In [4]: |train_targets Out[4]: array([15.2, 42.3, 50. , 21.1, 17.7, 18.5, 11.3, 15.6, 15.6, 14.4, 12.1, 17.9, 23.1, 19.9, 15.7, 8.8, 50., 22.5, 24.1, 27.5, 10.9, 30.8, 32.9, 24., 18.5, 13.3, 22.9, 34.7, 16.6, 17.5, 22.3, 16.1, 14.9, 23.1, 34.9, 25. , 13.9, 13.1, 20.4, 20. , 15.2, 24.7, 22.2, 16.7, 12.7, 15.6, 18.4, 21., 30.1, 15.1, 18.7, 9.6, 31.5, 24.8, 19.1, 22. , 14.5, 11. , 32. , 29.4, 20.3, 24.4, 14.6, 19.5, 14.1, 14.3, 15.6, 10.5, 6.3, 19.3, 19.3, 13.4, 36.4, 17.8, 13.5, 16.5, 8.3, 14.3, 16., 13.4, 28.6, 43.5, 20.2, 22., 23., 20.7, 12.5, 48.5, 14.6, 13.4, 23.7, 50., 21.7, 39.8, 38.7, 22.2, 34.9, 22.5, 31.1, 28.7, 46., 41.7, 21., 26.6, 15., 24.4, 13.3, 21.2, 11.7, 21.7, 19.4, 50., 22.8, 19.7, 24.7, 36.2, 14.2, 18.9, 18.3, 20.6, 24.6, 18.2, 8.7, 44., 10.4, 13.2, 21.2, 37., 30.7, 22.9, 20., 19.3, 31.7, 32., 23.1, 18.8, 10.9, 50., 19.6, 5., 14.4, 19.8, 13.8, 19.6, 23.9, 24.5, 25., 19.9, 17.2, 24.6, 13.5, 26.6, 21.4, 11.9, 22.6, 19.6, 8.5, 23.7, 23.1, 22.4, 20.5, 23.6, 18.4, 35.2, 23.1, 27.9, 20.6, 23.7, 28., 13.6, 27.1, 23.6, 20.6, 18.2, 21.7, 17.1, 8.4, 25.3, 13.8, 22.2, 18.4, 20.7, 31.6, 30.5, 20.3, 8.8, 19.2, 19.4, 23.1, 23., 14.8, 48.8, 22.6, 33.4, 21.1, 13.6, 32.2, 13.1, 23.4, 18.9, 23.9, 11.8, 23.3, 22.8, 19.6, 16.7, 13.4, 22.2, 20.4, 21.8, 26.4, 14.9, 24.1, 23.8, 12.3, 29.1, 21. , 19.5, 23.3, 23.8, 17.8, 11.5, 21.7, 19.9, 25., 33.4, 28.5, 21.4, 24.3, 27.5, 33.1, 16.2, 23.3, 48.3, 22.9, 22.8, 13.1, 12.7, 22.6, 15. , 15.3, 10.5, 24. , 18.5, 21.7, 19.5, 33.2, 23.2, 5. , 19.1, 12.7, 22.3, 10.2, 13.9, 16.3, 17., 20.1, 29.9, 17.2, 37.3, 45.4, 17.8, 23.2, 29., 22. , 18. , 17.4, 34.6, 20.1, 25. , 15.6, 24.8, 28.2, 21.2, 21.4, 23.8, 31., 26.2, 17.4, 37.9, 17.5, 20., 8.3, 23.9, 8.4, 13.8,

11.8, 24.4, 13.8, 19.4, 25.2, 19.4, 19.4, 29.1])

7.2, 11.7, 17.1, 21.6, 50. , 16.1, 20.4, 20.6, 21.4, 20.6, 36.5, 8.5, 24.8, 10.8, 21.9, 17.3, 18.9, 36.2, 14.9, 18.2, 33.3, 21.8, 19.7, 31.6, 24.8, 19.4, 22.8, 7.5, 44.8, 16.8, 18.7, 50. , 50. , 19.5, 20.1, 50. , 17.2, 20.8, 19.3, 41.3, 20.4, 20.5, 13.8, 16.5, 23.9, 20.6, 31.5, 23.3, 16.8, 14. , 33.8, 36.1, 12.8, 18.3, 18.7, 19.1, 29. , 30.1, 50. , 50. , 22. , 11.9, 37.6, 50. , 22.7, 20.8, 23.5, 27.9, 50. , 19.3, 23.9, 22.6, 15.2, 21.7, 19.2, 43.8, 20.3, 33.2, 19.9, 22.5, 32.7, 22. , 17.1, 19. , 15. , 16.1, 25.1, 23.7, 28.7, 37.2, 22.6, 16.4, 25. , 29.8, 22.1, 17.4, 18.1, 30.3, 17.5, 24.7, 12.6, 26.5, 28.7, 13.3, 10.4, 24.4, 23. , 20. , 17.8, 7.

```
In [5]: mean = train_data.mean(axis=0)
        train data -= mean
        std = train_data.std(axis=0)
        train_data /= std
        test_data -= mean
        test_data /= std
In [6]: from keras import models
        from keras import layers
        def build model():
            model = models.Sequential()
            model.add(layers.Dense(64, activation='relu',
            input_shape=(train_data.shape[1],)))
            model.add(layers.Dense(64, activation='relu'))
            model.add(layers.Dense(1))
            model.compile(optimizer='rmsprop', loss='mse', metrics=['mae'])
            return model
In [7]: import numpy as np
        k=4
        num_val_samples = len(train_data) // k
        num epochs = 100
        all scores = []
```

```
In [8]: for i in range(k):
             print('processing fold #', i)
             val data = train data[i * num val samples: (i + 1) * num val samples]
             val targets = train targets[i * num val samples: (i + 1) * num val samples]
             partial train data = np.concatenate(
             [train data[:i * num val samples],
             train data[(i + 1) * num val samples:]],
             axis=0)
             partial train targets = np.concatenate(
             [train targets[:i * num val samples],
             train targets[(i + 1) * num val samples:]],
             axis=0)
             model = build model()
             model.fit(partial train data, partial train targets,
                 epochs=num epochs, batch size=1, verbose=0)
             val mse, val mae = model.evaluate(val data, val targets, verbose=0)
             all scores.append(val mae)
         processing fold # 0
         processing fold # 1
         processing fold # 2
         processing fold # 3
 In [9]: all scores
Out[9]: [2.0706627368927, 2.364206552505493, 2.573894739151001, 2.4287617206573486]
In [10]: np.mean(all scores)
Out[10]: 2.3593814373016357
In [11]: num epochs = 500
```

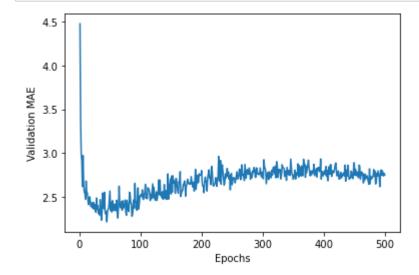
```
In [12]: all mae histories = []
         for i in range(k):
             print('processing fold #', i)
             val data = train data[i * num val samples: (i + 1) * num val samples]
             val targets = train targets[i * num val samples: (i + 1) * num val samples]
             partial train data = np.concatenate(
             [train data[:i * num val samples],
             train data[(i + 1) * num val samples:]],
             axis=0)
             partial train targets = np.concatenate(
                 [train targets[:i * num val samples],
                 train_targets[(i + 1) * num_val_samples:]],
                 axis=0)
             model = build model()
             history = model.fit(partial train data, partial train targets,
                                 validation data=(val data, val targets),
                                 epochs=num epochs, batch size=1, verbose=0)
             mae history = history.history['val mae']
             all mae histories.append(mae history)
         processing fold # 0
         processing fold # 1
         processing fold # 2
         processing fold # 3
```

np.mean([x[i] for x in all mae histories]) for i in range(num epochs)]

In [13]: average mae history = [

```
In [14]: 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()
```



```
In [15]:
    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()
```

