

Getting started with neural networks: Classification and regression

- ▼ Classifying movie reviews: A binary classification example
- ▼ The IMDB dataset

Loading the IMDB dataset

```
from tensorflow.keras.datasets import imdb
(train_data, train_labels), (test_data, test_labels) = imdb.load_data(
    num_words=10000)
```

```
train_data[0]
```

```
☐ [1,
    14,
    22,
    16,
    43,
    530,
    973,
    1622,
    1385,
    65,
    458,
    4468,
    66,
    3941,
    4,
    173,
    36,
    256,
    5,
    25,
    100,
    43,
    838,
    112,
    50,
    670,
    2,
    9,
```

```

35,
480,
284,
5,
150,
4,
172,
112,
167,
2,
336,
385,
39,
4,
172,
4536,
1111,
17,
546,
38,
13,
447,
4,
192,
50,
16,
6,
147,
2025,
11

```

```
train_labels[0]
```

```
1
```

```
max([max(sequence) for sequence in train_data])
```

```
9999
```

Decoding reviews back to text

```

word_index = imdb.get_word_index()
reverse_word_index = dict(
    [(value, key) for (key, value) in word_index.items()])
decoded_review = " ".join(
    [reverse_word_index.get(i - 3, "?") for i in train_data[0]])

```

```

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/imdb\_v1646592/1641221 [=====] - 0s 0us/step
1654784/1641221 [=====] - 0s 0us/step

```

▼ Preparing the data

Encoding the integer sequences via multi-hot encoding

```
import numpy as np
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)

x_train[0]

array([0., 1., 1., ..., 0., 0., 0.])

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

▼ Building your model

Model definition

Building the model using 1 Hidden Layer, 64 Hidden unit with tanh activation and mse loss function instead of binary_crossentropy

```
from tensorflow import keras
from tensorflow.keras import layers

model = keras.Sequential([
    layers.Dense(16, activation="relu"),
    layers.Dense(16, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
```

Compiling the model

```
model.compile(optimizer="rmsprop",
```

```
loss="binary_crossentropy",
metrics=["accuracy"])
```

▼ Validating your approach

Setting aside a validation set

```
x_val = x_train[:10000]
partial_x_train = x_train[10000:]
y_val = y_train[:10000]
partial_y_train = y_train[10000:]
```

Training your model

```
history = model.fit(partial_x_train,
                    partial_y_train,
                    epochs=20,
                    batch_size=512,
                    validation_data=(x_val, y_val))
```

```
Epoch 1/20
30/30 [=====] - 2s 49ms/step - loss: 0.5351 - accuracy: 0.7824
Epoch 2/20
30/30 [=====] - 1s 36ms/step - loss: 0.3262 - accuracy: 0.8993
Epoch 3/20
30/30 [=====] - 1s 36ms/step - loss: 0.2371 - accuracy: 0.9240
Epoch 4/20
30/30 [=====] - 1s 37ms/step - loss: 0.1866 - accuracy: 0.9401
Epoch 5/20
30/30 [=====] - 1s 36ms/step - loss: 0.1492 - accuracy: 0.9538
Epoch 6/20
30/30 [=====] - 1s 37ms/step - loss: 0.1237 - accuracy: 0.9613
Epoch 7/20
30/30 [=====] - 1s 36ms/step - loss: 0.1010 - accuracy: 0.9704
Epoch 8/20
30/30 [=====] - 2s 56ms/step - loss: 0.0842 - accuracy: 0.9763
Epoch 9/20
30/30 [=====] - 1s 48ms/step - loss: 0.0675 - accuracy: 0.9826
Epoch 10/20
30/30 [=====] - 1s 37ms/step - loss: 0.0572 - accuracy: 0.9856
Epoch 11/20
30/30 [=====] - 1s 36ms/step - loss: 0.0448 - accuracy: 0.9904
Epoch 12/20
30/30 [=====] - 1s 39ms/step - loss: 0.0384 - accuracy: 0.9917
Epoch 13/20
30/30 [=====] - 1s 36ms/step - loss: 0.0284 - accuracy: 0.9957
Epoch 14/20
30/30 [=====] - 1s 37ms/step - loss: 0.0239 - accuracy: 0.9955
Epoch 15/20
30/30 [=====] - 1s 36ms/step - loss: 0.0177 - accuracy: 0.9973
```

```

Epoch 16/20
30/30 [=====] - 1s 36ms/step - loss: 0.0163 - accuracy: 0.9966
Epoch 17/20
30/30 [=====] - 1s 36ms/step - loss: 0.0088 - accuracy: 0.9995
Epoch 18/20
30/30 [=====] - 1s 36ms/step - loss: 0.0091 - accuracy: 0.9993
Epoch 19/20
30/30 [=====] - 1s 37ms/step - loss: 0.0067 - accuracy: 0.9994
Epoch 20/20
30/30 [=====] - 1s 36ms/step - loss: 0.0056 - accuracy: 0.9993

```



```

history_dict = history.history
history_dict.keys()

dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy'])

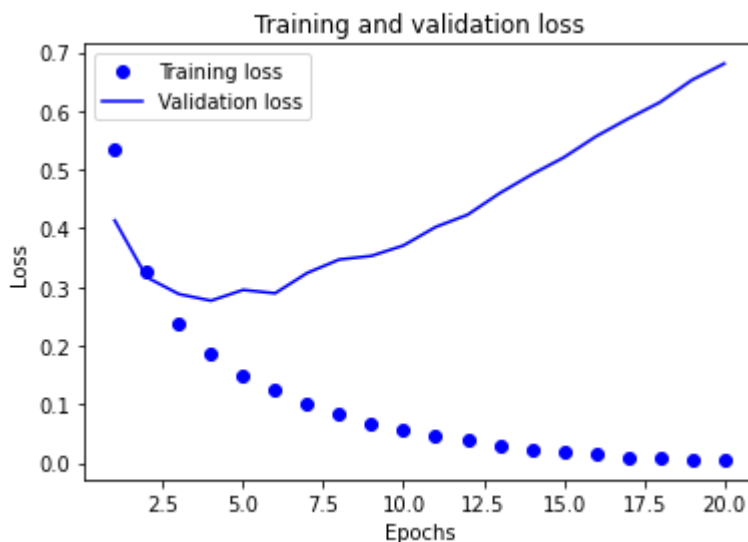
```

Plotting the training and validation loss

```

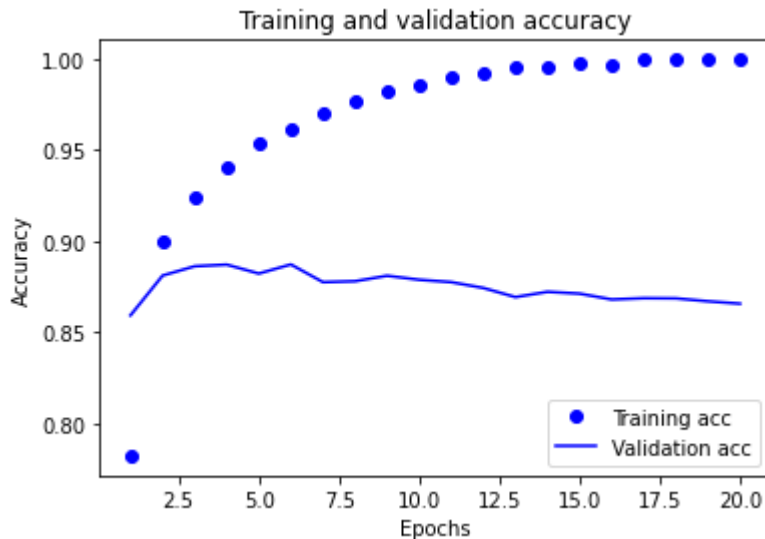
import matplotlib.pyplot as plt
history_dict = history.history
loss_values = history_dict["loss"]
val_loss_values = history_dict["val_loss"]
epochs = range(1, len(loss_values) + 1)
plt.plot(epochs, loss_values, "bo", label="Training loss")
plt.plot(epochs, val_loss_values, "b", label="Validation loss")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

```



Plotting the training and validation accuracy

```
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 a model from scratch

```
model = keras.Sequential([
    layers.Dense(16, activation="relu"),
    layers.Dense(16, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
model.compile(optimizer="rmsprop",
              loss="binary_crossentropy",
              metrics=["accuracy"])
model.fit(x_train, y_train, epochs=4, batch_size=512)
results = model.evaluate(x_test, y_test)
```

Epoch 1/4

49/49 [=====] - 4s 51ms/step - loss: 0.4436 - accuracy: 0.8265

Epoch 2/4

49/49 [=====] - 1s 29ms/step - loss: 0.2524 - accuracy: 0.9121

Epoch 3/4

49/49 [=====] - 1s 29ms/step - loss: 0.1962 - accuracy: 0.9308

Epoch 4/4

49/49 [=====] - 1s 28ms/step - loss: 0.1639 - accuracy: 0.9414

782/782 [=====] - 2s 2ms/step - loss: 0.3483 - accuracy: 0.8641



results

```
[0.34830668568611145, 0.864080011844635]
```

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