In [27]:

loading housing dataset

from keras.datasets import boston_housing
(train_data, train_targets), (test_data, test_targets)=
boston_housing.load_data()

In [28]:

exploratory

train data.shape

Out[28]:

(404, 13)

In [3]:

test_data.shape

Out[3]:

(102, 13)

In [33]:

need to normalize data

In [30]:

mean=train_data.mean(axis=0)
train data -= mean

In [31]:

std= train_data.std(axis=0)
train_data /= std

In [32]:

test_data -= mean
test_data /= std

In [34]:

building network: defining model

```
from keras import models
from keras import layers
def build model():
    model = models.Sequential()
    model.add(layers.Dense(64, activation='relu'))
    model.add(layers.Dense(1))
    model.compile(optimizer='rmsprop', loss='mse', metrics=['mae'])
    return model
In [36]:
# using K-fold to validate
import numpy as np
k=4
num val samples = len(train data) // k
num epochs= 100
all scores=[]
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 mae, 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 [37]:
# all validation scores
all scores
Out[37]:
[1.9309123754501343,
 2.3387398719787598,
 2.5432286262512207,
 2.34063196182250981
```

In [38]:

```
# mean of all scores
np.mean(all scores)
Out[38]:
2.288378208875656
In [40]:
history dict=history.history
history_dict.keys()
Out[40]:
dict keys(['val loss', 'val mae', 'loss', 'mae'])
In [41]:
# saving the validation logs at each fold
num epochs=500
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
In [42]:
# building history of successive mean k fold validation scores
average_mae_history= [np.mean([x[i] for x in all_mae_histories]) for i in
range(num epochs)]
In [60]:
```

```
# plotting validation scores

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()

14

12

10

6
```

In [67]:

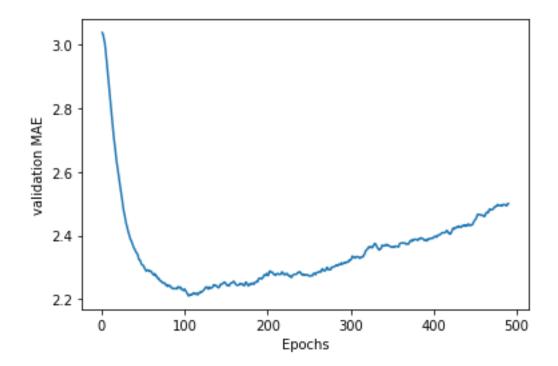
Plotting validation scores but removing first 10 points.

epochs

```
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()
```



In []:

training the final model

In [58]:

model = build model()

In [59]:

test_mae_score

Out[59]:

2.8621602058410645 # Mean absolute Error