

Introduction and motivation

- We developed classifiers to find any potential lung diseases from chest X-ray
- Built **independent binary classifiers** for each of the lung diseases
- Trained Logistic Regression and SVM models on top of **SIFT** and **HoG** features extracted from X-ray image

Dataset

- Published by National Institutes of Health (NIH) Clinical Center
- 100,000+ frontal-view X-ray images
- 32,717 unique patients, 14 lung diseases
- Each image has multi-label
- Images are gray scale of size 1024 x 1024

Features

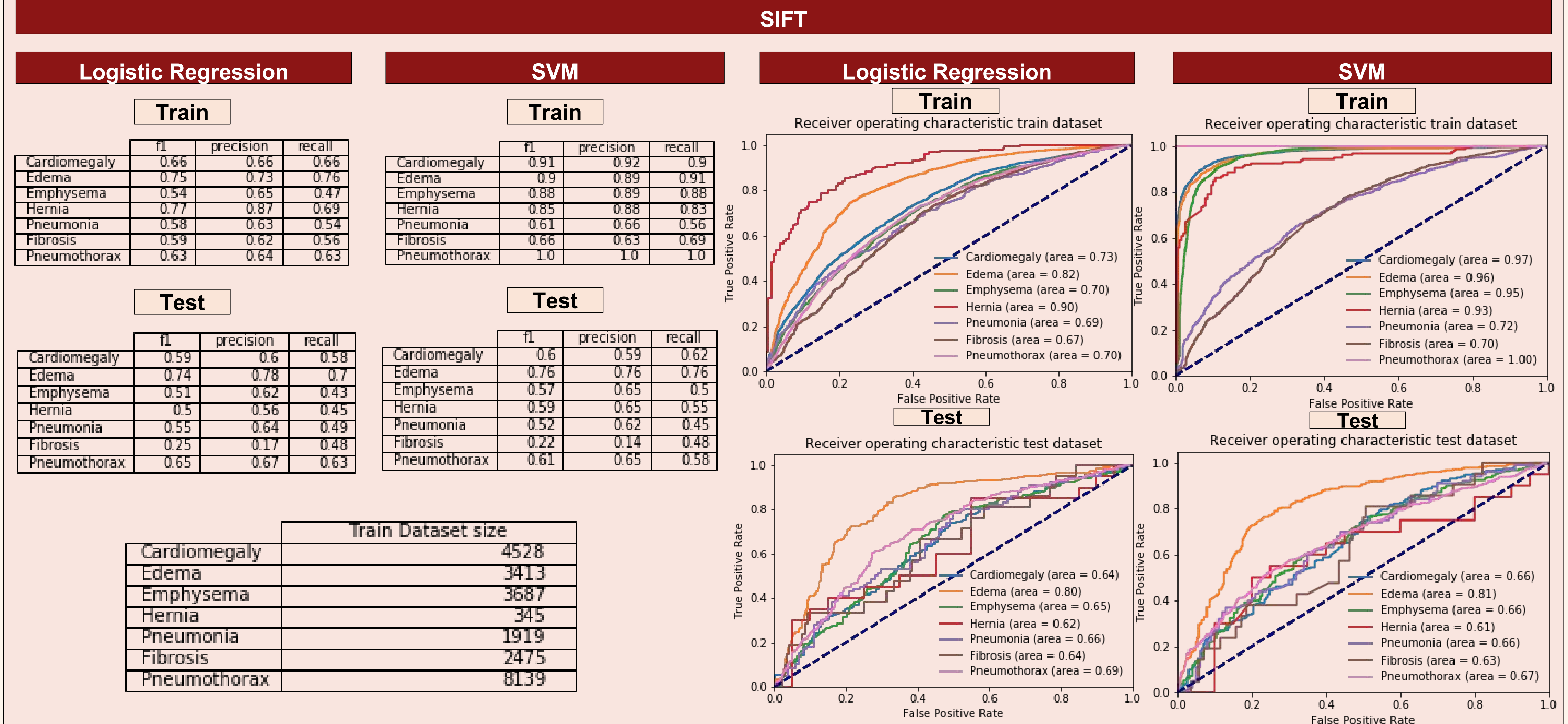
- **Scaled** 1024 x 1024 image to 224 x 224
- **Cropped** image to make lungs a focal point, resulting image size 180 x 200
- Applied histogram equalizer to **increase contrast** of the image
- Derived **SIFT** and **HoG** features. Both techniques helps in finding robust key-points in the image



Models

- [1] <http://scikit-learn.org/>
 [2] <https://kushalvyas.github.io/BOV.html>
 [3] https://docs.opencv.org/3.1.0/da/df5/tutorial_py_sift_intro.html

Results



Models

Logistic Regression

- Cost function:

$$J(\theta) = \frac{1}{2} \sum_{i=1}^m \left(\frac{1}{1 + e^{-\theta^T x}} - y^{(i)} \right)^2$$

SVM with Radial basis function

- Cost function:

$$J_{\lambda}(\alpha) = \frac{1}{m} \sum_{i=1}^m L(K^{(i)T} \alpha, y^{(i)}) + \frac{\lambda}{2} \alpha^T K \alpha$$

$$L(z, y) = \max\{0, 1 - yz\}$$

$$K(x, z) = \exp \left(- \frac{1}{2\tau^2} \|x - z\|_2^2 \right)$$

Discussion

- Logistic regression seems to be generalizing much better than SVM(Kernel) for SIFT. SVM is overfitting training data
- SIFT was able to detect various descriptors on the lung in the image, that proved useful as a features

Future

- Pre-trained CNN models for feature extraction
- Min-hashing to generate feature
- HoG with PCA
- Pre-processing to make lungs more focal