In [12]:

```
from sklearn.datasets import fetch_california_housing
import pandas as pd

california_dataset = fetch_california_housing()

df = pd.DataFrame(california_dataset.data, columns=california_dataset.feature_names)

df['MEDV'] = california_dataset.target

df.head(n=10)
```

Out[12]:

	Medinc	HouseAge	AveRooms	AveBedrms	Population	AveOccup	Latitude	Longitude	MEDV
0	8.3252	41.0	6.984127	1.023810	322.0	2.55556	37.88	-122.23	4.526
1	8.3014	21.0	6.238137	0.971880	2401.0	2.109842	37.86	-122.22	3.585
2	7.2574	52.0	8.288136	1.073446	496.0	2.802260	37.85	-122.24	3.521
3	5.6431	52.0	5.817352	1.073059	558.0	2.547945	37.85	-122.25	3.413
4	3.8462	52.0	6.281853	1.081081	565.0	2.181467	37.85	-122.25	3.422
5	4.0368	52.0	4.761658	1.103627	413.0	2.139896	37.85	-122.25	2.697
6	3.6591	52.0	4.931907	0.951362	1094.0	2.128405	37.84	-122.25	2.992
7	3.1200	52.0	4.797527	1.061824	1157.0	1.788253	37.84	-122.25	2.414
8	2.0804	42.0	4.294118	1.117647	1206.0	2.026891	37.84	-122.26	2.267
9	3.6912	52.0	4.970588	0.990196	1551.0	2.172269	37.84	-122.25	2.611

In [13]:

```
from sklearn.model_selection import train_test_split

X = df.loc[:, df.columns != 'MEDV']
y = df.loc[:, df.columns == 'MEDV']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=12
3)
```

In [16]:

```
from keras.models import Sequential
from keras.layers import Dense

model = Sequential()

model.add(Dense(128, input_shape=(8, ), activation='relu', name='dense_1'))
model.add(Dense(64, activation='relu', name='dense_2'))
model.add(Dense(1, activation='linear', name='dense_output'))

model.compile(optimizer='adam', loss='mse', metrics=['mae'])
model.summary()
```

Model: "sequential_2"

Layer (type)	Output Shape	Param #	
dense_1 (Dense)	(None, 128)	1152	
dense_2 (Dense)	(None, 64)	8256	
dense_output (Dense)	(None, 1)	65	

Total params: 9,473 Trainable params: 9,473 Non-trainable params: 0

In [17]:

```
history = model.fit(X train, y train, epochs=200, validation split=0.05, verbose = 1)
Epoch 1/200
al loss: 3.6336 - val mae: 1.0747
Epoch 2/200
loss: 2.1447 - val mae: 1.1562
Epoch 3/200
l loss: 5.5047 - val mae: 1.4988
Epoch 4/200
l loss: 272.4287 - val mae: 10.0643
Epoch 5/200
al loss: 1.0633 - val mae: 0.7675
Epoch 6/200
loss: 1.1112 - val mae: 0.7506
Epoch 7/200
loss: 0.8950 - val mae: 0.7865
Epoch 8/200
loss: 0.7706 - val mae: 0.6281
Epoch 9/200
loss: 1.9868 - val mae: 0.9981
Epoch 10/200
loss: 1.2546 - val mae: 0.9147
Epoch 11/200
loss: 2.1335 - val mae: 1.0505
Epoch 12/200
l loss: 60.2671 - val mae: 5.7113
Epoch 13/200
loss: 1.1230 - val mae: 0.7010
Epoch 14/200
loss: 0.8549 - val mae: 0.6264
Epoch 15/200
loss: 1.0464 - val mae: 0.6562
Epoch 16/200
loss: 1.2330 - val mae: 0.8025
Epoch 17/200
loss: 1.8198 - val mae: 0.9859
Epoch 18/200
loss: 0.9826 - val mae: 0.7114
Epoch 19/200
loss: 0.6033 - val mae: 0.5819
Epoch 20/200
loss: 1.1465 - val mae: 0.8297
Epoch 21/200
loss: 0.7251 - val mae: 0.6996
Epoch 22/200
```

```
loss: 0.9730 - val mae: 0.7315
Epoch 23/200
loss: 64.6038 - val mae: 6.0048
Epoch 24/200
loss: 1.1295 - val mae: 0.6116
Epoch 25/200
loss: 5.5252 - val mae: 1.5308
Epoch 26/200
loss: 1.0677 - val mae: 0.6192
Epoch 27/200
loss: 4.3800 - val mae: 1.5434
Epoch 28/200
loss: 0.6606 - val mae: 0.5842
Epoch 29/200
loss: 1.3038 - val mae: 0.8228
Epoch 30/200
loss: 1.1061 - val mae: 0.7517
Epoch 31/200
loss: 2.5993 - val mae: 1.1612
Epoch 32/200
loss: 1.6532 - val mae: 0.9182
Epoch 33/200
loss: 0.9641 - val mae: 0.6833
Epoch 34/200
loss: 1.5209 - val mae: 0.8640
Epoch 35/200
loss: 0.7368 - val mae: 0.6081
Epoch 36/200
loss: 0.8186 - val mae: 0.5578
Epoch 37/200
loss: 1.0016 - val mae: 0.6852
Epoch 38/200
loss: 2.5829 - val mae: 1.3008
Epoch 39/200
loss: 0.8783 - val mae: 0.5828
Epoch 40/200
loss: 0.8254 - val mae: 0.5824
Epoch 41/200
loss: 0.6828 - val mae: 0.5636
Epoch 42/200
loss: 1.3696 - val mae: 0.6318
Epoch 43/200
loss: 0.8475 - val mae: 0.7552
Epoch 44/200
loss: 1.3371 - val mae: 0.6481
Epoch 45/200
loss: 1.1540 - val mae: 0.5374
Epoch 46/200
```

```
loss: 0.9004 - val mae: 0.6784
Epoch 47/200
loss: 0.7388 - val mae: 0.6529
Epoch 48/200
loss: 0.7166 - val mae: 0.5741
Epoch 49/200
loss: 0.6016 - val mae: 0.5350
Epoch 50/200
loss: 1.1316 - val mae: 0.5446
Epoch 51/200
loss: 0.7251 - val mae: 0.6085
Epoch 52/200
loss: 1.0409 - val mae: 0.6128
Epoch 53/200
loss: 0.7529 - val mae: 0.6181
Epoch 54/200
loss: 1.4176 - val mae: 0.7583
Epoch 55/200
loss: 0.5892 - val mae: 0.5430
Epoch 56/200
loss: 0.6795 - val mae: 0.5586
Epoch 57/200
loss: 0.7089 - val mae: 0.5783
Epoch 58/200
loss: 0.6386 - val mae: 0.5165
Epoch 59/200
loss: 0.5429 - val mae: 0.5822
Epoch 60/200
loss: 0.5690 - val mae: 0.5634
Epoch 61/200
loss: 1.1495 - val mae: 0.5725
Epoch 62/200
loss: 0.9347 - val mae: 0.5776
Epoch 63/200
loss: 0.6300 - val mae: 0.5075
Epoch 64/200
loss: 0.6742 - val mae: 0.5108
Epoch 65/200
loss: 0.5569 - val mae: 0.5233
Epoch 66/200
loss: 0.7217 - val mae: 0.5194
Epoch 67/200
loss: 0.8204 - val mae: 0.6789
Epoch 68/200
loss: 0.9442 - val mae: 0.5347
Epoch 69/200
loss: 0.6028 - val mae: 0.5271
Epoch 70/200
```

```
loss: 0.6872 - val mae: 0.5398
Epoch 71/200
loss: 0.5774 - val mae: 0.5118
Epoch 72/200
loss: 0.5529 - val mae: 0.4969
Epoch 73/200
loss: 0.6069 - val mae: 0.5348
Epoch 74/200
loss: 0.7981 - val mae: 0.5367
Epoch 75/200
loss: 0.5283 - val mae: 0.5017
Epoch 76/200
loss: 0.5632 - val mae: 0.5310
Epoch 77/200
loss: 0.5870 - val mae: 0.4977
Epoch 78/200
loss: 0.6282 - val mae: 0.5119
Epoch 79/200
loss: 0.6481 - val mae: 0.5020
Epoch 80/200
loss: 0.5961 - val mae: 0.5201
Epoch 81/200
loss: 0.5197 - val mae: 0.5326
Epoch 82/200
loss: 0.6088 - val mae: 0.5023
Epoch 83/200
loss: 0.7235 - val mae: 0.5125
Epoch 84/200
loss: 0.6529 - val mae: 0.5115
Epoch 85/200
loss: 0.6999 - val mae: 0.6015
Epoch 86/200
loss: 0.5940 - val mae: 0.5047
Epoch 87/200
loss: 0.5760 - val mae: 0.5182
Epoch 88/200
loss: 0.5776 - val mae: 0.5006
Epoch 89/200
loss: 0.5701 - val mae: 0.4956
Epoch 90/200
loss: 0.6034 - val mae: 0.4917
Epoch 91/200
loss: 0.6698 - val mae: 0.5076
Epoch 92/200
loss: 0.5545 - val mae: 0.5464
Epoch 93/200
loss: 0.5625 - val mae: 0.5092
Epoch 94/200
```

```
loss: 0.5987 - val mae: 0.5104
Epoch 95/200
loss: 0.5772 - val mae: 0.5117
Epoch 96/200
loss: 0.6347 - val mae: 0.5831
Epoch 97/200
loss: 0.5627 - val mae: 0.4984
Epoch 98/200
loss: 0.6080 - val mae: 0.5309
Epoch 99/200
loss: 0.5727 - val mae: 0.5252
Epoch 100/200
loss: 0.5122 - val mae: 0.4914
Epoch 101/200
loss: 0.4818 - val mae: 0.4876
Epoch 102/200
loss: 0.5751 - val mae: 0.5089
Epoch 103/200
loss: 0.5145 - val mae: 0.4894
Epoch 104/200
loss: 0.5177 - val mae: 0.5497
Epoch 105/200
loss: 0.5110 - val mae: 0.5028
Epoch 106/200
loss: 0.5998 - val mae: 0.5352
Epoch 107/200
loss: 0.4690 - val mae: 0.5009
Epoch 108/200
loss: 0.4883 - val mae: 0.4966
Epoch 109/200
loss: 0.4850 - val mae: 0.4852
Epoch 110/200
loss: 0.5226 - val mae: 0.5069
Epoch 111/200
loss: 0.5317 - val mae: 0.5223
Epoch 112/200
loss: 0.4696 - val mae: 0.4864
Epoch 113/200
loss: 0.4428 - val mae: 0.4875
Epoch 114/200
_loss: 0.5001 - val_mae: 0.4989
Epoch 115/200
loss: 0.5296 - val mae: 0.5085
Epoch 116/200
loss: 0.5251 - val mae: 0.5437
Epoch 117/200
loss: 0.5100 - val mae: 0.5071
Epoch 118/200
```

```
loss: 0.4888 - val mae: 0.4859
Epoch 119/200
loss: 0.4461 - val mae: 0.4995
Epoch 120/200
loss: 0.4407 - val mae: 0.4935
Epoch 121/200
loss: 0.4782 - val mae: 0.5171
Epoch 122/200
loss: 0.4941 - val mae: 0.4939
Epoch 123/200
loss: 0.4820 - val mae: 0.4840
Epoch 124/200
loss: 0.5170 - val mae: 0.5504
Epoch 125/200
loss: 0.5128 - val mae: 0.5095
Epoch 126/200
loss: 0.4426 - val mae: 0.4717
Epoch 127/200
loss: 0.4471 - val mae: 0.4790
Epoch 128/200
loss: 0.5179 - val mae: 0.5175
Epoch 129/200
loss: 0.4705 - val mae: 0.4930
Epoch 130/200
loss: 0.5333 - val mae: 0.5134
Epoch 131/200
loss: 0.4668 - val mae: 0.4743
Epoch 132/200
loss: 0.4846 - val mae: 0.5301
Epoch 133/200
loss: 0.4450 - val mae: 0.4858
Epoch 134/200
loss: 0.4834 - val mae: 0.4879
Epoch 135/200
loss: 0.4714 - val mae: 0.4916
Epoch 136/200
loss: 0.4942 - val mae: 0.5043
Epoch 137/200
loss: 0.5654 - val mae: 0.5833
Epoch 138/200
loss: 0.4257 - val mae: 0.4817
Epoch 139/200
loss: 0.4943 - val mae: 0.4950
Epoch 140/200
loss: 0.4349 - val mae: 0.4781
Epoch 141/200
loss: 0.5183 - val mae: 0.5054
Epoch 142/200
```

```
loss: 0.4797 - val mae: 0.4857
Epoch 143/200
loss: 0.5472 - val mae: 0.5813
Epoch 144/200
loss: 0.4496 - val mae: 0.5084
Epoch 145/200
loss: 0.4634 - val mae: 0.4895
Epoch 146/200
loss: 0.4406 - val mae: 0.4820
Epoch 147/200
loss: 0.4364 - val mae: 0.4948
Epoch 148/200
loss: 0.4475 - val mae: 0.4913
Epoch 149/200
loss: 0.4820 - val mae: 0.4886
Epoch 150/200
loss: 0.5600 - val mae: 0.5205
Epoch 151/200
loss: 0.4479 - val mae: 0.4990
Epoch 152/200
loss: 0.4795 - val mae: 0.5069
Epoch 153/200
loss: 0.4755 - val mae: 0.4820
Epoch 154/200
loss: 0.4231 - val mae: 0.4786
Epoch 155/200
loss: 0.4497 - val mae: 0.4859
Epoch 156/200
loss: 0.5193 - val mae: 0.4873
Epoch 157/200
loss: 0.5529 - val mae: 0.5029
Epoch 158/200
loss: 0.4932 - val mae: 0.4988
Epoch 159/200
loss: 0.5491 - val mae: 0.5951
Epoch 160/200
loss: 0.4851 - val mae: 0.4788
Epoch 161/200
loss: 0.4349 - val mae: 0.4738
Epoch 162/200
loss: 0.4665 - val mae: 0.4787
Epoch 163/200
loss: 0.5585 - val mae: 0.5911
Epoch 164/200
loss: 0.4405 - val mae: 0.5014
Epoch 165/200
loss: 0.4315 - val mae: 0.4763
Epoch 166/200
```

```
loss: 0.5141 - val mae: 0.5025
Epoch 167/200
loss: 0.4409 - val mae: 0.4958
Epoch 168/200
_loss: 0.4711 - val_mae: 0.4814
Epoch 169/200
loss: 0.4455 - val mae: 0.5061
Epoch 170/200
loss: 0.4797 - val mae: 0.4964
Epoch 171/200
loss: 0.4408 - val mae: 0.4790
Epoch 172/200
loss: 0.4410 - val mae: 0.4775
Epoch 173/200
loss: 0.4422 - val mae: 0.4975
Epoch 174/200
loss: 0.4422 - val mae: 0.4673
Epoch 175/200
loss: 0.5323 - val mae: 0.5796
Epoch 176/200
loss: 0.4342 - val mae: 0.4654
Epoch 177/200
loss: 0.4298 - val mae: 0.4735
Epoch 178/200
loss: 0.4330 - val mae: 0.4895
Epoch 179/200
loss: 0.4210 - val mae: 0.4691
Epoch 180/200
loss: 0.5119 - val mae: 0.5350
Epoch 181/200
loss: 0.4809 - val mae: 0.4832
Epoch 182/200
loss: 0.4566 - val mae: 0.4754
Epoch 183/200
loss: 0.4307 - val mae: 0.4670
Epoch 184/200
loss: 0.4376 - val mae: 0.4981
Epoch 185/200
loss: 0.4654 - val mae: 0.4693
Epoch 186/200
loss: 0.5317 - val mae: 0.4654
Epoch 187/200
loss: 0.4786 - val mae: 0.5117
Epoch 188/200
loss: 0.6209 - val mae: 0.4692
Epoch 189/200
loss: 0.5805 - val mae: 0.5130
Epoch 190/200
```

```
loss: 0.5307 - val mae: 0.4697
Epoch 191/200
loss: 0.6056 - val mae: 0.5026
Epoch 192/200
loss: 0.5964 - val mae: 0.4815
Epoch 193/200
loss: 0.5342 - val mae: 0.4750
Epoch 194/200
loss: 0.5950 - val mae: 0.4731
Epoch 195/200
loss: 0.7073 - val mae: 0.5552
Epoch 196/200
loss: 0.6635 - val mae: 0.4834
Epoch 197/200
loss: 0.5831 - val mae: 0.4846
Epoch 198/200
loss: 0.7164 - val mae: 0.5204
Epoch 199/200
loss: 0.6445 - val mae: 0.5513
Epoch 200/200
loss: 0.6271 - val mae: 0.4747
In [18]:
mse nn, mae nn = model.evaluate(X test, y test)
print('Mean squared error on test data is: ', mse nn)
print('Mean absolute error on test data is: ', mae nn)
Mean squared error on test data is: 0.4117625951766968
Mean absolute error on test data is: 0.4550313353538513
```