dl1

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Boston House Price Prediction Name : Nidhi Kamath

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[26]: import numpy as np
      import tensorflow as tf
      from tensorflow import keras
      from tensorflow.keras import Sequential
      from tensorflow.keras.layers import Dense
      from sklearn import preprocessing
[38]: (X_train, Y_train), (X_test, Y_test) = keras.datasets.boston_housing.load_data()
[28]: print("Training data shape:", X_train.shape)
      print("Test data shape:", X_test.shape)
      print("Train output data shape:", Y_train.shape)
      print("Actual Test output data shape:", Y_test.shape)
     Training data shape: (404, 13)
     Test data shape: (102, 13)
     Train output data shape: (404,)
     Actual Test output data shape: (102,)
[29]: X_train = preprocessing.normalize(X_train)
      X_test = preprocessing.normalize(X_test)
[30]: model = Sequential([
          Dense(128, activation='relu', input_shape=(X_train.shape[1],)),
          Dense(64, activation='relu'),
          Dense(32, activation='relu'),
          Dense(1)
      ])
     /home/nidhi/.local/lib/python3.10/site-
     packages/keras/src/layers/core/dense.py:86: UserWarning: Do not pass an
     `input_shape`/`input_dim` argument to a layer. When using Sequential models,
     prefer using an `Input(shape)` object as the first layer in the model instead.
       super().__init__(activity_regularizer=activity_regularizer, **kwargs)
[31]: model.summary()
```

Model: "sequential_11"

[32]:

[33]:

Layer (type) Output Shape Param # dense_44 (Dense) (None, 128) 1,792 dense_45 (Dense) (None, 64) 8,256 dense_46 (Dense) (None, 32) 2,080 dense_47 (Dense) (None, 1) 33 Total params: 12,161 (47.50 KB)	
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Total params: 12,161 (47.50 KB)	
Trainable params: 12,161 (47.50 KB)	
Non-trainable params: 0 (0.00 B)	
<pre>model.compile(loss='mse', optimizer='rmsprop', metrics=['mae'])</pre>	
history = model.fit(X_train, Y_train, epochs=20, batch_size=1, verbose=1, open validation_data=(X_test, Y_test))	
Epoch 1/20	
1s 2ms/step -	
Loss: 257.4678 - mae: 12.2965 - val_loss: 79.8688 - val_mae: 6.3300 Epoch 2/20	
104/404 1s 1ms/step -	
Loss: 81.2376 - mae: 6.1475 - val_loss: 60.4605 - val_mae: 5.6386 Epoch 3/20	
104/404 1s 1ms/step -	
•	
104/404	
1s 1ms/step - Loss: 58.2065 - mae: 5.2785 - val_loss: 56.4742 - val_mae: 5.5420 Epoch 4/20	
1s 1ms/step - Loss: 58.2065 - mae: 5.2785 - val_loss: 56.4742 - val_mae: 5.5420 Epoch 4/20 1s 1ms/step - Loss: 63.3948 - mae: 5.5304 - val_loss: 55.2899 - val_mae: 5.4546	
1s 1ms/step - Loss: 58.2065 - mae: 5.2785 - val_loss: 56.4742 - val_mae: 5.5420 Epoch 4/20 1s 1ms/step - Loss: 63.3948 - mae: 5.5304 - val_loss: 55.2899 - val_mae: 5.4546 Epoch 5/20	
1s 1ms/step - loss: 58.2065 - mae: 5.2785 - val_loss: 56.4742 - val_mae: 5.5420 Epoch 4/20 104/404	
1s 1ms/step - Loss: 58.2065 - mae: 5.2785 - val_loss: 56.4742 - val_mae: 5.5420 Epoch 4/20 1s 1ms/step - Loss: 63.3948 - mae: 5.5304 - val_loss: 55.2899 - val_mae: 5.4546 Epoch 5/20 104/404	

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loss: 64.0252 - mae: 5.4795 - val loss: 61.3780 - val mae: 5.5002
     Epoch 8/20
     404/404
                         1s 1ms/step -
     loss: 48.8235 - mae: 4.5386 - val_loss: 52.6091 - val_mae: 5.1388
     Epoch 9/20
     404/404
                         1s 1ms/step -
     loss: 48.1088 - mae: 4.7712 - val loss: 59.7066 - val mae: 5.3833
     Epoch 10/20
     404/404
                         1s 2ms/step -
     loss: 56.3099 - mae: 5.0725 - val_loss: 48.1332 - val_mae: 4.9502
     Epoch 11/20
     404/404
                         1s 2ms/step -
     loss: 47.2650 - mae: 4.6020 - val_loss: 47.3174 - val_mae: 4.8223
     Epoch 12/20
     404/404
                         1s 2ms/step -
     loss: 36.3078 - mae: 4.0888 - val_loss: 77.9395 - val_mae: 7.7030
     Epoch 13/20
     404/404
                         1s 2ms/step -
     loss: 53.5331 - mae: 5.2450 - val_loss: 44.5192 - val_mae: 4.6327
     Epoch 14/20
     404/404
                         1s 2ms/step -
     loss: 50.3486 - mae: 4.8726 - val_loss: 43.0127 - val_mae: 4.5491
     Epoch 15/20
     404/404
                         1s 2ms/step -
     loss: 52.6901 - mae: 5.0294 - val_loss: 50.7954 - val_mae: 4.9646
     Epoch 16/20
     404/404
                         1s 2ms/step -
     loss: 38.6620 - mae: 4.1478 - val_loss: 45.2976 - val_mae: 4.6517
     Epoch 17/20
     404/404
                         1s 2ms/step -
     loss: 45.5645 - mae: 4.5826 - val_loss: 37.2901 - val_mae: 4.2180
     Epoch 18/20
     404/404
                         1s 2ms/step -
     loss: 32.5251 - mae: 4.0921 - val_loss: 37.0078 - val_mae: 4.1849
     Epoch 19/20
     404/404
                         1s 2ms/step -
     loss: 37.1011 - mae: 4.0821 - val loss: 56.6494 - val mae: 6.6335
     Epoch 20/20
     404/404
                         1s 2ms/step -
     loss: 30.4977 - mae: 3.9257 - val_loss: 43.6990 - val_mae: 5.5365
[35]: test_input1 = np.array([(0.0024119, 0.0, 0.01592969, 0.0, 0.00105285, 0.
       -01201967, 0.17945359, 0.00778265, 0.00782786, 0.6007879, 0.04109624, 0.
       →77671895, 0.03663436)])
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```
test_input2 = np.array([(4.07923050e-05, 1.54587284e-01, 3.80378407e-03, 0.0, 7.
       →77620881e-04, 1.42595058e-02, 2.94184285e-02, 1.17486336e-02, 3.
       →74757051e-03, 6.52077269e-01, 2.75446433e-02, 7.40857215e-01, 5.
       →82747215e-03)])
[36]: print("Actual output: 15.2")
      print("Predicted Output:", model.predict(test_input1))
      print("Actual output: 42.3")
     print("Predicted Output:", model.predict(test_input2))
     Actual output: 15.2
                    Os 63ms/step
     Predicted Output: [[22.11968]]
     Actual output: 42.3
     1/1
                    Os 19ms/step
     Predicted Output: [[35.09147]]
 []:
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