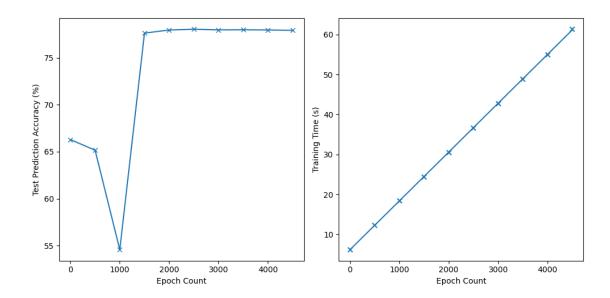
## epoch-count-analysis

The following code trains models on the Cat Recognition dataset and tests the model at regular Epoch Count intervals, and then plots graphs of Test Prediction Accuracy against Epoch Count and Training Time against Epoch Count.

```
[6]: 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 Epoch Counts from 1 to 5000, incremented by 500
     epoch_count_interval = 500
     epoch_counts = np.array(list(range(0, 5_000, epoch_count_interval)))
     test_prediction_accuracies = np.array([])
     training_times = np.array([])
     # Create model object
     model = Model(hidden_layers_shape=[100, 100],
                   train_dataset_size=209,
                   learning_rate=0.1,
                   use_relu=True)
     model.create_model_values()
     for index, epoch_count in enumerate(epoch_counts):
         clear_output(wait=True)
         display(f"Progress: {round(number=index/len(epoch_counts) * 100,__
      →ndigits=2)}%")
```

```
model.train(epoch_count=epoch_count_interval)
   model.test()
   test_prediction_accuracies = np.append(test_prediction_accuracies,
                                           model.test_prediction_accuracy)
   # Add training times cumulatively
   if len(training_times) != 0:
        training_times = np.append(training_times,
                                   training_times[-1] + model.training_time)
   else:
        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("Epoch Count")
axis[0].set_ylabel("Test Prediction Accuracy (%)")
# Plot regression line
axis[0].plot(epoch_counts, test_prediction_accuracies, marker='x')
# Determine gradient and y-intercept of training times regression line
m, c = np.polyfit(epoch_counts, training_times, deg=1)
print(f"Training Times Regression Line Gradient: {round(number=m, ndigits=2)}")
axis[1].set_xlabel("Epoch Count")
axis[1].set_ylabel("Training Time (s)")
# Plot scatter graph of epoch counts and training times
axis[1].scatter(epoch_counts, training_times, marker='x')
# Plot regression line
axis[1].plot(epoch_counts, m * epoch_counts + c)
plt.tight_layout()
plt.show()
```

'Progress: Complete'
Training Times Regression Line Gradient: 0.01



As shown above, as the epoch count increases so does both the test prediction accuracy and the training time taken.