Dense(64, activation="relu"),
```

```
layers.Dense(1, activation="sigmoid")
])
n n n
#Experiment 4
model = keras.Sequential([
    layers.Dense(128, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
#Experiment 5
11 11 11
model = keras.Sequential([
    layers.Dense(8, activation="relu"),
    layers.Dense(8, activation="relu"),
    layers.Dense(1, activation="sigmoid")
1)
11 11 11
#Experiment 6
model = keras.Sequential([
    layers.Dense(16, activation="relu"),
    layers.Dense(16, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
11 11 11
#Experiment 7
model = keras.Sequential([
    layers.Dense(32, activation="relu"),
    layers.Dense(32, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
11 11 11
#Experiment 8
n n n
model = keras.Sequential([
    layers.Dense(64, activation="relu"),
    layers.Dense(64, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
```

```
11 11 11
#Experiment 9
n n n
model = keras.Sequential([
    layers.Dense(128, activation="relu"),
    layers.Dense(128, activation="relu"),
    layers.Dense(1, activation="sigmoid")
7)
11 11 11
#Experiment 10
n n n
model = keras.Sequential([
    layers.Dense(256, activation="relu"),
    layers.Dense(256, activation="relu"),
    layers.Dense(1, activation="sigmoid")
1)
11 11 11
#Experiment 11
11 11 11
model = keras.Sequential([
    layers.Dense(8, activation="relu"),
    layers.Dense(8, activation="relu"),
    layers.Dense(8, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
11 11 11
#Experiment 12
model = keras.Sequential([
    layers.Dense(16, activation="relu"),
    layers.Dense(16, activation="relu"),
    layers.Dense(16, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
11 11 11
#Experiment 13
n n n
model = keras.Sequential([
    layers.Dense(32, activation="relu"),
    layers.Dense(32, activation="relu"),
    layers.Dense(32, activation="relu"),
    layers.Dense(1, activation="sigmoid")
```

```
1)
#Experiment 14
11 11 11
model = keras.Sequential([
    layers.Dense(64, activation="relu"),
    layers.Dense(64, activation="relu"),
    layers.Dense(64, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
11 11 11
#Experiment 15
model = keras.Sequential([
    layers.Dense(128, activation="relu"),
    layers.Dense(128, activation="relu"),
    layers.Dense(128, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
11 11 11
#Experiment 16,19, and 20
model = keras.Sequential([
    layers.Dense(64, activation="relu"),
    layers.Dense(64, activation="relu"),
    layers.Dense(64, activation="relu"),
    layers.Dense(64, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
11 11 11
#Experiment 17
11 11 11
model = keras.Sequential([
    layers.Dense(128, activation="relu"),
    layers.Dense(128, activation="relu"),
    layers.Dense(128, activation="relu"),
    layers.Dense(128, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
11 11 11
#Experiment 18
```

```
model = keras.Sequential([
    layers.Dense(64, activation="tanh"),
    layers.Dense(64, activation="tanh"),
    layers.Dense(64, activation="tanh"),
    layers.Dense(64, activation="tanh"),
    layers.Dense(1, activation="sigmoid")
7)
11 11 11
#Experiment 21 (0.2),22 (0.3), and 23 (0.5)
model = keras.Sequential([
    layers.Dense(64, activation="relu"),
    layers.Dropout(0.5),
    layers.Dense(64, activation="relu"),
    layers.Dropout(0.5),
    layers.Dense(64, activation="relu"),
    layers.Dropout(0.5),
    layers.Dense(64, activation="relu"),
    layers.Dropout(0.5),
    layers.Dense(1, activation="sigmoid")
])
HHHH
#Experiment 24 (0.00001), 25 (0.0001), and 26 (0.001)
model = keras.Sequential([
    layers.Dense(64, kernel_regularizer= l2(0.00001), activation="relu"),
    layers.Dense(64, kernel_regularizer= l2(0.00001), activation="relu"),
    layers.Dense(64, kernel regularizer= 12(0.00001), activation="relu"),
    layers.Dense(64, kernel_regularizer= l2(0.00001), activation="relu"),
    layers.Dense(1, activation="sigmoid")
1)
n n n
#Experiment 27
model = keras.Sequential([
    layers.Dense(64, kernel regularizer= 12(0.00001), activation="relu"),
    layers.Dropout(0.3),
    layers.Dense(64, kernel_regularizer= l2(0.00001), activation="relu"),
    layers.Dropout(0.3),
    layers.Dense(64, kernel_regularizer= l2(0.00001), activation="relu"),
    layers.Dropout(0.3),
    layers.Dense(64, kernel regularizer= 12(0.00001), activation="relu"),
    layers.Dropout(0.3),
    layers.Dense(1, activation="sigmoid")
```

```
#Final Model

model = keras.Sequential([
    layers.Dense(64, activation="relu"),
    layers.Dropout(0.3),
    layers.Dense(64, activation="relu"),
    layers.Dropout(0.3),
    layers.Dense(64, activation="relu"),
    layers.Dropout(0.3),
    layers.Dense(64, activation="relu"),
    layers.Dense(64, activation="relu"),
    layers.Dropout(0.3),
    layers.Dropout(0.3),
    layers.Dense(1, activation="sigmoid")
])
```

Compiling Model

```
[5]: model.compile(optimizer='rmsprop',loss="binary_crossentropy",u

metrics=["accuracy"])

#Used for Experiment 19

#model.compile(optimizer="rmsprop",loss="mse", metrics=["accuracy"])

#Used for Experiment 20

#model.compile(optimizer='adam', loss="binary_crossentropy",u

metrics=["accuracy"])
```

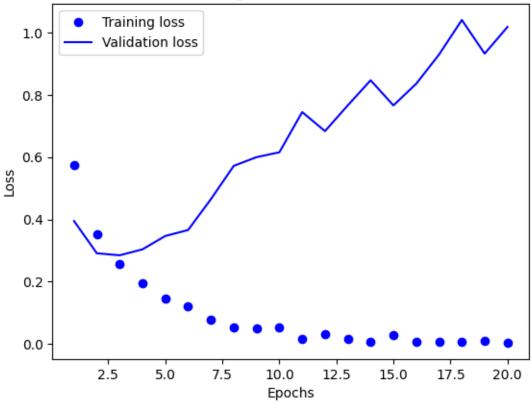
Training Model

```
accuracy: 0.8990 - loss: 0.2673 - val_accuracy: 0.8883 - val_loss: 0.2850
Epoch 4/20
30/30
                 1s 30ms/step -
accuracy: 0.9263 - loss: 0.2076 - val_accuracy: 0.8880 - val_loss: 0.3042
Epoch 5/20
30/30
                 1s 30ms/step -
accuracy: 0.9462 - loss: 0.1565 - val_accuracy: 0.8849 - val_loss: 0.3470
Epoch 6/20
30/30
                 2s 42ms/step -
accuracy: 0.9559 - loss: 0.1249 - val_accuracy: 0.8864 - val_loss: 0.3662
Epoch 7/20
30/30
                 2s 30ms/step -
accuracy: 0.9734 - loss: 0.0813 - val_accuracy: 0.8811 - val_loss: 0.4654
Epoch 8/20
30/30
                 1s 31ms/step -
accuracy: 0.9769 - loss: 0.0674 - val_accuracy: 0.8775 - val_loss: 0.5721
Epoch 9/20
30/30
                 2s 40ms/step -
accuracy: 0.9801 - loss: 0.0609 - val_accuracy: 0.8773 - val_loss: 0.6001
Epoch 10/20
30/30
                 1s 39ms/step -
accuracy: 0.9693 - loss: 0.1007 - val_accuracy: 0.8834 - val_loss: 0.6157
Epoch 11/20
30/30
                 1s 34ms/step -
accuracy: 0.9958 - loss: 0.0139 - val_accuracy: 0.8779 - val_loss: 0.7446
Epoch 12/20
30/30
                 1s 34ms/step -
accuracy: 0.9901 - loss: 0.0292 - val_accuracy: 0.8827 - val_loss: 0.6836
Epoch 13/20
30/30
                 1s 30ms/step -
accuracy: 0.9958 - loss: 0.0142 - val_accuracy: 0.8817 - val_loss: 0.7663
Epoch 14/20
30/30
                 2s 47ms/step -
accuracy: 0.9984 - loss: 0.0057 - val_accuracy: 0.8800 - val_loss: 0.8468
Epoch 15/20
30/30
                 1s 36ms/step -
accuracy: 0.9962 - loss: 0.0106 - val accuracy: 0.8744 - val loss: 0.7662
Epoch 16/20
30/30
                 1s 36ms/step -
accuracy: 0.9980 - loss: 0.0087 - val_accuracy: 0.8799 - val_loss: 0.8357
Epoch 17/20
30/30
                  1s 33ms/step -
accuracy: 0.9984 - loss: 0.0052 - val_accuracy: 0.8809 - val_loss: 0.9294
Epoch 18/20
30/30
                 1s 39ms/step -
accuracy: 0.9980 - loss: 0.0063 - val_accuracy: 0.8777 - val_loss: 1.0403
Epoch 19/20
30/30
                 1s 43ms/step -
```

Plotting taining and Validation loss

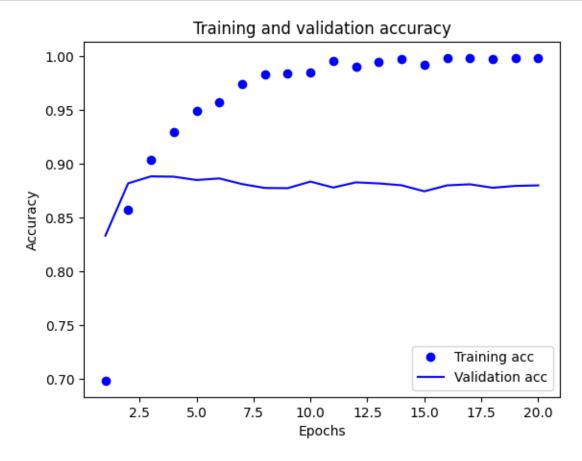
```
[7]: import matplotlib.pyplot as plt
history_dict = history.history
loss_values = history_dict["loss"]
val_loss_values = history_dict["val_loss"]
epochs = range(1, len(loss_values) + 1)
plt.plot(epochs, loss_values, "bo", label="Training loss")
plt.plot(epochs, val_loss_values, "b", label="Validation loss")
plt.title("Training and validation loss")
plt.xlabel("Epochs")
plt.ylabel("Loss")
plt.legend()
plt.show()
```

Training and validation loss



Plotting training and Validation accouracy

```
[8]: plt.clf()
    acc = history_dict["accuracy"]
    val_acc = history_dict["val_accuracy"]
    plt.plot(epochs, acc, "bo", label="Training acc")
    plt.plot(epochs, val_acc, "b", label="Validation acc")
    plt.title("Training and validation accuracy")
    plt.xlabel("Epochs")
    plt.ylabel("Accuracy")
    plt.legend()
    plt.show()
```



Retraining the model using number of epochs set before validation deteriorates

```
layers.Dense(1, activation="sigmoid")
])
n n n
#Experiment 2
model = keras.Sequential([
    layers.Dense(16, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
#Experiment 3
11 11 11
model = keras.Sequential([
    layers.Dense(64, activation="relu"),
    layers.Dense(1, activation="sigmoid")
1)
11 11 11
#Experiment 4
model = keras.Sequential([
    layers.Dense(128, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
11 11 11
#Experiment 5
model = keras.Sequential([
    layers.Dense(8, activation="relu"),
    layers.Dense(8, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
11 11 11
#Experiment 6
HHHH
model = keras.Sequential([
    layers.Dense(16, activation="relu"),
    layers.Dense(16, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
11 11 11
#Experiment 7
```

```
11 11 11
model = keras.Sequential([
    layers.Dense(32, activation="relu"),
    layers.Dense(32, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
11 11 11
#Experiment 8
model = keras.Sequential([
    layers.Dense(64, activation="relu"),
    layers.Dense(64, activation="relu"),
    layers.Dense(1, activation="sigmoid")
1)
11 II II
#Experiment 9
11 11 11
model = keras.Sequential([
    layers.Dense(128, activation="relu"),
    layers.Dense(128, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
11 11 11
#Experiment 10
11 11 11
model = keras.Sequential([
    layers.Dense(256, activation="relu"),
    layers.Dense(256, activation="relu"),
    layers.Dense(1, activation="sigmoid")
1)
11 11 11
#Experiment 11
model = keras.Sequential([
    layers.Dense(8, activation="relu"),
    layers.Dense(8, activation="relu"),
    layers.Dense(8, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
11 11 11
#Experiment 12
```

```
model = keras.Sequential([
    layers.Dense(16, activation="relu"),
    layers.Dense(16, activation="relu"),
    layers.Dense(16, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
11 11 11
#Experiment 13
model = keras.Sequential([
    layers.Dense(32, activation="relu"),
    layers.Dense(32, activation="relu"),
    layers.Dense(32, activation="relu"),
    layers.Dense(1, activation="sigmoid")
1)
11 11 11
#Experiment 14
model = keras.Sequential([
    layers.Dense(64, activation="relu"),
    layers.Dense(64, activation="relu"),
    layers.Dense(64, activation="relu"),
    layers.Dense(1, activation="sigmoid")
7)
11 11 11
#Experiment 15
11 11 11
model = keras.Sequential([
    layers.Dense(128, activation="relu"),
    layers.Dense(128, activation="relu"),
    layers.Dense(128, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
11 11 11
#Experiment 16,19, and 20
model = keras.Sequential([
    layers.Dense(64, activation="relu"),
    layers.Dense(64, activation="relu"),
    layers.Dense(64, activation="relu"),
    layers.Dense(64, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
```

```
11 11 11
#Experiment 17
model = keras.Sequential([
    layers.Dense(128, activation="relu"),
    layers.Dense(128, activation="relu"),
    layers.Dense(128, activation="relu"),
    layers.Dense(128, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
11 11 11
#Experiment 18
model = keras.Sequential([
    layers.Dense(64, activation="tanh"),
    layers.Dense(64, activation="tanh"),
    layers.Dense(64, activation="tanh"),
    layers.Dense(64, activation="tanh"),
    layers.Dense(1, activation="sigmoid")
])
H H H
#Experiment 21 (0.2),22 (0.3), and 23 (0.5)
model = keras.Sequential([
    layers.Dense(64, activation="relu"),
    layers.Dropout(0.5),
    layers.Dense(64, activation="relu"),
    layers.Dropout(0.5),
    layers.Dense(64, activation="relu"),
    layers.Dropout(0.5),
    layers.Dense(64, activation="relu"),
    layers.Dropout(0.5),
    layers.Dense(1, activation="sigmoid")
])
11 11 11
#Experiment 24 (0.00001), 25 (0.0001), and 26 (0.001)
model = keras.Sequential([
    layers.Dense(64, kernel_regularizer= l2(0.00001), activation="relu"),
    layers.Dense(64, kernel_regularizer= l2(0.00001), activation="relu"),
    layers.Dense(64, kernel_regularizer= l2(0.00001), activation="relu"),
    layers.Dense(64, kernel regularizer= 12(0.00001), activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
```

```
11 11 11
#Experiment 27
model = keras.Sequential([
    layers.Dense(64, kernel_regularizer= l2(0.00001), activation="relu"),
    layers.Dropout(0.3),
    layers.Dense(1, activation="sigmoid")
1)
11 II II
#Final Model
model = keras.Sequential([
    layers.Dense(64, activation="relu"),
    layers.Dropout(0.3),
    layers.Dense(64, activation="relu"),
    layers.Dropout(0.3),
    layers.Dense(64, activation="relu"),
    layers.Dropout(0.3),
    layers.Dense(64, activation="relu"),
    layers.Dropout(0.3),
    layers.Dense(1, activation="sigmoid")
])
model.compile(optimizer='rmsprop',loss="binary_crossentropy",_
 ⇔metrics=["accuracy"])
#Used for Experiment 19
#model.compile(optimizer="rmsprop",loss="mse", metrics=["accuracy"])
#Used for Experiment 20
#model.compile(optimizer=optimizer, loss="binary_crossentropy", __
→metrics=["accuracy"])
model.fit(x_train, y_train, epochs=3, batch_size=512)
results_test = model.evaluate(x_test, y_test)
results_val = model.evaluate(x_val, y_val)
results_train = model.evaluate(x_train, y_train)
print('Training results (loss,Acc)')
```

```
print(results_train)
      print('Val results (loss,Acc)')
      print(results_val)
      print('Test results (loss,Acc)')
      print(results_test)
     Epoch 1/3
     49/49
                       3s 17ms/step -
     accuracy: 0.6625 - loss: 0.5996
     Epoch 2/3
     49/49
                       1s 21ms/step -
     accuracy: 0.8784 - loss: 0.3153
     Epoch 3/3
     49/49
                       2s 29ms/step -
     accuracy: 0.9156 - loss: 0.2265
     782/782
                         3s 3ms/step -
     accuracy: 0.8874 - loss: 0.2966
     313/313
                         1s 3ms/step -
     accuracy: 0.9579 - loss: 0.1184
     782/782
                         3s 3ms/step -
     accuracy: 0.9578 - loss: 0.1203
     Training results (loss,Acc)
     [0.12189925462007523, 0.957319974899292]
     Val results (loss, Acc)
     [0.1203325092792511, 0.9587000012397766]
     Test results (loss, Acc)
     [0.29753419756889343, 0.8867200016975403]
     Predict
[11]: model.predict(x_test)
     782/782
                         2s 2ms/step
[11]: array([[0.07631286],
             [0.9999299],
             [0.908958],
             [0.06250498],
             [0.03814842],
             [0.79949695]], dtype=float32)
 []: !pip
```