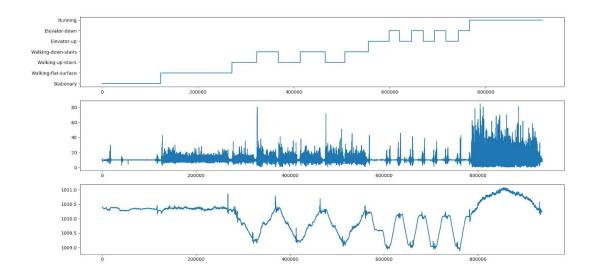
Step 1:

To get a larger version, you can run visualize\_data.py in the uploaded code and recreate the graph. There will also be an attached .png image in the uploaded homework files



The x-axis is always time in milliseconds, the first graph is activity vs time, the second is accelerometer magnitude vs time and the last one is barometric pressure vs time.

The data is my own stitched together activities. (dshen)

The mean and variance of the accelerometer magnitude seems like they should be effective for identifying stationary and running. It may also help in determining walking and moving up and down stairs. The range can also be useful for using the stairs and elevator. Medium range + low mean implies elevator usage. The slope of the barometric pressure seems like it will be effective at finding activities which have vertical movement like walking on stairs and using the elevator and whether the direction is up or down. Although there is a large change in barometric pressure during the run, this is because I made a mistake and did not run on a flat area. I thought I might try stitching together the activities randomly for feature extraction, but I realized there would be a problem because barometric pressure would start to suddenly jump huge amounts.

## Step 2:

The feature extraction code is in data\_extraction.py

Step 3-5 note: in the uploaded code, running train\_models.py will run all the code for steps 3-5. To isolate and run only one, comment out the necessary code at the bottom of the file

Step 3: the function for step 3 is located in train\_models.py, it is called step\_3\_classifier() I tried using the KNN classifier and decision tree classifiers.

First I made a stratified split of the data into 80% training and 20% testing.

I used stratified 5 fold cross validation on the training set to determine the best number of neighbors/best maximum depth for each classifier respectively.

For the KNN classifier, I found that the 1 nearest neighbor classifier had the highest overall accuracy.

After training on the entire training set and testing on the test set got these results Stationary

recall 0.8285714285714286

precision 0.8903508771929824

accuracy 0.9530154277699859

Walking-flat-surface

recall 0.7813620071684588

precision 0.8288973384030418

accuracy 0.9256661991584852

Walking-up-stairs

recall 0.8349514563106796

precision 0.76444444444445

accuracy 0.9389901823281908

Walking-down-stairs

recall 0.7760416666666666

precision 0.776041666666666

accuracy 0.9396914446002805

Elevator-up

recall 0.7070063694267515

precision 0.7655172413793103

accuracy 0.9438990182328191

Elevator-down

recall 0.7967479674796748

precision 0.7

accuracy 0.9530154277699859

Runnina

recall 0.9285714285714286

precision 0.8927038626609443

accuracy 0.9712482468443198

For the decision tree classifier I used a max depth of 31 and got these results:

Stationary

recall 0.8938775510204081

precision 0.9125

accuracy 0.9670406732117812

Walking-flat-surface

recall 0.9139784946236559

precision 0.9107142857142857

accuracy 0.9656381486676017

Walking-up-stairs

recall 0.8543689320388349

precision 0.8341232227488151

accuracy 0.9544179523141655

Walking-down-stairs

recall 0.77083333333333333

precision 0.783068783068783

accuracy 0.9403927068723703

Elevator-up

recall 0.7452229299363057

precision 0.740506329113924

accuracy 0.9431977559607293

Elevator-down

recall 0.6991869918699187

precision 0.7478260869565218

accuracy 0.9537166900420757

Running

recall 0.9330357142857143

precision 0.8969957081545065

accuracy 0.9726507713884993

The KNN classifier was better in some aspects, but the decision tree seemed better overall. The easiest to classify activities were stationary and running which isn't very surprising since they often have the most extreme values for both the accelerometer frequency and time domain features. Even though the barometric pressure seems like it should be a huge help for the stair and elevator values, I think the data collection for those might be somewhat messed up. I suspect other students recorded their values in a similar way to me and there were not many prolonged periods for the elevator data. Plus, going immediately up and down the elevator probably did not help since in real life a person would likely not immediately move up and down the elevator. There is another problem for walking up and down stairs where there was more than a few steps of flat-walking-distance between each flight of stairs between floors in the CS building. Finally, barometric pressure might not be providing useful data for activities like walking flat, stationary and running and could just be creating noise.

Step 4: you can run the code with the function step\_4\_classifier() located in train\_models.py Again I tried using both KNN and decision tree for both parts

Step 4a: KNN classifier best number of neighbors 26

Stationary recall 0.6913123844731978 precision 0.7540322580645161 accuracy 0.9051214707813526 Walking-flat-surface recall 0.4896 precision 0.392811296534018 accuracy 0.7399868680236376 Walking-up-stairs recall 0.40835266821345706 precision 0.55 accuracy 0.8690085357846355 Walking-down-stairs recall 0.3325581395348837 precision 0.5437262357414449 accuracy 0.8663821405121471 Elevator-up recall 0.6600660066006601 precision 0.45351473922902497 accuracy 0.8870650032829941 Elevator-down recall 0.45454545454545453 precision 0.55555555555556 accuracy 0.927774130006566 Running recall 0.8375527426160337

precision 0.7231329690346083 accuracy 0.9248194353250164

## Decision tree classifier

best max\_depth 9 Stationary recall 0.49722735674676527 precision 0.840625 accuracy 0.8939592908732764 Walking-flat-surface recall 0.3552 precision 0.5068493150684932 accuracy 0.7967826657912016 Walking-up-stairs recall 0.6218097447795824 precision 0.6077097505668935 accuracy 0.8896913985554826 Walking-down-stairs recall 0.7023255813953488 precision 0.41483516483516486 accuracy 0.8181221273801708 Elevator-up recall 0.7491749174917491 precision 0.49134199134199136 accuracy 0.8978988837820092 Elevator-down recall 0.6446280991735537 precision 0.639344262295082 accuracy 0.942875902823375 Running recall 0.8354430379746836 precision 0.9588377723970944 accuracy 0.968811556139199

Aside from stationary, it seems that the decision tree performs way better than the KNN classifier. The location mismatch definitely hurt the classifier

Part 4b.
KNN classifier
best number of neighbors 24

Stationary recall 0.706997084548105 precision 0.6170483460559797 accuracy 0.8770812928501469 Walking-flat-surface recall 0.3823146944083225 precision 0.3295964125560538 accuracy 0.7372673849167483 Walking-up-stairs recall 0.1480865224625624 precision 0.3852813852813853 accuracy 0.8398628795298727 Walking-down-stairs recall 0.3931947069943289 precision 0.2810810810810811 accuracy 0.7911361410381978 Elevator-up recall 0.36721991701244816 precision 0.5822368421052632 accuracy 0.8942213516160626 Elevator-down recall 0.24731182795698925 precision 0.30564784053156147 accuracy 0.8802644466209598 Running recall 0.8589147286821706 precision 0.6674698795180722 accuracy 0.9101371204701273

DT classifier best max\_depth 4

Stationary recall 0.7682215743440233 precision 0.7772861356932154 accuracy 0.9240940254652301 Walking-flat-surface recall 0.2925877763328999 precision 0.75757575757576 accuracy 0.8491674828599413 Walking-up-stairs recall 0.7487520798668885 precision 0.5005561735261401 accuracy 0.8530852105778648 Walking-down-stairs recall 0.6257088846880907 precision 0.47421203438395415 accuracy 0.8616552399608227 Elevator-up recall 0.46887966804979253 precision 0.578005115089514 accuracy 0.8969147894221352 Elevator-down recall 0.49193548387096775 precision 0.23735408560311283 accuracy 0.809745347698335 Running recall 0.5255813953488372 precision 0.9685714285714285 accuracy 0.9223800195886386

I feel like the decision tree was better again, but both classifiers were pretty terrible for many activities. I think training on the hand and testing on the pocket was worse than training on the pocket and testing on the hand. Not sure why this is the case, maybe because hand movement is less consistent from person to person. The obvious fix to location mismatch is to not mismatch the location. However if a classifier must work with both locations, then adding location data to the list of things to predict and doing some sort of multi label classification might help. Also making use of the gyroscope data might help since I imagine that there are different patterns in pocket orientation and orientation.

## Step 5:

Again I attempted both parts using both a KNN classifier and decision tree classifier

Time domain features only KNN classifier best number of neighbors 1

Stationary recall 0.9183673469387755 precision 0.9221311475409836 accuracy 0.9726507713884993 Walking-flat-surface recall 0.9068100358422939 precision 0.8815331010452961 accuracy 0.9579242636746143 Walking-up-stairs recall 0.8446601941747572 precision 0.8285714285714286 accuracy 0.9523141654978962 Walking-down-stairs

recall 0.765625

precision 0.7903225806451613 accuracy 0.94109396914446

Elevator-up

recall 0.6878980891719745

precision 0.72

accuracy 0.9361851332398317

Elevator-down

precision 0.6612903225806451

accuracy 0.9417952314165497

Running

recall 0.9285714285714286

accuracy 0.9768583450210379

This classifier performs similarly to the ones using all the features in step 3. It's even better in stationary and running than the step 3 classifiers

Time domain features only DT classifier best max\_depth 15

Stationary recall 0.8938775510204081 precision 0.9087136929460581 accuracy 0.9663394109396914 Walking-flat-surface recall 0.8960573476702509 precision 0.8802816901408451 accuracy 0.955820476858345 Walking-up-stairs recall 0.8155339805825242 precision 0.7813953488372093 accuracy 0.9403927068723703 Walking-down-stairs recall 0.78125 precision 0.819672131147541 accuracy 0.9474053295932678 Elevator-up recall 0.7006369426751592 precision 0.7534246575342466 accuracy 0.9417952314165497 Elevator-down recall 0.7235772357723578 precision 0.6953125 accuracy 0.9488078541374474 Running recall 0.9241071428571429 precision 0.9039301310043668 accuracy 0.9726507713884993

Again I feel like the decision tree is slightly better overall than the KNN classifier.

Frequency domain features only KNN classifier best number of neighbors 24

Stationary recall 0.4857142857142857 precision 0.44074074074074077 accuracy 0.8057503506311361 Walking-flat-surface recall 0.26523297491039427 precision 0.2846153846153846 accuracy 0.7258064516129032 Walking-up-stairs recall 0.5631067961165048 precision 0.42962962962964 accuracy 0.8288920056100981 Walking-down-stairs recall 0.484375 precision 0.34572490706319703 accuracy 0.8071528751753155 Elevator-up recall 0.46496815286624205 precision 0.553030303030303 accuracy 0.8997194950911641 Elevator-down recall 0.4634146341463415 precision 0.5588235294117647 accuracy 0.9221598877980365 Running recall 0.2857142857142857 precision 0.5203252032520326 accuracy 0.8464235624123422

Frequency domain features only DT classifier best max\_depth 10

Stationary recall 0.7020408163265306 precision 0.8557213930348259 accuracy 0.9284712482468443 Walking-flat-surface recall 0.7956989247311828 precision 0.6082191780821918 accuracy 0.8597475455820477 Walking-up-stairs recall 0.6796116504854369 precision 0.7608695652173914 accuracy 0.9228611500701263 Walking-down-stairs recall 0.661458333333333334 precision 0.5879629629629629 accuracy 0.8920056100981767 Elevator-up recall 0.6114649681528662 precision 0.64 accuracy 0.9193548387096774 Elevator-down recall 0.6260162601626016 precision 0.77 accuracy 0.9516129032258065 Running recall 0.5625 precision 0.6 accuracy 0.8723702664796634

This time the decision tree clearly performs far better than the KNN classifier. I'm not surprised, but this without the time domain features the classifiers aren't doing as well. I think the frequency domain features are also more susceptible to being thrown off by the poor data collection (up and down the elevator). Although I experimented a little bit with different spectral rolloff amounts 80-95%, I didn't feel like it made a huge difference.

It seems like the classifiers get the most out of the time domain features since the time domain features alone are able to make a classifier about as good as the step 3 all-feature classifiers. Plus the time domain features decision tree classifier performs better across all activities compared to either of the frequency domain classifiers.