In [1]:
```
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
import os
import tensorflow as tf
from tensorflow.keras.layers import Dense,Flatten
from tensorflow.keras.models import Sequential
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow import keras
from tensorflow.keras.callbacks import Callback
from tensorflow.keras.callbacks import ModelCheckpoint, Callback, EarlyStopping
from tensorflow.keras.callbacks import ModelCheckpoint
```

In [2]:
```
img_shape = (224, 224, 3)
resnet_model = Sequential()
pretrained_model= tf.keras.applications.ResNet50(include_top=False,
                   input_shape=img_shape,
                   weights='imagenet')
for layer in pretrained_model.layers:
        layer.trainable=False
```

In [3]:
```
classes = [
    'Class1.1', 'Class1.2', 'Class1.3', 'Class2.1', 'Class2.2', 'Class3.1',
    'Class3.2', 'Class4.1', 'Class4.2', 'Class5.1', 'Class5.2', 'Class5.3',
    'Class5.4', 'Class6.1', 'Class6.2', 'Class7.1', 'Class7.2', 'Class7.3',
    'Class8.1', 'Class8.2', 'Class8.3', 'Class8.4', 'Class8.5', 'Class8.6',
    'Class8.7', 'Class9.1', 'Class9.2', 'Class9.3', 'Class10.1', 'Class10.2',
    'Class10.3', 'Class11.1', 'Class11.2', 'Class11.3', 'Class11.4',
    'Class11.5', 'Class11.6'
] #37 vectors of Galaxy Zoo divided into 11 classes based on the 11 different questions and their responses


def append_ext(fn):
    '''
    This function is used to take the GalaxyID from the CSV and append .jpg to it in order to denote the image names.
    '''
    return fn + ".jpg"

traindf = pd.read_csv(os.path.join(r"D:\OneDrive\Major Project\Code\Galaxy_Morphology\Data\GalaxyZoo1\train", 'training_

traindf["id"] = traindf['GalaxyID'].astype(str).apply(append_ext) #Create a new column in the Data Frame called 'id' whi
traindf
```

Out[3]:

	GalaxyID	Class1.1	Class1.2	Class1.3	Class2.1	Class2.2	Class3.1	Class3.2	Class4.1	Class4.2	...	Class10.1	Class10.2	Class10.3
0	100008	0.383147	0.616853	0.000000	0.000000	0.616853	0.038452	0.578401	0.418398	0.198455	...	0.279952	0.138445	0.000000
1	100023	0.327001	0.663777	0.009222	0.031178	0.632599	0.467370	0.165229	0.591328	0.041271	...	0.000000	0.131378	0.459950
2	100053	0.765717	0.177352	0.056931	0.000000	0.177352	0.000000	0.177352	0.000000	0.177352	...	0.000000	0.000000	0.000000
3	100078	0.693377	0.238564	0.068059	0.000000	0.238564	0.109493	0.129071	0.189098	0.049466	...	0.094549	0.000000	0.094549
4	100090	0.933839	0.000000	0.066161	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...	0.000000	0.000000	0.000000
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
61573	999948	0.510379	0.489621	0.000000	0.059207	0.430414	0.000000	0.430414	0.226257	0.204157	...	0.226257	0.000000	0.000000
61574	999950	0.901216	0.098784	0.000000	0.000000	0.098784	0.000000	0.098784	0.000000	0.098784	...	0.000000	0.000000	0.000000
61575	999958	0.202841	0.777376	0.019783	0.116962	0.660414	0.067245	0.593168	0.140022	0.520391	...	0.000000	0.090673	0.049349
61576	999964	0.091000	0.909000	0.000000	0.045450	0.863550	0.022452	0.841098	0.795330	0.068220	...	0.068398	0.318132	0.408799

	GalaxyID	Class1.1	Class1.2	Class1.3	Class2.1	Class2.2	Class3.1	Class3.2	Class4.1	Class4.2	...	Class10.1	Class10.2	Class10.3
61577	999967	0.767000	0.140000	0.093000	0.000000	0.140000	0.000000	0.140000	0.023380	0.116620	...	0.023380	0.000000	0.000000

61578 rows × 39 columns

In [4]:
```
def convert_rgb_to_grayscale(image):
    return tf.image.rgb_to_grayscale(image)

datagenerator = ImageDataGenerator(
    fill_mode='nearest',
    cval=0,
    rescale=1/255,
    preprocessing_function=convert_rgb_to_grayscale,
    rotation_range=90,
    width_shift_range=0.1,
    height_shift_range=0.1,
    horizontal_flip=True,
    vertical_flip=True,
    validation_split=0.02)

train_generator = datagenerator.flow_from_dataframe(
    dataframe=traindf,
    directory="D:/OneDrive/Major Project/Code/Galaxy_Morphology/Data/GalaxyZoo1/train/images_training_rev1",
    x_col="id",
    y_col=classes,
    subset="training",
    batch_size=16,
    seed=123,
    shuffle=True,
    class_mode="raw",
    target_size=(224, 224))

validation_generator = datagenerator.flow_from_dataframe(
    dataframe=traindf,
    directory="D:/OneDrive/Major Project/Code/Galaxy_Morphology/Data/GalaxyZoo1/train/images_training_rev1",
    x_col="id",
    y_col=classes,
    subset="validation",
    batch_size=16,
    seed=123,
    shuffle=True,
    class_mode="raw",
    target_size=(224, 224))

STEP_SIZE_TRAIN = train_generator.n // train_generator.batch_size
STEP_SIZE_VALID = validation_generator.n // validation_generator.batch_size
```

```
Found 60347 validated image filenames.
Found 1231 validated image filenames.
```

In [5]:
```
resnet_model.add(pretrained_model)
resnet_model.add(Flatten())
resnet_model.add(Dense(len(classes), activation='softmax'))
```

In [6]:
```
print(resnet_model.summary())
```

```
Model: "sequential"
_________________________________________________________________
 Layer (type)                Output Shape              Param #
=================================================================
 resnet50 (Functional)       (None, 7, 7, 2048)        23587712

 flatten (Flatten)           (None, 100352)            0

 dense (Dense)               (None, 37)                3713061

=================================================================
Total params: 27,300,773
Trainable params: 3,713,061
Non-trainable params: 23,587,712
_________________________________________________________________
None
```

In [7]:
```
optimizer = keras.optimizers.Adam(learning_rate=0.001, decay=5e-4)

resnet_model.compile(optimizer, loss='mse', metrics=["accuracy"])
```

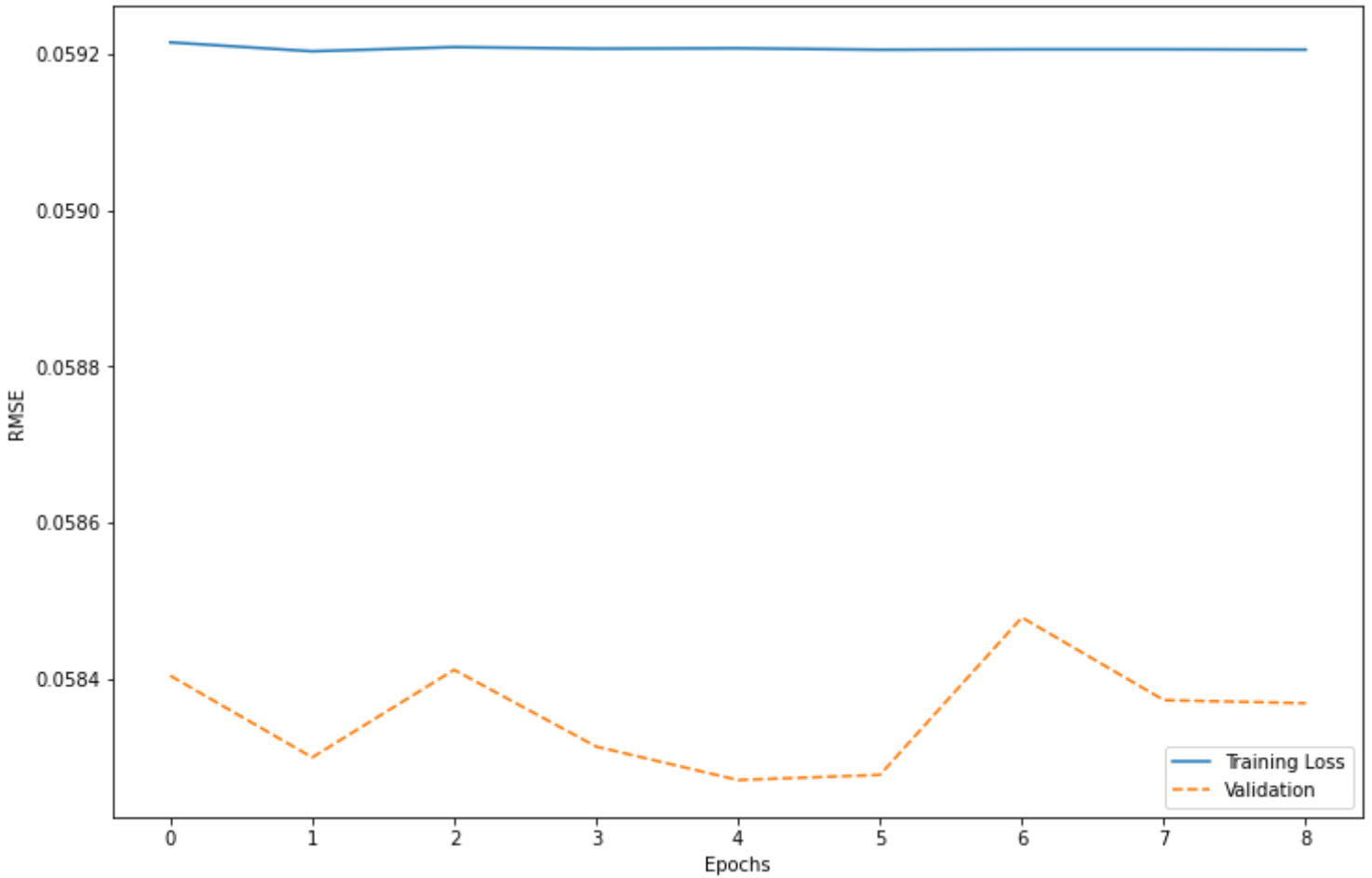
In [8]:
```
class LossHistory(Callback):
    def on_train_begin(self, logs={}):
        self.losses = []
        self.val_losses = []

    def on_batch_end(self, batch, logs={}):
        self.losses.append(logs.get('loss'))
        self.val_losses.append(logs.get('val_loss'))


early_stopping = EarlyStopping(
    monitor='val_loss', patience=4, verbose=1, mode='auto')

history = LossHistory()

checkpointer = ModelCheckpoint(
    filepath='D:/OneDrive/Major Project/Code/Galaxy_Morphology/Data/GalaxyZoo1/weights/GZ1_TL.hdf5', verbose=2, save_bes
```

In [9]:
```
hist = resnet_model.fit(
    train_generator,
    steps_per_epoch=STEP_SIZE_TRAIN,
    validation_data=validation_generator,
    validation_steps=STEP_SIZE_VALID,
    epochs=30,
    callbacks=[history, checkpointer, early_stopping])
```

```
Epoch 1/30
3771/3771 [==============================] - ETA: 0s - loss: 0.0592 - accuracy: 0.5949
Epoch 1: val_loss improved from inf to 0.05840, saving model to D:/OneDrive/Major Project/Code/Galaxy_Morphology/Dat
a/GalaxyZoo1/weights\GZ1_TL.hdf5
3771/3771 [==============================] - 2474s 652ms/step - loss: 0.0592 - accuracy: 0.5949 - val_loss: 0.0584 -
val_accuracy: 0.6308
Epoch 2/30
3771/3771 [==============================] - ETA: 0s - loss: 0.0592 - accuracy: 0.5954
Epoch 2: val_loss improved from 0.05840 to 0.05830, saving model to D:/OneDrive/Major Project/Code/Galaxy_Morpholog
y/Data/GalaxyZoo1/weights\GZ1_TL.hdf5
3771/3771 [==============================] - 1078s 286ms/step - loss: 0.0592 - accuracy: 0.5954 - val_loss: 0.0583 -
val_accuracy: 0.6308
Epoch 3/30
3771/3771 [==============================] - ETA: 0s - loss: 0.0592 - accuracy: 0.5954
Epoch 3: val_loss did not improve from 0.05830
3771/3771 [==============================] - 1073s 284ms/step - loss: 0.0592 - accuracy: 0.5954 - val_loss: 0.0584 -
val_accuracy: 0.6291
Epoch 4/30
3771/3771 [==============================] - ETA: 0s - loss: 0.0592 - accuracy: 0.5954
Epoch 4: val_loss did not improve from 0.05830
3771/3771 [==============================] - 1083s 287ms/step - loss: 0.0592 - accuracy: 0.5954 - val_loss: 0.0583 -
val_accuracy: 0.6316
Epoch 5/30
3771/3771 [==============================] - ETA: 0s - loss: 0.0592 - accuracy: 0.5954
Epoch 5: val_loss improved from 0.05830 to 0.05827, saving model to D:/OneDrive/Major Project/Code/Galaxy_Morpholog
y/Data/GalaxyZoo1/weights\GZ1_TL.hdf5
3771/3771 [==============================] - 1070s 284ms/step - loss: 0.0592 - accuracy: 0.5954 - val_loss: 0.0583 -
val_accuracy: 0.6299
Epoch 6/30
3771/3771 [==============================] - ETA: 0s - loss: 0.0592 - accuracy: 0.5954
Epoch 6: val_loss did not improve from 0.05827
3771/3771 [==============================] - 1073s 285ms/step - loss: 0.0592 - accuracy: 0.5954 - val_loss: 0.0583 -
val_accuracy: 0.6291
```

```
Epoch 7/30
3771/3771 [==============================] - ETA: 0s - loss: 0.0592 - accuracy: 0.5954
Epoch 7: val_loss did not improve from 0.05827
3771/3771 [==============================] - 1078s 286ms/step - loss: 0.0592 - accuracy: 0.5954 - val_loss: 0.0585 -
val_accuracy: 0.6291
Epoch 8/30
3771/3771 [==============================] - ETA: 0s - loss: 0.0592 - accuracy: 0.5954
Epoch 8: val_loss did not improve from 0.05827
3771/3771 [==============================] - 1077s 286ms/step - loss: 0.0592 - accuracy: 0.5954 - val_loss: 0.0584 -
val_accuracy: 0.6291
Epoch 9/30
3771/3771 [==============================] - ETA: 0s - loss: 0.0592 - accuracy: 0.5954
Epoch 9: val_loss did not improve from 0.05827
3771/3771 [==============================] - 1074s 285ms/step - loss: 0.0592 - accuracy: 0.5954 - val_loss: 0.0584 -
val_accuracy: 0.6291
Epoch 9: early stopping
```

In [10]:
```
plt.figure(figsize=(12, 8))
plt.plot(hist.epoch, hist.history['loss'], label='Training Loss')
plt.plot(
    hist.epoch, hist.history['val_loss'], label='Validation', linestyle='--')
plt.xlabel("Epochs")
plt.ylabel("RMSE")
plt.legend()
plt.show()
```



In [ ]: