```
[ ] import keras
    keras.__version__
```

```
'2.4.3'
```

```
[ ] from keras.datasets import imdb

    (train_data, train_labels), (test_data, test_labels) = imdb.load_data(num_words=10000)
```

```
Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/imdb.npz
17465344/17464789 [==============================] - 0s 0us/step
<string>:6: VisibleDeprecationWarning: Creating an ndarray from ragged nested sequences (which is a list-or-tuple of lists-or-tuples-or ndarrays with differ
/usr/local/lib/python3.6/dist-packages/tensorflow/python/keras/datasets/imdb.py:159: VisibleDeprecationWarning: Creating an ndarray from ragged nested sequer
  x_train, y_train = np.array(xs[:idx]), np.array(labels[:idx])
/usr/local/lib/python3.6/dist-packages/tensorflow/python/keras/datasets/imdb.py:160: VisibleDeprecationWarning: Creating an ndarray from ragged nested sequer
  x_test, y_test = np.array(xs[idx:]), np.array(labels[idx:])
```

```
train_data[0]
```

```
[1,
 14,
 22,
 16,
 43,
 530,
 973,
 1622,
 1385,
 65,
 458,
 4468,
 66,
 3941,
 4,
 173,
 36,
 256,
 5
```

```
[ ] train_labels[0]
```

```
1
```

```
[ ]
    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_word_index.json
1646592/1641221 [==============================] - 0s 0us/step
```

```
[ ] decoded_review
```

```
'? this film was just brilliant casting location scenery story direction everyone's really suited the part they played and you could just imagine being ther
e robert ? is an amazing actor and now the same being director ? father came from the same scottish island as myself so i loved the fact there was a real co
nnection with this film the witty remarks throughout the film were great it was just brilliant so much that i bought the film as soon as it was released for
? and would recommend it to everyone to watch and the fly fishing was amazing really cried at the end it was so sad and you know what they say if you cry at
a film it must have been good and this definitely was also ? to the two little boy's that played the ? of norman and paul they were just brilliant children
are often left out of the ? list i think because the stars that play them all grown up are such a big profile for the whole film but these children are amaz
ing and should be praised for what they have done don't you thi … '
```

```
import numpy as np

def vectorize_sequences(sequences, dimension=10000):

    results = np.zeros((len(sequences), dimension))
    for i, sequence in enumerate(sequences):
        results[i, sequence] = 1.  # set specific indices of results[i] to 1s
    return results
```

```
[ ] import numpy as np

    def vectorize_sequences(sequences, dimension=10000):

        results = np.zeros((len(sequences), dimension))
        for i, sequence in enumerate(sequences):
            results[i, sequence] = 1.  # set specific indices of results[i] to 1s
        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')
```

```
[ ] from keras import models
    from keras import layers

    model = models.Sequential()
    model.add(layers.Dense(16, activation='relu', input_shape=(10000,)))
    model.add(layers.Dense(16, activation='relu'))
    model.add(layers.Dense(1, activation='sigmoid'))
```

```
[ ] model.compile(optimizer='rmsprop',
                  loss='binary_crossentropy',
                  metrics=['accuracy'])
```

```
[ ] from keras import optimizers

    model.compile(optimizer=optimizers.RMSprop(lr=0.001),
                  loss='binary_crossentropy',
                  metrics=['accuracy'])
```

```
[ ] from keras import losses
    from keras import metrics

    model.compile(optimizer=optimizers.RMSprop(lr=0.001),
                  loss=losses.binary_crossentropy,
                  metrics=[metrics.binary_accuracy])
```

```
    x_val = x_train[:10000]
    partial_x_train = x_train[10000:]

    y_val = y_train[:10000]
    partial_y_train = y_train[10000:]
```

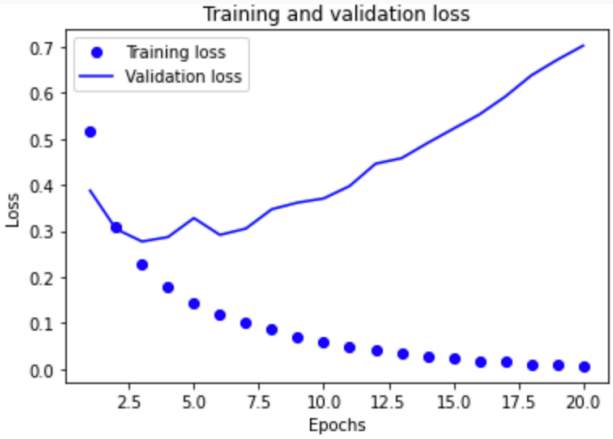
```
[ ] model.compile(optimizer='rmsprop',
                  loss='binary_crossentropy',
                  metrics=['acc'])
    history = model.fit(partial_x_train,
                        partial_y_train,
                        epochs=20,
                        batch_size=512,
                        validation_data=(x_val, y_val))
```

```
Epoch 1/20
30/30 [==============================] - 2s 55ms/step - loss: 0.5948 - acc: 0.6955 - val_loss: 0.3877 - val_acc: 0.8637
Epoch 2/20
30/30 [==============================] - 1s 35ms/step - loss: 0.3254 - acc: 0.8974 - val_loss: 0.3043 - val_acc: 0.8894
Epoch 3/20
30/30 [==============================] - 1s 47ms/step - loss: 0.2319 - acc: 0.9266 - val_loss: 0.2772 - val_acc: 0.8926
Epoch 4/20
30/30 [==============================] - 1s 35ms/step - loss: 0.1770 - acc: 0.9470 - val_loss: 0.2868 - val_acc: 0.8852
```

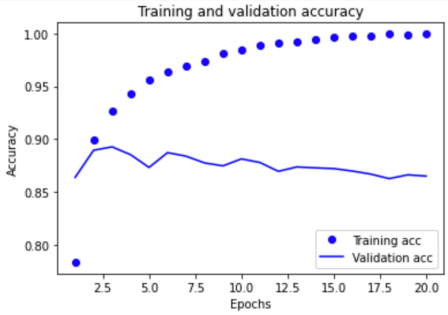
```
[ ] history_dict = history.history
    history_dict.keys()
```

```
dict_keys(['loss', 'acc', 'val_loss', 'val_acc'])
```

```
    import matplotlib.pyplot as plt

    acc = history.history['acc']
    val_acc = history.history['val_acc']
    loss = history.history['loss']
    val_loss = history.history['val_loss']

    epochs = range(1, len(acc) + 1)


    plt.plot(epochs, loss, 'bo', label='Training loss')

    plt.plot(epochs, val_loss, 'b', label='Validation loss')
    plt.title('Training and validation loss')
    plt.xlabel('Epochs')
    plt.ylabel('Loss')
    plt.legend()

    plt.show()
```




```
plt.clf()
acc_values = history_dict['acc']
val_acc_values = history_dict['val_acc']
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()
```




```
model = models.Sequential()
model.add(layers.Dense(16, activation='relu', input_shape=(10000,)))
model.add(layers.Dense(16, activation='relu'))
model.add(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)
results
```

```
Epoch 1/4
49/49 [==============================] - 2s 27ms/step - loss: 0.5391 - accuracy: 0.7429
Epoch 2/4
49/49 [==============================] - 1s 27ms/step - loss: 0.2648 - accuracy: 0.9136
Epoch 3/4
49/49 [==============================] - 1s 30ms/step - loss: 0.1966 - accuracy: 0.9353
Epoch 4/4
49/49 [==============================] - 1s 27ms/step - loss: 0.1615 - accuracy: 0.9444
782/782 [==============================] - 2s 2ms/step - loss: 0.3035 - accuracy: 0.8797
[0.3035437762737274, 0.8797199726104736]
```

```
model.predict(x_test)
```

```
array([[0.22093734],
       [0.9981879 ],
       [0.9459506 ],
       ...,
       [0.17986917],
       [0.05824137],
       [0.7571063 ]], dtype=float32)
```