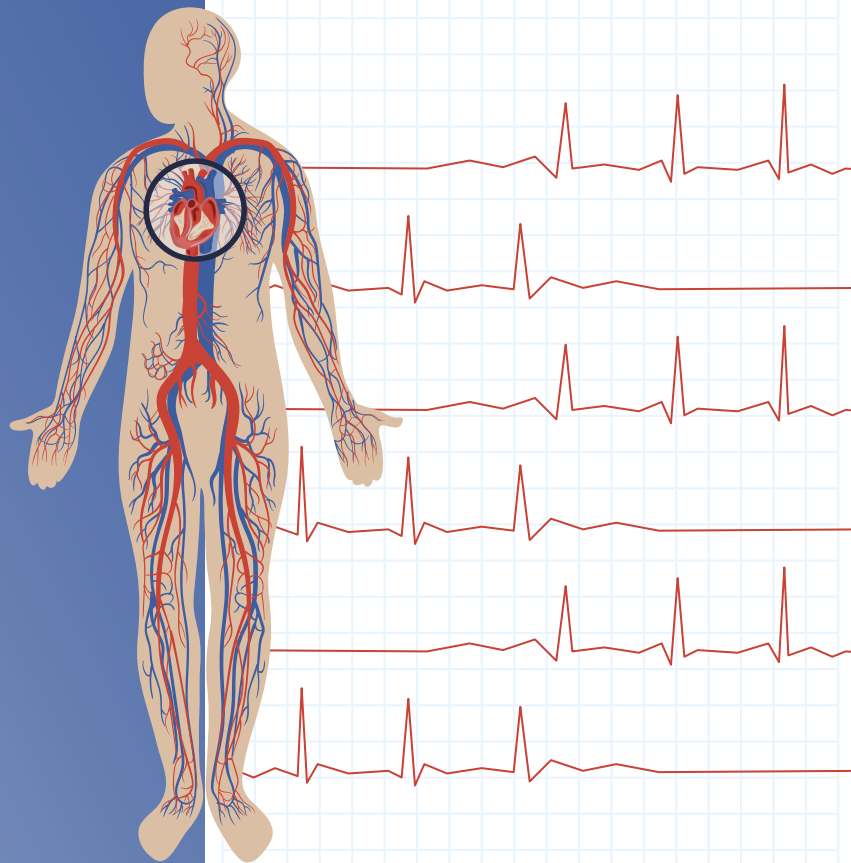
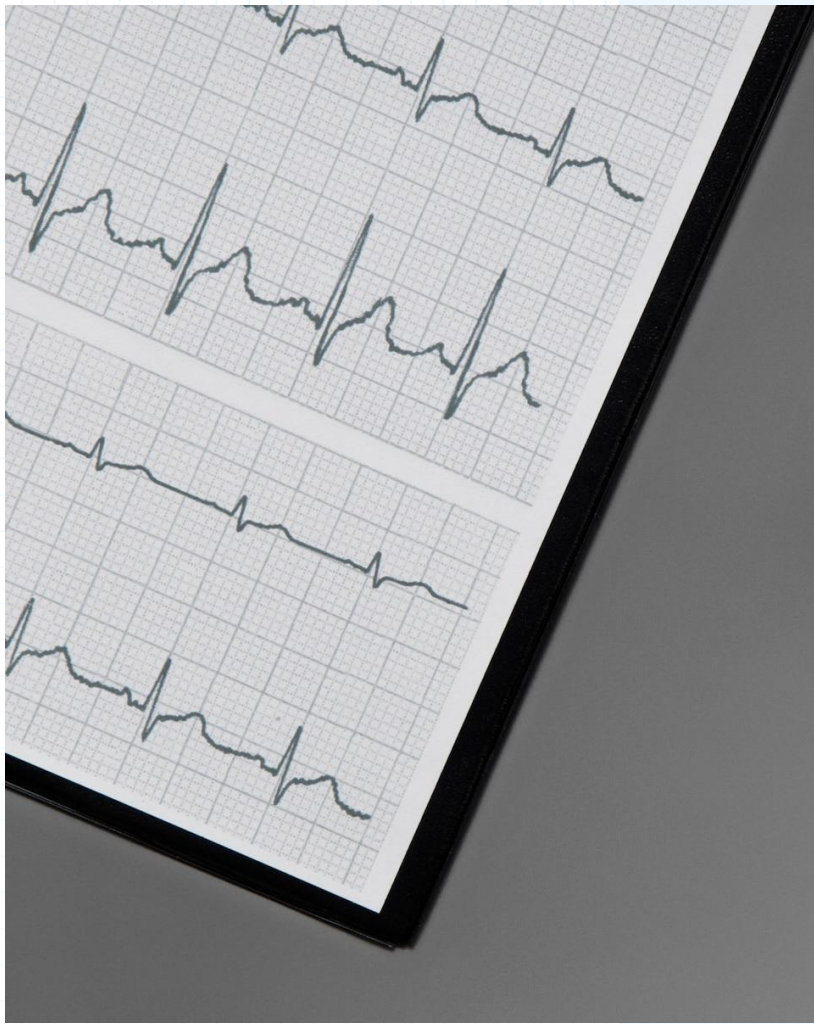


12 Lead EKG Image Classifier

By: Jose Gonzalez, Shannon
Williams, Nancy Ulloa, and Arle Alcid





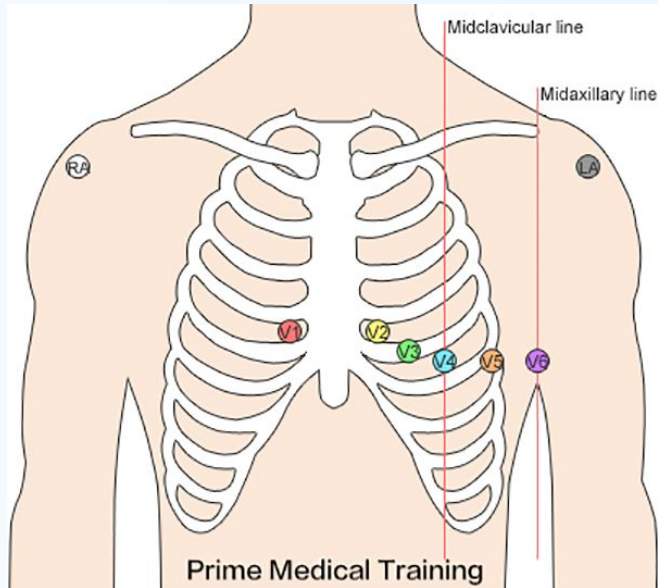
Purpose

The purpose of our project was to create a machine learning model that could classify EKGs as a healthy or arrhythmic heart rhythm

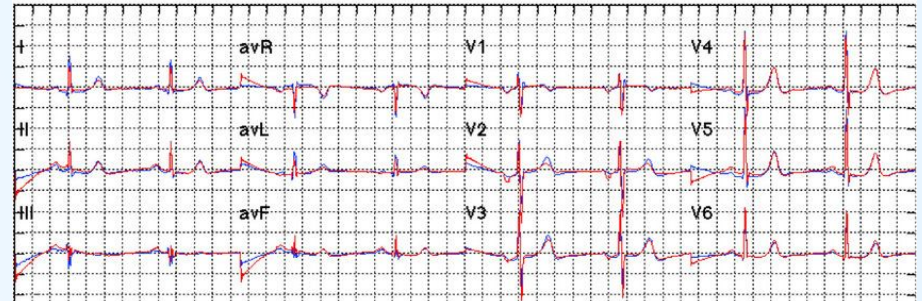


Introduction

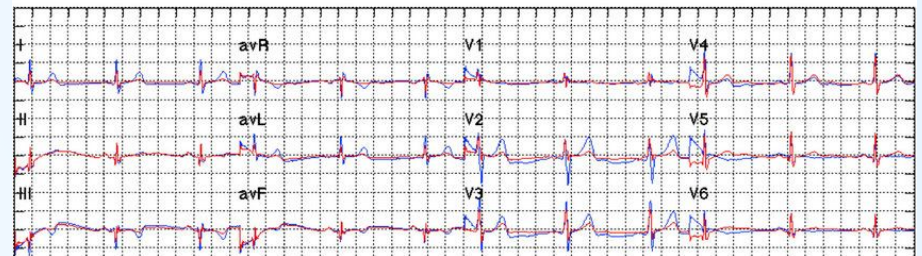
- 12 Lead Electrocardiogram (EKG)



A Healthy Control (HC) - 12 Lead ECG Signal Time Domain Plots



B Myocardial Infarction (MI) - 12 Lead ECG Signal Time Domain Plots



Data Retrieval

- PTB-XL
 - A large publicly available dataset
 - Over 21,700 test subjects
 - (<https://physionet.org/content/ptb-xl/1.0.3/#files-panel>)

| Folder Navigation: <base> | |
|---------------------------|--|
| Name | |
| records100 | |
| records500 | |
| LICENSE.txt | |
| RECORDS | |
| SHA256SUMS.txt | |
| example_physionet.py | |
| ptbxi_database.csv | |
| ptbxi_v102_changelog.txt | |
| ptbxi_v103_changelog.txt | |
| scp_statements.csv | |

| SCP-ECG Statement | Description |
|-------------------|------------------------------------|
| | normal ECG |
| | inferior myocardial infarction |
| | anteroseptal myocardial infarction |
| | left ventricular hypertrophy |

| Folder Navigation: <base> / records100 / 01000 | |
|--|------------------|
| Name | |
| ^ | Parent Directory |
| 01000_lr.dat | |
| 01000_lr.he | |
| 01001_lr.dat | |
| 01001_lr.he | |
| 01002_lr.dat | |
| 01002_lr.he | |

Data Cleaning



1st



Imported data to
MongoDB

2nd



Exported database to
our notebooks

3rd



Cleaned DataFrame
from NaNs/empty
columns

4th



Separated
"SCP_Codes" Column
and kept the "NORM"
or "ABNORM"

5th



For loop matching .hea
to file names in csv

6th



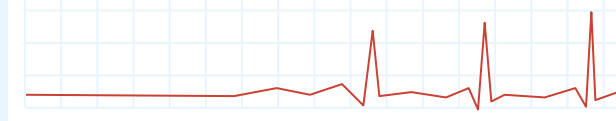
Waveform Database
(wfdb) package to
make EKGs

7th

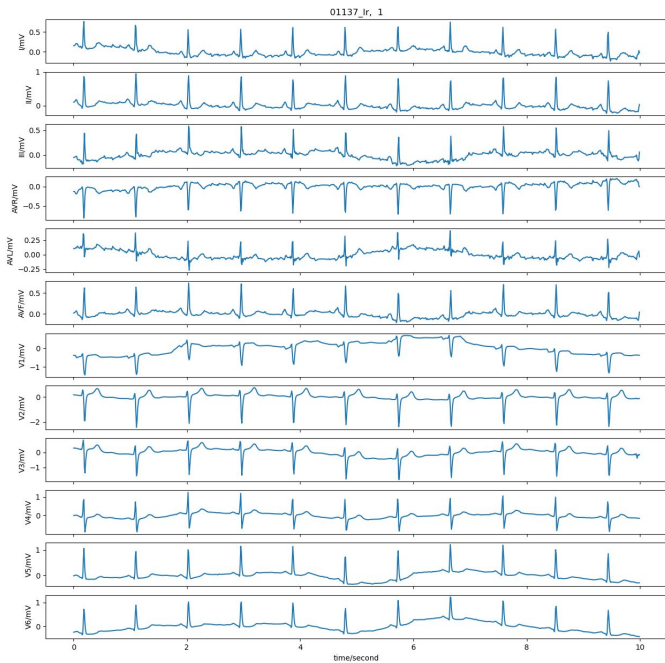


Saved 12 lead EKGs to
corresponding folders

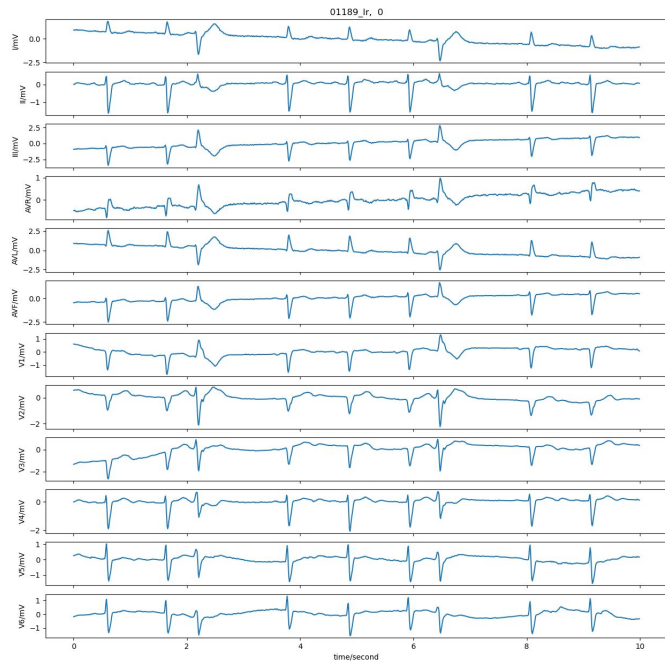
EKG Images



Normal



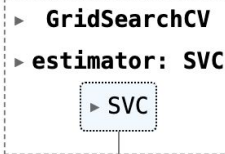
Abnormal



Scikit-Learn

- Limit size of model
 - Training size = 10%
 - Testing size = 10%
 - Shuffle
 - Stratify
- Training Classifier
 - Gamma (0.01, 0.001, 0.0001)
 - C (1, 10, 100)
- Test Performance
 - 52.27% of the samples were classified correctly

```
1 # Train Classifier
2 classifier = SVC()
3
4 parameters = [{'gamma': [0.01, 0.001, 0.0001],
5                      'C': [1, 10, 100, 1000]}]
6
7 grid_search = GridSearchCV(classifier, parameters)
8
9 grid_search.fit(X_train, y_train)
10
```



```
1 print(best_estimator)
```

SVC(C=1, gamma=0.01)

52.27272727272727% of samples were correctly classified



TensorFlow

On our second attempt we tried to create a model using tensorflow. The goal with this model was to get something that could actually read the images and classify them. This model proved to be inaccurate but it did allow us to get a better understanding of how the image processing model works for our optimizing face

```
train_ds = tf.keras.utils.image_dataset_from_directory(  
    data_dir,  
    validation_split=0.2,  
    subset="training",  
    seed=123,  
    image_size=(img_height, img_width),  
    batch_size=batch_size)
```

Found 434 files belonging to 2 classes.
Using 348 files for training.

Model: "sequential"

| Layer (type) | Output Shape | Param # |
|--------------------------------|----------------------|---------|
| ===== | | |
| rescaling_1 (Rescaling) | (None, 300, 300, 3) | 0 |
| conv2d (Conv2D) | (None, 298, 298, 32) | 896 |
| max_pooling2d (MaxPooling2D) | (None, 149, 149, 32) | 0 |
| conv2d_1 (Conv2D) | (None, 147, 147, 32) | 9248 |
| max_pooling2d_1 (MaxPooling2D) | (None, 73, 73, 32) | 0 |
| conv2d_2 (Conv2D) | (None, 71, 71, 32) | 9248 |
| max_pooling2d_2 (MaxPooling2D) | (None, 35, 35, 32) | 0 |
| flatten (Flatten) | (None, 39200) | 0 |
| dense (Dense) | (None, 128) | 5017728 |

...

Total params: 5037378 (19.22 MB)

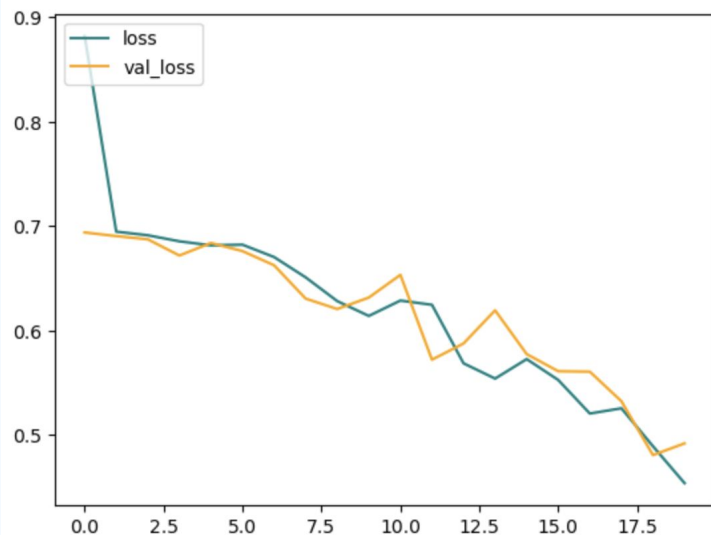
Trainable params: 5037378 (19.22 MB)

Non-trainable params: 0 (0.00 Byte)

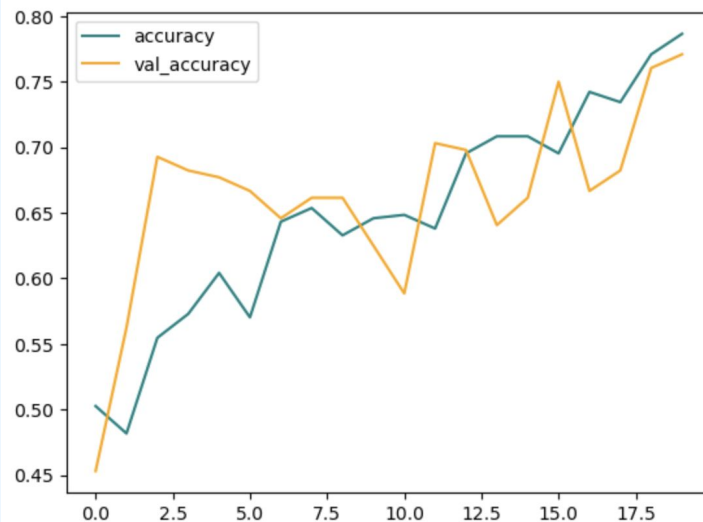
Optimization Model



Loss



Accuracy



```
print(pre.result(), re.result(), acc.result())
```

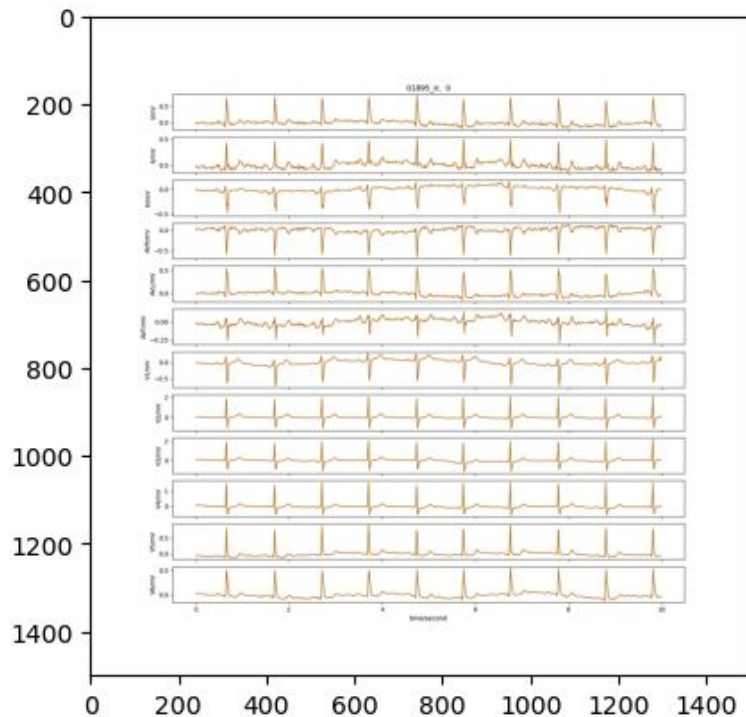
```
tf.Tensor(0.72727275, shape=(), dtype=float32) tf.Tensor(0.6956522, shape=(), dtype=float32) tf.Tensor(0.7291667, shape=(), dtype=float32)
```

Optimization Model Cont.

```
1 image_path = 'test_images/test_abnorm/01895_lr.png'
2 image_loaded = tf.keras.utils.load_img(image_path)
3 # Printing the obtained image.
4 image_loaded
5
6 img = cv2.imread(image_path)
7 plt.imshow(img)
8 plt.show()
9
```

```
: 1 if yhat > 0.5:
2     print(f'Predicted class is norm')
3 else:
4     print(f'Predicted class is abnorm')
```

Predicted class is abnorm





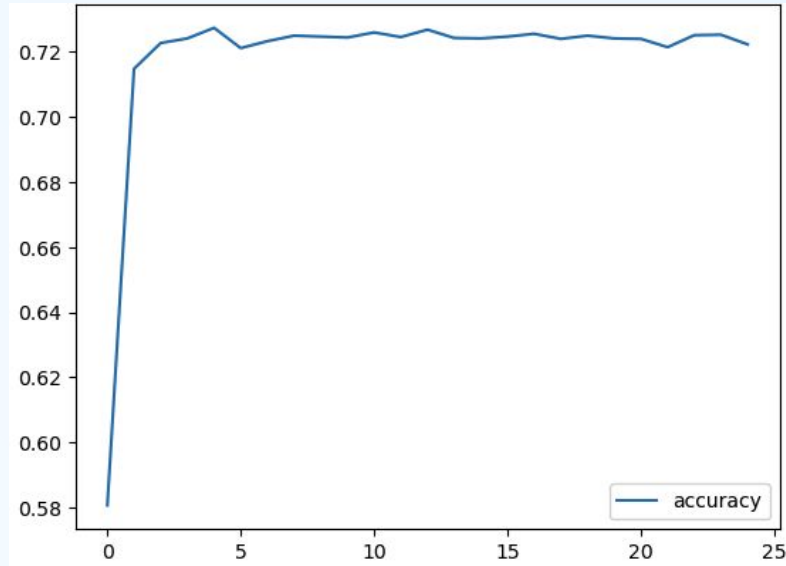
Neural Network Model

- A neural network algorithm was created in order to classify a normal or arrhythmic EKG, based off the patient's personal record
- Patient's most relevant demographics were kept for the training model

| | height | age | sex | weight | scp_codes_NORM |
|-----|--------|------|-----|--------|----------------|
| 100 | 172.0 | 48.0 | 1 | 72.0 | True |
| 115 | 174.0 | 49.0 | 1 | 74.0 | False |
| 135 | 167.0 | 52.0 | 0 | 72.0 | True |

Continued...

- 3 hidden layers
- 15, 10, 5 nodes per layer
- The model had a 89.7% accuracy rate



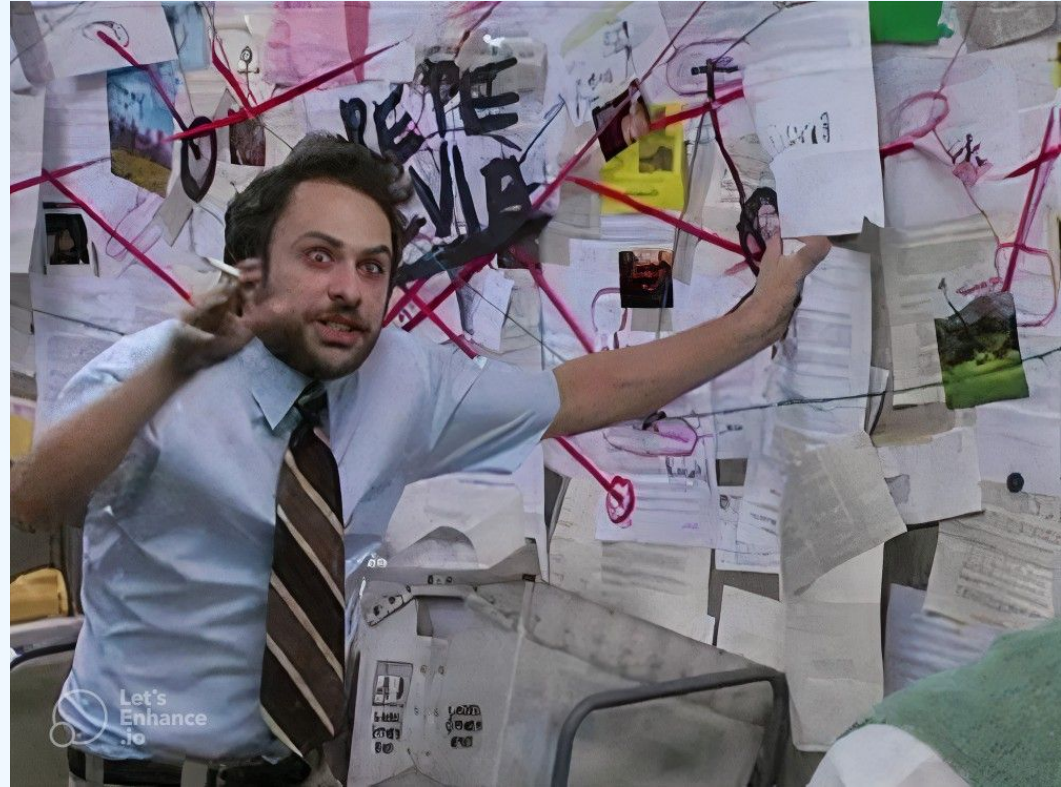


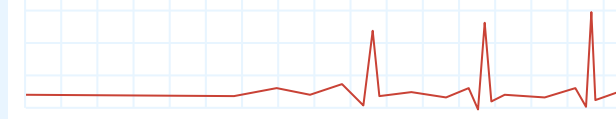
Limitations

- RAM – Our laptops melted
- To get the program to run we had to downscale the images and data might have been lost there.
- Lots of nuances throughout our abnormal classification



Questions?





Resources

- Goldberger, A., Amaral, L., Glass, L., Hausdorff, J., Ivanov, P. C., Mark, R., ... & Stanley, H. E. (2000). PhysioBank, PhysioToolkit, and PhysioNet: Components of a new research resource for complex physiologic signals. *Circulation* [Online]. 101 (23), pp. e215–e220.
- Wagner, Patrick, et al. "PTB-XL, a large publicly available electrocardiography dataset" (version 1.0.3). *PhysioNet* (2022), <https://doi.org/10.13026/kfzx-aw45>.
- Wagner, P., Strodthoff, N., Bousseljot, R.-D., Kreiseler, D., Lunze, F.I., Samek, W., Schaeffter, T. (2020), PTB-XL: A Large Publicly Available ECG Dataset. *Scientific Data*. <https://doi.org/10.1038/s41597-020-0495-6>

Medical Advice Disclaimer

DISCLAIMER: THIS WEBSITE DOES NOT PROVIDE MEDICAL ADVICE

The information, including but not limited to, text, graphics, images and other material contained on this website are for informational purposes only. No material on this site is intended to be a substitute for professional medical advice, diagnosis or treatment. Always seek the advice of your physician or other qualified health care provider with any questions you may have regarding a medical condition or treatment and before undertaking a new health care regimen, and never disregard professional medical advice or delay in seeking it because of something you have read on this website.