Analysis of the impact of neuron count on network performance and training time taken

The following code trains and tests models on the Cat Recognition dataset with a varying number of neurons in each layer, and then plots graphs of Test Prediction Accuracy against Neuron Count and Training Time against Neuron Count.

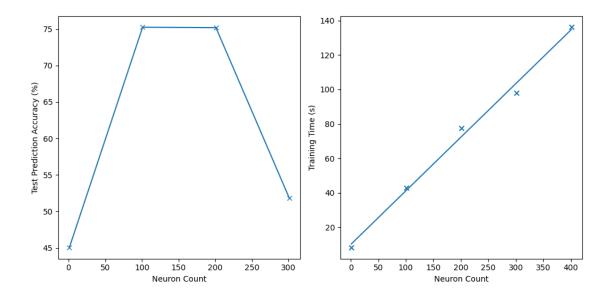
```
[1]: from IPython.display import clear_output, display
     import os
     import matplotlib.pyplot as plt
     import numpy as np
     from school_project.models.gpu.cat_recognition import CatRecognitionModel as_
      ⊶Model
     # Change to root directory of project
     os.chdir(os.getcwd())
     # Set width and height of figure
     plt.rcParams["figure.figsize"] = [10, 5]
     # Generate list of neuron counts from 1 to 501, incremented by 100
     neuron_counts = np.array(list(range(1, 501, 100)))
     layer count = 2
     test_prediction_accuracies = np.array([])
     training times = np.array([])
     for index, neuron_count in enumerate(neuron_counts):
         clear output(wait=True)
         display(f"Progress: {round(number=index/len(neuron_counts) * 100,__

ondigits=2)}%")
         model = Model(
                hidden_layers_shape=[neuron_count for layer in range(layer_count)],
                train_dataset_size=209,
                learning_rate=0.1,
                use relu=True
```

```
model.create_model_values()
   model.train(epoch_count=3_500)
   model.test()
   test_prediction_accuracies = np.append(test_prediction_accuracies,
                                            model.test_prediction_accuracy)
   training_times = np.append(training_times,
                                model.training time)
clear_output(wait=True)
display("Progress: Complete")
figure, axis = plt.subplots(nrows=1, ncols=2)
axis[0].set_xlabel("Neuron Count")
axis[0].set_ylabel("Test Prediction Accuracy (%)")
axis[0].plot(neuron_counts, test_prediction_accuracies, marker='x')
# Determine gradient and y-intercept of training times regression line
m, c = np.polyfit(neuron_counts, training_times, deg=1)
print(f"Training Times Regression Line Gradient: {round(number=m, ndigits=2)}")
axis[1].set_xlabel("Neuron Count")
axis[1].set_ylabel("Training Time (s)")
# Plot scatter graph of neuron counts and training times
axis[1].scatter(neuron_counts, training_times, marker='x')
# Plot regression line
axis[1].plot(neuron_counts, m * neuron_counts + c)
plt.tight_layout()
plt.show()
```

'Progress: Complete'

Training Times Regression Line Gradient: 0.31



As shown above, as the neuron count of each layer increases so does the training time taken and the test prediction accuracy at first. However, as the neuron count continued to increase the prediction accuracy began to drop greatly (after 200 neurons in this case). This is most likely due to the model overfitting and learning the training dataset too closely, causing it to fail on the new inputs of the test dataset.