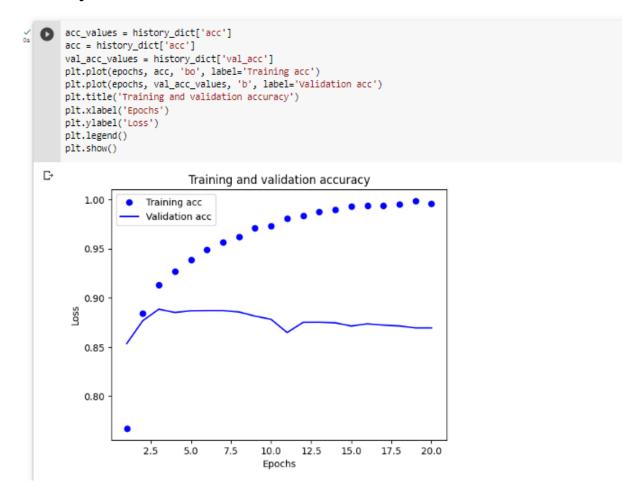
Sania Bibi

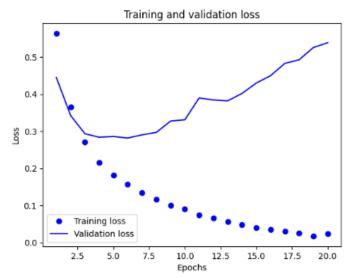
Task # 20

Binary classification



```
+ code + Text
```

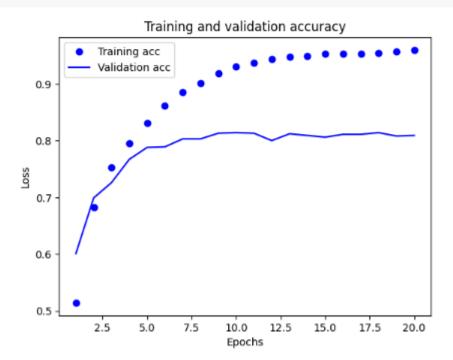
```
[21] val_loss_values = history_dict['val_loss']
    epochs = range(1, len(acc) + 1)
    plt.plot(epochs, loss_values, 'bo', label='Training loss')
    plt.plot(epochs, val_loss_values, 'b', label='Validation loss')
    plt.title('Training and validation loss')
    plt.xlabel('Epochs')
    plt.ylabel('Loss')
    plt.legend()
    plt.show()
```



Multiclass classification

```
+ Code + Text
\equiv
      loss = history.history['loss']
val_loss = history.history['val_loss']
epochs = range(1, len(loss) + 1)
Q
                  plt.plot(epochs, loss, 'bo', label='Training loss')
plt.plot(epochs, val_loss, 'b', label='Validation loss')
\{X\}
                  plt.title('Training and validation loss')
                 plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()
                  plt.show()
                                                      Training and validation loss
                                                                                                    Training loss
                                                                                                    Validation loss
                       2.0
                   Loss
                       1.0
                       0.5
<>
                                     2.5
                                                                      10.0
                                                                                 12.5
                                                                                            15.0
                                                                                                       17.5
                                                 5.0
                                                            7.5
                                                                      Epochs
\equiv
```

```
plt.clf()
acc = history.history['accuracy']
val_acc = history.history['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('Loss')
plt.legend()
plt.show()
```



Regression

```
uh·meau/[v[r] ioi v rii arr_mae_urscorres]) ioi r rii rauße/umm_ebocus/]
[18] import matplotlib.pyplot as plt plt.plot(range(1, len(average_mae_history) + 1), average_mae_history)
         plt.xlabel('Epochs')
         plt.ylabel('Validation MAE')
         plt.show()
              4.5
              4.0
          Validation MAE
              3.5
              3.0
              2.5
                     Ó
                                 100
                                               200
                                                             300
                                                                           400
                                                                                         500
                                                    Epochs
```

```
+ code + lex
```

```
def smooth_curve(points, factor=0.9):
    smoothed_points = []
    for point in points:
        if smoothed_points:
            previous = smoothed_points[-1]
            smoothed_points.append(previous * factor + point * (1 - factor))
        else:
            smoothed_points.append(point)
        return smoothed_points
smooth_mae_history = smooth_curve(average_mae_history[10:])
plt.plot(range(1, len(smooth_mae_history) + 1), smooth_mae_history)
plt.xlabel('Epochs')
plt.ylabel('Validation MAE')
plt.show()
```

