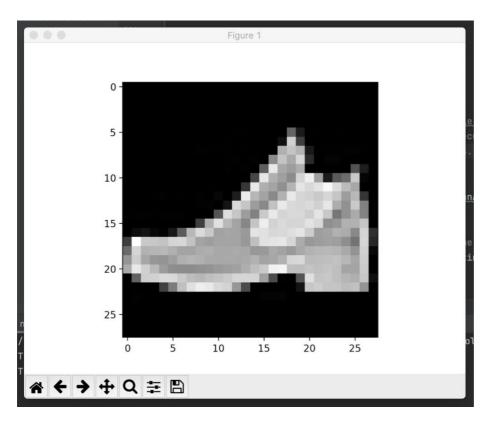
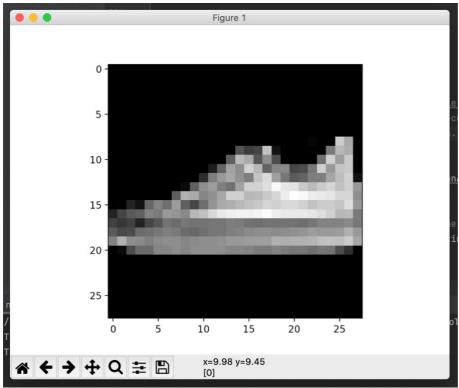
David Irwin R00109532

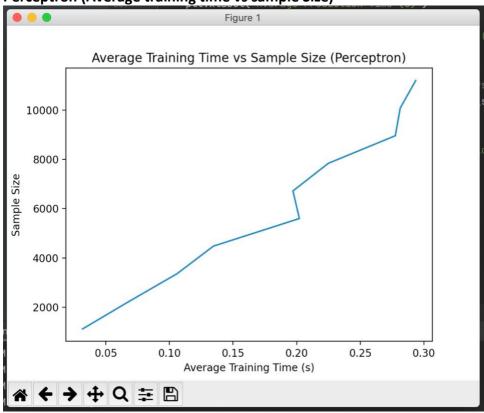
Task 1



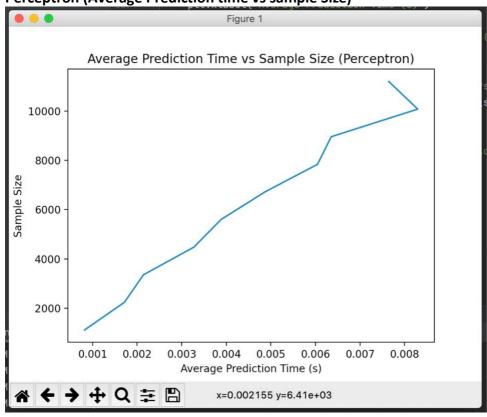


Task 3

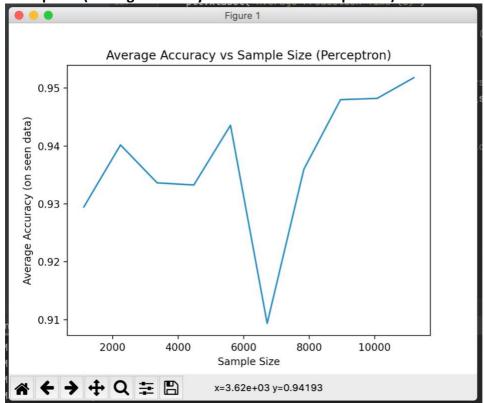
Perceptron (Average training time vs sample Size)



Perceptron (Average Prediction time vs sample Size)



Perceptron (Average accuracy on seen data vs sample size)



Accuracy of model (on unseen data, sample size = 11200)

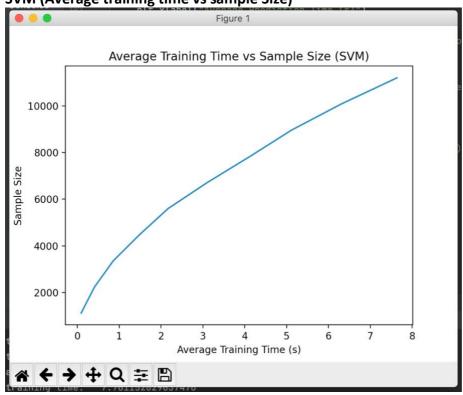
Accuracy (on unseen data): 0.9592857142857143

Training times when sample size = 11200

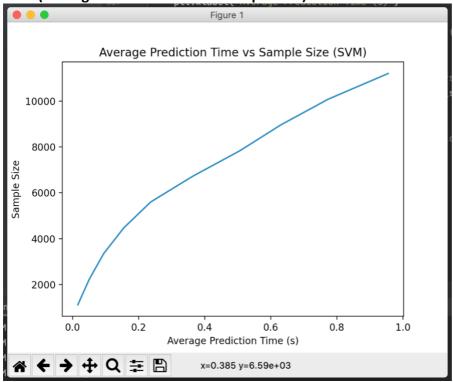
Minimum training time: 0.19684505462646484 Minimum testing time: 0.007031917572021484 Minimum accuracy: 0.9321428571428572 Maximum training time: 0.5936670303344727 Maximum testing time: 0.013043880462646484 Maximum accuracy: 0.9696428571428571 Average training time: 0.31124613285064695 Average testing time: 0.007950448989868164 Average accuracy: 0.9517857142857142

Task 4

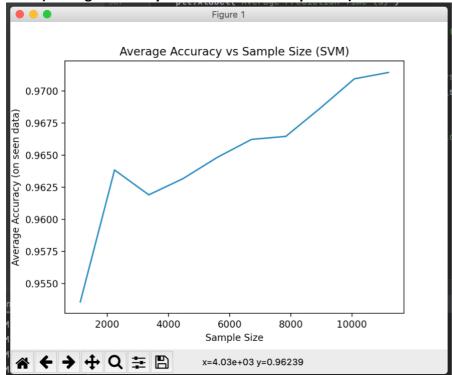
SVM (Average training time vs sample Size)



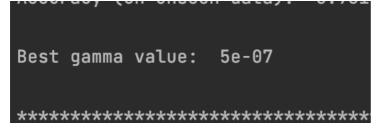




SVM (Average accuracy on seen data vs sample size)



Best gamma value (determined by highest mean accuracy on seen data when sample size = 3000)



Accuracy of model (on unseen data, sample size = 11200)

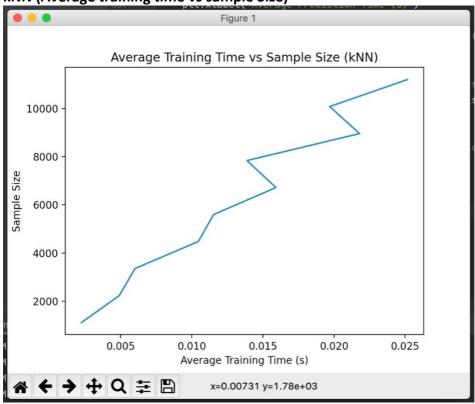
Accuracy (on unseen data): 0.9739285714285715

Training times when sample size = 11200

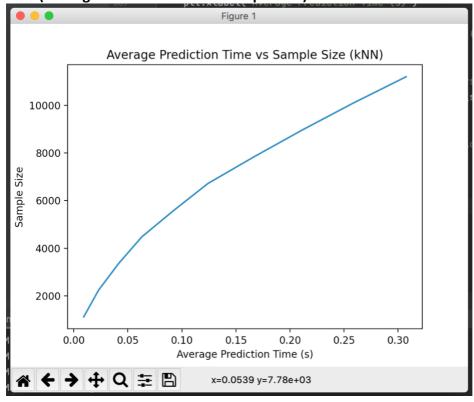
Minimum	training time:	7.537694215774536
Minimum	testing time:	0.9308781623840332
Minimum	accuracy:	0.9669642857142857
Maximum	training time:	7.873740911483765
Maximum	testing time:	1.0243539810180664
Maximum	accuracy:	0.9794642857142857
Average	training time:	7.725786089897156
Average	testing time:	0.9703830003738403
Average	accuracy:	0.9714285714285713

Task 5

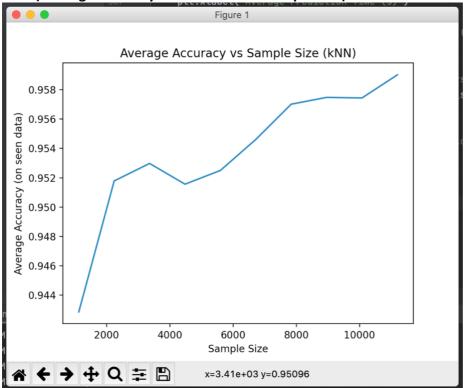
kNN (Average training time vs sample Size)







kNN (Average accuracy on seen data vs sample size)



Best k value (from a choice of 1 to 10) (determined by highest mean accuracy on seen data when sample size = 3000)

Best k value: 7

Accuracy of model (on unseen data, sample size = 11200)

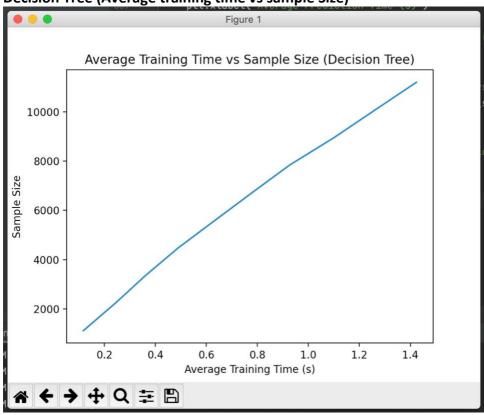
Accuracy (on unseen data): 0.9628571428571429

Training times when sample size = 11200

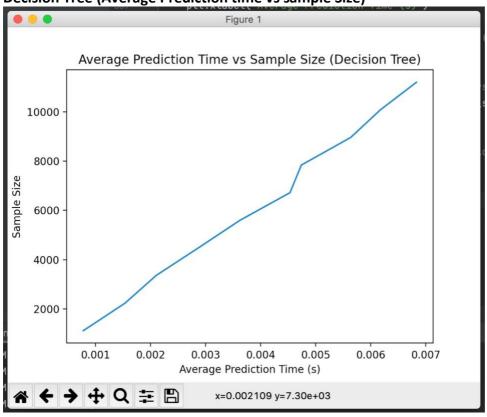
raining times when sample size = 11200				
Minimum	training time:	0.024376869201660156		
Minimum	testing time:	0.3010427951812744		
Minimum	accuracy:	0.9526785714285714		
Maximum	training time:	0.027286052703857422		
Maximum	testing time:	0.3645608425140381		
Maximum	accuracy:	0.96875		
Average	training time:	0.025330209732055665		
Average	testing time:	0.32100620269775393		
Average	accuracy:	0.9590178571428571		

Task 6

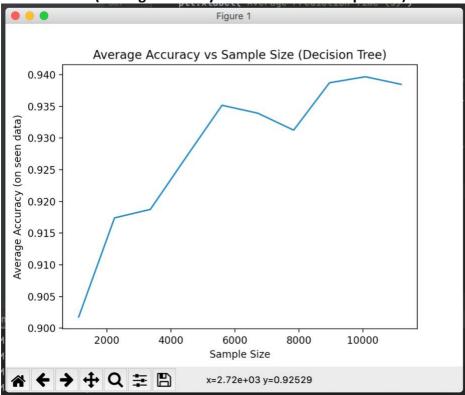
Decision Tree (Average training time vs sample Size)



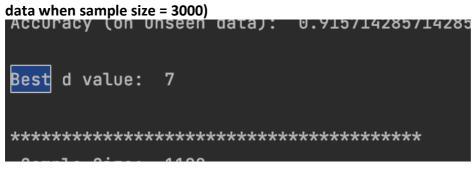
Decision Tree (Average Prediction time vs sample Size)



Decision Tree (Average accuracy on seen data vs sample size)



Best depth value (from a choice of 1 - 10) (determined by highest mean accuracy on seen data when sample size = 3000)



Accuracy of model (on unseen data, sample size = 11200)

Accuracy (on unseen data): 0.9414285714285714

Training times when sample size = 11200

Minimum training time: 1.2494759559631348

Minimum testing time: 0.006673336029052734

Minimum accuracy: 0.925

Maximum training time: 1.294084072113037

Maximum testing time: 0.007397890090942383

Maximum accuracy: 0.9526785714285714

Average training time: 1.267964768409729

Average testing time: 0.006956148147583008

Average accuracy: 0.9401785714285713

Task 7

Trends

Perceptron

Average training time seems to increase largely linearly to the sample size. It can be inferred that the average training time is directly correlated to the sample size. Average prediction time seems to increase largely linearly to the sample size. It can be inferred that the average training time is directly correlated to the sample size.

I think this is the case because as the dataset grows, the number of calculations required also grows proportionately.

Average accuracy on seen data also seems to increase linearly. Inferring that average accuracy has a direct correlation with sample size.

I think this is the case because as the dataset grows, it has more x-y pairs to learn from.

Note the dip between 6,000 & 8,000. My assumption here is that because I take a sample randomly, I believe that maybe there is an unbalance between the number of sneaker rows in the sample and the number of ankle boot rows in the dataset. And that this imbalance means that one label gets predicted wrong a lot. This dip is also present in the other models (except from the decision tree).

SVM

After sample equals 4000 **average training time** seems to increase largely linearly to the sample size. It can be inferred that the average training time is directly correlated to the sample size.

After sample equals 4000 **Average prediction time** seems to increase largely linearly to the sample size. It can be inferred that the average training time is directly correlated to the sample size.

I think this is the case because as the dataset grows, the number of calculations required also grows proportionately.

Average accuracy on seen data also seems to increase linearly after sample size 3,000. Inferring that average accuracy has a direct correlation with sample size.

kNN

Seems to follow the same trends as the previous models. However, it's accuracy sharply increases going to 2,000.

Decision Tree

Also seems to follow the same trends as model 2 & 3. This model however is more linear than the rest for training & testing time.

Classifier Ranking

The accuracy of the four models on unseen data is as follows:

Model	Accuracy
Perceptron	0.9593
SVM	0.9739
kNN	0.9629
Decision tree	0.9414

The average time to train & test the four models on seen data is as follows (sample size = 11400):

Model	Train	Test
Perceptron	0.3112	0.0080
SVM	7.7258	0.9703
kNN	0.0253	0.3210
Decision tree	1.2680	0.0070

As the accuracy seem to be very similar, I need to look at the training time / testing time to see the trade-off. I will rank the models in terms of lowest training / testing times.

- 1. Perceptron
- 2. kNN
- 3. Decision Tree
- 4. SVM

One could argue that because the training / testing times of the Perceptron & kNN are very similar, that kNN could be ranked as number 1. However, I decided to stick with the above rankings.

Note that SVM has the highest accuracy, yet I ranked it lowest. This is because for the small gain in accuracy, it takes way too much time compared to the other models. However, the SVM can have much better training / testing times in real life as it can very effectively utilise Nvidia GPUs for training which will make it much faster. All above models were completely CPU bound for this assignment.