# Getting started with neural networks: Classification and regression

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

2, 9,

## 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,
```

```
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,
train_labels[0]
     1
max([max(sequence) for sequence in train data])
```

#### **Decoding reviews back to text**

9999

# 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(64, activation="tanh"),
    layers.Dropout(0.5),
    layers.Dense(1, activation="sigmoid")
])
```

# Compiling the model

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

```
loss="mean_squared_error",
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
 Epoch 2/20
 Epoch 3/20
 Epoch 4/20
 Epoch 5/20
 Epoch 6/20
 Epoch 7/20
 Epoch 8/20
 Epoch 9/20
 Epoch 10/20
 Epoch 11/20
 Epoch 12/20
 Epoch 13/20
 Epoch 14/20
 30/30 [============= ] - 2s 57ms/step - loss: 0.0116 - accuracy: 0.9893
 Epoch 15/20
```

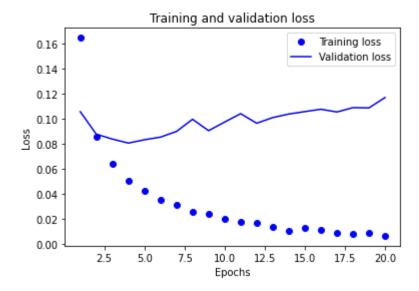
30/30 [============== ] - 2s 57ms/step - loss: 0.0108 - accuracy: 0.9893

```
history_dict = history.history
history_dict.keys()

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

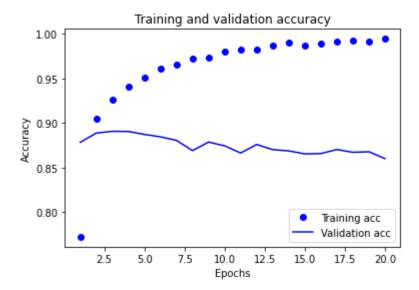
### Plotting the training and validation loss

```
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

layers.Dense(64, activation="tanh"),

model = keras.Sequential([

```
layers.Dropout(0.5),
   layers.Dense(1, activation="sigmoid")
])
model.compile(optimizer="rmsprop",
            loss="mean squared error",
            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 [======
    Epoch 2/4
    49/49 [======
                    ============ ] - 3s 58ms/step - loss: 0.0691 - accuracy: 0.9146
    Epoch 3/4
    49/49 [=====
                          ========] - 2s 44ms/step - loss: 0.0537 - accuracy: 0.9334
    Epoch 4/4
                                =====] - 2s 44ms/step - loss: 0.0438 - accuracy: 0.9456
    49/49 [=====
                          =========] - 2s 3ms/step - loss: 0.1010 - accuracy: 0.8652
```

results

[0.10104542225599289, 0.8651999831199646]

▼ Using a trained model to generate predictions on new data

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