

# Assignment\_1

February 27, 2022

```
[129]: # Importing the required libraries
```

```
[130]: import matplotlib
import matplotlib.pyplot as plt
import numpy as np
from keras import models
from keras import layers
```

```
[131]: # Loading the IMDB dataset using tensorflow and keras
```

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

```
[133]: # Decoding reviews back to text
```

```
[134]: 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]])
```

```
[135]: # Vectorized the dataset by Creating an all-zero matrix of shape
↳ (len(sequences), dimension) and Sets specific indices of results[i] to 1s
```

```
[136]: 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
```

```
[137]: # Vectorizing Training and Test Dataset
```

```
[138]: 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")
```

```
[139]: # Model defination
```

```
[140]: from tensorflow import keras
model = keras.Sequential([
    layers.Dense(16, activation="relu"),
    layers.Dense(16, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
```

```
[141]: # Model Compilation
```

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

```
[143]: # Setting aside a validation set
```

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

```
[145]: # Model Training
```

```
[146]: 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 [=====] - 6s 152ms/step - loss: 0.5336 - accuracy: 0.7844 - val\_loss: 0.4159 - val\_accuracy: 0.8505

Epoch 2/20

30/30 [=====] - 1s 21ms/step - loss: 0.3238 - accuracy: 0.9024 - val\_loss: 0.3150 - val\_accuracy: 0.8853

Epoch 3/20

30/30 [=====] - 1s 21ms/step - loss: 0.2369 - accuracy: 0.9251 - val\_loss: 0.2834 - val\_accuracy: 0.8895

Epoch 4/20

30/30 [=====] - 1s 21ms/step - loss: 0.1892 - accuracy: 0.9381 - val\_loss: 0.2771 - val\_accuracy: 0.8895

Epoch 5/20

30/30 [=====] - 1s 20ms/step - loss: 0.1550 - accuracy: 0.9509 - val\_loss: 0.2816 - val\_accuracy: 0.8860

Epoch 6/20

30/30 [=====] - 1s 20ms/step - loss: 0.1299 - accuracy: 0.9589 - val\_loss: 0.2924 - val\_accuracy: 0.8856

```

Epoch 7/20
30/30 [=====] - 1s 21ms/step - loss: 0.1093 - accuracy:
0.9677 - val_loss: 0.2999 - val_accuracy: 0.8843
Epoch 8/20
30/30 [=====] - 1s 21ms/step - loss: 0.0921 - accuracy:
0.9728 - val_loss: 0.3373 - val_accuracy: 0.8761
Epoch 9/20
30/30 [=====] - 1s 21ms/step - loss: 0.0792 - accuracy:
0.9777 - val_loss: 0.3348 - val_accuracy: 0.8812
Epoch 10/20
30/30 [=====] - 1s 20ms/step - loss: 0.0638 - accuracy:
0.9835 - val_loss: 0.3708 - val_accuracy: 0.8781
Epoch 11/20
30/30 [=====] - 1s 20ms/step - loss: 0.0522 - accuracy:
0.9873 - val_loss: 0.4151 - val_accuracy: 0.8732
Epoch 12/20
30/30 [=====] - 1s 21ms/step - loss: 0.0448 - accuracy:
0.9889 - val_loss: 0.4396 - val_accuracy: 0.8677
Epoch 13/20
30/30 [=====] - 1s 22ms/step - loss: 0.0374 - accuracy:
0.9920 - val_loss: 0.4429 - val_accuracy: 0.8730
Epoch 14/20
30/30 [=====] - 1s 20ms/step - loss: 0.0297 - accuracy:
0.9938 - val_loss: 0.4723 - val_accuracy: 0.8742
Epoch 15/20
30/30 [=====] - 1s 21ms/step - loss: 0.0247 - accuracy:
0.9956 - val_loss: 0.5005 - val_accuracy: 0.8731
Epoch 16/20
30/30 [=====] - 1s 21ms/step - loss: 0.0188 - accuracy:
0.9980 - val_loss: 0.5368 - val_accuracy: 0.8699
Epoch 17/20
30/30 [=====] - 1s 21ms/step - loss: 0.0163 - accuracy:
0.9979 - val_loss: 0.5681 - val_accuracy: 0.8705
Epoch 18/20
30/30 [=====] - 1s 20ms/step - loss: 0.0140 - accuracy:
0.9980 - val_loss: 0.6037 - val_accuracy: 0.8696
Epoch 19/20
30/30 [=====] - 1s 20ms/step - loss: 0.0101 - accuracy:
0.9987 - val_loss: 0.6494 - val_accuracy: 0.8637
Epoch 20/20
30/30 [=====] - 1s 22ms/step - loss: 0.0088 - accuracy:
0.9987 - val_loss: 0.6895 - val_accuracy: 0.8624

```

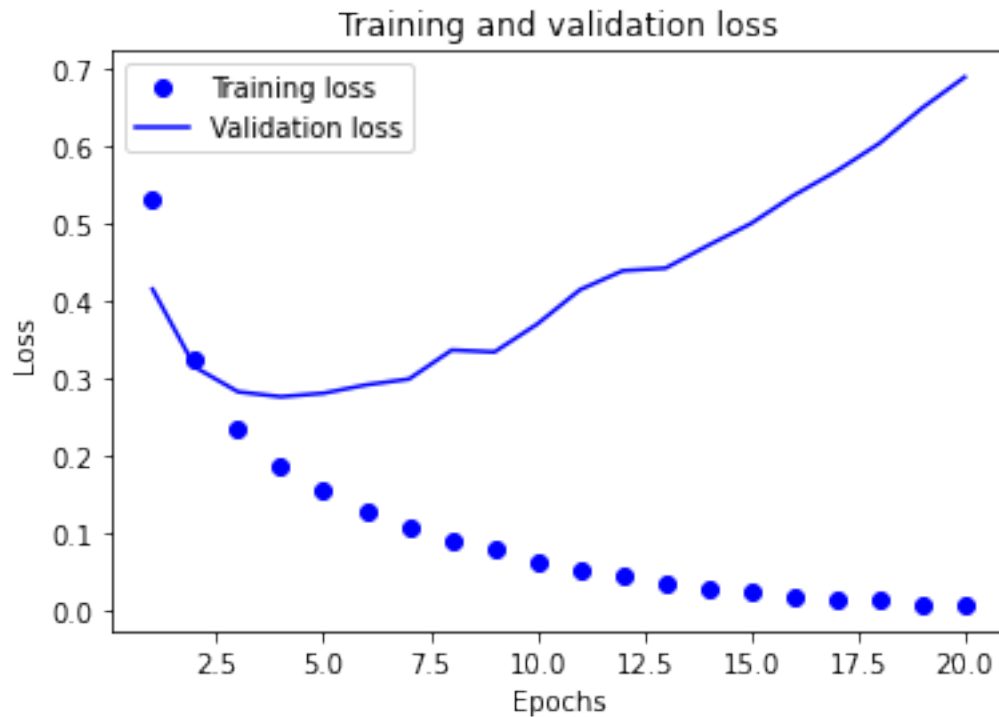
```
[147]: # Plotting of Training and Validation Loss
```

```
[148]: 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()

```

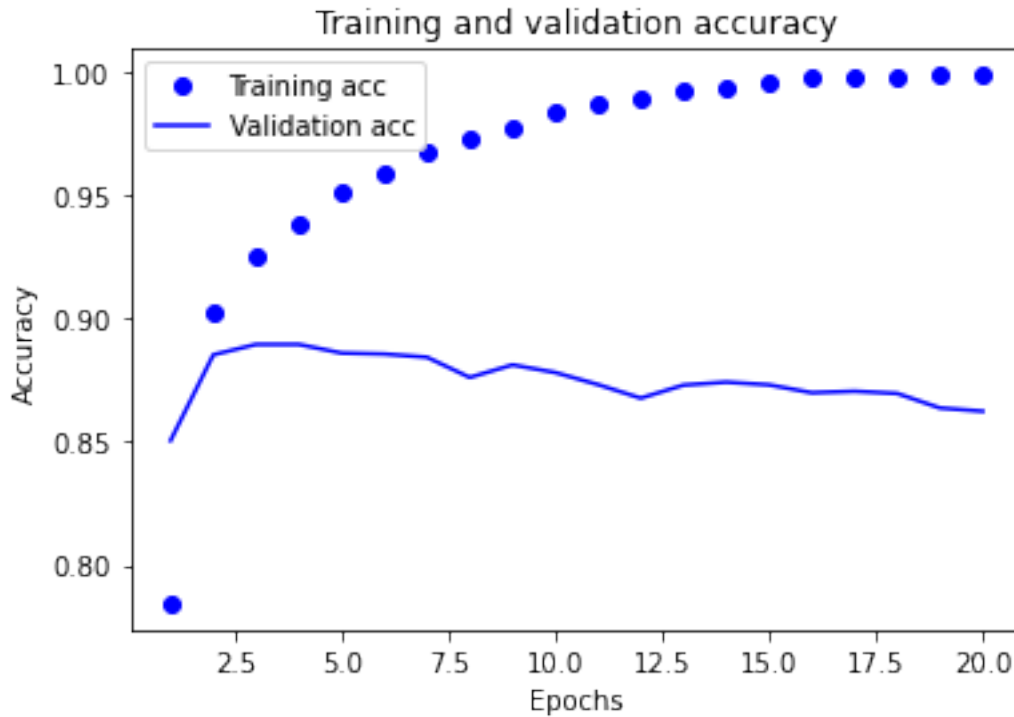


[149]: *#Plotting Training and Validation accuracy*

```

[150]: 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()

```



```
[151]: # Retraining Model from begining
```

```
[152]: 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 [=====] - 2s 18ms/step - loss: 0.4724 - accuracy: 0.8111

Epoch 2/4

49/49 [=====] - 1s 21ms/step - loss: 0.2698 - accuracy: 0.9076

Epoch 3/4

49/49 [=====] - 1s 15ms/step - loss: 0.2050 - accuracy: 0.9284

Epoch 4/4

49/49 [=====] - 1s 14ms/step - loss: 0.1725 - accuracy:

```
0.9391
782/782 [=====] - 2s 1ms/step - loss: 0.2914 -
accuracy: 0.8844
```

```
[153]: # ASSIGNMENT
# You used two hidden layers. Try using one or three hidden layers, and see how
↳ doing so affects validation and test accuracy.
```

```
[154]: model_1 = keras.Sequential([
    layers.Dense(16, activation="relu"),
    layers.Dense(16, activation="relu"),
    layers.Dense(16, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])

model_2 = keras.Sequential([
    layers.Dense(16, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
```

```
[155]: model_1.compile(optimizer="rmsprop",
    loss="binary_crossentropy",
    metrics=["accuracy"])

model_2.compile(optimizer="rmsprop",
    loss="binary_crossentropy",
    metrics=["accuracy"])
```

```
[156]: # Model Training
```

```
[157]: history_1 = model_1.fit(partial_x_train,
    partial_y_train,
    epochs=20,
    batch_size=512,
    validation_data=(x_val, y_val))

history_2 = model_2.fit(partial_x_train,
    partial_y_train,
    epochs=20,
    batch_size=512,
    validation_data=(x_val, y_val))
```

```
Epoch 1/20
30/30 [=====] - 3s 76ms/step - loss: 0.5439 - accuracy:
0.7541 - val_loss: 0.4034 - val_accuracy: 0.8453
Epoch 2/20
30/30 [=====] - 1s 23ms/step - loss: 0.3021 - accuracy:
0.9047 - val_loss: 0.3501 - val_accuracy: 0.8569
```

Epoch 3/20  
30/30 [=====] - 1s 21ms/step - loss: 0.2168 - accuracy: 0.9280 - val\_loss: 0.3633 - val\_accuracy: 0.8541

Epoch 4/20  
30/30 [=====] - 1s 21ms/step - loss: 0.1627 - accuracy: 0.9471 - val\_loss: 0.3552 - val\_accuracy: 0.8654

Epoch 5/20  
30/30 [=====] - 1s 22ms/step - loss: 0.1346 - accuracy: 0.9557 - val\_loss: 0.3249 - val\_accuracy: 0.8744

Epoch 6/20  
30/30 [=====] - 1s 20ms/step - loss: 0.1094 - accuracy: 0.9668 - val\_loss: 0.3214 - val\_accuracy: 0.8822

Epoch 7/20  
30/30 [=====] - 1s 21ms/step - loss: 0.0888 - accuracy: 0.9732 - val\_loss: 0.3395 - val\_accuracy: 0.8804

Epoch 8/20  
30/30 [=====] - 1s 21ms/step - loss: 0.0688 - accuracy: 0.9811 - val\_loss: 0.3799 - val\_accuracy: 0.8782

Epoch 9/20  
30/30 [=====] - 1s 20ms/step - loss: 0.0643 - accuracy: 0.9822 - val\_loss: 0.4152 - val\_accuracy: 0.8771

Epoch 10/20  
30/30 [=====] - 1s 20ms/step - loss: 0.0533 - accuracy: 0.9860 - val\_loss: 0.4221 - val\_accuracy: 0.8769

Epoch 11/20  
30/30 [=====] - 1s 20ms/step - loss: 0.0374 - accuracy: 0.9917 - val\_loss: 0.5415 - val\_accuracy: 0.8642

Epoch 12/20  
30/30 [=====] - 1s 20ms/step - loss: 0.0254 - accuracy: 0.9967 - val\_loss: 0.6101 - val\_accuracy: 0.8609

Epoch 13/20  
30/30 [=====] - 1s 19ms/step - loss: 0.0265 - accuracy: 0.9943 - val\_loss: 0.5272 - val\_accuracy: 0.8730

Epoch 14/20  
30/30 [=====] - 1s 19ms/step - loss: 0.0277 - accuracy: 0.9927 - val\_loss: 0.5554 - val\_accuracy: 0.8690

Epoch 15/20  
30/30 [=====] - 1s 20ms/step - loss: 0.0118 - accuracy: 0.9986 - val\_loss: 0.6204 - val\_accuracy: 0.8689

Epoch 16/20  
30/30 [=====] - 1s 20ms/step - loss: 0.0185 - accuracy: 0.9952 - val\_loss: 0.6552 - val\_accuracy: 0.8684

Epoch 17/20  
30/30 [=====] - 1s 20ms/step - loss: 0.0066 - accuracy: 0.9991 - val\_loss: 0.7529 - val\_accuracy: 0.8661

Epoch 18/20  
30/30 [=====] - 1s 21ms/step - loss: 0.0209 - accuracy: 0.9947 - val\_loss: 0.7473 - val\_accuracy: 0.8661

Epoch 19/20  
30/30 [=====] - 1s 28ms/step - loss: 0.0031 - accuracy: 0.9996 - val\_loss: 0.7709 - val\_accuracy: 0.8658

Epoch 20/20  
30/30 [=====] - 1s 23ms/step - loss: 0.0161 - accuracy: 0.9954 - val\_loss: 0.8077 - val\_accuracy: 0.8643

Epoch 1/20  
30/30 [=====] - 2s 34ms/step - loss: 0.5036 - accuracy: 0.7919 - val\_loss: 0.3921 - val\_accuracy: 0.8660

Epoch 2/20  
30/30 [=====] - 1s 22ms/step - loss: 0.3197 - accuracy: 0.9040 - val\_loss: 0.3221 - val\_accuracy: 0.8832

Epoch 3/20  
30/30 [=====] - 1s 23ms/step - loss: 0.2477 - accuracy: 0.9235 - val\_loss: 0.2945 - val\_accuracy: 0.8888

Epoch 4/20  
30/30 [=====] - 1s 24ms/step - loss: 0.2026 - accuracy: 0.9376 - val\_loss: 0.2783 - val\_accuracy: 0.8917

Epoch 5/20  
30/30 [=====] - 1s 21ms/step - loss: 0.1732 - accuracy: 0.9454 - val\_loss: 0.2790 - val\_accuracy: 0.8892

Epoch 6/20  
30/30 [=====] - 1s 20ms/step - loss: 0.1491 - accuracy: 0.9559 - val\_loss: 0.2754 - val\_accuracy: 0.8901

Epoch 7/20  
30/30 [=====] - 1s 21ms/step - loss: 0.1311 - accuracy: 0.9619 - val\_loss: 0.2857 - val\_accuracy: 0.8844

Epoch 8/20  
30/30 [=====] - 1s 21ms/step - loss: 0.1153 - accuracy: 0.9681 - val\_loss: 0.2870 - val\_accuracy: 0.8864

Epoch 9/20  
30/30 [=====] - 1s 21ms/step - loss: 0.1010 - accuracy: 0.9736 - val\_loss: 0.3017 - val\_accuracy: 0.8867

Epoch 10/20  
30/30 [=====] - 1s 21ms/step - loss: 0.0903 - accuracy: 0.9769 - val\_loss: 0.3122 - val\_accuracy: 0.8848

Epoch 11/20  
30/30 [=====] - 1s 20ms/step - loss: 0.0806 - accuracy: 0.9795 - val\_loss: 0.3174 - val\_accuracy: 0.8836

Epoch 12/20  
30/30 [=====] - 1s 21ms/step - loss: 0.0703 - accuracy: 0.9835 - val\_loss: 0.3344 - val\_accuracy: 0.8835

Epoch 13/20  
30/30 [=====] - 1s 22ms/step - loss: 0.0615 - accuracy: 0.9871 - val\_loss: 0.3497 - val\_accuracy: 0.8768

Epoch 14/20  
30/30 [=====] - 1s 20ms/step - loss: 0.0551 - accuracy: 0.9878 - val\_loss: 0.3610 - val\_accuracy: 0.8780



```

Epoch 15/20
30/30 [=====] - 1s 20ms/step - loss: 0.0487 - accuracy:
0.9901 - val_loss: 0.3786 - val_accuracy: 0.8790
Epoch 16/20
30/30 [=====] - 1s 20ms/step - loss: 0.0427 - accuracy:
0.9923 - val_loss: 0.4054 - val_accuracy: 0.8694
Epoch 17/20
30/30 [=====] - 1s 19ms/step - loss: 0.0376 - accuracy:
0.9935 - val_loss: 0.4157 - val_accuracy: 0.8724
Epoch 18/20
30/30 [=====] - 1s 20ms/step - loss: 0.0326 - accuracy:
0.9949 - val_loss: 0.4305 - val_accuracy: 0.8755
Epoch 19/20
30/30 [=====] - 1s 20ms/step - loss: 0.0293 - accuracy:
0.9957 - val_loss: 0.4481 - val_accuracy: 0.8722
Epoch 20/20
30/30 [=====] - 1s 19ms/step - loss: 0.0251 - accuracy:
0.9970 - val_loss: 0.4673 - val_accuracy: 0.8711

```

```

[158]: model_1.summary()
       model_2.summary()

```

```
Model: "sequential_18"
```

Layer (type)	Output Shape	Param #
dense_54 (Dense)	(None, 16)	160016
dense_55 (Dense)	(None, 16)	272
dense_56 (Dense)	(None, 16)	272
dense_57 (Dense)	(None, 1)	17

```

=====
Total params: 160,577
Trainable params: 160,577
Non-trainable params: 0

```

```
Model: "sequential_19"
```

Layer (type)	Output Shape	Param #
dense_58 (Dense)	(None, 16)	160016
dense_59 (Dense)	(None, 1)	17

Total params: 160,033  
Trainable params: 160,033  
Non-trainable params: 0

---

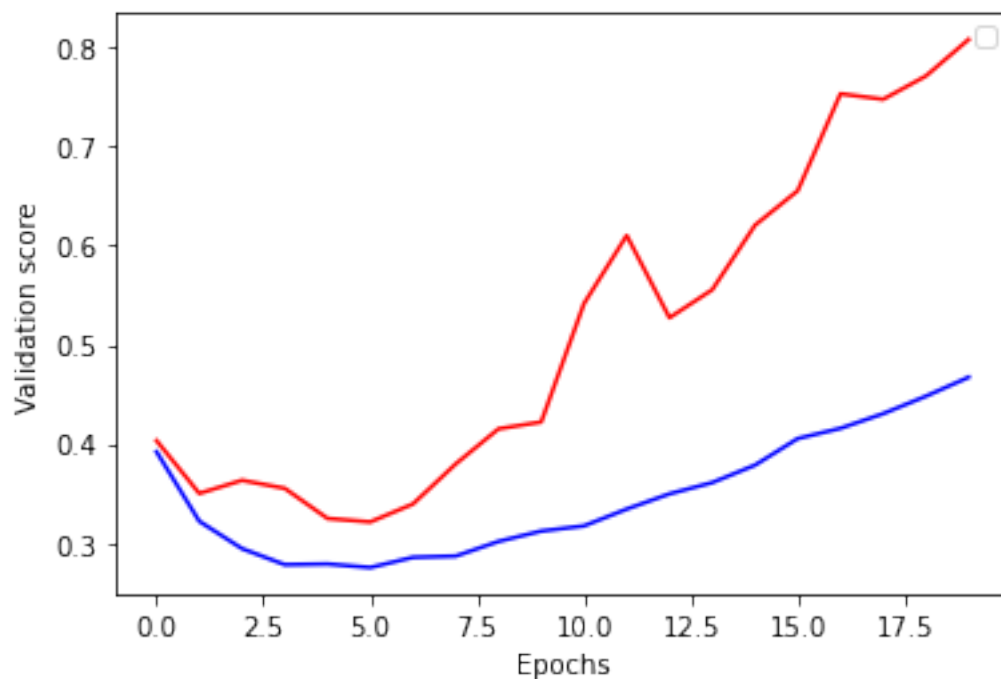
```
[159]: # Plotting the training and validation loss
```

```
[160]: history_dict_1 = history_1.history
history_dict_2 = history_2.history

plt.plot(history_1.history['val_loss'], 'r', history_2.history['val_loss'], 'b')
plt.xlabel('Epochs')
plt.ylabel('Validation score')
plt.legend()
```

No handles with labels found to put in legend.

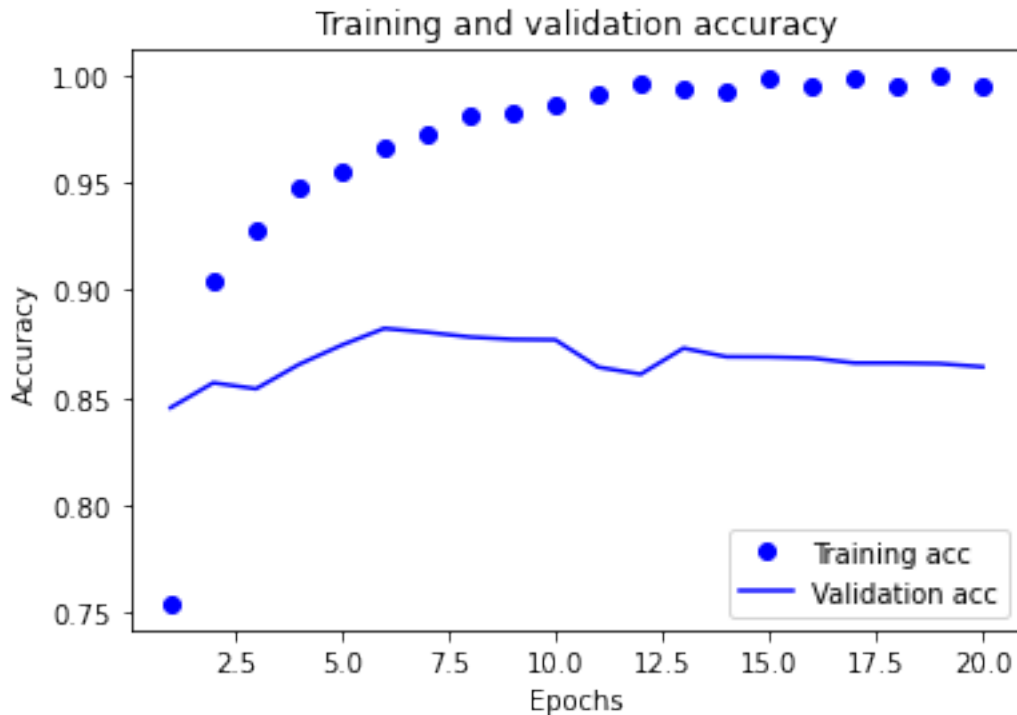
```
[160]: <matplotlib.legend.Legend at 0x193c1002df0>
```



```
[161]: # Plotting Training and Validation accuracy
```

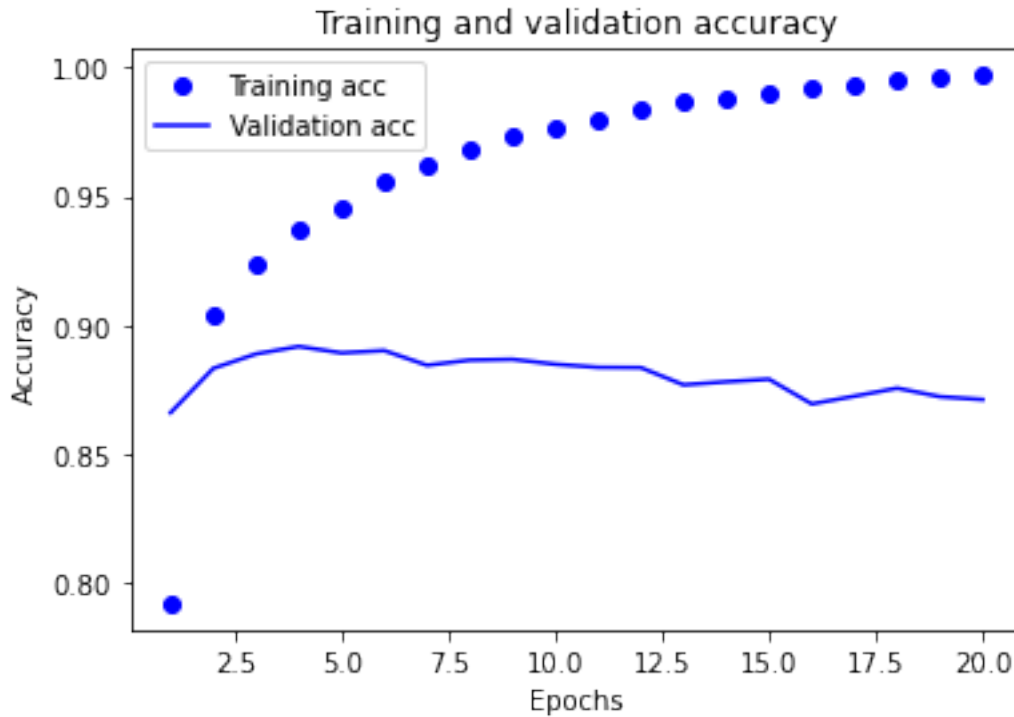
```
[162]: plt.clf()
acc = history_dict_1["accuracy"]
val_acc = history_dict_1["val_accuracy"]
plt.plot(epochs, acc, "bo", label="Training acc")
```

```
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()
```



```
[163]: # plot_loss
# (history_dict_1.history['loss'], history_dict_1.history['val_loss'])
```

```
[164]: plt.clf()
acc = history_dict_2["accuracy"]
val_acc = history_dict_2["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()
```



[165]: *# 2. Try using layers with more hidden units or fewer hidden units: 32 units, 64 units, and so on.*

```
[166]: model_3 = keras.Sequential([
        layers.Dense(32, activation="relu"),
        layers.Dense(64, activation="relu"),
        layers.Dense(1, activation="sigmoid")
    ])
```

```
[167]: model_3.compile(optimizer="rmsprop",
        loss="binary_crossentropy",
        metrics=["accuracy"])
```

```
[168]: history_3 = model_3.fit(partial_x_train,
        partial_y_train,
        epochs=20,
        batch_size=512,
        validation_data=(x_val, y_val))
```

Epoch 1/20

30/30 [=====] - 2s 37ms/step - loss: 0.4925 - accuracy: 0.7789 - val\_loss: 0.3674 - val\_accuracy: 0.8545

Epoch 2/20

30/30 [=====] - 1s 31ms/step - loss: 0.2745 - accuracy:

0.9039 - val\_loss: 0.2831 - val\_accuracy: 0.8924  
Epoch 3/20  
30/30 [=====] - 1s 23ms/step - loss: 0.1991 - accuracy:  
0.9304 - val\_loss: 0.2907 - val\_accuracy: 0.8840  
Epoch 4/20  
30/30 [=====] - 1s 28ms/step - loss: 0.1546 - accuracy:  
0.9467 - val\_loss: 0.2973 - val\_accuracy: 0.8829  
Epoch 5/20  
30/30 [=====] - 1s 33ms/step - loss: 0.1258 - accuracy:  
0.9565 - val\_loss: 0.3024 - val\_accuracy: 0.8858  
Epoch 6/20  
30/30 [=====] - 1s 31ms/step - loss: 0.0943 - accuracy:  
0.9707 - val\_loss: 0.3209 - val\_accuracy: 0.8830  
Epoch 7/20  
30/30 [=====] - 1s 44ms/step - loss: 0.0768 - accuracy:  
0.9774 - val\_loss: 0.3499 - val\_accuracy: 0.8803  
Epoch 8/20  
30/30 [=====] - 1s 32ms/step - loss: 0.0622 - accuracy:  
0.9813 - val\_loss: 0.3874 - val\_accuracy: 0.8802  
Epoch 9/20  
30/30 [=====] - 1s 25ms/step - loss: 0.0457 - accuracy:  
0.9873 - val\_loss: 0.4173 - val\_accuracy: 0.8782  
Epoch 10/20  
30/30 [=====] - 1s 25ms/step - loss: 0.0382 - accuracy:  
0.9883 - val\_loss: 0.4583 - val\_accuracy: 0.8743  
Epoch 11/20  
30/30 [=====] - 1s 26ms/step - loss: 0.0253 - accuracy:  
0.9943 - val\_loss: 0.5374 - val\_accuracy: 0.8628  
Epoch 12/20  
30/30 [=====] - 1s 25ms/step - loss: 0.0140 - accuracy:  
0.9982 - val\_loss: 0.5763 - val\_accuracy: 0.8662  
Epoch 13/20  
30/30 [=====] - 1s 24ms/step - loss: 0.0259 - accuracy:  
0.9916 - val\_loss: 0.5655 - val\_accuracy: 0.8716  
Epoch 14/20  
30/30 [=====] - 1s 26ms/step - loss: 0.0053 - accuracy:  
0.9999 - val\_loss: 0.6009 - val\_accuracy: 0.8723  
Epoch 15/20  
30/30 [=====] - 1s 27ms/step - loss: 0.0228 - accuracy:  
0.9933 - val\_loss: 0.6289 - val\_accuracy: 0.8692  
Epoch 16/20  
30/30 [=====] - 1s 25ms/step - loss: 0.0029 - accuracy:  
0.9999 - val\_loss: 0.6596 - val\_accuracy: 0.8703  
Epoch 17/20  
30/30 [=====] - 1s 25ms/step - loss: 0.0021 - accuracy:  
0.9999 - val\_loss: 0.7398 - val\_accuracy: 0.8666  
Epoch 18/20  
30/30 [=====] - 1s 26ms/step - loss: 0.0211 - accuracy:

```
0.9948 - val_loss: 0.7671 - val_accuracy: 0.8675
Epoch 19/20
30/30 [=====] - 1s 30ms/step - loss: 9.4256e-04 -
accuracy: 0.9999 - val_loss: 0.7942 - val_accuracy: 0.8680
Epoch 20/20
30/30 [=====] - 1s 24ms/step - loss: 7.2426e-04 -
accuracy: 0.9999 - val_loss: 0.8358 - val_accuracy: 0.8663
```

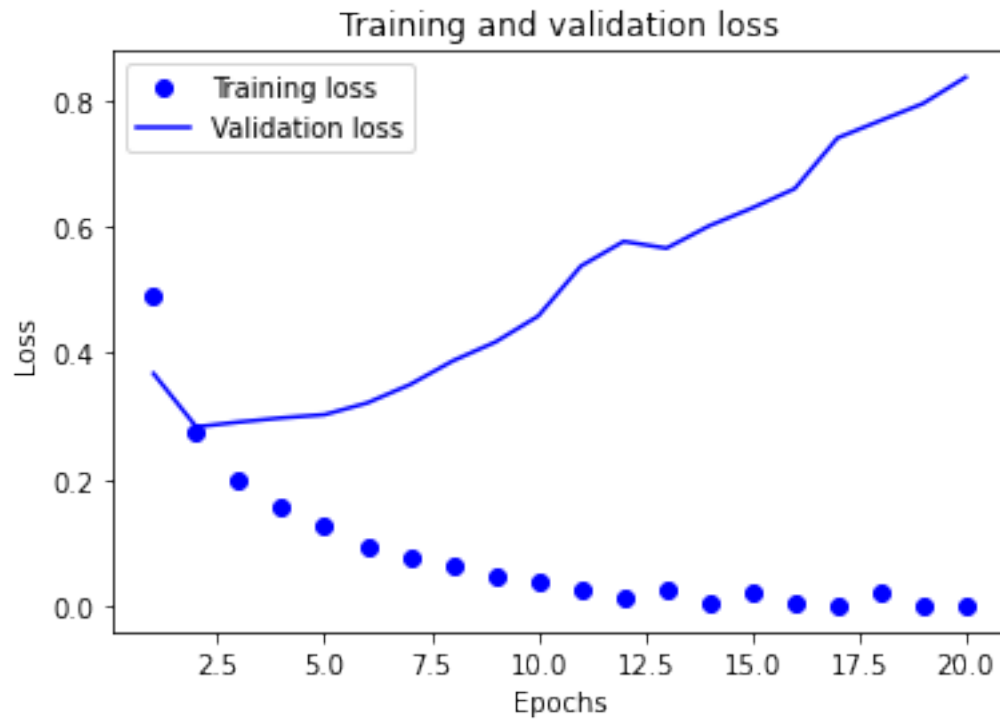
```
[169]: model_3.summary()
```

```
Model: "sequential_20"
```

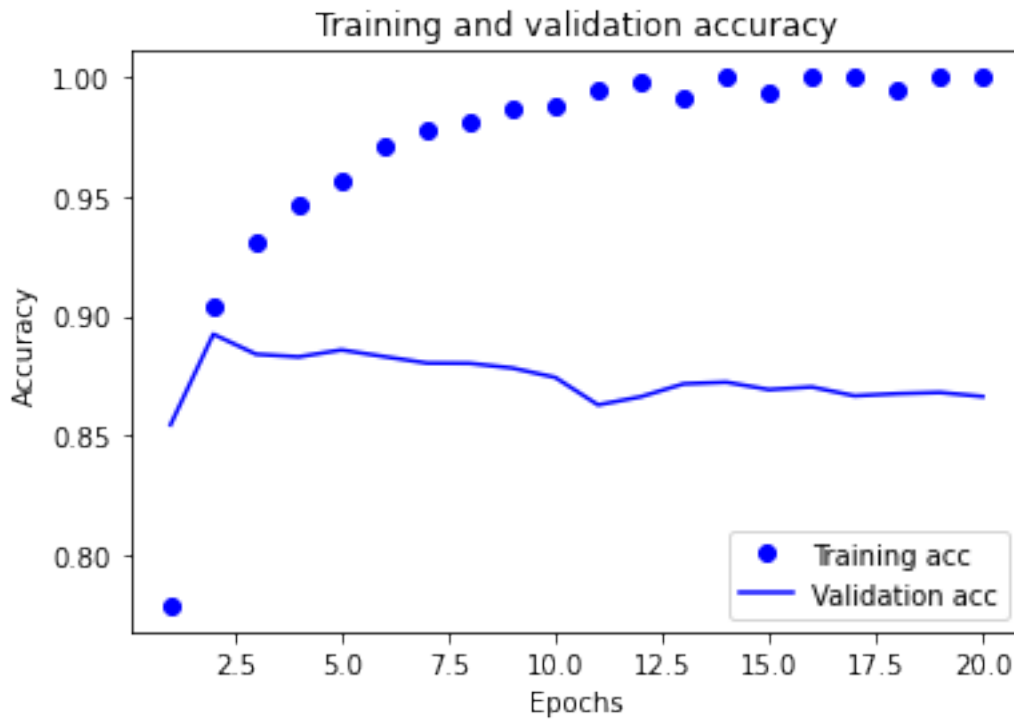
Layer (type)	Output Shape	Param #
dense_60 (Dense)	(None, 32)	320032
dense_61 (Dense)	(None, 64)	2112
dense_62 (Dense)	(None, 1)	65

```
=====  
Total params: 322,209  
Trainable params: 322,209  
Non-trainable params: 0  
=====
```

```
[170]: history_dict_3 = history_3.history
loss_values = history_dict_3["loss"]
val_loss_values = history_dict_3["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()
```



```
[171]: plt.clf()
acc = history_dict_3["accuracy"]
val_acc = history_dict_3["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()
```



```
[172]: # 3.Try using the 'mse' loss function instead of 'binary_crossentropy
```

```
[173]: model_4 = keras.Sequential([
        layers.Dense(16, activation="relu"),
        layers.Dense(16, activation="relu"),
        layers.Dense(1, activation="sigmoid")
    ])
```

```
[174]: model_4.compile(optimizer="rmsprop",
                    loss="mse",
                    metrics=["accuracy"])
```

```
[175]: # Training your model
```

```
history_4 = model_4.fit(partial_x_train,
                        partial_y_train,
                        epochs=20,
                        batch_size=512,
                        validation_data=(x_val, y_val))
```

Epoch 1/20

30/30 [=====] - 2s 36ms/step - loss: 0.1648 - accuracy: 0.7859 - val\_loss: 0.1126 - val\_accuracy: 0.8724

Epoch 2/20



30/30 [=====] - 1s 25ms/step - loss: 0.0873 - accuracy: 0.9042 - val\_loss: 0.0952 - val\_accuracy: 0.8814  
Epoch 3/20  
30/30 [=====] - 1s 22ms/step - loss: 0.0627 - accuracy: 0.9317 - val\_loss: 0.0864 - val\_accuracy: 0.8877  
Epoch 4/20  
30/30 [=====] - 1s 24ms/step - loss: 0.0495 - accuracy: 0.9463 - val\_loss: 0.0824 - val\_accuracy: 0.8891  
Epoch 5/20  
30/30 [=====] - 1s 25ms/step - loss: 0.0405 - accuracy: 0.9571 - val\_loss: 0.0836 - val\_accuracy: 0.8864  
Epoch 6/20  
30/30 [=====] - 1s 24ms/step - loss: 0.0332 - accuracy: 0.9661 - val\_loss: 0.0846 - val\_accuracy: 0.8842  
Epoch 7/20  
30/30 [=====] - 1s 23ms/step - loss: 0.0279 - accuracy: 0.9735 - val\_loss: 0.0897 - val\_accuracy: 0.8764  
Epoch 8/20  
30/30 [=====] - 1s 26ms/step - loss: 0.0223 - accuracy: 0.9801 - val\_loss: 0.0878 - val\_accuracy: 0.8818  
Epoch 9/20  
30/30 [=====] - 1s 22ms/step - loss: 0.0187 - accuracy: 0.9841 - val\_loss: 0.0911 - val\_accuracy: 0.8805  
Epoch 10/20  
30/30 [=====] - 1s 20ms/step - loss: 0.0146 - accuracy: 0.9886 - val\_loss: 0.0953 - val\_accuracy: 0.8726  
Epoch 11/20  
30/30 [=====] - 1s 21ms/step - loss: 0.0134 - accuracy: 0.9896 - val\_loss: 0.0946 - val\_accuracy: 0.8779  
Epoch 12/20  
30/30 [=====] - 1s 20ms/step - loss: 0.0110 - accuracy: 0.9913 - val\_loss: 0.0967 - val\_accuracy: 0.8768  
Epoch 13/20  
30/30 [=====] - 1s 20ms/step - loss: 0.0096 - accuracy: 0.9923 - val\_loss: 0.0986 - val\_accuracy: 0.8745  
Epoch 14/20  
30/30 [=====] - 1s 20ms/step - loss: 0.0072 - accuracy: 0.9947 - val\_loss: 0.1007 - val\_accuracy: 0.8703  
Epoch 15/20  
30/30 [=====] - 1s 21ms/step - loss: 0.0067 - accuracy: 0.9948 - val\_loss: 0.1021 - val\_accuracy: 0.8723  
Epoch 16/20  
30/30 [=====] - 1s 24ms/step - loss: 0.0061 - accuracy: 0.9950 - val\_loss: 0.1041 - val\_accuracy: 0.8680  
Epoch 17/20  
30/30 [=====] - 1s 22ms/step - loss: 0.0039 - accuracy: 0.9971 - val\_loss: 0.1085 - val\_accuracy: 0.8672  
Epoch 18/20

```

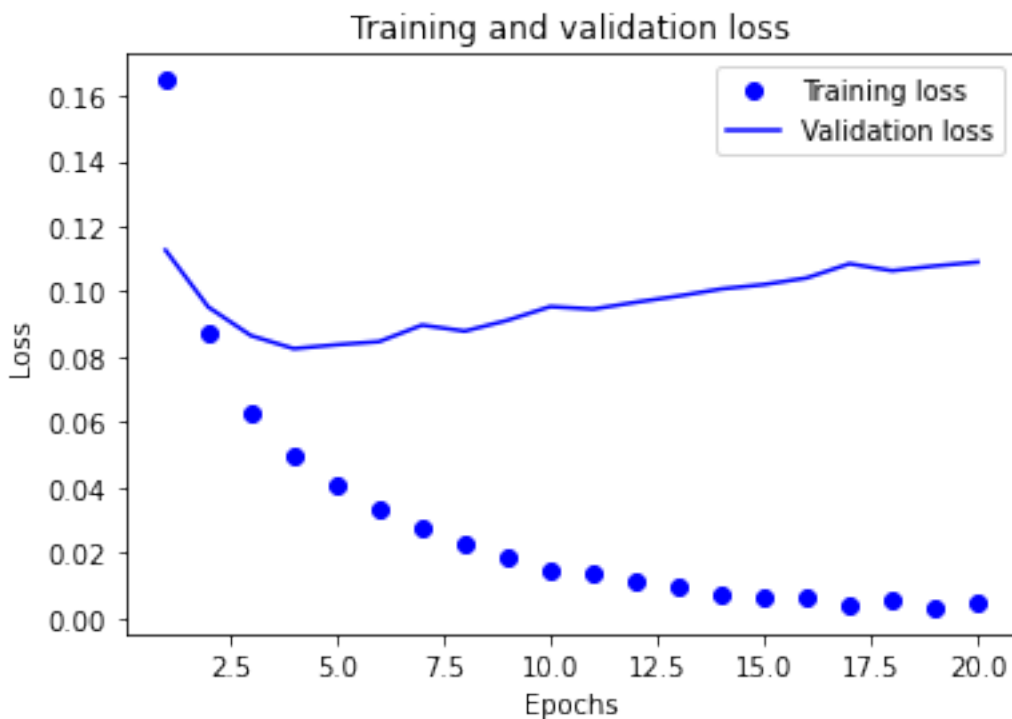
30/30 [=====] - 1s 28ms/step - loss: 0.0054 - accuracy:
0.9949 - val_loss: 0.1063 - val_accuracy: 0.8675
Epoch 19/20
30/30 [=====] - 1s 22ms/step - loss: 0.0031 - accuracy:
0.9974 - val_loss: 0.1078 - val_accuracy: 0.8664
Epoch 20/20
30/30 [=====] - 1s 23ms/step - loss: 0.0048 - accuracy:
0.9949 - val_loss: 0.1090 - val_accuracy: 0.8664

```

```

[176]: history_dict_4 = history_4.history
loss_values = history_dict_4["loss"]
val_loss_values = history_dict_4["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()

```

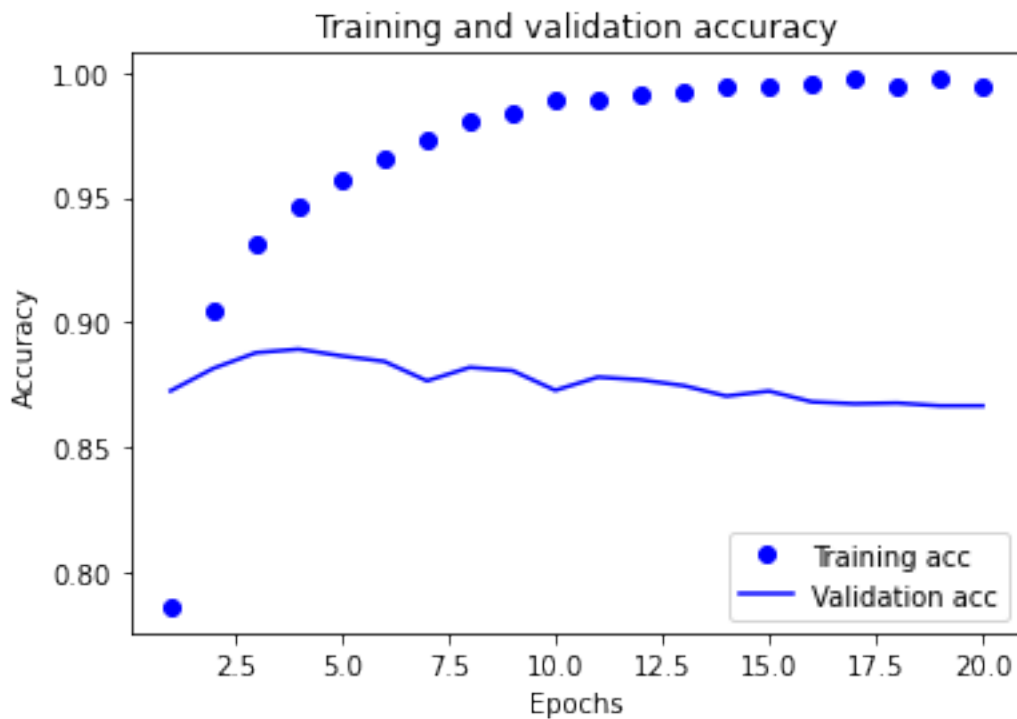


```

[ ]: # Plotting the training and validation accuracy

```

```
[177]: plt.clf()
acc = history_dict_4["accuracy"]
val_acc = history_dict_4["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()
```



```
[178]: # 4. Try using the tanh activation (an activation that was popular in the early
↳ days of neural networks) instead of 'relu'
```

```
[179]: model_5 = keras.Sequential([
    layers.Dense(16, activation="tanh"),
    layers.Dense(16, activation="tanh"),
    layers.Dense(1, activation="sigmoid")
])
```

```
[180]: model_5.compile(optimizer="rmsprop",
    loss="mse",
    metrics=["accuracy"])
```

```
[181]: history_5 = model_5.fit(partial_x_train,
                                partial_y_train,
                                epochs=20,
                                batch_size=512,
                                validation_data=(x_val, y_val))
```

Epoch 1/20

30/30 [=====] - 2s 46ms/step - loss: 0.1591 - accuracy: 0.7988 - val\_loss: 0.1084 - val\_accuracy: 0.8737

Epoch 2/20

30/30 [=====] - 1s 22ms/step - loss: 0.0791 - accuracy: 0.9123 - val\_loss: 0.0996 - val\_accuracy: 0.8643

Epoch 3/20

30/30 [=====] - 1s 19ms/step - loss: 0.0549 - accuracy: 0.9363 - val\_loss: 0.0819 - val\_accuracy: 0.8910

Epoch 4/20

30/30 [=====] - 1s 19ms/step - loss: 0.0413 - accuracy: 0.9527 - val\_loss: 0.0865 - val\_accuracy: 0.8824

Epoch 5/20

30/30 [=====] - 1s 19ms/step - loss: 0.0311 - accuracy: 0.9643 - val\_loss: 0.0891 - val\_accuracy: 0.8789

Epoch 6/20

30/30 [=====] - 1s 20ms/step - loss: 0.0233 - accuracy: 0.9743 - val\_loss: 0.0929 - val\_accuracy: 0.8805

Epoch 7/20

30/30 [=====] - 1s 20ms/step - loss: 0.0189 - accuracy: 0.9797 - val\_loss: 0.0961 - val\_accuracy: 0.8770

Epoch 8/20

30/30 [=====] - 1s 21ms/step - loss: 0.0153 - accuracy: 0.9839 - val\_loss: 0.1024 - val\_accuracy: 0.8750

Epoch 9/20

30/30 [=====] - 1s 22ms/step - loss: 0.0135 - accuracy: 0.9856 - val\_loss: 0.1032 - val\_accuracy: 0.8756

Epoch 10/20

30/30 [=====] - 1s 19ms/step - loss: 0.0115 - accuracy: 0.9881 - val\_loss: 0.1075 - val\_accuracy: 0.8722

Epoch 11/20

30/30 [=====] - 1s 19ms/step - loss: 0.0093 - accuracy: 0.9901 - val\_loss: 0.1077 - val\_accuracy: 0.8733

Epoch 12/20

30/30 [=====] - 1s 21ms/step - loss: 0.0091 - accuracy: 0.9901 - val\_loss: 0.1099 - val\_accuracy: 0.8712

Epoch 13/20

30/30 [=====] - 1s 20ms/step - loss: 0.0101 - accuracy: 0.9885 - val\_loss: 0.1115 - val\_accuracy: 0.8711

Epoch 14/20

30/30 [=====] - 1s 18ms/step - loss: 0.0054 - accuracy: 0.9947 - val\_loss: 0.1131 - val\_accuracy: 0.8700

```

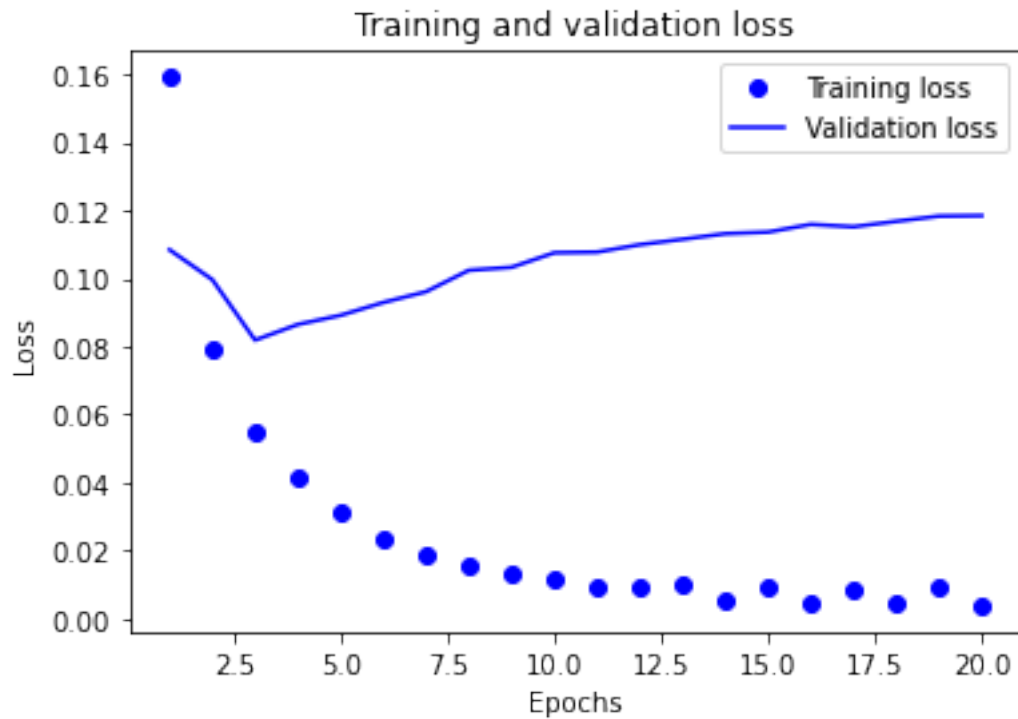
Epoch 15/20
30/30 [=====] - 1s 20ms/step - loss: 0.0091 - accuracy:
0.9892 - val_loss: 0.1136 - val_accuracy: 0.8695
Epoch 16/20
30/30 [=====] - 1s 18ms/step - loss: 0.0049 - accuracy:
0.9952 - val_loss: 0.1158 - val_accuracy: 0.8664
Epoch 17/20
30/30 [=====] - 1s 19ms/step - loss: 0.0088 - accuracy:
0.9901 - val_loss: 0.1151 - val_accuracy: 0.8688
Epoch 18/20
30/30 [=====] - 1s 18ms/step - loss: 0.0043 - accuracy:
0.9957 - val_loss: 0.1167 - val_accuracy: 0.8687
Epoch 19/20
30/30 [=====] - 1s 18ms/step - loss: 0.0091 - accuracy:
0.9897 - val_loss: 0.1183 - val_accuracy: 0.8665
Epoch 20/20
30/30 [=====] - 1s 18ms/step - loss: 0.0039 - accuracy:
0.9961 - val_loss: 0.1184 - val_accuracy: 0.8657

```

```

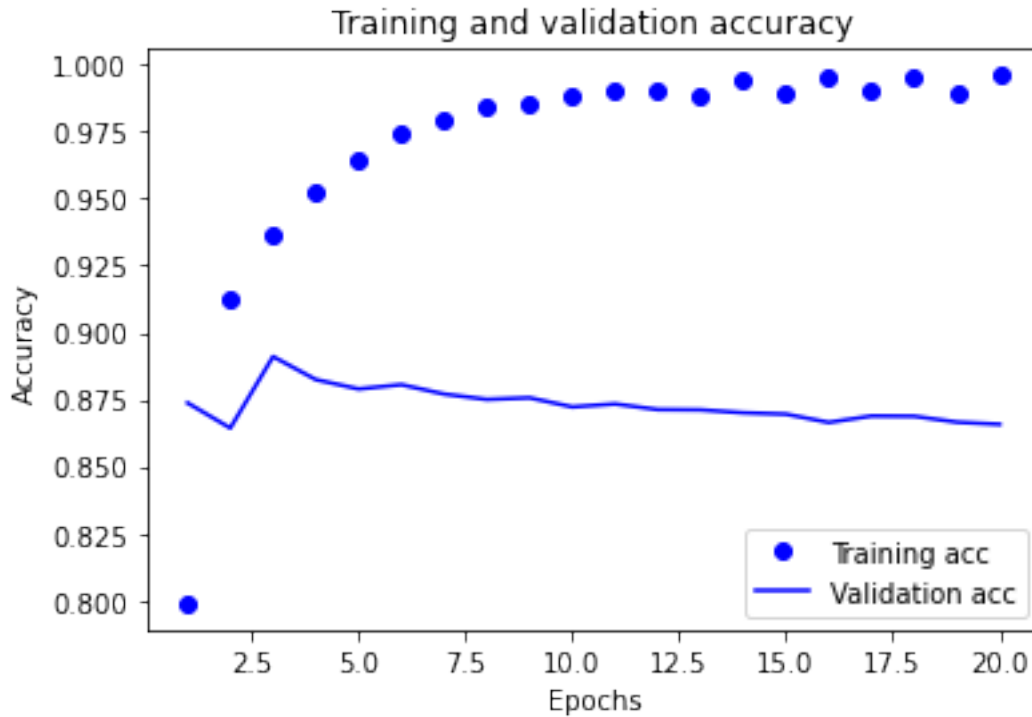
[182]: history_dict_5 = history_5.history
loss_values = history_dict_5["loss"]
val_loss_values = history_dict_5["val_loss"]
epochs = range(1, len(loss_values) + 1)
plt.plot(epochs, loss_values, "bo", label="Training loss")
plt.plot(epochs, val_loss_values, "b", label="Validation loss")
plt.title("Training and validation loss")
plt.xlabel("Epochs")
plt.ylabel("Loss")
plt.legend()
plt.show()

```



```
[183]: # Plotting the training and validation accuracy
```

```
[184]: plt.clf()
acc = history_dict_5["accuracy"]
val_acc = history_dict_5["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()
```



[185]: *# 5. Use any technique we studied in class, and these include regularization, ↵  
 ↳ dropout, etc., to get your model to perform better on validation.*

```
[186]: model_6 = keras.Sequential([
    #layers.Dropout(0.2),
    layers.Dense(20, activation="relu"),
    layers.Dropout(0.2),
    layers.Dense(15, activation="relu"),
    layers.Dense(1, activation="sigmoid")
])
```

[187]: *# Model completion*

```
[188]: model_6.compile(optimizer="rmsprop",
    loss="binary_crossentropy",
    metrics=["accuracy"])
```

```
[189]: history_6 = model_6.fit(partial_x_train,
    partial_y_train,
    epochs=20,
    batch_size=512,
    validation_data=(x_val, y_val))
```

Epoch 1/20

30/30 [=====] - 2s 34ms/step - loss: 0.5248 - accuracy: 0.7677 - val\_loss: 0.3721 - val\_accuracy: 0.8752  
Epoch 2/20  
30/30 [=====] - 1s 23ms/step - loss: 0.3218 - accuracy: 0.8903 - val\_loss: 0.3197 - val\_accuracy: 0.8767  
Epoch 3/20  
30/30 [=====] - 1s 31ms/step - loss: 0.2458 - accuracy: 0.9128 - val\_loss: 0.2827 - val\_accuracy: 0.8882  
Epoch 4/20  
30/30 [=====] - 1s 27ms/step - loss: 0.1965 - accuracy: 0.9333 - val\_loss: 0.3174 - val\_accuracy: 0.8735  
Epoch 5/20  
30/30 [=====] - 1s 23ms/step - loss: 0.1611 - accuracy: 0.9455 - val\_loss: 0.2805 - val\_accuracy: 0.8880  
Epoch 6/20  
30/30 [=====] - 1s 23ms/step - loss: 0.1294 - accuracy: 0.9569 - val\_loss: 0.2979 - val\_accuracy: 0.8861  
Epoch 7/20  
30/30 [=====] - 1s 27ms/step - loss: 0.1122 - accuracy: 0.9627 - val\_loss: 0.3078 - val\_accuracy: 0.8854  
Epoch 8/20  
30/30 [=====] - 1s 27ms/step - loss: 0.0870 - accuracy: 0.9725 - val\_loss: 0.3272 - val\_accuracy: 0.8843  
Epoch 9/20  
30/30 [=====] - 1s 28ms/step - loss: 0.0724 - accuracy: 0.9793 - val\_loss: 0.3511 - val\_accuracy: 0.8823  
Epoch 10/20  
30/30 [=====] - 1s 23ms/step - loss: 0.0579 - accuracy: 0.9821 - val\_loss: 0.3787 - val\_accuracy: 0.8785  
Epoch 11/20  
30/30 [=====] - 1s 21ms/step - loss: 0.0513 - accuracy: 0.9851 - val\_loss: 0.4036 - val\_accuracy: 0.8810  
Epoch 12/20  
30/30 [=====] - 1s 24ms/step - loss: 0.0380 - accuracy: 0.9903 - val\_loss: 0.4448 - val\_accuracy: 0.8745  
Epoch 13/20  
30/30 [=====] - 1s 24ms/step - loss: 0.0294 - accuracy: 0.9924 - val\_loss: 0.4983 - val\_accuracy: 0.8741  
Epoch 14/20  
30/30 [=====] - 1s 27ms/step - loss: 0.0232 - accuracy: 0.9955 - val\_loss: 0.5193 - val\_accuracy: 0.8722  
Epoch 15/20  
30/30 [=====] - 1s 27ms/step - loss: 0.0201 - accuracy: 0.9955 - val\_loss: 0.5475 - val\_accuracy: 0.8748  
Epoch 16/20  
30/30 [=====] - 1s 27ms/step - loss: 0.0135 - accuracy: 0.9976 - val\_loss: 0.6287 - val\_accuracy: 0.8613  
Epoch 17/20



```

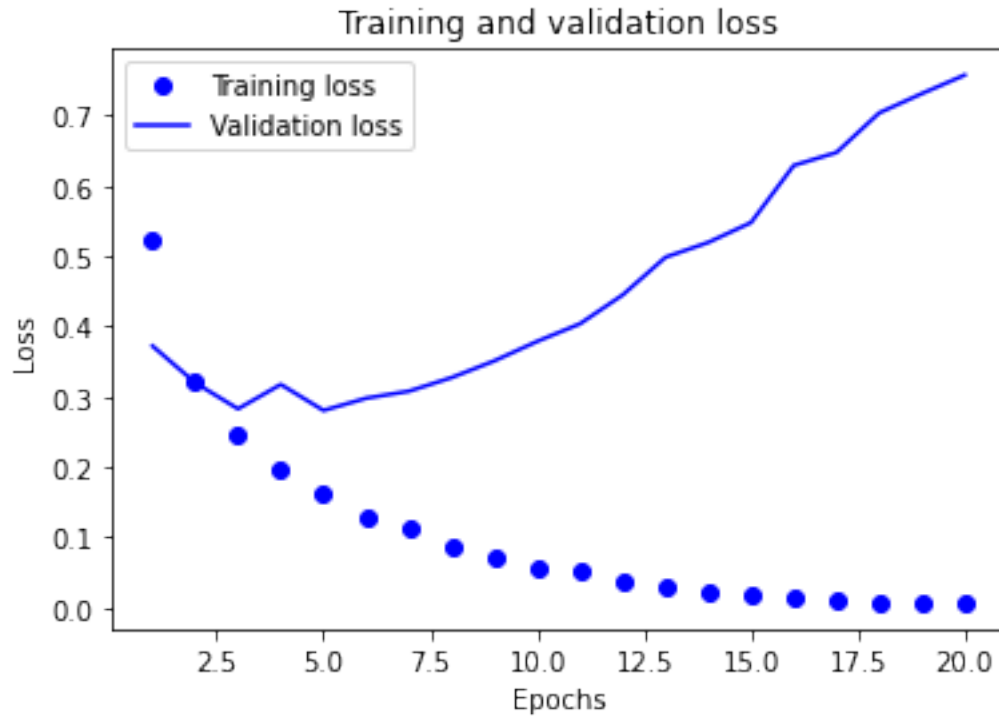
30/30 [=====] - 1s 25ms/step - loss: 0.0099 - accuracy:
0.9987 - val_loss: 0.6472 - val_accuracy: 0.8707
Epoch 18/20
30/30 [=====] - 1s 26ms/step - loss: 0.0084 - accuracy:
0.9985 - val_loss: 0.7035 - val_accuracy: 0.8725
Epoch 19/20
30/30 [=====] - 1s 20ms/step - loss: 0.0079 - accuracy:
0.9986 - val_loss: 0.7310 - val_accuracy: 0.8735
Epoch 20/20
30/30 [=====] - 1s 24ms/step - loss: 0.0072 - accuracy:
0.9985 - val_loss: 0.7573 - val_accuracy: 0.8720

```

```

[190]: history_dict_6 = history_6.history
loss_values = history_dict_6["loss"]
val_loss_values = history_dict_6["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()

```



```
[191]: # Plotting the training and validation accuracy
```

```
[192]: plt.clf()
acc = history_dict_6["accuracy"]
val_acc = history_dict_6["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()
```

