Week 1: Multiple Output Models using the Keras Functional API

Welcome to the first programming assignment of the course! Your task will be to use the Keras functional API to train a model to predict two outputs. For this lab, you will use the <u>Wine Quality Dataset</u> from the **UCI machine learning repository**. It has separate datasets for red wine and white wine.

Normally, the wines are classified into one of the quality ratings specified in the attributes. In this exercise, you will combine the two datasets to predict the wine quality and whether the wine is red or white solely from the attributes.

You will model wine quality estimations as a regression problem and wine type detection as a binary classification problem.

Please complete sections that are marked (TODO)

Imports

```
In [1]:
```

```
import tensorflow as tf
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Dense, Input

import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.metrics import confusion_matrix
import itertools
import utils
```

Load Dataset

You will now download the dataset from the UCI Machine Learning Repository.

Pre-process the white wine dataset (TODO)

You will add a new column named <code>is_red</code> in your dataframe to indicate if the wine is white or red.

• In the white wine dataset, you will fill the column is red with zeros (0).

In [4]:

```
## Please uncomment all lines in this cell and replace those marked with `# YOUR CODE HERE`.
## You can select all lines in this code cell with Ctrl+A (Windows/Linux) or Cmd+A (Mac), then pre
ss Ctrl+/ (Windows/Linux) or Cmd+/ (Mac) to uncomment.

# URL of the white wine dataset
URL = 'http://archive.ics.uci.edu/ml/machine-learning-databases/wine-quality/winequality-
white.csv'

# load the dataset from the URL
white_df = pd.read_csv(URL, sep=";")

# fill the `is_red` column with zeros.
white_df["is_red"] = 0

# keep only the first of duplicate items
white_df = white_df.drop_duplicates(keep='first')
```

```
# in case you want to inspect the unit tests being used for each graded function.
utils.test_white_df(white_df)
All public tests passed
In [6]:
print(white df.alcohol[0])
print(white df.alcohol[100])
# EXPECTED OUTPUT
# 8.8
# 9.1
8.8
9.1
Pre-process the red wine dataset (TODO)
 • In the red wine dataset, you will fill in the column is red with ones (1).
In [7]:
## Please uncomment all lines in this cell and replace those marked with `# YOUR CODE HERE`.
## You can select all lines in this code cell with Ctrl+A (Windows/Linux) or Cmd+A (Mac), then pre
ss Ctrl+/ (Windows/Linux) or Cmd+/ (Mac) to uncomment.
# URL of the red wine dataset
URL = 'http://archive.ics.uci.edu/ml/machine-learning-databases/wine-quality/winequality-red.csv'
# load the dataset from the URL
red df = pd.read csv(URL, sep=";")
# fill the `is red` column with ones.
red_df["is_red"] = 1
# keep only the first of duplicate items
red_df = red_df.drop_duplicates(keep='first')
In [8]:
utils.test_red_df(red_df)
All public tests passed
In [9]:
print(red_df.alcohol[0])
print(red df.alcohol[100])
# EXPECTED OUTPUT
# 9.4
# 10.2
9.4
```

You can click `File -> Open` in the menu above and open the `utils.py` file

Concatenate the datasets

10.2

Next, concatenate the red and white wine dataframes.

```
In [10]:

df = pd.concat([red df, white df], ignore index=True)
```

```
In [11]:
```

```
print(df.alcohol[0])
print(df.alcohol[100])

# EXPECTED OUTPUT
# 9.4
# 9.5
```

9.4

9.5

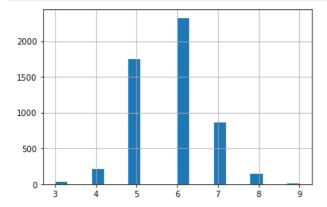
In [13]:

```
# NOTE: In a real-world scenario, you should shuffle the data.
# YOU ARE NOT going to do that here because we want to test
# with deterministic data. But if you want the code to do it,
# it's in the commented line below:
#df = df.iloc[np.random.permutation(len(df))]
```

This will chart the quality of the wines.

In [14]:

```
df['quality'].hist(bins=20);
```



Imbalanced data (TODO)

You can see from the plot above that the wine quality dataset is imbalanced.

- Since there are very few observations with quality equal to 3, 4, 8 and 9, you can drop these observations from your dataset.
- You can do this by removing data belonging to all classes except those > 4 and < 8.

In [15]:

```
# Please uncomment all lines in this cell and replace those marked with `# YOUR CODE HERE`.
# You can select all lines in this code cell with Ctrl+A (Windows/Linux) or Cmd+A (Mac), then pres
s Ctrl+/ (Windows/Linux) or Cmd+/ (Mac) to uncomment.

# get data with wine quality greater than 4 and less than 8
df = df[(df['quality'] > 4) & (df['quality'] < 8)]

# reset index and drop the old one
df = df.reset_index(drop=True)</pre>
```

In [16]:

```
utils.test_df_drop(df)
```

```
All public tests passed
```

In [17]:

```
print(df.alcohol[0])
print(df.alcohol[100])

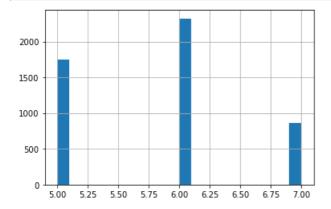
# EXPECTED OUTPUT
# 9.4
# 10.9
9.4
```

10.9

You can plot again to see the new range of data and quality

In [18]:

```
df['quality'].hist(bins=20);
```



Train Test Split (TODO)

Next, you can split the datasets into training, test and validation datasets.

- The data frame should be split 80:20 into train and test sets.
- The resulting train should then be split 80:20 into train and val sets.
- The train_test_split parameter test_size takes a float value that ranges between 0. and 1, and represents the proportion of the dataset that is allocated to the test set. The rest of the data is allocated to the training set.

In [19]:

```
# Please uncomment all lines in this cell and replace those marked with `# YOUR CODE HERE`.
# You can select all lines in this code cell with Ctrl+A (Windows/Linux) or Cmd+A (Mac), then pres
s Ctrl+/ (Windows/Linux) or Cmd+/ (Mac) to uncomment.

# Please do not change the random_state parameter. This is needed for grading.

# split df into 80:20 train and test sets
train, test = train_test_split(df, test_size=.2, random_state = 1)

# split train into 80:20 train and val sets
train, val = train_test_split(train, test_size=.2, random_state = 1)
```

In [20]:

```
utils.test_data_sizes(train.size, test.size, val.size)
```

Here's where you can explore the training stats. You can pop the labels 'is_red' and 'quality' from the data as these will be used as the labels

```
In [21]:
```

```
train_stats = train.describe()
train_stats.pop('is_red')
train_stats.pop('quality')
train_stats = train_stats.transpose()
```

Explore the training stats!

In [22]:

```
train_stats
```

Out[22]:

	count	mean	std	min	25%	50%	75%	max
fixed acidity	3155.0	7.221616	1.325297	3.80000	6.40000	7.00000	7.7000	15.60000
volatile acidity	3155.0	0.338929	0.162476	0.08000	0.23000	0.29000	0.4000	1.24000
citric acid	3155.0	0.321569	0.147970	0.00000	0.25000	0.31000	0.4000	1.66000
residual sugar	3155.0	5.155911	4.639632	0.60000	1.80000	2.80000	7.6500	65.80000
chlorides	3155.0	0.056976	0.036802	0.01200	0.03800	0.04700	0.0660	0.61100
free sulfur dioxide	3155.0	30.388590	17.236784	1.00000	17.00000	28.00000	41.0000	131.00000
total sulfur dioxide	3155.0	115.062282	56.706617	6.00000	75.00000	117.00000	156.0000	344.00000
density	3155.0	0.994633	0.003005	0.98711	0.99232	0.99481	0.9968	1.03898
рН	3155.0	3.223201	0.161272	2.72000	3.11000	3.21000	3.3300	4.01000
sulphates	3155.0	0.534051	0.149149	0.22000	0.43000	0.51000	0.6000	1.95000
alcohol	3155.0	10.504466	1.154654	8.50000	9.50000	10.30000	11.3000	14.00000

Get the labels (TODO)

The features and labels are currently in the same dataframe.

- You will want to store the label columns is red and quality separately from the feature columns.
- The following function, format_output, gets these two columns from the dataframe (it's given to you).
- format output also formats the data into numpy arrays.
- Please use the format output and apply it to the train, val and test sets to get dataframes for the labels.

In [23]:

```
def format_output(data):
    is_red = data.pop('is_red')
    is_red = np.array(is_red)
    quality = data.pop('quality')
    quality = np.array(quality)
    return (quality, is_red)
```

In [24]:

```
# Please uncomment all lines in this cell and replace those marked with `# YOUR CODE HERE`.
# You can select all lines in this code cell with Ctrl+A (Windows/Linux) or Cmd+A (Mac), then pres
s Ctrl+/ (Windows/Linux) or Cmd+/ (Mac) to uncomment.

# format the output of the train set
train_Y = format_output(train)

# format the output of the val set
val_Y = format_output(val)
```

```
# format the output of the test set
test_Y = format_output(test)
```

In [25]:

```
utils.test_format_output(df, train_Y, val_Y, test_Y)
```

```
All public tests passed
```

Notice that after you get the labels, the train, val and test dataframes no longer contain the label columns, and contain just the feature columns.

• This is because you used .pop in the format_output function.

In [26]:

```
train.head()
```

Out[26]:

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	alcohol
225	7.5	0.65	0.18	7.0	0.088	27.0	94.0	0.99915	3.38	0.77	9.4
3557	6.3	0.27	0.29	12.2	0.044	59.0	196.0	0.99782	3.14	0.40	8.8
3825	8.8	0.27	0.25	5.0	0.024	52.0	99.0	0.99250	2.87	0.49	11.4
1740	6.4	0.45	0.07	1.1	0.030	10.0	131.0	0.99050	2.97	0.28	10.8
1221	7.2	0.53	0.13	2.0	0.058	18.0	22.0	0.99573	3.21	0.68	9.9

Normalize the data (TODO)

Next, you can normalize the data, x, using the formula: $x_{\text{norm}} = \frac{x - \mu}{sgma}$

- The norm function is defined for you.
- Please apply the norm function to normalize the dataframes that contains the feature columns of train, val and test sets.

In [27]:

```
def norm(x):
    return (x - train_stats['mean']) / train_stats['std']
```

In [28]:

```
# Please uncomment all lines in this cell and replace those marked with `# YOUR CODE HERE`.
# You can select all lines in this code cell with Ctrl+A (Windows/Linux) or Cmd+A (Mac), then pres
s Ctrl+/ (Windows/Linux) or Cmd+/ (Mac) to uncomment.

# # normalize the train set
norm_train_X = norm(train)

# # normalize the val set
norm_val_X = norm(val)

# # normalize the test set
norm_test_X = norm(test)
```

In [29]:

```
utils.test_norm(norm_train_X, norm_val_X, norm_test_X, train, val, test)
```

```
All public tests passed
```

Define the Model (TODO)

Define the model using the functional API. The base model will be 2 <code>Dense</code> layers of 128 neurons each, and have the <code>'relu'</code> activation.

• Check out the documentation for tf.keras.layers.Dense

In [30]:

```
# Please uncomment all lines in this cell and replace those marked with `# YOUR CODE HERE`.
# You can select all lines in this code cell with Ctrl+A (Windows/Linux) or Cmd+A (Mac), then pres
s Ctrl+/ (Windows/Linux) or Cmd+/ (Mac) to uncomment.

def base_model(inputs):

    # connect a Dense layer with 128 neurons and a relu activation
    x = Dense(128, activation='relu')(inputs)

# connect another Dense layer with 128 neurons and a relu activation
x = Dense(128, activation='relu')(x)
return x
```

In [31]:

```
utils.test_base_model(base_model)
```

All public tests passed

Define output layers of the model (TODO)

You will add output layers to the base model.

• The model will need two outputs.

One output layer will predict wine quality, which is a numeric value.

- Define a Dense layer with 1 neuron.
- Since this is a regression output, the activation can be left as its default value None.

The other output layer will predict the wine type, which is either red 1 or not red 0 (white).

- Define a Dense layer with 1 neuron.
- Since there are two possible categories, you can use a sigmoid activation for binary classification.

Define the Model

- Define the Model object, and set the following parameters:
 - inputs : pass in the inputs to the model as a list.
 - outputs: pass in a list of the outputs that you just defined: wine quality, then wine type.
 - **Note**: please list the wine quality before wine type in the outputs, as this will affect the calculated loss if you choose the other order.

In [34]:

```
# Please uncomment all lines in this cell and replace those marked with `# YOUR CODE HERE`.
# You can select all lines in this code cell with Ctrl+A (Windows/Linux) or Cmd+A (Mac), then pres
s Ctrl+/ (Windows/Linux) or Cmd+/ (Mac) to uncomment.

def final_model(inputs):
    # get the base model
    x = base_model(inputs)
# connect the output Dense layer for regression
    wine_quality = Dense(units='1', name='wine_quality')(x)
```

```
# connect the output Dense layer for classification. this will use a sigmoid activation.
wine_type = Dense(units='1', activation='sigmoid', name='wine_type')(x)

# define the model using the input and output layers
model = Model(inputs=inputs, outputs=[wine_quality, wine_type])
return model
```

```
In [35]:
```

```
utils.test_final_model(final_model)
```

All public tests passed

Compiling the Model

Next, compile the model. When setting the loss parameter of <code>model.compile</code> , you're setting the loss for each of the two outputs (wine quality and wine type).

To set more than one loss, use a dictionary of key-value pairs.

- You can look at the docs for the losses here.
 - **Note**: For the desired spelling, please look at the "Functions" section of the documentation and not the "classes" section on that same page.
- wine_type: Since you will be performing binary classification on wine type, you should use the binary crossentropy loss function for it. Please pass this in as a string.
 - Hint, this should be all lowercase. In the documentation, you'll see this under the "Functions" section, not the "Classes" section.
- wine_quality: since this is a regression output, use the mean squared error. Please pass it in as a string, all lowercase.
 - Hint: You may notice that there are two aliases for mean squared error. Please use the shorter name.

You will also set the metric for each of the two outputs. Again, to set metrics for two or more outputs, use a dictionary with key value pairs.

- The metrics documentation is linked here.
- For the wine type, please set it to accuracy as a string, all lowercase.
- For wine quality, please use the root mean squared error. Instead of a string, you'll set it to an instance of the class RootMeanSquaredError, which belongs to the tf.keras.metrics module.

Note: If you see the error message

Exception: wine quality loss function is incorrect.

Please also check your other losses and metrics, as the error may be caused by the other three key-value pairs and not the wine
quality loss.

In [39]:

```
utils.test_model_compile(model)
```

All public tests passed

Training the Model

Fit the model to the training inputs and outputs.

- Check the documentation for model.fit.
- Remember to use the normalized training set as inputs.
- For the validation data, please use the normalized validation set.

```
In [42]:
# Please uncomment all lines in this cell and replace those marked with `# YOUR CODE HERE`.
# You can select all lines in this code cell with Ctrl+A (Windows/Linux) or Cmd+A (Mac), then pres
s Ctrl+/ (Windows/Linux) or Cmd+/ (Mac) to uncomment.
history = model.fit(norm train_X, train_Y,
               epochs = 180, validation data=(norm val X, val Y))
Train on 3155 samples, validate on 789 samples
Epoch 1/180
23.9919 - wine_type_loss: 0.6763 - wine_quality_root_mean_squared_error: 4.9030 -
wine type accuracy: 0.5971 - val loss: 16.0329 - val wine quality loss: 15.4115 -
val_wine_type_loss: 0.6436 - val_wine_quality_root_mean_squared_error: 3.9230 -
val_wine_type_accuracy: 0.7529
Epoch 2/180
.4268 - wine_type_loss: 0.5990 - wine_quality_root_mean_squared_error: 3.0742 -
wine type accuracy: 0.7930 - val loss: 5.4150 - val wine quality loss: 4.9202 -
val wine type loss: 0.5619 - val wine quality root mean squared error: 2.2032 -
val_wine_type_accuracy: 0.8200
Epoch 3/180
2639 - wine type loss: 0.5019 - wine quality root mean squared error: 1.8078 - wine type accuracy:
0.8390 - val loss: 2.8972 - val wine quality loss: 2.5012 - val wine type loss: 0.4544 -
val_wine_quality_root_mean_squared_error: 1.5632 - val_wine_type_accuracy: 0.8504
Epoch 4/180
2134 - wine_type_loss: 0.3981 - wine_quality_root_mean_squared_error: 1.4885 - wine_type_accuracy:
0.8764 - val loss: 2.3650 - val wine quality loss: 2.0381 - val wine type loss: 0.3648 -
val_wine_quality_root_mean_squared_error: 1.4144 - val_wine_type_accuracy: 0.8872
Epoch 5/180
8403 - wine_type_loss: 0.3148 - wine_quality_root_mean_squared_error: 1.3558 - wine_type_accuracy:
0.9242 - val_loss: 2.0153 - val_wine_quality_loss: 1.7500 - val_wine_type_loss: 0.2877 -
val_wine_quality_root_mean_squared_error: 1.3145 - val_wine_type_accuracy: 0.9392
Epoch 6/180
6217 - wine_type_loss: 0.2504 - wine_quality_root_mean_squared_error: 1.2737 - wine_type accuracy:
0.9601 - val loss: 1.7766 - val wine quality loss: 1.5615 - val wine type loss: 0.2285 -
val wine quality root mean squared error: 1.2443 - val wine type accuracy: 0.9708
Epoch 7/180
4650 - wine_type_loss: 0.1995 - wine_quality_root_mean_squared_error: 1.2112 - wine_type_accuracy:
0.9743 - val_loss: 1.5879 - val_wine_quality_loss: 1.4140 - val_wine_type_loss: 0.1821 -
val_wine_quality_root_mean_squared_error: 1.1857 - val_wine_type_accuracy: 0.9823
Epoch 8/180
3456 - wine_type_loss: 0.1608 - wine_quality_root_mean_squared_error: 1.1613 - wine_type_accuracy:
0.9797 - val loss: 1.4438 - val wine quality loss: 1.3006 - val wine type loss: 0.1486 -
val_wine_quality_root_mean_squared_error: 1.1381 - val_wine_type_accuracy: 0.9848
Epoch 9/180
2534 - wine_type_loss: 0.1330 - wine_quality_root_mean_squared_error: 1.1201 - wine_type_accuracy:
0.9851 - val loss: 1.3339 - val wine quality loss: 1.2139 - val wine type loss: 0.1224 -
val_wine_quality_root_mean_squared_error: 1.1007 - val_wine_type_accuracy: 0.9873
Epoch 10/180
                                                    . . . . . .
```

```
1710 - wine_type_loss: 0.1123 - wine_quality_root_mean_squared_error: 1.0820 - wine_type_accuracy:
0.9864 - val_loss: 1.2373 - val_wine_quality_loss: 1.1344 - val_wine_type_loss: 0.1038
val wine quality root mean squared error: 1.0647 - val wine type accuracy: 0.9899
Epoch 11/180
1008 - wine type_loss: 0.0968 - wine_quality_root_mean_squared_error: 1.0498 - wine_type_accuracy:
0.9873 - val_loss: 1.1451 - val_wine_quality_loss: 1.0553 - val_wine_type_loss: 0.0903 -
val_wine_quality_root_mean_squared_error: 1.0270 - val_wine_type_accuracy: 0.9911
Epoch 12/180
0349 - wine type loss: 0.0850 - wine quality root mean squared error: 1.0179 - wine type accuracy:
0.9880 - val_loss: 1.0753 - val_wine_quality_loss: 0.9959 - val_wine_type_loss: 0.0796 -
val wine quality root mean squared error: 0.9978 - val wine type accuracy: 0.9911
Epoch 13/180
9796 - wine type loss: 0.0761 - wine quality root mean squared error: 0.9867 - wine type accuracy:
0.9883 - val loss: 1.0020 - val wine quality loss: 0.9310 - val wine type loss: 0.0711 -
val wine quality root mean squared error: 0.9648 - val wine type accuracy: 0.9911
Epoch 14/180
9159 - wine_type_loss: 0.0689 - wine_quality_root_mean_squared_error: 0.9570 - wine_type_accuracy:
0.9886 - val loss: 0.9426 - val wine quality loss: 0.8775 - val wine type loss: 0.0641 -
val_wine_quality_root_mean_squared_error: 0.9372 - val_wine_type_accuracy: 0.9911
Epoch 15/180
8611 - wine_type_loss: 0.0630 - wine_quality_root_mean_squared_error: 0.9280 - wine_type_accuracy:
0.9895 - val loss: 0.8936 - val wine quality loss: 0.8334 - val wine type loss: 0.0586 -
val wine quality root mean squared error: 0.9137 - val wine type accuracy: 0.9924
Epoch 16/180
8104 - wine type loss: 0.0582 - wine quality root mean squared error: 0.9007 - wine type accuracy:
0.9905 - val_loss: 0.8345 - val_wine_quality_loss: 0.7788 - val_wine_type_loss: 0.0544 -
val wine quality root mean squared error: 0.8831 - val wine type accuracy: 0.9924
Epoch 17/180
7630 - wine_type_loss: 0.0544 - wine_quality_root_mean_squared_error: 0.8730 - wine_type_accuracy:
0.9908 - val_loss: 0.7855 - val_wine_quality_loss: 0.7332 - val_wine_type_loss: 0.0510 -
val_wine_quality_root_mean_squared_error: 0.8569 - val_wine_type_accuracy: 0.9924
Epoch 18/180
7207 - wine_type_loss: 0.0514 - wine_quality_root_mean_squared_error: 0.8479 - wine_type_accuracy:
0.9908 - val_loss: 0.7378 - val_wine_quality_loss: 0.6882 - val_wine_type_loss: 0.0480 -
val_wine_quality_root_mean_squared_error: 0.8304 - val_wine_type_accuracy: 0.9924
Epoch 19/180
6794 - wine_type_loss: 0.0487 - wine_quality_root_mean_squared_error: 0.8249 - wine_type_accuracy:
0.9914 - val loss: 0.7013 - val wine quality loss: 0.6539 - val wine type loss: 0.0456 -
val wine quality root mean squared error: 0.8096 - val wine type accuracy: 0.9924
Epoch 20/180
6449 - wine_type_loss: 0.0466 - wine_quality_root_mean_squared_error: 0.8035 - wine_type_accuracy:
0.9924 - val_loss: 0.6591 - val_wine_quality_loss: 0.6139 - val_wine_type_loss: 0.0436
val_wine_quality_root_mean_squared_error: 0.7844 - val_wine_type_accuracy: 0.9924
Epoch 21/180
6091 - wine_type_loss: 0.0447 - wine_quality_root_mean_squared_error: 0.7809 - wine_type_accuracy:
0.9918 - val_loss: 0.6263 - val_wine_quality_loss: 0.5824 - val_wine_type_loss: 0.0420 \cdot
val wine quality root mean squared error: 0.7642 - val wine type accuracy: 0.9924
Epoch 22/180
5790 - wine type loss: 0.0433 - wine quality root mean squared error: 0.7610 - wine type accuracy:
0.9921 - val_loss: 0.5972 - val_wine_quality_loss: 0.5548 - val_wine_type_loss: 0.0406 -
val wine quality root mean squared error: 0.7459 - val wine type accuracy: 0.9924
Epoch 23/180
5512 - wine type loss: 0.0418 - wine quality root mean squared error: 0.7430 - wine type accuracy:
0.9924 - val_loss: 0.5759 - val_wine_quality_loss: 0.5344 - val_wine_type_loss: 0.0393 -
val_wine_quality_root_mean_squared_error: 0.7323 - val_wine_type_accuracy: 0.9924
Epoch 24/180
5246 - wine type loss: 0.0407 - wine quality_root_mean_squared_error: 0.7241 - wine_type_accuracy:
0.9924 - val_loss: 0.5440 - val_wine_quality_loss: 0.5037 - val_wine_type_loss: 0.0384 -
val_wine_quality_root_mean_squared_error: 0.7108 - val_wine_type_accuracy: 0.9937
Epoch 25/180
5004 - wine_type_loss: 0.0395 - wine_quality_root_mean_squared_error: 0.7072 - wine_type_accuracy:
```

```
0.9930 - val loss: 0.5194 - val wine quality loss: 0.4801 - val wine type loss: 0.0375 -
val wine quality root mean squared error: 0.6939 - val wine type accuracy: 0.9937
Epoch 26/180
4791 - wine_type_loss: 0.0388 - wine_quality_root_mean_squared_error: 0.6922 - wine_type_accuracy:
0.9933 - val loss: 0.4963 - val wine quality loss: 0.4575 - val wine type loss: 0.0367 - val wine type
val_wine_quality_root_mean_squared_error: 0.6777 - val_wine_type accuracy: 0.9937
Epoch 27/180
4586 - wine type loss: 0.0377 - wine quality root mean squared error: 0.6779 - wine type accuracy:
0.9933 - val loss: 0.4864 - val wine quality loss: 0.4481 - val wine type loss: 0.0361 -
val_wine_quality_root_mean_squared_error: 0.6708 - val_wine_type_accuracy: 0.9937
Epoch 28/180
4414 - wine_type_loss: 0.0370 - wine_quality_root_mean_squared_error: 0.6644 - wine_type_accuracy:
0.9937 - val loss: 0.4711 - val wine quality loss: 0.4335 - val wine type loss: 0.0353 -
val_wine_quality_root_mean_squared_error: 0.6599 - val_wine_type_accuracy: 0.9949
Epoch 29/180
4271 - wine_type_loss: 0.0363 - wine_quality_root_mean_squared_error: 0.6540 - wine_type_accuracy:
0.9940 - val loss: 0.4579 - val wine quality loss: 0.4208 - val wine type loss: 0.0349 -
val_wine_quality_root_mean_squared_error: 0.6501 - val_wine_type_accuracy: 0.9949
Epoch 30/180
4137 - wine_type_loss: 0.0357 - wine_quality_root_mean_squared_error: 0.6431 - wine_type_accuracy:
0.9940 - val_loss: 0.4363 - val_wine_quality_loss: 0.3998 - val_wine_type_loss: 0.0344 -
val_wine_quality_root_mean_squared_error: 0.6337 - val_wine_type_accuracy: 0.9949
Epoch 31/180
3998 - wine type_loss: 0.0351 - wine_quality_root_mean_squared_error: 0.6323 - wine_type_accuracy:
0.9940 - val_loss: 0.4273 - val_wine_quality_loss: 0.3911 - val_wine_type_loss: 0.0340 -
val wine quality_root_mean_squared_error: 0.6269 - val_wine_type_accuracy: 0.9949
Epoch 32/180
3903 - wine type loss: 0.0348 - wine quality root mean squared error: 0.6235 - wine type accuracy:
0.9943 - val_loss: 0.4175 - val_wine_quality_loss: 0.3820 - val_wine_type_loss: 0.0337 -
val_wine_quality_root_mean_squared_error: 0.6193 - val_wine_type_accuracy: 0.9949
Epoch 33/180
3814 - wine type loss: 0.0341 - wine quality root mean squared error: 0.6176 - wine type accuracy:
0.9943 - val_loss: 0.4090 - val_wine_quality_loss: 0.3739 - val_wine_type_loss: 0.0332 -
val_wine_quality_root_mean_squared_error: 0.6128 - val_wine_type_accuracy: 0.9949
Epoch 34/180
3716 - wine type loss: 0.0341 - wine quality root mean squared error: 0.6098 - wine type accuracy:
0.9943 - val loss: 0.4136 - val wine quality loss: 0.3785 - val wine type loss: 0.0330 -
val_wine_quality_root_mean_squared_error: 0.6167 - val_wine_type_accuracy: 0.9949
Epoch 35/180
3671 - wine type loss: 0.0334 - wine quality root mean squared error: 0.6054 - wine type accuracy:
0.9943 - val loss: 0.3999 - val wine quality loss: 0.3651 - val wine type loss: 0.0327 -
val_wine_quality_root_mean_squared_error: 0.6057 - val_wine_type_accuracy: 0.9949
Epoch 36/180
3578 - wine_type_loss: 0.0331 - wine_quality_root_mean_squared_error: 0.5984 - wine_type_accuracy:
0.9940 - val_loss: 0.3885 - val_wine_quality_loss: 0.3544 - val_wine_type_loss: 0.0324 -
val wine quality root mean squared error: 0.5965 - val wine type accuracy: 0.9949
Epoch 37/180
3542 - wine_type_loss: 0.0327 - wine_quality_root_mean_squared_error: 0.5953 - wine_type_accuracy:
0.9943 - val loss: 0.3805 - val wine quality loss: 0.3465 - val wine type loss: 0.0322 -
val wine quality root mean squared error: 0.5899 - val wine type accuracy: 0.9949
Epoch 38/180
3472 - wine type loss: 0.0324 - wine quality root mean squared error: 0.5898 - wine type accuracy:
0.9940 - val_loss: 0.3800 - val_wine_quality_loss: 0.3465 - val_wine_type_loss: 0.0320 -
val_wine_quality_root_mean_squared_error: 0.5897 - val_wine_type_accuracy: 0.9949
Epoch 39/180
3448 - wine type_loss: 0.0320 - wine_quality_root_mean_squared_error: 0.5876 - wine_type_accuracy:
0.9940 - val_loss: 0.3773 - val_wine_quality_loss: 0.3438 - val_wine_type_loss: 0.0317 -
val_wine_quality_root_mean_squared_error: 0.5875 - val_wine_type_accuracy: 0.9949
Epoch 40/180
3400 - wine_type_loss: 0.0317 - wine_quality_root_mean_squared_error: 0.5835 - wine_type_accuracy:
0.9946 - val loss: 0.3696 - val wine quality loss: 0.3364 - val wine type loss: 0.0314 -
val wine quality root mean squared error: 0.5812 - val wine type accuracy: 0.9949
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Epoch 41/180
3365 - wine type_loss: 0.0313 - wine_quality_root_mean_squared_error: 0.5805 - wine_type_accuracy:
0.9946 - val_loss: 0.3706 - val_wine_quality_loss: 0.3373 - val_wine_type_loss: 0.0313
val wine quality root mean squared error: 0.5821 - val wine type accuracy: 0.9949
Epoch 42/180
3320 - wine type loss: 0.0310 - wine quality root mean squared error: 0.5762 - wine type accuracy:
0.9946 - val_loss: 0.3709 - val_wine_quality_loss: 0.3378 - val_wine_type_loss: 0.0311 -
val_wine_quality_root_mean_squared_error: 0.5826 - val_wine_type_accuracy: 0.9949
Epoch 43/180
3302 - wine type loss: 0.0309 - wine quality root mean squared error: 0.5753 - wine type accuracy:
0.9946 - val_loss: 0.3727 - val_wine_quality_loss: 0.3401 - val_wine_type_loss: 0.0311 -
val_wine_quality_root_mean_squared_error: 0.5842 - val_wine_type_accuracy: 0.9949
Epoch 44/180
3289 - wine type loss: 0.0305 - wine quality_root_mean_squared_error: 0.5731 - wine_type_accuracy:
0.9946 - val loss: 0.3683 - val wine quality loss: 0.3358 - val wine type loss: 0.0308 -
val wine quality root mean squared error: 0.5806 - val wine type accuracy: 0.9949
Epoch 45/180
3248 - wine_type_loss: 0.0303 - wine_quality_root_mean_squared_error: 0.5704 - wine_type_accuracy:
0.9946 - val loss: 0.3609 - val wine quality loss: 0.3285 - val wine type loss: 0.0307 -
val_wine_quality_root_mean_squared_error: 0.5743 - val_wine_type_accuracy: 0.9949
Epoch 46/180
3228 - wine type loss: 0.0300 - wine quality root mean squared error: 0.5684 - wine type accuracy:
0.9949 - val loss: 0.3585 - val wine quality loss: 0.3264 - val wine type loss: 0.0306 -
val wine quality root mean squared error: 0.5723 - val wine type accuracy: 0.9949
Epoch 47/180
3202 - wine_type_loss: 0.0299 - wine_quality_root_mean_squared_error: 0.5660 - wine_type_accuracy:
0.9946 - val_loss: 0.3582 - val_wine_quality_loss: 0.3261 - val_wine_type_loss: 0.0305 -
val wine quality root mean squared error: 0.5721 - val wine type accuracy: 0.9949
Epoch 48/180
3194 - wine type loss: 0.0296 - wine quality root mean squared error: 0.5647 - wine type accuracy:
0.9949 - val_loss: 0.3607 - val_wine_quality_loss: 0.3287 - val_wine_type_loss: 0.0303 -
val_wine_quality_root_mean_squared_error: 0.5744 - val wine type accuracy: 0.9949
3166 - wine_type_loss: 0.0294 - wine_quality_root_mean_squared_error: 0.5626 - wine_type_accuracy:
0.9943 - val_loss: 0.3602 - val_wine_quality_loss: 0.3286 - val_wine_type_loss: 0.0302
val_wine_quality_root_mean_squared_error: 0.5741 - val_wine_type_accuracy: 0.9949
Epoch 50/180
3164 - wine_type_loss: 0.0291 - wine_quality_root_mean_squared_error: 0.5625 - wine_type_accuracy:
0.9949 - val loss: 0.3549 - val wine quality loss: 0.3232 - val wine type loss: 0.0302 -
val wine quality root mean squared error: 0.5695 - val wine type accuracy: 0.9949
Epoch 51/180
3150 - wine_type_loss: 0.0290 - wine_quality_root_mean_squared_error: 0.5606 - wine_type_accuracy:
0.9946 - val_loss: 0.3551 - val_wine_quality_loss: 0.3236 - val_wine_type_loss: 0.0301
val_wine_quality_root_mean_squared_error: 0.5698 - val_wine_type_accuracy: 0.9949
Epoch 52/180
3119 - wine_type_loss: 0.0287 - wine_quality_root_mean_squared_error: 0.5585 - wine_type_accuracy:
0.9952 - val_loss: 0.3577 - val_wine_quality_loss: 0.3263 - val_wine_type_loss: 0.0299 -
val wine quality root mean squared error: 0.5722 - val wine type accuracy: 0.9949
Epoch 53/180
3109 - wine_type_loss: 0.0286 - wine_quality_root_mean_squared_error: 0.5573 - wine_type_accuracy:
0.9952 - val_loss: 0.3550 - val_wine_quality_loss: 0.3236 - val_wine_type_loss: 0.0298 -
val wine quality_root_mean_squared_error: 0.5700 - val_wine_type_accuracy: 0.9949
Epoch 54/180
3117 - wine type loss: 0.0284 - wine quality root mean squared error: 0.5573 - wine type accuracy:
0.9952 - val_loss: 0.3543 - val_wine_quality_loss: 0.3230 - val_wine_type_loss: 0.0297 -
val_wine_quality_root_mean_squared_error: 0.5693 - val_wine_type_accuracy: 0.9949
Epoch 55/180
3090 - wine type loss: 0.0281 - wine quality root mean squared error: 0.5562 - wine type accuracy:
0.9952 - val loss: 0.3550 - val wine quality loss: 0.3240 - val wine type loss: 0.0296 -
val_wine_quality_root_mean_squared_error: 0.5701 - val_wine_type_accuracy: 0.9949
Epoch 56/180
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3085 - wine type loss: 0.0281 - wine quality root mean squared error: 0.5551 - wine type accuracy:
0.9952 - val loss: 0.3532 - val wine quality loss: 0.3223 - val wine type loss: 0.0296 -
val wine quality_root_mean_squared_error: 0.5686 - val_wine_type_accuracy: 0.9949
Epoch 57/180
3062 - wine type loss: 0.0278 - wine quality root mean squared error: 0.5534 - wine type accuracy:
0.9949 - val_loss: 0.3565 - val_wine_quality_loss: 0.3255 - val_wine_type_loss: 0.0294 -
val_wine_quality_root_mean_squared_error: 0.5715 - val_wine_type_accuracy: 0.9949
Epoch 58/180
3047 - wine type loss: 0.0276 - wine quality root mean squared error: 0.5522 - wine type accuracy:
0.9952 - val_loss: 0.3613 - val_wine_quality_loss: 0.3302 - val_wine_type_loss: 0.0295 -
val_wine_quality_root_mean_squared_error: 0.5757 - val_wine_type_accuracy: 0.9949
Epoch 59/180
3044 - wine_type_loss: 0.0275 - wine_quality_root_mean_squared_error: 0.5518 - wine_type_accuracy:
0.9952 - val loss: 0.3633 - val wine quality loss: 0.3329 - val wine type loss: 0.0293 -
val_wine_quality_root_mean_squared_error: 0.5776 - val_wine_type_accuracy: 0.9949
Epoch 60/180
3035 - wine_type_loss: 0.0273 - wine_quality_root_mean_squared_error: 0.5511 - wine_type_accuracy:
0.9949 - val loss: 0.3524 - val wine quality loss: 0.3218 - val wine type loss: 0.0293 -
val_wine_quality_root_mean_squared_error: 0.5682 - val_wine_type_accuracy: 0.9949
Epoch 61/180
3024 - wine_type_loss: 0.0271 - wine_quality_root_mean_squared_error: 0.5503 - wine_type_accuracy:
0.9952 - val_loss: 0.3551 - val_wine_quality_loss: 0.3247 - val_wine_type_loss: 0.0292
val_wine_quality_root_mean_squared_error: 0.5706 - val_wine_type_accuracy: 0.9949
Epoch 62/180
3018 - wine type_loss: 0.0269 - wine_quality_root_mean_squared_error: 0.5488 - wine_type_accuracy:
0.9956 - val_loss: 0.3554 - val_wine_quality_loss: 0.3247 - val_wine_type_loss: 0.0292 -
val_wine_quality_root_mean_squared_error: 0.5708 - val_wine_type_accuracy: 0.9949
Epoch 63/180
3009 - wine type loss: 0.0283 - wine quality root mean squared error: 0.5490 - wine type accuracy:
0.9952 - val_loss: 0.3488 - val_wine_quality_loss: 0.3186 - val_wine_type_loss: 0.0290 -
val_wine_quality_root_mean_squared_error: 0.5652 - val_wine_type_accuracy: 0.9949
Epoch 64/180
2999 - wine type loss: 0.0267 - wine quality root mean squared error: 0.5473 - wine type accuracy:
0.9956 - val_loss: 0.3555 - val_wine_quality_loss: 0.3253 - val_wine_type_loss: 0.0291 -
val_wine_quality_root_mean_squared_error: 0.5710 - val_wine_type_accuracy: 0.9949
Epoch 65/180
2985 - wine type loss: 0.0265 - wine quality root mean squared error: 0.5465 - wine type accuracy:
0.9956 - val loss: 0.3589 - val wine quality loss: 0.3285 - val wine type loss: 0.0290 -
val_wine_quality_root_mean_squared_error: 0.5740 - val_wine_type_accuracy: 0.9949
Epoch 66/180
2969 - wine type loss: 0.0264 - wine quality root mean squared error: 0.5451 - wine type accuracy:
0.9952 - val loss: 0.3597 - val wine quality loss: 0.3296 - val wine type loss: 0.0289 -
val_wine_quality_root_mean_squared_error: 0.5748 - val_wine_type_accuracy: 0.9949
Epoch 67/180
2974 - wine_type_loss: 0.0262 - wine_quality_root_mean_squared_error: 0.5446 - wine_type_accuracy:
0.9956 - val_loss: 0.3465 - val_wine_quality_loss: 0.3165 - val_wine_type_loss: 0.0288 -
val_wine_quality_root_mean_squared_error: 0.5633 - val_wine_type_accuracy: 0.9949
Epoch 68/180
2963 - wine type loss: 0.0261 - wine quality root mean squared error: 0.5449 - wine type accuracy:
0.9952 - val_loss: 0.3528 - val_wine_quality_loss: 0.3230 - val_wine_type_loss: 0.0287 -
val wine quality root mean squared error: 0.5689 - val wine type accuracy: 0.9949
Epoch 69/180
2972 - wine type loss: 0.0261 - wine quality root mean squared error: 0.5450 - wine type accuracy:
0.9952 - val loss: 0.3503 - val wine quality loss: 0.3203 - val wine type loss: 0.0287 -
val_wine_quality_root_mean_squared_error: 0.5668 - val_wine_type_accuracy: 0.9949
Epoch 70/180
2958 - wine_type_loss: 0.0258 - wine_quality_root_mean_squared_error: 0.5440 - wine_type_accuracy:
0.9956 - val_loss: 0.3455 - val_wine_quality_loss: 0.3157 - val_wine_type_loss: 0.0287 -
val_wine_quality_root_mean_squared_error: 0.5625 - val_wine_type_accuracy: 0.9949
Epoch 71/180
3155/3155 [============] - 0s 72us/sample - loss: 0.3201 - wine_quality_loss: 0.
2947 - wine_type_loss: 0.0257 - wine_quality_root_mean_squared_error: 0.5425 - wine_type_accuracy:
0.9952 - val loss: 0.3477 - val wine quality loss: 0.3182 - val wine type loss: 0.0286
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val wine quality root mean squared error: 0.5646 - val wine type accuracy: 0.9949
Epoch 72/180
2946 - wine type loss: 0.0256 - wine quality root mean squared error: 0.5431 - wine type accuracy:
0.9952 - val_loss: 0.3486 - val_wine_quality_loss: 0.3190 - val_wine_type_loss: 0.0285
val wine quality root mean squared_error: 0.5655 - val_wine_type_accuracy: 0.9949
Epoch 73/180
2935 - wine type loss: 0.0254 - wine quality root mean squared error: 0.5414 - wine type accuracy:
0.9956 - val_loss: 0.3491 - val_wine_quality_loss: 0.3196 - val_wine_type_loss: 0.0286 -
val_wine_quality_root_mean_squared_error: 0.5659 - val_wine_type_accuracy: 0.9949
Epoch 74/180
2925 - wine type loss: 0.0253 - wine quality root mean squared error: 0.5412 - wine type accuracy:
0.9952 - val_loss: 0.3495 - val_wine_quality_loss: 0.3200 - val_wine_type_loss: 0.0285 -
val_wine_quality_root_mean_squared_error: 0.5662 - val_wine_type_accuracy: 0.9949
Epoch 75/180
2930 - wine type loss: 0.0252 - wine quality_root_mean_squared_error: 0.5414 - wine_type_accuracy:
0.9952 - val loss: 0.3546 - val wine quality loss: 0.3252 - val wine type loss: 0.0284 -
val wine quality root mean squared error: 0.5709 - val wine type accuracy: 0.9949
Epoch 76/180
2930 - wine_type_loss: 0.0252 - wine_quality_root_mean_squared_error: 0.5405 - wine_type_accuracy:
0.9956 - val loss: 0.3553 - val wine quality loss: 0.3259 - val wine type loss: 0.0283 -
val_wine_quality_root_mean_squared_error: 0.5715 - val_wine_type_accuracy: 0.9949
Epoch 77/180
2911 - wine type loss: 0.0250 - wine quality root mean squared error: 0.5394 - wine type accuracy:
0.9956 - val_loss: 0.3473 - val_wine_quality_loss: 0.3181 - val_wine_type_loss: 0.0282 -
val wine quality root mean squared error: 0.5645 - val wine type accuracy: 0.9949
Epoch 78/180
2895 - wine type loss: 0.0261 - wine quality root mean squared error: 0.5384 - wine type accuracy:
0.9956 - val loss: 0.3543 - val_wine_quality_loss: 0.3248 - val_wine_type_loss: 0.0282 -
val_wine_quality_root_mean_squared_error: 0.5708 - val_wine_type_accuracy: 0.9949
Epoch 79/180
2894 - wine_type_loss: 0.0247 - wine_quality_root_mean_squared_error: 0.5381 - wine_type_accuracy:
0.9952 - val_loss: 0.3514 - val_wine_quality_loss: 0.3221 - val_wine_type_loss: 0.0281 -
val_wine_quality_root_mean_squared_error: 0.5683 - val_wine_type_accuracy: 0.9949
Epoch 80/180
2878 - wine_type_loss: 0.0246 - wine_quality_root_mean_squared_error: 0.5367 - wine_type_accuracy:
0.9956 - val_loss: 0.3477 - val_wine_quality_loss: 0.3184 - val_wine_type_loss: 0.0281
val_wine_quality_root_mean_squared_error: 0.5650 - val_wine_type_accuracy: 0.9949
Epoch 81/180
2895 - wine_type_loss: 0.0245 - wine_quality_root_mean_squared_error: 0.5374 - wine_type_accuracy:
0.9952 - val_loss: 0.3436 - val_wine_quality_loss: 0.3147 - val_wine_type_loss: 0.0280 -
val wine quality root mean squared error: 0.5615 - val wine type accuracy: 0.9949
Epoch 82/180
2878 - wine_type_loss: 0.0256 - wine_quality_root_mean_squared_error: 0.5366 - wine_type_accuracy:
0.9952 - val_loss: 0.3474 - val_wine_quality_loss: 0.3185 - val_wine_type_loss: 0.0279 -
val_wine_quality_root_mean_squared_error: 0.5649 - val_wine_type_accuracy: 0.9949
Epoch 83/180
2865 - wine type_loss: 0.0242 - wine_quality_root_mean_squared_error: 0.5359 - wine_type_accuracy:
0.9952 - val_loss: 0.3496 - val_wine_quality_loss: 0.3206 - val_wine_type_loss: 0.0279 -
val wine quality root_mean_squared_error: 0.5669 - val_wine_type_accuracy: 0.9949
Epoch 84/180
2866 - wine type loss: 0.0241 - wine quality root mean squared error: 0.5359 - wine type accuracy:
0.9952 - val_loss: 0.3472 - val_wine_quality_loss: 0.3185 - val_wine_type_loss: 0.0278 -
val_wine_quality_root_mean_squared_error: 0.5649 - val_wine_type_accuracy: 0.9949
Epoch 85/180
3155/3155 [============] - Os 89us/sample - loss: 0.3085 - wine quality loss: 0.
2846 - wine_type_loss: 0.0240 - wine_quality_root_mean_squared_error: 0.5333 - wine_type_accuracy:
0.9956 - val_loss: 0.3466 - val_wine_quality_loss: 0.3180 - val_wine_type_loss: 0.0278 -
val_wine_quality_root_mean_squared_error: 0.5643 - val_wine_type_accuracy: 0.9949
Epoch 86/180
2851 - wine_type_loss: 0.0239 - wine_quality_root_mean_squared_error: 0.5344 - wine_type_accuracy:
0.9956 - val loss: 0.3450 - val wine quality loss: 0.3163 - val wine type loss: 0.0277 -
val_wine_quality_root_mean_squared_error: 0.5630 - val_wine_type_accuracy: 0.9949
Epoch 87/180
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2851 - wine type loss: 0.0241 - wine quality root mean squared error: 0.5340 - wine type accuracy:
0.9956 - val loss: 0.3432 - val wine quality loss: 0.3146 - val wine type loss: 0.0278 -
val_wine_quality_root_mean_squared_error: 0.5613 - val_wine_type_accuracy: 0.9949
Epoch 88/180
2844 - wine type loss: 0.0237 - wine quality root mean squared error: 0.5335 - wine type accuracy:
0.9952 - val loss: 0.3455 - val wine quality loss: 0.3169 - val wine type loss: 0.0276 -
val wine quality root mean squared error: 0.5635 - val wine type accuracy: 0.9949
Epoch 89/180
2842 - wine type loss: 0.0236 - wine quality root mean squared error: 0.5336 - wine type accuracy:
0.9956 - val_loss: 0.3443 - val_wine_quality_loss: 0.3158 - val_wine_type_loss: 0.0276 -
val_wine_quality_root_mean_squared_error: 0.5624 - val_wine_type_accuracy: 0.9949
Epoch 90/180
2819 - wine_type_loss: 0.0235 - wine_quality_root_mean_squared_error: 0.5311 - wine_type_accuracy:
0.9956 - val loss: 0.3449 - val wine quality loss: 0.3164 - val wine type loss: 0.0276 -
val wine quality root mean squared error: 0.5629 - val wine type accuracy: 0.9949
Epoch 91/180
2826 - wine type_loss: 0.0234 - wine_quality_root_mean_squared_error: 0.5312 - wine_type_accuracy:
0.9952 - val_loss: 0.3475 - val_wine_quality_loss: 0.3188 - val_wine_type_loss: 0.0275
val wine quality root mean squared error: 0.5654 - val wine type accuracy: 0.9949
Epoch 92/180
2812 - wine_type_loss: 0.0234 - wine_quality_root_mean_squared_error: 0.5310 - wine_type_accuracy:
0.9952 - val_loss: 0.3437 - val_wine_quality_loss: 0.3153 - val_wine_type_loss: 0.0274 -
val wine quality root mean squared error: 0.5621 - val wine type accuracy: 0.9949
Epoch 93/180
2815 - wine type loss: 0.0232 - wine quality root mean squared error: 0.5306 - wine type accuracy:
0.9956 - val_loss: 0.3414 - val_wine_quality_loss: 0.3131 - val_wine_type_loss: 0.0274 -
val_wine_quality_root_mean_squared_error: 0.5601 - val_wine_type_accuracy: 0.9949
Epoch 94/180
2816 - wine type loss: 0.0231 - wine quality root mean squared error: 0.5306 - wine type accuracy:
0.9956 - val_loss: 0.3457 - val_wine_quality_loss: 0.3174 - val_wine_type_loss: 0.0274 -
val wine quality root mean squared error: 0.5639 - val wine type accuracy: 0.9949
Epoch 95/180
2810 - wine type loss: 0.0230 - wine quality root mean squared error: 0.5308 - wine type accuracy:
0.9952 - val_loss: 0.3448 - val_wine_quality_loss: 0.3166 - val_wine_type_loss: 0.0273 -
val_wine_quality_root_mean_squared_error: 0.5631 - val_wine_type_accuracy: 0.9949
Epoch 96/180
2801 - wine type loss: 0.0241 - wine quality root mean squared error: 0.5293 - wine type accuracy:
0.9952 - val loss: 0.3440 - val wine quality loss: 0.3158 - val wine type loss: 0.0272 -
val_wine_quality_root_mean_squared_error: 0.5626 - val_wine_type_accuracy: 0.9949
Epoch 97/180
2782 - wine type loss: 0.0230 - wine quality root mean squared error: 0.5272 - wine type accuracy:
0.9956 - val loss: 0.3529 - val_wine_quality_loss: 0.3248 - val_wine_type_loss: 0.0273 -
val wine quality root mean squared error: 0.5703 - val wine type accuracy: 0.9949
Epoch 98/180
2800 - wine type loss: 0.0227 - wine quality root mean squared error: 0.5292 - wine type accuracy:
0.9956 - val_loss: 0.3456 - val_wine_quality_loss: 0.3177 - val_wine_type_loss: 0.0272 -
val wine quality root mean squared error: 0.5640 - val wine type accuracy: 0.9949
Epoch 99/180
2793 - wine type loss: 0.0226 - wine quality root mean squared error: 0.5285 - wine type accuracy:
0.9956 - val_loss: 0.3521 - val_wine_quality_loss: 0.3241 - val_wine_type_loss: 0.0271 -
val wine quality root mean squared error: 0.5698 - val wine type accuracy: 0.9949
Epoch 100/180
2794 - wine type loss: 0.0225 - wine quality root mean squared error: 0.5288 - wine type accuracy:
0.9956 - val loss: 0.3426 - val wine quality loss: 0.3146 - val wine type loss: 0.0271 -
val_wine_quality_root_mean_squared_error: 0.5614 - val_wine_type_accuracy: 0.9949
Epoch 101/180
2778 - wine_type_loss: 0.0225 - wine_quality_root_mean_squared_error: 0.5273 - wine_type_accuracy:
0.9956 - val_loss: 0.3399 - val_wine_quality_loss: 0.3120 - val_wine_type_loss: 0.0270 -
val_wine_quality_root_mean_squared_error: 0.5590 - val_wine_type_accuracy: 0.9949
Epoch 102/180
2762 - wine_type_loss: 0.0223 - wine_quality_root_mean_squared_error: 0.5261 - wine_type_accuracy:
```

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0.9959 - val_loss: 0.3451 - val_wine_quality_loss: 0.3172 - val_wine_type_loss: 0.0270 -
val wine quality root mean squared error: 0.5637 - val wine type accuracy: 0.9949
Epoch 103/180
2772 - wine type loss: 0.0222 - wine quality root mean squared error: 0.5262 - wine type accuracy:
0.9959 - val_loss: 0.3440 - val_wine_quality_loss: 0.3160 - val_wine_type_loss: 0.0270 -
val wine quality root mean squared error: 0.5627 - val wine type accuracy: 0.9949
Epoch 104/180
2753 - wine type loss: 0.0222 - wine quality root mean squared error: 0.5253 - wine type accuracy:
0.9959 - val_loss: 0.3456 - val_wine_quality_loss: 0.3177 - val_wine_type_loss: 0.0269 -
val wine quality_root_mean_squared_error: 0.5642 - val_wine_type_accuracy: 0.9949
Epoch 105/180
2757 - wine_type_loss: 0.0221 - wine_quality_root_mean_squared_error: 0.5256 - wine_type_accuracy:
0.9959 - val loss: 0.3510 - val wine quality loss: 0.3235 - val wine type loss: 0.0269 -
val_wine_quality_root_mean_squared_error: 0.5689 - val_wine_type_accuracy: 0.9949
Epoch 106/180
2764 - wine type loss: 0.0220 - wine quality root mean squared error: 0.5258 - wine type accuracy:
0.9959 - val loss: 0.3436 - val wine quality loss: 0.3159 - val wine type loss: 0.0269 -
val wine quality root mean squared error: 0.5625 - val wine type accuracy: 0.9949
Epoch 107/180
2765 - wine_type_loss: 0.0219 - wine_quality_root_mean_squared_error: 0.5250 - wine_type_accuracy:
0.9959 - val loss: 0.3485 - val wine quality loss: 0.3209 - val wine type loss: 0.0269 -
val_wine_quality_root_mean_squared_error: 0.5669 - val_wine_type_accuracy: 0.9949
Epoch 108/180
2741 - wine_type_loss: 0.0218 - wine_quality_root_mean_squared_error: 0.5231 - wine_type_accuracy:
0.9959 - val_loss: 0.3455 - val_wine_quality_loss: 0.3180 - val_wine_type_loss: 0.0268 -
val wine quality root mean squared error: 0.5642 - val wine type accuracy: 0.9949
Epoch 109/180
2743 - wine type loss: 0.0217 - wine quality root mean squared error: 0.5242 - wine type accuracy:
0.9959 - val loss: 0.3428 - val wine quality loss: 0.3154 - val wine type loss: 0.0268 -
val_wine_quality_root_mean_squared_error: 0.5618 - val_wine_type_accuracy: 0.9949
Epoch 110/180
2736 - wine_type_loss: 0.0217 - wine_quality_root_mean_squared_error: 0.5233 - wine_type_accuracy:
0.9959 - val_loss: 0.3394 - val_wine_quality_loss: 0.3119 - val_wine_type_loss: 0.0268 -
val_wine_quality_root_mean_squared_error: 0.5589 - val_wine_type_accuracy: 0.9949
Epoch 111/180
2741 - wine_type_loss: 0.0216 - wine_quality_root_mean_squared_error: 0.5232 - wine_type_accuracy:
0.9959 - val loss: 0.3450 - val wine quality loss: 0.3176 - val wine type loss: 0.0267
val_wine_quality_root_mean_squared_error: 0.5639 - val_wine_type_accuracy: 0.9949
Epoch 112/180
2738 - wine_type_loss: 0.0215 - wine_quality_root_mean_squared_error: 0.5229 - wine_type_accuracy:
0.9959 - val_loss: 0.3495 - val_wine_quality_loss: 0.3220 - val_wine_type_loss: 0.0267
val wine quality_root_mean_squared_error: 0.5679 - val_wine_type_accuracy: 0.9949
Epoch 113/180
2729 - wine_type_loss: 0.0214 - wine_quality_root_mean_squared_error: 0.5225 - wine_type_accuracy:
0.9962 - val_loss: 0.3491 - val_wine_quality_loss: 0.3214 - val_wine_type_loss: 0.0266 -
val_wine_quality_root_mean_squared_error: 0.5675 - val_wine_type_accuracy: 0.9949
Epoch 114/180
2720 - wine_type_loss: 0.0214 - wine_quality_root_mean_squared_error: 0.5214 - wine_type_accuracy:
0.9959 - val_loss: 0.3418 - val_wine_quality_loss: 0.3146 - val_wine_type_loss: 0.0265 -
val_wine_quality_root_mean_squared_error: 0.5612 - val_wine_type_accuracy: 0.9949
Epoch 115/180
2731 - wine type loss: 0.0212 - wine quality root mean squared error: 0.5225 - wine type accuracy:
0.9959 - val_loss: 0.3407 - val_wine_quality_loss: 0.3134 - val_wine_type_loss: 0.0266 -
val_wine_quality_root_mean_squared_error: 0.5602 - val_wine_type_accuracy: 0.9949
Epoch 116/180
2723 - wine_type_loss: 0.0214 - wine_quality_root_mean_squared_error: 0.5219 - wine type accuracy:
0.9962 - val_loss: 0.3413 - val_wine_quality_loss: 0.3139 - val_wine_type_loss: 0.0265 -
val_wine_quality_root_mean_squared_error: 0.5608 - val_wine_type_accuracy: 0.9949
Epoch 117/180
2716 - wine_type_loss: 0.0211 - wine_quality_root_mean_squared_error: 0.5208 - wine_type_accuracy:
0.9959 - val loss: 0.3592 - val wine quality loss: 0.3319 - val wine type loss: 0.0265 -
val wine quality root mean squared error: 0.5765 - val wine type accuracy: 0.9949
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Epoch 118/180
2708 - wine type_loss: 0.0210 - wine_quality_root_mean_squared_error: 0.5207 - wine_type_accuracy:
0.9959 - val_loss: 0.3414 - val_wine_quality_loss: 0.3141 - val_wine_type_loss: 0.0265 -
val wine quality root mean squared error: 0.5608 - val wine type accuracy: 0.9949
Epoch 119/180
2713 - wine_type_loss: 0.0209 - wine_quality_root_mean_squared_error: 0.5208 - wine_type_accuracy:
0.9959 - val loss: 0.3442 - val wine quality loss: 0.3168 - val wine type loss: 0.0264 -
val wine quality root mean squared error: 0.5634 - val wine type accuracy: 0.9949
Epoch 120/180
2709 - wine_type_loss: 0.0209 - wine_quality_root_mean_squared_error: 0.5204 - wine_type_accuracy:
0.9962 - val_loss: 0.3380 - val_wine_quality_loss: 0.3109 - val_wine_type_loss: 0.0264 -
val_wine_quality_root_mean_squared_error: 0.5579 - val_wine_type_accuracy: 0.9949
Epoch 121/180
2690 - wine_type_loss: 0.0209 - wine_quality_root_mean_squared_error: 0.5190 - wine_type_accuracy:
0.9959 - val_loss: 0.3380 - val_wine_quality_loss: 0.3108 - val_wine_type_loss: 0.0264 -
val wine quality root mean squared error: 0.5579 - val wine type accuracy: 0.9949
Epoch 122/180
2694 - wine_type_loss: 0.0207 - wine_quality_root_mean_squared_error: 0.5192 - wine_type_accuracy:
0.9962 - val_loss: 0.3548 - val_wine_quality_loss: 0.3274 - val_wine_type_loss: 0.0264
val wine quality root mean squared error: 0.5728 - val wine type accuracy: 0.9949
Epoch 123/180
2693 - wine_type_loss: 0.0207 - wine_quality_root_mean_squared_error: 0.5188 - wine_type_accuracy:
0.9962 - val_loss: 0.3604 - val_wine_quality_loss: 0.3332 - val_wine_type_loss: 0.0264
val wine quality root mean squared error: 0.5776 - val wine type accuracy: 0.9949
Epoch 124/180
2691 - wine type loss: 0.0206 - wine quality root mean squared error: 0.5187 - wine type accuracy:
0.9962 - val_loss: 0.3477 - val_wine_quality_loss: 0.3202 - val_wine_type_loss: 0.0264 -
val_wine_quality_root_mean_squared_error: 0.5665 - val_wine_type_accuracy: 0.9949
Epoch 125/180
2668 - wine type loss: 0.0206 - wine quality root mean squared error: 0.5166 - wine type accuracy:
0.9962 - val_loss: 0.3473 - val_wine_quality_loss: 0.3202 - val_wine_type_loss: 0.0263 -
val wine quality root mean squared error: 0.5663 - val wine type accuracy: 0.9949
Epoch 126/180
2682 - wine type loss: 0.0205 - wine quality root mean squared error: 0.5182 - wine type accuracy:
0.9962 - val_loss: 0.3487 - val_wine_quality_loss: 0.3213 - val_wine_type_loss: 0.0264 -
val_wine_quality_root_mean_squared_error: 0.5674 - val_wine_type_accuracy: 0.9949
Epoch 127/180
3155/3155 [=============] - Os 90us/sample - loss: 0.2884 - wine_quality_loss: 0.
2676 - wine type loss: 0.0216 - wine quality root mean squared error: 0.5176 - wine type accuracy:
0.9962 - val loss: 0.3392 - val wine quality loss: 0.3123 - val wine type loss: 0.0263 -
val_wine_quality_root_mean_squared_error: 0.5591 - val_wine_type_accuracy: 0.9949
Epoch 128/180
2661 - wine type loss: 0.0203 - wine quality root mean squared error: 0.5161 - wine type accuracy:
0.9962 - val loss: 0.3540 - val wine quality loss: 0.3271 - val wine type loss: 0.0262 -
val wine quality root mean squared error: 0.5722 - val wine type accuracy: 0.9949
Epoch 129/180
2669 - wine_type_loss: 0.0202 - wine_quality_root_mean_squared_error: 0.5164 - wine_type_accuracy:
0.9962 - val_loss: 0.3405 - val_wine_quality_loss: 0.3137 - val_wine_type_loss: 0.0262 -
val_wine_quality_root_mean_squared_error: 0.5604 - val_wine_type_accuracy: 0.9949
Epoch 130/180
2662 - wine_type_loss: 0.0201 - wine_quality_root_mean_squared_error: 0.5160 - wine_type_accuracy:
0.9962 - val loss: 0.3395 - val wine quality loss: 0.3127 - val wine type loss: 0.0262 -
val_wine_quality_root_mean_squared_error: 0.5594 - val_wine_type_accuracy: 0.9949
Epoch 131/180
2653 - wine type loss: 0.0201 - wine quality root mean squared error: 0.5155 - wine type accuracy:
0.9962 - val loss: 0.3418 - val wine quality loss: 0.3148 - val wine type loss: 0.0261 -
val_wine_quality_root_mean_squared_error: 0.5616 - val_wine_type_accuracy: 0.9949
Epoch 132/180
2658 - wine_type_loss: 0.0200 - wine_quality_root_mean_squared_error: 0.5155 - wine_type_accuracy:
0.9962 - val_loss: 0.3470 - val_wine_quality_loss: 0.3203 - val_wine_type_loss: 0.0261 -
val_wine_quality_root_mean_squared_error: 0.5662 - val_wine_type_accuracy: 0.9949
Epoch 133/180
```

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2660 - wine_type_loss: 0.0199 - wine_quality_root_mean_squared_error: 0.5160 - wine_type accuracy:
0.9962 - val_loss: 0.3399 - val_wine_quality_loss: 0.3131 - val_wine_type_loss: 0.0261 -
val wine quality root mean squared error: 0.5599 - val wine type accuracy: 0.9949
Epoch 134/180
2667 - wine type loss: 0.0198 - wine quality root mean squared error: 0.5160 - wine type accuracy:
0.9962 - val_loss: 0.3501 - val_wine_quality_loss: 0.3230 - val_wine_type_loss: 0.0261 -
val wine quality root mean squared error: 0.5689 - val wine type accuracy: 0.9949
Epoch 135/180
2649 - wine type loss: 0.0198 - wine quality root mean squared error: 0.5147 - wine type accuracy:
0.9962 - val_loss: 0.3512 - val_wine_quality_loss: 0.3241 - val_wine_type_loss: 0.0261 -
val_wine_quality_root_mean_squared_error: 0.5698 - val_wine_type_accuracy: 0.9949
Epoch 136/180
2647 - wine_type_loss: 0.0197 - wine_quality_root_mean_squared_error: 0.5146 - wine_type_accuracy:
0.9962 - val loss: 0.3406 - val wine quality loss: 0.3136 - val wine type loss: 0.0261 -
val_wine_quality_root_mean_squared_error: 0.5605 - val_wine_type_accuracy: 0.9949
Epoch 137/180
2649 - wine_type_loss: 0.0197 - wine_quality_root_mean_squared_error: 0.5142 - wine_type_accuracy:
0.9962 - val loss: 0.3479 - val wine quality loss: 0.3210 - val wine type loss: 0.0261 -
val wine quality_root_mean_squared_error: 0.5669 - val_wine_type_accuracy: 0.9949
Epoch 138/180
2630 - wine type loss: 0.0196 - wine quality root mean squared error: 0.5125 - wine type accuracy:
0.9962 - val_loss: 0.3446 - val_wine_quality_loss: 0.3175 - val_wine_type_loss: 0.0261 -
val_wine_quality_root_mean_squared_error: 0.5640 - val_wine_type_accuracy: 0.9949
Epoch 139/180
2636 - wine_type_loss: 0.0195 - wine_quality_root_mean_squared_error: 0.5133 - wine_type_accuracy:
0.9962 - val_loss: 0.3444 - val_wine_quality_loss: 0.3175 - val_wine_type_loss: 0.0261 -
val_wine_quality_root_mean_squared_error: 0.5639 - val_wine_type_accuracy: 0.9949
Epoch 140/180
2634 - wine type loss: 0.0195 - wine quality root mean squared error: 0.5133 - wine type accuracy:
0.9962 - val loss: 0.3502 - val wine quality loss: 0.3233 - val wine type loss: 0.0261 -
val_wine_quality_root_mean_squared_error: 0.5690 - val_wine_type_accuracy: 0.9949
Epoch 141/180
2616 - wine_type_loss: 0.0194 - wine_quality_root_mean_squared_error: 0.5118 - wine_type_accuracy:
0.9962 - val_loss: 0.3438 - val_wine_quality_loss: 0.3169 - val_wine_type_loss: 0.0261 -
val_wine_quality_root_mean_squared_error: 0.5633 - val_wine_type_accuracy: 0.9949
Epoch 142/180
2616 - wine_type_loss: 0.0195 - wine_quality_root_mean_squared_error: 0.5111 - wine_type_accuracy:
0.9962 - val_loss: 0.3421 - val_wine_quality_loss: 0.3157 - val_wine_type_loss: 0.0261 -
val wine quality root mean squared error: 0.5619 - val wine type accuracy: 0.9949
Epoch 143/180
2621 - wine_type_loss: 0.0207 - wine_quality_root_mean_squared_error: 0.5116 - wine_type_accuracy:
0.9962 - val_loss: 0.3441 - val_wine_quality_loss: 0.3174 - val_wine_type_loss: 0.0261 -
val wine quality root mean squared error: 0.5636 - val wine type accuracy: 0.9949
Epoch 144/180
2621 - wine_type_loss: 0.0192 - wine_quality_root_mean_squared_error: 0.5118 - wine_type_accuracy:
0.9962 - val_loss: 0.3425 - val_wine_quality_loss: 0.3156 - val_wine_type_loss: 0.0260 -
val_wine_quality_root_mean_squared_error: 0.5622 - val_wine_type_accuracy: 0.9949
Epoch 145/180
2611 - wine type loss: 0.0191 - wine quality root mean squared error: 0.5111 - wine type accuracy:
0.9962 - val_loss: 0.3461 - val_wine_quality_loss: 0.3196 - val_wine_type_loss: 0.0260 -
val_wine_quality_root_mean_squared_error: 0.5655 - val_wine_type_accuracy: 0.9949
Epoch 146/180
2604 - wine type loss: 0.0191 - wine quality root mean squared error: 0.5102 - wine type accuracy:
0.9962 - val_loss: 0.3511 - val_wine_quality_loss: 0.3242 - val_wine_type_loss: 0.0260 -
val_wine_quality_root_mean_squared_error: 0.5699 - val_wine_type_accuracy: 0.9949
Epoch 147/180
2616 - wine type loss: 0.0191 - wine quality root mean squared error: 0.5111 - wine type accuracy:
0.9962 - val_loss: 0.3402 - val_wine_quality_loss: 0.3136 - val_wine_type_loss: 0.0260 -
val_wine_quality_root_mean_squared_error: 0.5602 - val_wine_type_accuracy: 0.9949
Epoch 148/180
2602 - wine_type_loss: 0.0189 - wine_quality_root_mean_squared_error: 0.5098 - wine_type_accuracy:
```

0.9962 - val loss: 0.3537 - val wine quality loss: 0.3270 - val wine type loss: 0.0260 -

```
val wine quality root mean squared error: 0.5722 - val wine type accuracy: 0.9949
Epoch 149/180
2603 - wine type loss: 0.0189 - wine quality root mean squared error: 0.5097 - wine type accuracy:
0.9962 - val_loss: 0.3401 - val_wine_quality_loss: 0.3132 - val_wine_type_loss: 0.0261 -
val wine quality root mean squared error: 0.5601 - val wine type accuracy: 0.9949
Epoch 150/180
2597 - wine_type_loss: 0.0188 - wine_quality_root_mean_squared_error: 0.5092 - wine_type_accuracy:
0.9962 - val loss: 0.3414 - val wine quality loss: 0.3146 - val wine type loss: 0.0261 -
val_wine_quality_root_mean_squared_error: 0.5613 - val wine type accuracy: 0.9949
Epoch 151/180
2590 - wine_type_loss: 0.0188 - wine_quality_root_mean_squared_error: 0.5087 - wine_type_accuracy:
0.9962 - val_loss: 0.3515 - val_wine_quality_loss: 0.3248 - val_wine_type_loss: 0.0260
val_wine_quality_root_mean_squared_error: 0.5702 - val_wine_type_accuracy: 0.9949
Epoch 152/180
2585 - wine_type_loss: 0.0187 - wine_quality_root_mean_squared_error: 0.5081 - wine_type_accuracy:
0.9962 - val_loss: 0.3455 - val_wine_quality_loss: 0.3186 - val_wine_type_loss: 0.0260 -
val wine quality root_mean_squared_error: 0.5649 - val_wine_type_accuracy: 0.9949
Epoch 153/180
2582 - wine_type_loss: 0.0187 - wine_quality_root_mean_squared_error: 0.5085 - wine_type_accuracy:
0.9962 - val_loss: 0.3395 - val_wine_quality_loss: 0.3128 - val_wine_type_loss: 0.0260
val wine quality root mean squared error: 0.5596 - val wine type accuracy: 0.9949
Epoch 154/180
2578 - wine_type_loss: 0.0186 - wine_quality_root_mean_squared_error: 0.5081 - wine_type_accuracy:
0.9962 - val_loss: 0.3462 - val_wine_quality_loss: 0.3195 - val_wine_type_loss: 0.0260 -
val wine quality root mean squared error: 0.5656 - val wine type accuracy: 0.9949
Epoch 155/180
2590 - wine type loss: 0.0186 - wine quality root mean squared error: 0.5087 - wine type accuracy:
0.9962 - val_loss: 0.3427 - val_wine_quality_loss: 0.3159 - val_wine_type_loss: 0.0260 -
val_wine_quality_root_mean_squared_error: 0.5624 - val_wine_type_accuracy: 0.9949
Epoch 156/180
2578 - wine type loss: 0.0185 - wine quality root mean squared error: 0.5080 - wine type accuracy:
0.9962 - val loss: 0.3428 - val wine quality loss: 0.3161 - val wine type loss: 0.0260 -
val_wine_quality_root_mean_squared_error: 0.5626 - val_wine_type_accuracy: 0.9949
Epoch 157/180
2577 - wine_type_loss: 0.0185 - wine_quality_root_mean_squared_error: 0.5073 - wine_type_accuracy:
0.9962 - val loss: 0.3391 - val wine quality loss: 0.3123 - val wine type loss: 0.0260 -
val_wine_quality_root_mean_squared_error: 0.5592 - val_wine_type_accuracy: 0.9949
Epoch 158/180
2567 - wine_type_loss: 0.0183 - wine_quality_root_mean_squared_error: 0.5064 - wine_type_accuracy:
0.9962 - val loss: 0.3422 - val wine quality loss: 0.3155 - val wine type loss: 0.0261 -
val_wine_quality_root_mean_squared_error: 0.5620 - val_wine_type_accuracy: 0.9949
Epoch 159/180
2558 - wine type loss: 0.0183 - wine quality root mean squared error: 0.5058 - wine type accuracy:
0.9962 - val loss: 0.3482 - val wine quality loss: 0.3212 - val wine type loss: 0.0261 -
val wine quality root mean squared error: 0.5672 - val wine type accuracy: 0.9949
Epoch 160/180
2560 - wine_type_loss: 0.0183 - wine_quality_root_mean_squared_error: 0.5067 - wine_type_accuracy:
0.9962 - val_loss: 0.3372 - val_wine_quality_loss: 0.3105 - val_wine_type_loss: 0.0261 -
val_wine_quality_root_mean_squared_error: 0.5575 - val_wine_type_accuracy: 0.9949
Epoch 161/180
2564 - wine_type_loss: 0.0182 - wine_quality_root_mean_squared_error: 0.5063 - wine_type_accuracy:
0.9962 - val loss: 0.3439 - val wine quality loss: 0.3170 - val wine type loss: 0.0261 -
val_wine_quality_root_mean_squared_error: 0.5634 - val_wine_type_accuracy: 0.9949
Epoch 162/180
2559 - wine_type_loss: 0.0182 - wine_quality_root_mean_squared_error: 0.5058 - wine_type_accuracy:
0.9962 - val_loss: 0.3469 - val_wine_quality_loss: 0.3200 - val_wine_type_loss: 0.0261 -
val_wine_quality_root_mean_squared_error: 0.5661 - val_wine_type_accuracy: 0.9949
Epoch 163/180
2546 - wine_type_loss: 0.0181 - wine_quality_root_mean_squared_error: 0.5047 - wine_type_accuracy:
0.9962 - val_loss: 0.3527 - val_wine_quality_loss: 0.3258 - val_wine_type_loss: 0.0260 -
val_wine_quality_root_mean_squared_error: 0.5713 - val_wine_type_accuracy: 0.9949
```

Epoch 164/180

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2545 - wine type_loss: 0.0181 - wine_quality_root_mean_squared_error: 0.5048 - wine_type_accuracy:
0.9962 - val_loss: 0.3422 - val_wine_quality_loss: 0.3154 - val_wine_type_loss: 0.0260 -
val_wine_quality_root_mean_squared_error: 0.5620 - val_wine_type_accuracy: 0.9949
Epoch 165/180
2554 - wine type loss: 0.0180 - wine quality root mean squared error: 0.5054 - wine type accuracy:
0.9962 - val_loss: 0.3482 - val_wine_quality_loss: 0.3215 - val_wine_type_loss: 0.0260 -
val wine quality root mean squared error: 0.5674 - val wine type accuracy: 0.9949
Epoch 166/180
2542 - wine type loss: 0.0180 - wine quality root mean squared error: 0.5039 - wine type accuracy:
0.9962 - val loss: 0.3491 - val wine quality loss: 0.3222 - val wine type loss: 0.0259 -
val_wine_quality_root_mean_squared_error: 0.5682 - val_wine_type_accuracy: 0.9949
Epoch 167/180
2538 - wine_type_loss: 0.0179 - wine_quality_root_mean_squared_error: 0.5038 - wine_type_accuracy:
0.9962 - val loss: 0.3403 - val wine quality loss: 0.3134 - val wine type loss: 0.0261 -
val_wine_quality_root_mean_squared_error: 0.5603 - val_wine_type_accuracy: 0.9949
Epoch 168/180
2537 - wine_type_loss: 0.0178 - wine_quality_root_mean_squared_error: 0.5032 - wine_type_accuracy:
0.9962 - val loss: 0.3508 - val wine quality loss: 0.3240 - val wine type loss: 0.0261 -
val wine quality root mean squared error: 0.5696 - val wine type accuracy: 0.9949
Epoch 169/180
2532 - wine type loss: 0.0178 - wine quality root mean squared error: 0.5035 - wine type accuracy:
0.9962 - val loss: 0.3627 - val wine quality loss: 0.3358 - val wine type loss: 0.0260 -
val_wine_quality_root_mean_squared_error: 0.5799 - val_wine_type_accuracy: 0.9949
Epoch 170/180
2539 - wine type loss: 0.0178 - wine quality root mean squared error: 0.5038 - wine type accuracy:
0.9962 - val_loss: 0.3461 - val_wine_quality_loss: 0.3193 - val_wine_type_loss: 0.0260 -
val_wine_quality_root_mean_squared_error: 0.5654 - val_wine_type_accuracy: 0.9937
Epoch 171/180
2525 - wine type loss: 0.0176 - wine quality root mean squared error: 0.5025 - wine type accuracy:
0.9962 - val_loss: 0.3524 - val_wine_quality_loss: 0.3252 - val_wine_type_loss: 0.0262 -
val_wine_quality_root_mean_squared_error: 0.5708 - val_wine_type_accuracy: 0.9937
Epoch 172/180
2531 - wine_type_loss: 0.0176 - wine_quality_root_mean_squared_error: 0.5034 - wine_type_accuracy:
0.9962 - val_loss: 0.3427 - val_wine_quality_loss: 0.3160 - val_wine_type_loss: 0.0260 -
val_wine_quality_root_mean_squared_error: 0.5625 - val_wine_type_accuracy: 0.9949
Epoch 173/180
2520 - wine_type_loss: 0.0176 - wine_quality_root_mean_squared_error: 0.5019 - wine_type_accuracy:
0.9962 - val_loss: 0.3471 - val_wine_quality_loss: 0.3202 - val_wine_type_loss: 0.0261
val_wine_quality_root_mean_squared_error: 0.5663 - val_wine_type_accuracy: 0.9949
Epoch 174/180
2516 - wine_type_loss: 0.0175 - wine_quality_root_mean_squared_error: 0.5021 - wine_type_accuracy:
0.9962 - val_loss: 0.3415 - val_wine_quality_loss: 0.3148 - val_wine_type_loss: 0.0260
val wine quality root mean squared error: 0.5614 - val wine type accuracy: 0.9949
Epoch 175/180
2514 - wine type loss: 0.0174 - wine_quality_root_mean_squared_error: 0.5012 - wine_type_accuracy:
0.9962 - val_loss: 0.3540 - val_wine_quality_loss: 0.3273 - val_wine_type_loss: 0.0260 -
val_wine_quality_root_mean_squared_error: 0.5724 - val_wine_type_accuracy: 0.9937
Epoch 176/180
2517 - wine type loss: 0.0174 - wine quality root mean squared error: 0.5019 - wine type accuracy:
0.9962 - val_loss: 0.3446 - val_wine_quality_loss: 0.3176 - val_wine_type_loss: 0.0261 -
val_wine_quality_root_mean_squared_error: 0.5641 - val_wine_type_accuracy: 0.9937
Epoch 177/180
2514 - wine type loss: 0.0174 - wine quality root mean squared error: 0.5012 - wine type accuracy:
0.9962 - val_loss: 0.3435 - val_wine_quality_loss: 0.3167 - val_wine_type_loss: 0.0260 -
val_wine_quality_root_mean_squared_error: 0.5631 - val_wine_type_accuracy: 0.9937
Epoch 178/180
2502 - wine_type_loss: 0.0173 - wine_quality_root_mean_squared error: 0.5006 - wine type accuracy:
0.9962 - val_loss: 0.3412 - val_wine_quality_loss: 0.3142 - val_wine_type_loss: 0.0261 -
val_wine_quality_root_mean_squared_error: 0.5610 - val_wine_type_accuracy: 0.9937
Epoch 179/180
```

2510 - wine type loss: 0.0173 - wine quality root mean squared error: 0.5012 - wine type accuracy:

```
0.9962 - val_loss: 0.3440 - val_wine_quality_loss: 0.3171 - val_wine_type_loss: 0.0260 -
                                                                          _0160_00001001.
val wine quality root mean squared error: 0.5636 - val wine type accuracy: 0.9937
Epoch 180/180
2505 - wine type loss: 0.0189 - wine quality root mean squared error: 0.5005 - wine type accuracy:
0.9962 - val_loss: 0.3397 - val_wine_quality_loss: 0.3129 - val_wine_type_loss: 0.0260 -
val_wine_quality_root_mean_squared_error: 0.5597 - val_wine_type_accuracy: 0.9937
In [43]:
utils.test history(history)
All public tests passed
In [44]:
# Gather the training metrics
loss, wine quality loss, wine type loss, wine quality rmse, wine type accuracy = model.evaluate(x=n
orm_val_X, y=val_Y)
print()
print(f'loss: {loss}')
print(f'wine quality loss: {wine quality loss}')
print(f'wine_type_loss: {wine_type_loss}')
print(f'wine_quality_rmse: {wine_quality_rmse}')
print(f'wine type accuracy: {wine type accuracy}')
# EXPECTED VALUES
# ~ 0.30 - 0.38
# ~ 0.30 - 0.38
# ~ 0.018 - 0.030
# ~ 0.50 - 0.62
# ~ 0.97 - 1.0
# Example:
#0.3657050132751465
#0.3463745415210724
#0.019330406561493874
#0.5885359048843384
#0.9974651336669922
29 - wine type loss: 0.0260 - wine quality root mean squared error: 0.5597 - wine type accuracy: 0
.9937
loss: 0.339655585766443
wine_quality_loss: 0.31292524933815
wine_type_loss: 0.026018548756837845
wine_quality_rmse: 0.5597356557846069
wine_type_accuracy: 0.9936628937721252
Analyze the Model Performance
```

Note that the model has two outputs. The output at index 0 is quality and index 1 is wine type

So, round the quality predictions to the nearest integer.

```
In [45]:
```

```
predictions = model.predict(norm_test_X)
quality_pred = predictions[0]
type_pred = predictions[1]
```

```
In [46]:
```

```
print(quality_pred[0])
# EXPECTED OUTPUT
# 5.6 - 6.0
```

```
[5.7119555]
In [47]:
print(type_pred[0])
print(type_pred[944])
# EXPECTED OUTPUT
# A number close to zero
# A number close to or equal to 1

[0.00012767]
[0.999998]
```

Plot Utilities

We define a few utilities to visualize the model performance.

```
In [48]:
```

```
def plot_metrics(metric_name, title, ylim=5):
    plt.title(title)
    plt.ylim(0,ylim)
    plt.plot(history.history[metric_name],color='blue',label=metric_name)
    plt.plot(history.history['val_' + metric_name],color='green',label='val_' + metric_name)
```

In [49]:

```
def plot confusion matrix(y true, y pred, title='', labels=[0,1]):
   cm = confusion matrix(y true, y pred)
   fig = plt.figure()
   ax = fig.add_subplot(111)
   cax = ax.matshow(cm)
    plt.title('Confusion matrix of the classifier')
   fig.colorbar(cax)
   ax.set xticklabels([''] + labels)
   ax.set_yticklabels([''] + labels)
   plt.xlabel('Predicted')
    plt.ylabel('True')
   fmt = 'd'
   thresh = cm.max() / 2.
    for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
          plt.text(j, i, format(cm[i, j], fmt),
                  horizontalalignment="center",
                  color="black" if cm[i, j] > thresh else "white")
    plt.show()
```

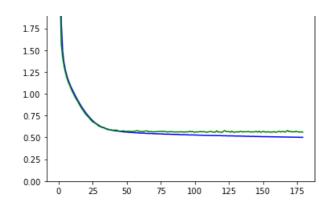
In [50]:

```
def plot_diff(y_true, y_pred, title = ''):
    plt.scatter(y_true, y_pred)
    plt.title(title)
    plt.xlabel('True Values')
    plt.ylabel('Predictions')
    plt.axis('equal')
    plt.axis('square')
    plt.plot([-100, 100], [-100, 100])
    return plt
```

Plots for Metrics

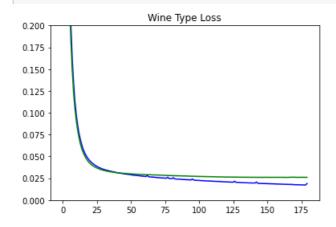
```
In [51]:
```

```
plot_metrics('wine_quality_root_mean_squared_error', 'RMSE', ylim=2)
```



In [52]:

```
plot_metrics('wine_type_loss', 'Wine Type Loss', ylim=0.2)
```

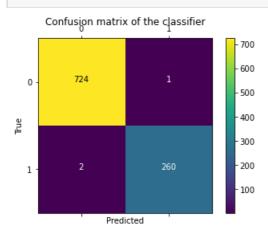


Plots for Confusion Matrix

Plot the confusion matrices for wine type. You can see that the model performs well for prediction of wine type from the confusion matrix and the loss metrics.

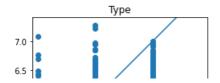
In [53]:

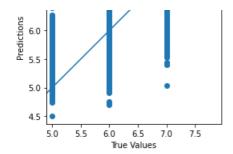
```
plot_confusion_matrix(test_Y[1], np.round(type_pred), title='Wine Type', labels = [0, 1])
```



In [54]:

```
scatter_plot = plot_diff(test_Y[0], quality_pred, title='Type')
```





In []: