

# Change Log

## Overview

- This model aims to be able to predict the quality of wine based off of different factors about the wine.
- I obtained my dataset from the University of California Irvine's website (<https://archive.ics.uci.edu/dataset/186/wine+quality>)
- There are 11 columns about the wine and it's makeup and then a quality rating at the end for my target value.
- The red wine dataset has 1600 entries and the white wine dataset has 4900 entries.
- This is a relatively small dataset but I wanted to use a simple enough dataset and prediction whilst I get used to using Neural Networks and figuring out how they work.
- I am planning to use a Feedforward Neural Network for this as it is a simple regression task and it should be a good fit for the data that I have.

## Learnings & Findings

- I was reading the tensorflow docs here (<https://www.tensorflow.org/tutorials/quickstart/beginner>) and a GeeksForGeeks article here (<https://www.geeksforgeeks.org/feedforward-neural-network/>) to get an idea of how to structure the code for this model.
- I have always used Sci-Kit learn for all of my models previously, however I was doing some research and found that Tensorflow seems to be a better fit for Neural Networks as it gives you a lot more control over the model in comparison to Sci-Kit Learns version.
- I would like to see if there is a difference in what makes a white wine good quality in comparison to red wine, this is something I'd like to investigate further on in the notebook, I want to run this model on Red, White and a combination of the two to see the results.
- Learned about adam optimiser from here (<https://www.geeksforgeeks.org/adam-optimizer-in-tensorflow/>)
- I have a MAE rating of 0.57 now so I want to see if I can improve this by using Standard Scaling
- I then moved on to replicating the same process with the red wine dataset, with even better success at a 0.51 MAE score.
- I then wanted to see what the score would be like for the combination of both and this was also a success.

## Results

- I now have a good and much clearer understanding of how even a simple enough Neural Network works and how it is structured.
- These datasets worked well in making predictions in the end with them all only being about .5 off in mst scenarios on the quality score.
- I can now take these learnings and take on a more complex model with more obscure data to challenge myself in using Neural Networks.
- Red wine and Combined ended up being a little bit better on the predictions but not by a massive amount.

```
import tensorflow as tf
from tensorflow import keras
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split
import pandas as pd

white_df = pd.read_csv('winequality-white.csv', sep=';')
red_df = pd.read_csv('winequality-red.csv', sep=';')

print(white_df.head())
print(red_df.head())
```

	fixed acidity	volatile acidity	citric acid	residual sugar
0	7.0	0.27	0.36	20.7
1	6.3	0.30	0.34	1.6
2	8.1	0.28	0.40	6.9
3	7.2	0.23	0.32	8.5
4	7.2	0.23	0.32	8.5

	free sulfur dioxide	total sulfur dioxide	density	pH	sulphates
0	45.0	170.0	1.0010	3.00	0.45
1	14.0	132.0	0.9940	3.30	0.49
2	30.0	97.0	0.9951	3.26	0.44
3	47.0	186.0	0.9956	3.19	0.40
4	47.0	186.0	0.9956	3.19	0.40

	alcohol	quality
0	8.8	6
1	9.5	6
2	10.1	6
3	9.9	6
4	9.9	6

	fixed acidity	volatile acidity	citric acid	residual sugar
0	7.4	0.70	0.00	1.9
1	7.8	0.88	0.00	2.6

2	7.8	0.76	0.04	2.3
0.092				
3	11.2	0.28	0.56	1.9
0.075				
4	7.4	0.70	0.00	1.9
0.076				
	free sulfur dioxide	total sulfur dioxide	density	pH sulphates
\				
0	11.0	34.0	0.9978	3.51 0.56
1	25.0	67.0	0.9968	3.20 0.68
2	15.0	54.0	0.9970	3.26 0.65
3	17.0	60.0	0.9980	3.16 0.58
4	11.0	34.0	0.9978	3.51 0.56
	alcohol	quality		
0	9.4	5		
1	9.8	5		
2	9.8	5		
3	9.8	6		
4	9.4	5		

I am just importing my two datasets here using pandas and importing the necessary libraries for later on.

```
X = white_df.drop('quality', axis=1)
y = white_df['quality']

X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.2, random_state=42)
```

I'm just starting off with white at random to see how it goes, I will eventually be doing this model on all of the datasets

```
model = keras.Sequential([
    keras.layers.Input(shape=(X_train.shape[1],)),
    keras.layers.Dense(64, activation='relu'),
    keras.layers.Dense(32, activation='relu'),
    keras.layers.Dense(1, activation='linear')
])
```

Here I am defining the model using Keras and Tensorflow.

Using Sequential here tells it that we are using a Feed Forward Neural Network. Firstly I added my input layer which just contains my training X data. Then I am creating 2 hidden layers on top

of Tensorflows already built model with 64 neurons and then the next one is the same but with 32 neurons. The relu activation function keeps the positive values but turns the negative ones into 0. The final layer I added then is the output layer which just has one neuron and I used linear because this is a regression model rather than classification.

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

In this code block I am compiling the model. An optimiser controls how the model updates the weights for the Neural Network. The adam optimiser is usually used because it adjusts the learning rates automatically.

I am then using Mean Square Error and Mean Absolute error as these are very good for regression as they track how much error is occurring in the predictions that the model is making.

```
model.fit(X_train, y_train, epochs=100, batch_size=16,
validation_split=0.2)
```

```
Epoch 1/100
196/196 _____ 1s 951us/step - loss: 18.0223 - mae:
2.8530 - val_loss: 1.0642 - val_mae: 0.8004
Epoch 2/100
196/196 _____ 0s 671us/step - loss: 0.7668 - mae:
0.6834 - val_loss: 0.6459 - val_mae: 0.6253
Epoch 3/100
196/196 _____ 0s 665us/step - loss: 0.7435 - mae:
0.6764 - val_loss: 0.6571 - val_mae: 0.6319
Epoch 4/100
196/196 _____ 0s 665us/step - loss: 0.6736 - mae:
0.6427 - val_loss: 0.6431 - val_mae: 0.6277
Epoch 5/100
196/196 _____ 0s 667us/step - loss: 0.6657 - mae:
0.6253 - val_loss: 0.6236 - val_mae: 0.6170
Epoch 6/100
196/196 _____ 0s 670us/step - loss: 0.7041 - mae:
0.6524 - val_loss: 0.6539 - val_mae: 0.6369
Epoch 7/100
196/196 _____ 0s 668us/step - loss: 0.6596 - mae:
0.6354 - val_loss: 0.8880 - val_mae: 0.7370
Epoch 8/100
196/196 _____ 0s 671us/step - loss: 0.7306 - mae:
0.6730 - val_loss: 0.7512 - val_mae: 0.6706
Epoch 9/100
196/196 _____ 0s 675us/step - loss: 0.8951 - mae:
0.7438 - val_loss: 0.8290 - val_mae: 0.7329
Epoch 10/100
196/196 _____ 0s 662us/step - loss: 0.7282 - mae:
0.6680 - val_loss: 0.8485 - val_mae: 0.7414
Epoch 11/100
196/196 _____ 0s 652us/step - loss: 0.8454 - mae:
0.7257 - val_loss: 0.6002 - val_mae: 0.6045
```

Epoch 12/100  
196/196 \_\_\_\_\_ 0s 672us/step - loss: 0.6253 - mae:  
0.6161 - val\_loss: 1.1739 - val\_mae: 0.8697  
Epoch 13/100  
196/196 \_\_\_\_\_ 0s 678us/step - loss: 0.7739 - mae:  
0.6910 - val\_loss: 0.6033 - val\_mae: 0.6055  
Epoch 14/100  
196/196 \_\_\_\_\_ 0s 665us/step - loss: 0.6306 - mae:  
0.6242 - val\_loss: 0.5875 - val\_mae: 0.5991  
Epoch 15/100  
196/196 \_\_\_\_\_ 0s 665us/step - loss: 0.6528 - mae:  
0.6401 - val\_loss: 1.0507 - val\_mae: 0.8101  
Epoch 16/100  
196/196 \_\_\_\_\_ 0s 687us/step - loss: 0.8030 - mae:  
0.7125 - val\_loss: 0.6653 - val\_mae: 0.6288  
Epoch 17/100  
196/196 \_\_\_\_\_ 0s 675us/step - loss: 0.7328 - mae:  
0.6614 - val\_loss: 1.0034 - val\_mae: 0.8203  
Epoch 18/100  
196/196 \_\_\_\_\_ 0s 649us/step - loss: 0.7551 - mae:  
0.6815 - val\_loss: 0.5862 - val\_mae: 0.6046  
Epoch 19/100  
196/196 \_\_\_\_\_ 0s 670us/step - loss: 0.6716 - mae:  
0.6457 - val\_loss: 0.6041 - val\_mae: 0.6121  
Epoch 20/100  
196/196 \_\_\_\_\_ 0s 667us/step - loss: 0.6108 - mae:  
0.6149 - val\_loss: 0.6138 - val\_mae: 0.6125  
Epoch 21/100  
196/196 \_\_\_\_\_ 0s 665us/step - loss: 0.6788 - mae:  
0.6458 - val\_loss: 0.8042 - val\_mae: 0.6926  
Epoch 22/100  
196/196 \_\_\_\_\_ 0s 660us/step - loss: 0.6961 - mae:  
0.6573 - val\_loss: 0.7104 - val\_mae: 0.6691  
Epoch 23/100  
196/196 \_\_\_\_\_ 0s 662us/step - loss: 0.6294 - mae:  
0.6218 - val\_loss: 0.6867 - val\_mae: 0.6579  
Epoch 24/100  
196/196 \_\_\_\_\_ 0s 665us/step - loss: 0.6541 - mae:  
0.6289 - val\_loss: 0.8090 - val\_mae: 0.6959  
Epoch 25/100  
196/196 \_\_\_\_\_ 0s 670us/step - loss: 0.6615 - mae:  
0.6309 - val\_loss: 0.5892 - val\_mae: 0.5958  
Epoch 26/100  
196/196 \_\_\_\_\_ 0s 666us/step - loss: 0.6461 - mae:  
0.6353 - val\_loss: 0.7365 - val\_mae: 0.6627  
Epoch 27/100  
196/196 \_\_\_\_\_ 0s 650us/step - loss: 0.7638 - mae:  
0.6865 - val\_loss: 0.6348 - val\_mae: 0.6117  
Epoch 28/100

```
196/196 _____ 0s 665us/step - loss: 0.6033 - mae:
0.6071 - val_loss: 0.6371 - val_mae: 0.6169
Epoch 29/100
196/196 _____ 0s 663us/step - loss: 0.5712 - mae:
0.5904 - val_loss: 0.5971 - val_mae: 0.5989
Epoch 30/100
196/196 _____ 0s 649us/step - loss: 0.5954 - mae:
0.6052 - val_loss: 0.7430 - val_mae: 0.6630
Epoch 31/100
196/196 _____ 0s 665us/step - loss: 0.6271 - mae:
0.6157 - val_loss: 0.9087 - val_mae: 0.7482
Epoch 32/100
196/196 _____ 0s 655us/step - loss: 0.5961 - mae:
0.5978 - val_loss: 0.5844 - val_mae: 0.5948
Epoch 33/100
196/196 _____ 0s 669us/step - loss: 0.6520 - mae:
0.6419 - val_loss: 0.5891 - val_mae: 0.6010
Epoch 34/100
196/196 _____ 0s 706us/step - loss: 0.6027 - mae:
0.6088 - val_loss: 0.6887 - val_mae: 0.6375
Epoch 35/100
196/196 _____ 0s 645us/step - loss: 0.6369 - mae:
0.6268 - val_loss: 0.7299 - val_mae: 0.6623
Epoch 36/100
196/196 _____ 0s 634us/step - loss: 0.6361 - mae:
0.6236 - val_loss: 0.7137 - val_mae: 0.6476
Epoch 37/100
196/196 _____ 0s 643us/step - loss: 0.6561 - mae:
0.6301 - val_loss: 0.6781 - val_mae: 0.6354
Epoch 38/100
196/196 _____ 0s 677us/step - loss: 0.5829 - mae:
0.5903 - val_loss: 0.6223 - val_mae: 0.6168
Epoch 39/100
196/196 _____ 0s 637us/step - loss: 0.6167 - mae:
0.6077 - val_loss: 0.7516 - val_mae: 0.6870
Epoch 40/100
196/196 _____ 0s 895us/step - loss: 0.5916 - mae:
0.6002 - val_loss: 0.6376 - val_mae: 0.6164
Epoch 41/100
196/196 _____ 0s 668us/step - loss: 0.6046 - mae:
0.6089 - val_loss: 0.6201 - val_mae: 0.6145
Epoch 42/100
196/196 _____ 0s 662us/step - loss: 0.6359 - mae:
0.6207 - val_loss: 0.5978 - val_mae: 0.6032
Epoch 43/100
196/196 _____ 0s 679us/step - loss: 0.5849 - mae:
0.6003 - val_loss: 0.5875 - val_mae: 0.5940
Epoch 44/100
196/196 _____ 0s 663us/step - loss: 0.5799 - mae:
```

0.5984 - val\_loss: 0.6294 - val\_mae: 0.6113  
Epoch 45/100  
196/196 \_\_\_\_\_ 0s 668us/step - loss: 0.5856 - mae:  
0.5960 - val\_loss: 0.6630 - val\_mae: 0.6229  
Epoch 46/100  
196/196 \_\_\_\_\_ 0s 662us/step - loss: 0.6639 - mae:  
0.6383 - val\_loss: 0.5816 - val\_mae: 0.5912  
Epoch 47/100  
196/196 \_\_\_\_\_ 0s 668us/step - loss: 0.5771 - mae:  
0.5916 - val\_loss: 0.6882 - val\_mae: 0.6321  
Epoch 48/100  
196/196 \_\_\_\_\_ 0s 643us/step - loss: 0.6298 - mae:  
0.6201 - val\_loss: 0.5840 - val\_mae: 0.5935  
Epoch 49/100  
196/196 \_\_\_\_\_ 0s 662us/step - loss: 0.6003 - mae:  
0.6070 - val\_loss: 0.6098 - val\_mae: 0.6040  
Epoch 50/100  
196/196 \_\_\_\_\_ 0s 667us/step - loss: 0.5900 - mae:  
0.5922 - val\_loss: 0.5984 - val\_mae: 0.5930  
Epoch 51/100  
196/196 \_\_\_\_\_ 0s 672us/step - loss: 0.6271 - mae:  
0.6235 - val\_loss: 0.5870 - val\_mae: 0.5934  
Epoch 52/100  
196/196 \_\_\_\_\_ 0s 691us/step - loss: 0.5669 - mae:  
0.5910 - val\_loss: 0.7179 - val\_mae: 0.6525  
Epoch 53/100  
196/196 \_\_\_\_\_ 0s 665us/step - loss: 0.6035 - mae:  
0.6099 - val\_loss: 0.7189 - val\_mae: 0.6561  
Epoch 54/100  
196/196 \_\_\_\_\_ 0s 669us/step - loss: 0.6292 - mae:  
0.6223 - val\_loss: 0.6151 - val\_mae: 0.6127  
Epoch 55/100  
196/196 \_\_\_\_\_ 0s 671us/step - loss: 0.5426 - mae:  
0.5797 - val\_loss: 0.5993 - val\_mae: 0.5976  
Epoch 56/100  
196/196 \_\_\_\_\_ 0s 662us/step - loss: 0.5953 - mae:  
0.5956 - val\_loss: 0.5869 - val\_mae: 0.5959  
Epoch 57/100  
196/196 \_\_\_\_\_ 0s 651us/step - loss: 0.5927 - mae:  
0.6039 - val\_loss: 0.6083 - val\_mae: 0.6086  
Epoch 58/100  
196/196 \_\_\_\_\_ 0s 674us/step - loss: 0.6074 - mae:  
0.6108 - val\_loss: 0.5972 - val\_mae: 0.5995  
Epoch 59/100  
196/196 \_\_\_\_\_ 0s 652us/step - loss: 0.6185 - mae:  
0.6087 - val\_loss: 0.5945 - val\_mae: 0.5934  
Epoch 60/100  
196/196 \_\_\_\_\_ 0s 668us/step - loss: 0.5833 - mae:  
0.5953 - val\_loss: 0.5757 - val\_mae: 0.5881

Epoch 61/100  
196/196 \_\_\_\_\_ 0s 671us/step - loss: 0.5785 - mae:  
0.5933 - val\_loss: 0.9049 - val\_mae: 0.7395  
Epoch 62/100  
196/196 \_\_\_\_\_ 0s 665us/step - loss: 0.5796 - mae:  
0.5967 - val\_loss: 0.6315 - val\_mae: 0.6090  
Epoch 63/100  
196/196 \_\_\_\_\_ 0s 664us/step - loss: 0.5593 - mae:  
0.5868 - val\_loss: 0.5977 - val\_mae: 0.5965  
Epoch 64/100  
196/196 \_\_\_\_\_ 0s 670us/step - loss: 0.5358 - mae:  
0.5749 - val\_loss: 0.6351 - val\_mae: 0.6235  
Epoch 65/100  
196/196 \_\_\_\_\_ 0s 660us/step - loss: 0.5897 - mae:  
0.5936 - val\_loss: 0.5864 - val\_mae: 0.5918  
Epoch 66/100  
196/196 \_\_\_\_\_ 0s 667us/step - loss: 0.5636 - mae:  
0.5899 - val\_loss: 0.6123 - val\_mae: 0.6106  
Epoch 67/100  
196/196 \_\_\_\_\_ 0s 669us/step - loss: 0.5885 - mae:  
0.6012 - val\_loss: 0.5818 - val\_mae: 0.5899  
Epoch 68/100  
196/196 \_\_\_\_\_ 0s 666us/step - loss: 0.6060 - mae:  
0.6054 - val\_loss: 0.5815 - val\_mae: 0.5908  
Epoch 69/100  
196/196 \_\_\_\_\_ 0s 668us/step - loss: 0.5880 - mae:  
0.6033 - val\_loss: 0.5862 - val\_mae: 0.5938  
Epoch 70/100  
196/196 \_\_\_\_\_ 0s 679us/step - loss: 0.6027 - mae:  
0.6055 - val\_loss: 0.5908 - val\_mae: 0.5981  
Epoch 71/100  
196/196 \_\_\_\_\_ 0s 663us/step - loss: 0.5504 - mae:  
0.5765 - val\_loss: 0.6048 - val\_mae: 0.6012  
Epoch 72/100  
196/196 \_\_\_\_\_ 0s 832us/step - loss: 0.5286 - mae:  
0.5732 - val\_loss: 0.7828 - val\_mae: 0.7077  
Epoch 73/100  
196/196 \_\_\_\_\_ 0s 673us/step - loss: 0.5633 - mae:  
0.5917 - val\_loss: 0.6731 - val\_mae: 0.6319  
Epoch 74/100  
196/196 \_\_\_\_\_ 0s 671us/step - loss: 0.5439 - mae:  
0.5773 - val\_loss: 0.5843 - val\_mae: 0.5912  
Epoch 75/100  
196/196 \_\_\_\_\_ 0s 642us/step - loss: 0.5659 - mae:  
0.5875 - val\_loss: 0.6061 - val\_mae: 0.6009  
Epoch 76/100  
196/196 \_\_\_\_\_ 0s 653us/step - loss: 0.6054 - mae:  
0.6104 - val\_loss: 0.6117 - val\_mae: 0.6029  
Epoch 77/100



```
196/196 _____ 0s 664us/step - loss: 0.5693 - mae:
0.5867 - val_loss: 0.5856 - val_mae: 0.5967
Epoch 78/100
196/196 _____ 0s 662us/step - loss: 0.5594 - mae:
0.5917 - val_loss: 0.5865 - val_mae: 0.5914
Epoch 79/100
196/196 _____ 0s 700us/step - loss: 0.5524 - mae:
0.5787 - val_loss: 0.5905 - val_mae: 0.5986
Epoch 80/100
196/196 _____ 0s 668us/step - loss: 0.5419 - mae:
0.5821 - val_loss: 0.7119 - val_mae: 0.6654
Epoch 81/100
196/196 _____ 0s 888us/step - loss: 0.5820 - mae:
0.5967 - val_loss: 0.5837 - val_mae: 0.5951
Epoch 82/100
196/196 _____ 0s 665us/step - loss: 0.5530 - mae:
0.5826 - val_loss: 0.6018 - val_mae: 0.5985
Epoch 83/100
196/196 _____ 0s 674us/step - loss: 0.6151 - mae:
0.6158 - val_loss: 0.5956 - val_mae: 0.5960
Epoch 84/100
196/196 _____ 0s 662us/step - loss: 0.5662 - mae:
0.5928 - val_loss: 0.5814 - val_mae: 0.5911
Epoch 85/100
196/196 _____ 0s 670us/step - loss: 0.5676 - mae:
0.5880 - val_loss: 0.6348 - val_mae: 0.6209
Epoch 86/100
196/196 _____ 0s 667us/step - loss: 0.5683 - mae:
0.5878 - val_loss: 0.6955 - val_mae: 0.6577
Epoch 87/100
196/196 _____ 0s 716us/step - loss: 0.6165 - mae:
0.6040 - val_loss: 0.5708 - val_mae: 0.5850
Epoch 88/100
196/196 _____ 0s 690us/step - loss: 0.5594 - mae:
0.5853 - val_loss: 0.5786 - val_mae: 0.5869
Epoch 89/100
196/196 _____ 0s 646us/step - loss: 0.5384 - mae:
0.5751 - val_loss: 0.6137 - val_mae: 0.6087
Epoch 90/100
196/196 _____ 0s 690us/step - loss: 0.5697 - mae:
0.5893 - val_loss: 0.5829 - val_mae: 0.5922
Epoch 91/100
196/196 _____ 0s 683us/step - loss: 0.5412 - mae:
0.5788 - val_loss: 0.5707 - val_mae: 0.5865
Epoch 92/100
196/196 _____ 0s 674us/step - loss: 0.5400 - mae:
0.5806 - val_loss: 0.5989 - val_mae: 0.5963
Epoch 93/100
196/196 _____ 0s 737us/step - loss: 0.5370 - mae:
```

```

0.5780 - val_loss: 0.6160 - val_mae: 0.6126
Epoch 94/100
196/196 _____ 0s 735us/step - loss: 0.5619 - mae:
0.5846 - val_loss: 0.7307 - val_mae: 0.6813
Epoch 95/100
196/196 _____ 0s 638us/step - loss: 0.5521 - mae:
0.5831 - val_loss: 0.5754 - val_mae: 0.5878
Epoch 96/100
196/196 _____ 0s 706us/step - loss: 0.5304 - mae:
0.5721 - val_loss: 0.5689 - val_mae: 0.5833
Epoch 97/100
196/196 _____ 0s 744us/step - loss: 0.5194 - mae:
0.5633 - val_loss: 0.6094 - val_mae: 0.6085
Epoch 98/100
196/196 _____ 0s 882us/step - loss: 0.5718 - mae:
0.5905 - val_loss: 0.5821 - val_mae: 0.5892
Epoch 99/100
196/196 _____ 0s 701us/step - loss: 0.5580 - mae:
0.5868 - val_loss: 0.5836 - val_mae: 0.5902
Epoch 100/100
196/196 _____ 0s 776us/step - loss: 0.5423 - mae:
0.5796 - val_loss: 0.5721 - val_mae: 0.5859

<keras.src.callbacks.history.History at 0x178deff90>

```

In this code block I am now fitting my model to the data and it gives me back the results for each of the specified metrics that I gave it beforehand.

```

loss, mae = model.evaluate(X_test, y_test)
print(f"Model MAE: {mae:.4f}")

31/31 _____ 0s 748us/step - loss: 0.5459 - mae: 0.5856
Model MAE: 0.5759

```

ChatGPT gave me this block just to check the overall MAE for the model which is at 0.57, this means that on average it is 0.57 away from the actual rating which on a scale of 1 to 10 is quite a good rating but I'd like to see if I can improve it.

My first step that I'm going to take to see if it improves it is using Standard Scaling, this is because features in this dataset are scaled differently and the model might be able to predict better if they are all scaled similarly.

```

scaler = StandardScaler()
X = white_df.drop('quality', axis=1)
y = white_df['quality']

X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.2, random_state=42)

```

```
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
```

Here I am scaling the x training and test values to see if this will help with the accuracy rating of my model.

```
model = keras.Sequential([
    keras.layers.Input(shape=(X_train.shape[1],)),
    keras.layers.Dense(64, activation='relu'),
    keras.layers.Dense(32, activation='relu'),
    keras.layers.Dense(1, activation='linear')
])

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

model.fit(X_train, y_train, epochs=100, batch_size=16,
validation_split=0.2)
```

Epoch 1/100  
196/196 ————— 1s 949us/step - loss: 15.4253 - mae: 3.3969 - val\_loss: 2.2578 - val\_mae: 1.1803  
Epoch 2/100  
196/196 ————— 0s 675us/step - loss: 2.1597 - mae: 1.1228 - val\_loss: 1.4029 - val\_mae: 0.9099  
Epoch 3/100  
196/196 ————— 0s 706us/step - loss: 1.2295 - mae: 0.8644 - val\_loss: 0.9535 - val\_mae: 0.7487  
Epoch 4/100  
196/196 ————— 0s 643us/step - loss: 0.8830 - mae: 0.7260 - val\_loss: 0.7221 - val\_mae: 0.6483  
Epoch 5/100  
196/196 ————— 0s 653us/step - loss: 0.6984 - mae: 0.6483 - val\_loss: 0.6225 - val\_mae: 0.6028  
Epoch 6/100  
196/196 ————— 0s 634us/step - loss: 0.5895 - mae: 0.5950 - val\_loss: 0.5848 - val\_mae: 0.5956  
Epoch 7/100  
196/196 ————— 0s 638us/step - loss: 0.5323 - mae: 0.5649 - val\_loss: 0.5554 - val\_mae: 0.5763  
Epoch 8/100  
196/196 ————— 0s 631us/step - loss: 0.5006 - mae: 0.5538 - val\_loss: 0.5651 - val\_mae: 0.5759  
Epoch 9/100  
196/196 ————— 0s 627us/step - loss: 0.4876 - mae: 0.5511 - val\_loss: 0.5439 - val\_mae: 0.5647  
Epoch 10/100  
196/196 ————— 0s 631us/step - loss: 0.5072 - mae: 0.5593 - val\_loss: 0.5762 - val\_mae: 0.5821  
Epoch 11/100

```
196/196 _____ 0s 634us/step - loss: 0.4770 - mae:
0.5397 - val_loss: 0.5264 - val_mae: 0.5555
Epoch 12/100
196/196 _____ 0s 627us/step - loss: 0.4430 - mae:
0.5260 - val_loss: 0.5328 - val_mae: 0.5600
Epoch 13/100
196/196 _____ 0s 633us/step - loss: 0.4723 - mae:
0.5413 - val_loss: 0.5711 - val_mae: 0.5826
Epoch 14/100
196/196 _____ 0s 638us/step - loss: 0.4593 - mae:
0.5375 - val_loss: 0.5640 - val_mae: 0.5782
Epoch 15/100
196/196 _____ 0s 632us/step - loss: 0.4902 - mae:
0.5480 - val_loss: 0.5203 - val_mae: 0.5482
Epoch 16/100
196/196 _____ 0s 656us/step - loss: 0.4455 - mae:
0.5237 - val_loss: 0.5052 - val_mae: 0.5413
Epoch 17/100
196/196 _____ 0s 690us/step - loss: 0.4701 - mae:
0.5283 - val_loss: 0.5172 - val_mae: 0.5495
Epoch 18/100
196/196 _____ 0s 700us/step - loss: 0.4741 - mae:
0.5426 - val_loss: 0.5259 - val_mae: 0.5554
Epoch 19/100
196/196 _____ 0s 635us/step - loss: 0.4350 - mae:
0.5151 - val_loss: 0.5242 - val_mae: 0.5547
Epoch 20/100
196/196 _____ 0s 645us/step - loss: 0.4494 - mae:
0.5274 - val_loss: 0.5155 - val_mae: 0.5490
Epoch 21/100
196/196 _____ 0s 656us/step - loss: 0.4658 - mae:
0.5281 - val_loss: 0.5110 - val_mae: 0.5439
Epoch 22/100
196/196 _____ 0s 645us/step - loss: 0.4141 - mae:
0.5037 - val_loss: 0.5384 - val_mae: 0.5626
Epoch 23/100
196/196 _____ 0s 676us/step - loss: 0.4257 - mae:
0.5068 - val_loss: 0.5176 - val_mae: 0.5458
Epoch 24/100
196/196 _____ 0s 686us/step - loss: 0.4353 - mae:
0.5165 - val_loss: 0.5113 - val_mae: 0.5455
Epoch 25/100
196/196 _____ 0s 683us/step - loss: 0.4385 - mae:
0.5054 - val_loss: 0.5148 - val_mae: 0.5460
Epoch 26/100
196/196 _____ 0s 650us/step - loss: 0.4045 - mae:
0.4945 - val_loss: 0.5218 - val_mae: 0.5575
Epoch 27/100
196/196 _____ 0s 635us/step - loss: 0.4109 - mae:
```

0.5014 - val\_loss: 0.5377 - val\_mae: 0.5637  
Epoch 28/100  
196/196 \_\_\_\_\_ 0s 695us/step - loss: 0.4391 - mae:  
0.5237 - val\_loss: 0.5364 - val\_mae: 0.5598  
Epoch 29/100  
196/196 \_\_\_\_\_ 0s 707us/step - loss: 0.4077 - mae:  
0.5009 - val\_loss: 0.5200 - val\_mae: 0.5492  
Epoch 30/100  
196/196 \_\_\_\_\_ 0s 675us/step - loss: 0.4062 - mae:  
0.5031 - val\_loss: 0.5272 - val\_mae: 0.5526  
Epoch 31/100  
196/196 \_\_\_\_\_ 0s 697us/step - loss: 0.3993 - mae:  
0.4962 - val\_loss: 0.5191 - val\_mae: 0.5505  
Epoch 32/100  
196/196 \_\_\_\_\_ 0s 977us/step - loss: 0.4147 - mae:  
0.5020 - val\_loss: 0.5086 - val\_mae: 0.5482  
Epoch 33/100  
196/196 \_\_\_\_\_ 0s 660us/step - loss: 0.4094 - mae:  
0.4985 - val\_loss: 0.5523 - val\_mae: 0.5745  
Epoch 34/100  
196/196 \_\_\_\_\_ 0s 643us/step - loss: 0.4301 - mae:  
0.5093 - val\_loss: 0.5192 - val\_mae: 0.5523  
Epoch 35/100  
196/196 \_\_\_\_\_ 0s 647us/step - loss: 0.3858 - mae:  
0.4850 - val\_loss: 0.5295 - val\_mae: 0.5577  
Epoch 36/100  
196/196 \_\_\_\_\_ 0s 666us/step - loss: 0.4195 - mae:  
0.5085 - val\_loss: 0.5099 - val\_mae: 0.5415  
Epoch 37/100  
196/196 \_\_\_\_\_ 0s 674us/step - loss: 0.4127 - mae:  
0.4962 - val\_loss: 0.5366 - val\_mae: 0.5618  
Epoch 38/100  
196/196 \_\_\_\_\_ 0s 652us/step - loss: 0.3817 - mae:  
0.4829 - val\_loss: 0.5338 - val\_mae: 0.5634  
Epoch 39/100  
196/196 \_\_\_\_\_ 0s 673us/step - loss: 0.4083 - mae:  
0.4968 - val\_loss: 0.5207 - val\_mae: 0.5527  
Epoch 40/100  
196/196 \_\_\_\_\_ 0s 669us/step - loss: 0.3867 - mae:  
0.4861 - val\_loss: 0.5305 - val\_mae: 0.5510  
Epoch 41/100  
196/196 \_\_\_\_\_ 0s 648us/step - loss: 0.3704 - mae:  
0.4727 - val\_loss: 0.5259 - val\_mae: 0.5574  
Epoch 42/100  
196/196 \_\_\_\_\_ 0s 640us/step - loss: 0.3756 - mae:  
0.4839 - val\_loss: 0.5421 - val\_mae: 0.5661  
Epoch 43/100  
196/196 \_\_\_\_\_ 0s 631us/step - loss: 0.3805 - mae:  
0.4810 - val\_loss: 0.5173 - val\_mae: 0.5459

Epoch 44/100  
196/196 \_\_\_\_\_ 0s 632us/step - loss: 0.3858 - mae:  
0.4864 - val\_loss: 0.5325 - val\_mae: 0.5591  
Epoch 45/100  
196/196 \_\_\_\_\_ 0s 626us/step - loss: 0.3700 - mae:  
0.4716 - val\_loss: 0.5397 - val\_mae: 0.5644  
Epoch 46/100  
196/196 \_\_\_\_\_ 0s 632us/step - loss: 0.3547 - mae:  
0.4715 - val\_loss: 0.5236 - val\_mae: 0.5567  
Epoch 47/100  
196/196 \_\_\_\_\_ 0s 628us/step - loss: 0.3856 - mae:  
0.4886 - val\_loss: 0.5135 - val\_mae: 0.5472  
Epoch 48/100  
196/196 \_\_\_\_\_ 0s 707us/step - loss: 0.3551 - mae:  
0.4678 - val\_loss: 0.5162 - val\_mae: 0.5507  
Epoch 49/100  
196/196 \_\_\_\_\_ 0s 639us/step - loss: 0.3572 - mae:  
0.4684 - val\_loss: 0.5663 - val\_mae: 0.5756  
Epoch 50/100  
196/196 \_\_\_\_\_ 0s 660us/step - loss: 0.3697 - mae:  
0.4746 - val\_loss: 0.5322 - val\_mae: 0.5601  
Epoch 51/100  
196/196 \_\_\_\_\_ 0s 663us/step - loss: 0.3470 - mae:  
0.4617 - val\_loss: 0.5285 - val\_mae: 0.5563  
Epoch 52/100  
196/196 \_\_\_\_\_ 0s 663us/step - loss: 0.3656 - mae:  
0.4665 - val\_loss: 0.5564 - val\_mae: 0.5703  
Epoch 53/100  
196/196 \_\_\_\_\_ 0s 647us/step - loss: 0.3484 - mae:  
0.4640 - val\_loss: 0.5329 - val\_mae: 0.5603  
Epoch 54/100  
196/196 \_\_\_\_\_ 0s 640us/step - loss: 0.3756 - mae:  
0.4845 - val\_loss: 0.5344 - val\_mae: 0.5579  
Epoch 55/100  
196/196 \_\_\_\_\_ 0s 648us/step - loss: 0.3490 - mae:  
0.4643 - val\_loss: 0.5347 - val\_mae: 0.5614  
Epoch 56/100  
196/196 \_\_\_\_\_ 0s 636us/step - loss: 0.3350 - mae:  
0.4558 - val\_loss: 0.5474 - val\_mae: 0.5687  
Epoch 57/100  
196/196 \_\_\_\_\_ 0s 632us/step - loss: 0.3555 - mae:  
0.4668 - val\_loss: 0.5274 - val\_mae: 0.5547  
Epoch 58/100  
196/196 \_\_\_\_\_ 0s 651us/step - loss: 0.3423 - mae:  
0.4521 - val\_loss: 0.5476 - val\_mae: 0.5691  
Epoch 59/100  
196/196 \_\_\_\_\_ 0s 635us/step - loss: 0.3529 - mae:  
0.4602 - val\_loss: 0.5289 - val\_mae: 0.5573  
Epoch 60/100

196/196 \_\_\_\_\_ 0s 640us/step - loss: 0.3440 - mae:  
0.4554 - val\_loss: 0.5252 - val\_mae: 0.5546  
Epoch 61/100  
196/196 \_\_\_\_\_ 0s 632us/step - loss: 0.3234 - mae:  
0.4496 - val\_loss: 0.5553 - val\_mae: 0.5679  
Epoch 62/100  
196/196 \_\_\_\_\_ 0s 629us/step - loss: 0.3522 - mae:  
0.4654 - val\_loss: 0.5509 - val\_mae: 0.5771  
Epoch 63/100  
196/196 \_\_\_\_\_ 0s 667us/step - loss: 0.3489 - mae:  
0.4581 - val\_loss: 0.5363 - val\_mae: 0.5637  
Epoch 64/100  
196/196 \_\_\_\_\_ 0s 699us/step - loss: 0.3324 - mae:  
0.4485 - val\_loss: 0.5455 - val\_mae: 0.5669  
Epoch 65/100  
196/196 \_\_\_\_\_ 0s 751us/step - loss: 0.3466 - mae:  
0.4638 - val\_loss: 0.5372 - val\_mae: 0.5615  
Epoch 66/100  
196/196 \_\_\_\_\_ 0s 792us/step - loss: 0.3566 - mae:  
0.4652 - val\_loss: 0.5352 - val\_mae: 0.5619  
Epoch 67/100  
196/196 \_\_\_\_\_ 0s 650us/step - loss: 0.3255 - mae:  
0.4472 - val\_loss: 0.5551 - val\_mae: 0.5632  
Epoch 68/100  
196/196 \_\_\_\_\_ 0s 696us/step - loss: 0.3512 - mae:  
0.4611 - val\_loss: 0.5603 - val\_mae: 0.5728  
Epoch 69/100  
196/196 \_\_\_\_\_ 0s 718us/step - loss: 0.3486 - mae:  
0.4641 - val\_loss: 0.5215 - val\_mae: 0.5501  
Epoch 70/100  
196/196 \_\_\_\_\_ 0s 709us/step - loss: 0.3180 - mae:  
0.4429 - val\_loss: 0.5297 - val\_mae: 0.5580  
Epoch 71/100  
196/196 \_\_\_\_\_ 0s 934us/step - loss: 0.3393 - mae:  
0.4561 - val\_loss: 0.5421 - val\_mae: 0.5633  
Epoch 72/100  
196/196 \_\_\_\_\_ 0s 687us/step - loss: 0.3255 - mae:  
0.4457 - val\_loss: 0.5216 - val\_mae: 0.5494  
Epoch 73/100  
196/196 \_\_\_\_\_ 0s 682us/step - loss: 0.3216 - mae:  
0.4438 - val\_loss: 0.5433 - val\_mae: 0.5655  
Epoch 74/100  
196/196 \_\_\_\_\_ 0s 686us/step - loss: 0.3295 - mae:  
0.4476 - val\_loss: 0.5428 - val\_mae: 0.5609  
Epoch 75/100  
196/196 \_\_\_\_\_ 0s 864us/step - loss: 0.3455 - mae:  
0.4585 - val\_loss: 0.5801 - val\_mae: 0.5847  
Epoch 76/100  
196/196 \_\_\_\_\_ 0s 721us/step - loss: 0.3342 - mae:

0.4492 - val\_loss: 0.5294 - val\_mae: 0.5550  
Epoch 77/100  
196/196 \_\_\_\_\_ 0s 666us/step - loss: 0.3209 - mae:  
0.4452 - val\_loss: 0.5825 - val\_mae: 0.5854  
Epoch 78/100  
196/196 \_\_\_\_\_ 0s 671us/step - loss: 0.3223 - mae:  
0.4445 - val\_loss: 0.5524 - val\_mae: 0.5686  
Epoch 79/100  
196/196 \_\_\_\_\_ 0s 665us/step - loss: 0.3112 - mae:  
0.4350 - val\_loss: 0.5344 - val\_mae: 0.5578  
Epoch 80/100  
196/196 \_\_\_\_\_ 0s 688us/step - loss: 0.3159 - mae:  
0.4373 - val\_loss: 0.5326 - val\_mae: 0.5560  
Epoch 81/100  
196/196 \_\_\_\_\_ 0s 716us/step - loss: 0.3185 - mae:  
0.4432 - val\_loss: 0.5481 - val\_mae: 0.5672  
Epoch 82/100  
196/196 \_\_\_\_\_ 0s 694us/step - loss: 0.3251 - mae:  
0.4458 - val\_loss: 0.5719 - val\_mae: 0.5815  
Epoch 83/100  
196/196 \_\_\_\_\_ 0s 692us/step - loss: 0.3169 - mae:  
0.4395 - val\_loss: 0.5431 - val\_mae: 0.5644  
Epoch 84/100  
196/196 \_\_\_\_\_ 0s 669us/step - loss: 0.3116 - mae:  
0.4345 - val\_loss: 0.5460 - val\_mae: 0.5674  
Epoch 85/100  
196/196 \_\_\_\_\_ 0s 686us/step - loss: 0.3031 - mae:  
0.4335 - val\_loss: 0.5384 - val\_mae: 0.5609  
Epoch 86/100  
196/196 \_\_\_\_\_ 0s 698us/step - loss: 0.2955 - mae:  
0.4227 - val\_loss: 0.5375 - val\_mae: 0.5588  
Epoch 87/100  
196/196 \_\_\_\_\_ 0s 956us/step - loss: 0.2949 - mae:  
0.4237 - val\_loss: 0.6170 - val\_mae: 0.5972  
Epoch 88/100  
196/196 \_\_\_\_\_ 0s 673us/step - loss: 0.3008 - mae:  
0.4332 - val\_loss: 0.5399 - val\_mae: 0.5617  
Epoch 89/100  
196/196 \_\_\_\_\_ 0s 669us/step - loss: 0.2937 - mae:  
0.4270 - val\_loss: 0.5457 - val\_mae: 0.5653  
Epoch 90/100  
196/196 \_\_\_\_\_ 0s 668us/step - loss: 0.2929 - mae:  
0.4208 - val\_loss: 0.5360 - val\_mae: 0.5580  
Epoch 91/100  
196/196 \_\_\_\_\_ 0s 664us/step - loss: 0.2792 - mae:  
0.4151 - val\_loss: 0.5745 - val\_mae: 0.5739  
Epoch 92/100  
196/196 \_\_\_\_\_ 0s 731us/step - loss: 0.3052 - mae:  
0.4322 - val\_loss: 0.5424 - val\_mae: 0.5597



```

Epoch 93/100
196/196 ————— 0s 858us/step - loss: 0.2905 - mae:
0.4177 - val_loss: 0.5322 - val_mae: 0.5577
Epoch 94/100
196/196 ————— 0s 759us/step - loss: 0.2951 - mae:
0.4256 - val_loss: 0.5569 - val_mae: 0.5736
Epoch 95/100
196/196 ————— 0s 664us/step - loss: 0.2826 - mae:
0.4111 - val_loss: 0.5893 - val_mae: 0.5937
Epoch 96/100
196/196 ————— 0s 762us/step - loss: 0.3061 - mae:
0.4330 - val_loss: 0.5457 - val_mae: 0.5606
Epoch 97/100
196/196 ————— 0s 869us/step - loss: 0.2853 - mae:
0.4173 - val_loss: 0.5523 - val_mae: 0.5683
Epoch 98/100
196/196 ————— 0s 690us/step - loss: 0.2772 - mae:
0.4109 - val_loss: 0.5519 - val_mae: 0.5701
Epoch 99/100
196/196 ————— 0s 694us/step - loss: 0.2976 - mae:
0.4267 - val_loss: 0.5499 - val_mae: 0.5616
Epoch 100/100
196/196 ————— 0s 644us/step - loss: 0.2686 - mae:
0.4084 - val_loss: 0.5489 - val_mae: 0.5643

<keras.src.callbacks.history.History at 0x178de6410>

loss, mae = model.evaluate(X_test, y_test)
print(f"Model MAE: {mae:.4f}")

31/31 ————— 0s 722us/step - loss: 0.5539 - mae: 0.5711
Model MAE: 0.5584

```

I also ended up trying a few different epoch amounts and I seemed to have the best results whilst using 100 epochs.

As can be seen here this dropped the MAE score from .57 to .55 which is encouraging to see as it was still fitting well beforehand.

I am now going to try this on the red wine dataset.

```

scaler = StandardScaler()
X = red_df.drop('quality', axis=1)
y = red_df['quality']

X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.2, random_state=42)

```

```

X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)

model = keras.Sequential([
    keras.layers.Input(shape=(X_train.shape[1],)),
    keras.layers.Dense(64, activation='relu'),
    keras.layers.Dense(32, activation='relu'),
    keras.layers.Dense(1, activation='linear')
])

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

model.fit(X_train, y_train, epochs=100, batch_size=16,
validation_split=0.2)

Epoch 1/100
64/64 ━━━━━━━━━━━ 0s 2ms/step - loss: 27.6665 - mae: 5.0815 -
val_loss: 4.7705 - val_mae: 1.9128
Epoch 2/100
64/64 ━━━━━━━━━━━ 0s 1ms/step - loss: 3.0794 - mae: 1.4374 -
val_loss: 2.0831 - val_mae: 1.1222
Epoch 3/100
64/64 ━━━━━━━━━━━ 0s 1ms/step - loss: 2.1349 - mae: 1.1196 -
val_loss: 1.7931 - val_mae: 1.0498
Epoch 4/100
64/64 ━━━━━━━━━━━ 0s 969us/step - loss: 1.7248 - mae: 1.0241
- val_loss: 1.6002 - val_mae: 0.9924
Epoch 5/100
64/64 ━━━━━━━━━━━ 0s 1ms/step - loss: 1.4155 - mae: 0.9429 -
val_loss: 1.4399 - val_mae: 0.9326
Epoch 6/100
64/64 ━━━━━━━━━━━ 0s 1ms/step - loss: 1.2647 - mae: 0.8827 -
val_loss: 1.2695 - val_mae: 0.8813
Epoch 7/100
64/64 ━━━━━━━━━━━ 0s 984us/step - loss: 1.0398 - mae: 0.8043
- val_loss: 1.1428 - val_mae: 0.8269
Epoch 8/100
64/64 ━━━━━━━━━━━ 0s 994us/step - loss: 1.0090 - mae: 0.7943
- val_loss: 1.0298 - val_mae: 0.7878
Epoch 9/100
64/64 ━━━━━━━━━━━ 0s 1ms/step - loss: 0.8240 - mae: 0.7210 -
val_loss: 0.9358 - val_mae: 0.7460
Epoch 10/100
64/64 ━━━━━━━━━━━ 0s 993us/step - loss: 0.8480 - mae: 0.7299
- val_loss: 0.8707 - val_mae: 0.7347
Epoch 11/100
64/64 ━━━━━━━━━━━ 0s 972us/step - loss: 0.7314 - mae: 0.6666
- val_loss: 0.7584 - val_mae: 0.6693
Epoch 12/100
64/64 ━━━━━━━━━━━ 0s 991us/step - loss: 0.6894 - mae: 0.6501

```

```
- val_loss: 0.7365 - val_mae: 0.6685
Epoch 13/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 979us/step - loss: 0.6197 - mae: 0.6132
- val_loss: 0.6522 - val_mae: 0.6162
Epoch 14/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 987us/step - loss: 0.5580 - mae: 0.5820
- val_loss: 0.6195 - val_mae: 0.6041
Epoch 15/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 997us/step - loss: 0.5284 - mae: 0.5706
- val_loss: 0.5843 - val_mae: 0.5929
Epoch 16/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 988us/step - loss: 0.4944 - mae: 0.5510
- val_loss: 0.5541 - val_mae: 0.5831
Epoch 17/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 989us/step - loss: 0.5229 - mae: 0.5744
- val_loss: 0.5146 - val_mae: 0.5398
Epoch 18/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 991us/step - loss: 0.4684 - mae: 0.5335
- val_loss: 0.4888 - val_mae: 0.5288
Epoch 19/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 1ms/step - loss: 0.4780 - mae: 0.5398 -
val_loss: 0.4682 - val_mae: 0.5326
Epoch 20/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 934us/step - loss: 0.4329 - mae: 0.5156
- val_loss: 0.4541 - val_mae: 0.5171
Epoch 21/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 996us/step - loss: 0.4067 - mae: 0.4997
- val_loss: 0.4530 - val_mae: 0.5150
Epoch 22/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 931us/step - loss: 0.3962 - mae: 0.4862
- val_loss: 0.4519 - val_mae: 0.5211
Epoch 23/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 922us/step - loss: 0.4010 - mae: 0.5022
- val_loss: 0.4436 - val_mae: 0.5155
Epoch 24/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 1ms/step - loss: 0.4566 - mae: 0.5309 -
val_loss: 0.4224 - val_mae: 0.5024
Epoch 25/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 987us/step - loss: 0.3867 - mae: 0.4874
- val_loss: 0.4424 - val_mae: 0.5237
Epoch 26/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 1ms/step - loss: 0.3965 - mae: 0.5013 -
val_loss: 0.4291 - val_mae: 0.5069
Epoch 27/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 996us/step - loss: 0.3855 - mae: 0.4858
- val_loss: 0.4081 - val_mae: 0.5011
Epoch 28/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 991us/step - loss: 0.3651 - mae: 0.4771
- val_loss: 0.4062 - val_mae: 0.4908
```

```
Epoch 29/100
64/64 _____ 0s 983us/step - loss: 0.3844 - mae: 0.4867
- val_loss: 0.3971 - val_mae: 0.4940
Epoch 30/100
64/64 _____ 0s 935us/step - loss: 0.3473 - mae: 0.4588
- val_loss: 0.4176 - val_mae: 0.4975
Epoch 31/100
64/64 _____ 0s 1ms/step - loss: 0.3816 - mae: 0.4913 -
val_loss: 0.3924 - val_mae: 0.4851
Epoch 32/100
64/64 _____ 0s 986us/step - loss: 0.3716 - mae: 0.4751
- val_loss: 0.3816 - val_mae: 0.4755
Epoch 33/100
64/64 _____ 0s 914us/step - loss: 0.3529 - mae: 0.4530
- val_loss: 0.4030 - val_mae: 0.4845
Epoch 34/100
64/64 _____ 0s 932us/step - loss: 0.3979 - mae: 0.4901
- val_loss: 0.3849 - val_mae: 0.4746
Epoch 35/100
64/64 _____ 0s 919us/step - loss: 0.3412 - mae: 0.4570
- val_loss: 0.3791 - val_mae: 0.4746
Epoch 36/100
64/64 _____ 0s 984us/step - loss: 0.3489 - mae: 0.4552
- val_loss: 0.3650 - val_mae: 0.4674
Epoch 37/100
64/64 _____ 0s 980us/step - loss: 0.3221 - mae: 0.4388
- val_loss: 0.3874 - val_mae: 0.4806
Epoch 38/100
64/64 _____ 0s 927us/step - loss: 0.3445 - mae: 0.4512
- val_loss: 0.3631 - val_mae: 0.4606
Epoch 39/100
64/64 _____ 0s 989us/step - loss: 0.3705 - mae: 0.4682
- val_loss: 0.3924 - val_mae: 0.4771
Epoch 40/100
64/64 _____ 0s 979us/step - loss: 0.3148 - mae: 0.4354
- val_loss: 0.3791 - val_mae: 0.4752
Epoch 41/100
64/64 _____ 0s 1ms/step - loss: 0.3369 - mae: 0.4465 -
val_loss: 0.3909 - val_mae: 0.4853
Epoch 42/100
64/64 _____ 0s 1ms/step - loss: 0.3248 - mae: 0.4501 -
val_loss: 0.3972 - val_mae: 0.4908
Epoch 43/100
64/64 _____ 0s 992us/step - loss: 0.3379 - mae: 0.4516
- val_loss: 0.3739 - val_mae: 0.4636
Epoch 44/100
64/64 _____ 0s 983us/step - loss: 0.3036 - mae: 0.4233
- val_loss: 0.3734 - val_mae: 0.4687
Epoch 45/100
```

64/64 \_\_\_\_\_ 0s 995us/step - loss: 0.3117 - mae: 0.4375  
- val\_loss: 0.3928 - val\_mae: 0.4876  
Epoch 46/100  
64/64 \_\_\_\_\_ 0s 918us/step - loss: 0.3277 - mae: 0.4468  
- val\_loss: 0.3882 - val\_mae: 0.4758  
Epoch 47/100  
64/64 \_\_\_\_\_ 0s 2ms/step - loss: 0.3147 - mae: 0.4366 -  
val\_loss: 0.3898 - val\_mae: 0.4759  
Epoch 48/100  
64/64 \_\_\_\_\_ 0s 984us/step - loss: 0.2997 - mae: 0.4225  
- val\_loss: 0.4157 - val\_mae: 0.4970  
Epoch 49/100  
64/64 \_\_\_\_\_ 0s 925us/step - loss: 0.3050 - mae: 0.4350  
- val\_loss: 0.3627 - val\_mae: 0.4622  
Epoch 50/100  
64/64 \_\_\_\_\_ 0s 939us/step - loss: 0.3240 - mae: 0.4380  
- val\_loss: 0.4057 - val\_mae: 0.4843  
Epoch 51/100  
64/64 \_\_\_\_\_ 0s 989us/step - loss: 0.3008 - mae: 0.4193  
- val\_loss: 0.3632 - val\_mae: 0.4712  
Epoch 52/100  
64/64 \_\_\_\_\_ 0s 995us/step - loss: 0.2943 - mae: 0.4220  
- val\_loss: 0.4489 - val\_mae: 0.5154  
Epoch 53/100  
64/64 \_\_\_\_\_ 0s 985us/step - loss: 0.3250 - mae: 0.4395  
- val\_loss: 0.3870 - val\_mae: 0.4801  
Epoch 54/100  
64/64 \_\_\_\_\_ 0s 988us/step - loss: 0.2833 - mae: 0.4170  
- val\_loss: 0.3836 - val\_mae: 0.4768  
Epoch 55/100  
64/64 \_\_\_\_\_ 0s 960us/step - loss: 0.3187 - mae: 0.4368  
- val\_loss: 0.3902 - val\_mae: 0.4883  
Epoch 56/100  
64/64 \_\_\_\_\_ 0s 992us/step - loss: 0.3238 - mae: 0.4399  
- val\_loss: 0.3731 - val\_mae: 0.4641  
Epoch 57/100  
64/64 \_\_\_\_\_ 0s 1ms/step - loss: 0.2694 - mae: 0.4043 -  
val\_loss: 0.3605 - val\_mae: 0.4500  
Epoch 58/100  
64/64 \_\_\_\_\_ 0s 1ms/step - loss: 0.2582 - mae: 0.3923 -  
val\_loss: 0.3610 - val\_mae: 0.4571  
Epoch 59/100  
64/64 \_\_\_\_\_ 0s 1ms/step - loss: 0.2672 - mae: 0.4035 -  
val\_loss: 0.3858 - val\_mae: 0.4755  
Epoch 60/100  
64/64 \_\_\_\_\_ 0s 1ms/step - loss: 0.2943 - mae: 0.4196 -  
val\_loss: 0.3663 - val\_mae: 0.4546  
Epoch 61/100  
64/64 \_\_\_\_\_ 0s 1ms/step - loss: 0.3024 - mae: 0.4222 -

```
val_loss: 0.3650 - val_mae: 0.4599
Epoch 62/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 1ms/step - loss: 0.2474 - mae: 0.3899 -
val_loss: 0.3659 - val_mae: 0.4544
Epoch 63/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 997us/step - loss: 0.2745 - mae: 0.4076
- val_loss: 0.3543 - val_mae: 0.4515
Epoch 64/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 1ms/step - loss: 0.2952 - mae: 0.4198 -
val_loss: 0.3775 - val_mae: 0.4714
Epoch 65/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 1ms/step - loss: 0.2786 - mae: 0.4060 -
val_loss: 0.3914 - val_mae: 0.4862
Epoch 66/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 1ms/step - loss: 0.2663 - mae: 0.3966 -
val_loss: 0.3668 - val_mae: 0.4578
Epoch 67/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 1ms/step - loss: 0.2717 - mae: 0.3911 -
val_loss: 0.3771 - val_mae: 0.4643
Epoch 68/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 1ms/step - loss: 0.2548 - mae: 0.3878 -
val_loss: 0.3711 - val_mae: 0.4611
Epoch 69/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 1ms/step - loss: 0.2663 - mae: 0.3947 -
val_loss: 0.3612 - val_mae: 0.4578
Epoch 70/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 1ms/step - loss: 0.2810 - mae: 0.4078 -
val_loss: 0.4048 - val_mae: 0.4826
Epoch 71/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 950us/step - loss: 0.2899 - mae: 0.4181
- val_loss: 0.4525 - val_mae: 0.5155
Epoch 72/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 1ms/step - loss: 0.2780 - mae: 0.3990 -
val_loss: 0.4233 - val_mae: 0.4987
Epoch 73/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 948us/step - loss: 0.2621 - mae: 0.3937
- val_loss: 0.3689 - val_mae: 0.4687
Epoch 74/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 932us/step - loss: 0.2464 - mae: 0.3856
- val_loss: 0.4029 - val_mae: 0.4841
Epoch 75/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 990us/step - loss: 0.2763 - mae: 0.4112
- val_loss: 0.3995 - val_mae: 0.4738
Epoch 76/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 931us/step - loss: 0.2574 - mae: 0.3902
- val_loss: 0.3838 - val_mae: 0.4719
Epoch 77/100
64/64 ━━━━━━━━━━━━━━━━━ 0s 979us/step - loss: 0.2652 - mae: 0.3917
- val_loss: 0.4049 - val_mae: 0.4869
```

Epoch 78/100  
64/64 \_\_\_\_\_ 0s 982us/step - loss: 0.2340 - mae: 0.3729  
- val\_loss: 0.3734 - val\_mae: 0.4606  
Epoch 79/100  
64/64 \_\_\_\_\_ 0s 919us/step - loss: 0.2408 - mae: 0.3796  
- val\_loss: 0.3895 - val\_mae: 0.4698  
Epoch 80/100  
64/64 \_\_\_\_\_ 0s 991us/step - loss: 0.2319 - mae: 0.3609  
- val\_loss: 0.3825 - val\_mae: 0.4672  
Epoch 81/100  
64/64 \_\_\_\_\_ 0s 990us/step - loss: 0.2441 - mae: 0.3786  
- val\_loss: 0.3945 - val\_mae: 0.4764  
Epoch 82/100  
64/64 \_\_\_\_\_ 0s 994us/step - loss: 0.2503 - mae: 0.3794  
- val\_loss: 0.4267 - val\_mae: 0.5026  
Epoch 83/100  
64/64 \_\_\_\_\_ 0s 988us/step - loss: 0.2414 - mae: 0.3818  
- val\_loss: 0.4494 - val\_mae: 0.5066  
Epoch 84/100  
64/64 \_\_\_\_\_ 0s 963us/step - loss: 0.2613 - mae: 0.3986  
- val\_loss: 0.3870 - val\_mae: 0.4607  
Epoch 85/100  
64/64 \_\_\_\_\_ 0s 977us/step - loss: 0.2229 - mae: 0.3582  
- val\_loss: 0.4057 - val\_mae: 0.4864  
Epoch 86/100  
64/64 \_\_\_\_\_ 0s 1ms/step - loss: 0.2322 - mae: 0.3709 -  
val\_loss: 0.3940 - val\_mae: 0.4708  
Epoch 87/100  
64/64 \_\_\_\_\_ 0s 1ms/step - loss: 0.2221 - mae: 0.3614 -  
val\_loss: 0.3933 - val\_mae: 0.4657  
Epoch 88/100  
64/64 \_\_\_\_\_ 0s 1ms/step - loss: 0.2296 - mae: 0.3586 -  
val\_loss: 0.3871 - val\_mae: 0.4696  
Epoch 89/100  
64/64 \_\_\_\_\_ 0s 996us/step - loss: 0.2172 - mae: 0.3607  
- val\_loss: 0.4095 - val\_mae: 0.4847  
Epoch 90/100  
64/64 \_\_\_\_\_ 0s 935us/step - loss: 0.2189 - mae: 0.3585  
- val\_loss: 0.4341 - val\_mae: 0.4872  
Epoch 91/100  
64/64 \_\_\_\_\_ 0s 988us/step - loss: 0.2091 - mae: 0.3510  
- val\_loss: 0.3861 - val\_mae: 0.4631  
Epoch 92/100  
64/64 \_\_\_\_\_ 0s 995us/step - loss: 0.2273 - mae: 0.3653  
- val\_loss: 0.4007 - val\_mae: 0.4823  
Epoch 93/100  
64/64 \_\_\_\_\_ 0s 963us/step - loss: 0.2156 - mae: 0.3646  
- val\_loss: 0.4272 - val\_mae: 0.4979  
Epoch 94/100

```

64/64 _____ 0s 980us/step - loss: 0.2044 - mae: 0.3464
- val_loss: 0.4197 - val_mae: 0.4910
Epoch 95/100
64/64 _____ 0s 998us/step - loss: 0.2180 - mae: 0.3585
- val_loss: 0.4065 - val_mae: 0.4744
Epoch 96/100
64/64 _____ 0s 1000us/step - loss: 0.2059 - mae: 0.3484
- val_loss: 0.4263 - val_mae: 0.4861
Epoch 97/100
64/64 _____ 0s 977us/step - loss: 0.2012 - mae: 0.3508
- val_loss: 0.4398 - val_mae: 0.5188
Epoch 98/100
64/64 _____ 0s 932us/step - loss: 0.2423 - mae: 0.3793
- val_loss: 0.4421 - val_mae: 0.4939
Epoch 99/100
64/64 _____ 0s 978us/step - loss: 0.2166 - mae: 0.3596
- val_loss: 0.4107 - val_mae: 0.4681
Epoch 100/100
64/64 _____ 0s 929us/step - loss: 0.2160 - mae: 0.3512
- val_loss: 0.4443 - val_mae: 0.5055

<keras.src.callbacks.history.History at 0x17b166210>

loss, mae = model.evaluate(X_test, y_test)
print(f"Model MAE: {mae:.4f}")

10/10 _____ 0s 1ms/step - loss: 0.4428 - mae: 0.5176
Model MAE: 0.5184

```

As can be seen by the code block above the red wine is predicting better as it has a MAE rating of 0.51 which is better than the white wine rating, I now want to combine the 2 datasets and see the results of this.

```

combined_df = pd.concat([white_df, red_df])

combined_df.to_csv('winequality-combined.csv', index=False)

```

Here I'm just combining both of the datasets and then saving them to a new csv file.

```

scaler = StandardScaler()
X = combined_df.drop('quality', axis=1)
y = combined_df['quality']

X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.2, random_state=42)

X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)

```



```

model = keras.Sequential([
    keras.layers.Input(shape=(X_train.shape[1],)),
    keras.layers.Dense(64, activation='relu'),
    keras.layers.Dense(32, activation='relu'),
    keras.layers.Dense(1, activation='linear')
])

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

model.fit(X_train, y_train, epochs=100, batch_size=16,
validation_split=0.2)

Epoch 1/100
260/260 _____ 1s 864us/step - loss: 13.7221 - mae:
3.0679 - val_loss: 1.5738 - val_mae: 0.9810
Epoch 2/100
260/260 _____ 0s 661us/step - loss: 1.5737 - mae:
0.9689 - val_loss: 1.1133 - val_mae: 0.8191
Epoch 3/100
260/260 _____ 0s 653us/step - loss: 1.0925 - mae:
0.7970 - val_loss: 0.8281 - val_mae: 0.7094
Epoch 4/100
260/260 _____ 0s 634us/step - loss: 0.7683 - mae:
0.6830 - val_loss: 0.6570 - val_mae: 0.6211
Epoch 5/100
260/260 _____ 0s 631us/step - loss: 0.6175 - mae:
0.6167 - val_loss: 0.5960 - val_mae: 0.5962
Epoch 6/100
260/260 _____ 0s 630us/step - loss: 0.5330 - mae:
0.5702 - val_loss: 0.5485 - val_mae: 0.5697
Epoch 7/100
260/260 _____ 0s 626us/step - loss: 0.5323 - mae:
0.5689 - val_loss: 0.5450 - val_mae: 0.5669
Epoch 8/100
260/260 _____ 0s 630us/step - loss: 0.5275 - mae:
0.5646 - val_loss: 0.5397 - val_mae: 0.5626
Epoch 9/100
260/260 _____ 0s 634us/step - loss: 0.4803 - mae:
0.5443 - val_loss: 0.7002 - val_mae: 0.6488
Epoch 10/100
260/260 _____ 0s 637us/step - loss: 0.4754 - mae:
0.5400 - val_loss: 0.5377 - val_mae: 0.5632
Epoch 11/100
260/260 _____ 0s 627us/step - loss: 0.4853 - mae:
0.5422 - val_loss: 0.5041 - val_mae: 0.5416
Epoch 12/100
260/260 _____ 0s 651us/step - loss: 0.4745 - mae:
0.5337 - val_loss: 0.5102 - val_mae: 0.5460
Epoch 13/100
260/260 _____ 0s 622us/step - loss: 0.4549 - mae:

```

0.5251 - val\_loss: 0.5061 - val\_mae: 0.5408  
Epoch 14/100  
260/260 \_\_\_\_\_ 0s 625us/step - loss: 0.4793 - mae:  
0.5405 - val\_loss: 0.5054 - val\_mae: 0.5442  
Epoch 15/100  
260/260 \_\_\_\_\_ 0s 624us/step - loss: 0.4649 - mae:  
0.5368 - val\_loss: 0.5266 - val\_mae: 0.5587  
Epoch 16/100  
260/260 \_\_\_\_\_ 0s 630us/step - loss: 0.4617 - mae:  
0.5280 - val\_loss: 0.5204 - val\_mae: 0.5510  
Epoch 17/100  
260/260 \_\_\_\_\_ 0s 622us/step - loss: 0.4609 - mae:  
0.5339 - val\_loss: 0.5055 - val\_mae: 0.5430  
Epoch 18/100  
260/260 \_\_\_\_\_ 0s 634us/step - loss: 0.4392 - mae:  
0.5189 - val\_loss: 0.5387 - val\_mae: 0.5642  
Epoch 19/100  
260/260 \_\_\_\_\_ 0s 629us/step - loss: 0.4514 - mae:  
0.5217 - val\_loss: 0.5108 - val\_mae: 0.5396  
Epoch 20/100  
260/260 \_\_\_\_\_ 0s 650us/step - loss: 0.4299 - mae:  
0.5152 - val\_loss: 0.5423 - val\_mae: 0.5680  
Epoch 21/100  
260/260 \_\_\_\_\_ 0s 627us/step - loss: 0.4300 - mae:  
0.5160 - val\_loss: 0.5388 - val\_mae: 0.5643  
Epoch 22/100  
260/260 \_\_\_\_\_ 0s 648us/step - loss: 0.4287 - mae:  
0.5193 - val\_loss: 0.5101 - val\_mae: 0.5461  
Epoch 23/100  
260/260 \_\_\_\_\_ 0s 637us/step - loss: 0.4297 - mae:  
0.5169 - val\_loss: 0.5177 - val\_mae: 0.5539  
Epoch 24/100  
260/260 \_\_\_\_\_ 0s 736us/step - loss: 0.4248 - mae:  
0.5073 - val\_loss: 0.5252 - val\_mae: 0.5584  
Epoch 25/100  
260/260 \_\_\_\_\_ 0s 986us/step - loss: 0.4176 - mae:  
0.5069 - val\_loss: 0.5048 - val\_mae: 0.5448  
Epoch 26/100  
260/260 \_\_\_\_\_ 0s 619us/step - loss: 0.4184 - mae:  
0.5051 - val\_loss: 0.5340 - val\_mae: 0.5609  
Epoch 27/100  
260/260 \_\_\_\_\_ 0s 644us/step - loss: 0.4288 - mae:  
0.5138 - val\_loss: 0.5207 - val\_mae: 0.5517  
Epoch 28/100  
260/260 \_\_\_\_\_ 0s 608us/step - loss: 0.4246 - mae:  
0.5097 - val\_loss: 0.5009 - val\_mae: 0.5396  
Epoch 29/100  
260/260 \_\_\_\_\_ 0s 602us/step - loss: 0.4110 - mae:  
0.5008 - val\_loss: 0.5754 - val\_mae: 0.5841

Epoch 30/100  
260/260 ————— 0s 605us/step - loss: 0.4298 - mae:  
0.5133 - val\_loss: 0.5297 - val\_mae: 0.5543  
Epoch 31/100  
260/260 ————— 0s 605us/step - loss: 0.4309 - mae:  
0.5107 - val\_loss: 0.5119 - val\_mae: 0.5486  
Epoch 32/100  
260/260 ————— 0s 614us/step - loss: 0.4171 - mae:  
0.5055 - val\_loss: 0.4930 - val\_mae: 0.5359  
Epoch 33/100  
260/260 ————— 0s 630us/step - loss: 0.4186 - mae:  
0.5088 - val\_loss: 0.5006 - val\_mae: 0.5421  
Epoch 34/100  
260/260 ————— 0s 624us/step - loss: 0.4021 - mae:  
0.4970 - val\_loss: 0.5262 - val\_mae: 0.5606  
Epoch 35/100  
260/260 ————— 0s 619us/step - loss: 0.4068 - mae:  
0.4983 - val\_loss: 0.4976 - val\_mae: 0.5362  
Epoch 36/100  
260/260 ————— 0s 606us/step - loss: 0.4018 - mae:  
0.4966 - val\_loss: 0.5246 - val\_mae: 0.5578  
Epoch 37/100  
260/260 ————— 0s 610us/step - loss: 0.4117 - mae:  
0.5009 - val\_loss: 0.4906 - val\_mae: 0.5286  
Epoch 38/100  
260/260 ————— 0s 599us/step - loss: 0.3941 - mae:  
0.4914 - val\_loss: 0.5229 - val\_mae: 0.5497  
Epoch 39/100  
260/260 ————— 0s 598us/step - loss: 0.3905 - mae:  
0.4899 - val\_loss: 0.5045 - val\_mae: 0.5463  
Epoch 40/100  
260/260 ————— 0s 592us/step - loss: 0.3937 - mae:  
0.4965 - val\_loss: 0.5133 - val\_mae: 0.5468  
Epoch 41/100  
260/260 ————— 0s 600us/step - loss: 0.4088 - mae:  
0.5011 - val\_loss: 0.5242 - val\_mae: 0.5585  
Epoch 42/100  
260/260 ————— 0s 614us/step - loss: 0.3815 - mae:  
0.4804 - val\_loss: 0.5354 - val\_mae: 0.5613  
Epoch 43/100  
260/260 ————— 0s 597us/step - loss: 0.3991 - mae:  
0.4947 - val\_loss: 0.5380 - val\_mae: 0.5678  
Epoch 44/100  
260/260 ————— 0s 621us/step - loss: 0.3905 - mae:  
0.4908 - val\_loss: 0.5334 - val\_mae: 0.5599  
Epoch 45/100  
260/260 ————— 0s 630us/step - loss: 0.3969 - mae:  
0.4878 - val\_loss: 0.5068 - val\_mae: 0.5404  
Epoch 46/100

260/260 ————— 0s 610us/step - loss: 0.3908 - mae:  
0.4899 - val\_loss: 0.5315 - val\_mae: 0.5604  
Epoch 47/100  
260/260 ————— 0s 813us/step - loss: 0.3969 - mae:  
0.4929 - val\_loss: 0.5003 - val\_mae: 0.5399  
Epoch 48/100  
260/260 ————— 0s 603us/step - loss: 0.3825 - mae:  
0.4859 - val\_loss: 0.5013 - val\_mae: 0.5404  
Epoch 49/100  
260/260 ————— 0s 600us/step - loss: 0.3820 - mae:  
0.4825 - val\_loss: 0.5157 - val\_mae: 0.5493  
Epoch 50/100  
260/260 ————— 0s 593us/step - loss: 0.3805 - mae:  
0.4821 - val\_loss: 0.4987 - val\_mae: 0.5438  
Epoch 51/100  
260/260 ————— 0s 600us/step - loss: 0.3843 - mae:  
0.4800 - val\_loss: 0.4859 - val\_mae: 0.5335  
Epoch 52/100  
260/260 ————— 0s 595us/step - loss: 0.3773 - mae:  
0.4754 - val\_loss: 0.5411 - val\_mae: 0.5600  
Epoch 53/100  
260/260 ————— 0s 642us/step - loss: 0.3710 - mae:  
0.4725 - val\_loss: 0.5063 - val\_mae: 0.5420  
Epoch 54/100  
260/260 ————— 0s 607us/step - loss: 0.3743 - mae:  
0.4801 - val\_loss: 0.5546 - val\_mae: 0.5650  
Epoch 55/100  
260/260 ————— 0s 612us/step - loss: 0.3757 - mae:  
0.4834 - val\_loss: 0.4967 - val\_mae: 0.5369  
Epoch 56/100  
260/260 ————— 0s 622us/step - loss: 0.3763 - mae:  
0.4836 - val\_loss: 0.4981 - val\_mae: 0.5403  
Epoch 57/100  
260/260 ————— 0s 626us/step - loss: 0.3575 - mae:  
0.4670 - val\_loss: 0.5047 - val\_mae: 0.5394  
Epoch 58/100  
260/260 ————— 0s 681us/step - loss: 0.3569 - mae:  
0.4689 - val\_loss: 0.5015 - val\_mae: 0.5445  
Epoch 59/100  
260/260 ————— 0s 594us/step - loss: 0.3661 - mae:  
0.4770 - val\_loss: 0.5591 - val\_mae: 0.5761  
Epoch 60/100  
260/260 ————— 0s 597us/step - loss: 0.3778 - mae:  
0.4823 - val\_loss: 0.5104 - val\_mae: 0.5495  
Epoch 61/100  
260/260 ————— 0s 599us/step - loss: 0.3644 - mae:  
0.4675 - val\_loss: 0.5217 - val\_mae: 0.5444  
Epoch 62/100  
260/260 ————— 0s 593us/step - loss: 0.3778 - mae:

0.4814 - val\_loss: 0.5155 - val\_mae: 0.5554  
Epoch 63/100  
260/260 \_\_\_\_\_ 0s 631us/step - loss: 0.3671 - mae:  
0.4772 - val\_loss: 0.5037 - val\_mae: 0.5425  
Epoch 64/100  
260/260 \_\_\_\_\_ 0s 777us/step - loss: 0.3606 - mae:  
0.4647 - val\_loss: 0.5074 - val\_mae: 0.5442  
Epoch 65/100  
260/260 \_\_\_\_\_ 0s 620us/step - loss: 0.3625 - mae:  
0.4713 - val\_loss: 0.5002 - val\_mae: 0.5409  
Epoch 66/100  
260/260 \_\_\_\_\_ 0s 625us/step - loss: 0.3554 - mae:  
0.4658 - val\_loss: 0.5255 - val\_mae: 0.5476  
Epoch 67/100  
260/260 \_\_\_\_\_ 0s 595us/step - loss: 0.3628 - mae:  
0.4741 - val\_loss: 0.5034 - val\_mae: 0.5456  
Epoch 68/100  
260/260 \_\_\_\_\_ 0s 602us/step - loss: 0.3612 - mae:  
0.4715 - val\_loss: 0.5156 - val\_mae: 0.5452  
Epoch 69/100  
260/260 \_\_\_\_\_ 0s 594us/step - loss: 0.3597 - mae:  
0.4652 - val\_loss: 0.5291 - val\_mae: 0.5615  
Epoch 70/100  
260/260 \_\_\_\_\_ 0s 590us/step - loss: 0.3557 - mae:  
0.4679 - val\_loss: 0.5478 - val\_mae: 0.5622  
Epoch 71/100  
260/260 \_\_\_\_\_ 0s 593us/step - loss: 0.3383 - mae:  
0.4550 - val\_loss: 0.5194 - val\_mae: 0.5577  
Epoch 72/100  
260/260 \_\_\_\_\_ 0s 636us/step - loss: 0.3545 - mae:  
0.4681 - val\_loss: 0.5354 - val\_mae: 0.5644  
Epoch 73/100  
260/260 \_\_\_\_\_ 0s 623us/step - loss: 0.3648 - mae:  
0.4687 - val\_loss: 0.5034 - val\_mae: 0.5394  
Epoch 74/100  
260/260 \_\_\_\_\_ 0s 600us/step - loss: 0.3357 - mae:  
0.4538 - val\_loss: 0.5178 - val\_mae: 0.5507  
Epoch 75/100  
260/260 \_\_\_\_\_ 0s 832us/step - loss: 0.3535 - mae:  
0.4593 - val\_loss: 0.5240 - val\_mae: 0.5591  
Epoch 76/100  
260/260 \_\_\_\_\_ 0s 608us/step - loss: 0.3488 - mae:  
0.4567 - val\_loss: 0.5172 - val\_mae: 0.5535  
Epoch 77/100  
260/260 \_\_\_\_\_ 0s 601us/step - loss: 0.3399 - mae:  
0.4538 - val\_loss: 0.5095 - val\_mae: 0.5494  
Epoch 78/100  
260/260 \_\_\_\_\_ 0s 605us/step - loss: 0.3389 - mae:  
0.4548 - val\_loss: 0.4944 - val\_mae: 0.5302

Epoch 79/100  
260/260 ————— 0s 668us/step - loss: 0.3497 - mae: 0.4614 - val\_loss: 0.5429 - val\_mae: 0.5617  
Epoch 80/100  
260/260 ————— 0s 689us/step - loss: 0.3632 - mae: 0.4733 - val\_loss: 0.5116 - val\_mae: 0.5439  
Epoch 81/100  
260/260 ————— 0s 747us/step - loss: 0.3463 - mae: 0.4527 - val\_loss: 0.5056 - val\_mae: 0.5429  
Epoch 82/100  
260/260 ————— 0s 655us/step - loss: 0.3698 - mae: 0.4660 - val\_loss: 0.5280 - val\_mae: 0.5564  
Epoch 83/100  
260/260 ————— 0s 691us/step - loss: 0.3431 - mae: 0.4557 - val\_loss: 0.5490 - val\_mae: 0.5741  
Epoch 84/100  
260/260 ————— 0s 610us/step - loss: 0.3544 - mae: 0.4658 - val\_loss: 0.5376 - val\_mae: 0.5618  
Epoch 85/100  
260/260 ————— 0s 606us/step - loss: 0.3508 - mae: 0.4614 - val\_loss: 0.5127 - val\_mae: 0.5430  
Epoch 86/100  
260/260 ————— 0s 890us/step - loss: 0.3335 - mae: 0.4477 - val\_loss: 0.5315 - val\_mae: 0.5642  
Epoch 87/100  
260/260 ————— 0s 629us/step - loss: 0.3301 - mae: 0.4475 - val\_loss: 0.5080 - val\_mae: 0.5460  
Epoch 88/100  
260/260 ————— 0s 610us/step - loss: 0.3166 - mae: 0.4425 - val\_loss: 0.5078 - val\_mae: 0.5460  
Epoch 89/100  
260/260 ————— 0s 672us/step - loss: 0.3383 - mae: 0.4533 - val\_loss: 0.5136 - val\_mae: 0.5488  
Epoch 90/100  
260/260 ————— 0s 670us/step - loss: 0.3456 - mae: 0.4606 - val\_loss: 0.5277 - val\_mae: 0.5511  
Epoch 91/100  
260/260 ————— 0s 611us/step - loss: 0.3388 - mae: 0.4516 - val\_loss: 0.5104 - val\_mae: 0.5459  
Epoch 92/100  
260/260 ————— 0s 595us/step - loss: 0.3272 - mae: 0.4508 - val\_loss: 0.5050 - val\_mae: 0.5447  
Epoch 93/100  
260/260 ————— 0s 603us/step - loss: 0.3357 - mae: 0.4495 - val\_loss: 0.5482 - val\_mae: 0.5715  
Epoch 94/100  
260/260 ————— 0s 659us/step - loss: 0.3320 - mae: 0.4474 - val\_loss: 0.5197 - val\_mae: 0.5484  
Epoch 95/100

```

260/260 _____ 0s 618us/step - loss: 0.3286 - mae:
0.4469 - val_loss: 0.5118 - val_mae: 0.5449
Epoch 96/100
260/260 _____ 0s 641us/step - loss: 0.3313 - mae:
0.4506 - val_loss: 0.5401 - val_mae: 0.5663
Epoch 97/100
260/260 _____ 0s 679us/step - loss: 0.3378 - mae:
0.4542 - val_loss: 0.5109 - val_mae: 0.5477
Epoch 98/100
260/260 _____ 0s 855us/step - loss: 0.3320 - mae:
0.4455 - val_loss: 0.5107 - val_mae: 0.5420
Epoch 99/100
260/260 _____ 0s 654us/step - loss: 0.3457 - mae:
0.4564 - val_loss: 0.5051 - val_mae: 0.5413
Epoch 100/100
260/260 _____ 0s 652us/step - loss: 0.3355 - mae:
0.4512 - val_loss: 0.5129 - val_mae: 0.5445

<keras.src.callbacks.history.History at 0x178dd63d0>

loss, mae = model.evaluate(X_test, y_test)
print(f"Model MAE: {mae:.4f}")

41/41 _____ 0s 731us/step - loss: 0.4374 - mae: 0.5128
Model MAE: 0.5122

```

The combined dataset is also performing well similarly to the red wine as it is sitting at 0.51 on the MAE score.