# Detection of Pneumonia from Chest X-ray Images

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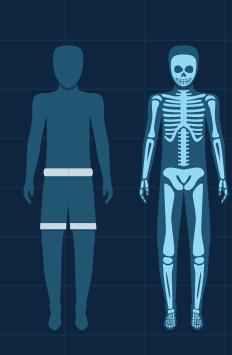
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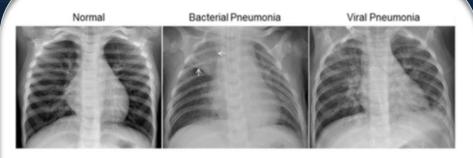
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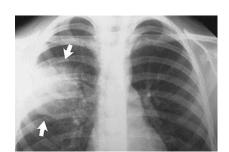


#### Introduction

- Pneumonia has been directly responsible for a huge number of deaths all across the globally.
- Early detection of pneumonia is essential for effective treatment and improved patient outcomes.
- Developing accurate pneumonia detection algorithms requires large amounts of high-quality labeled data, which can be difficult to obtain.
- Machine learning has emerged as a powerful tool for detecting and diagnosing pneumonia from medical images such as chest X-rays.

#### Introduction

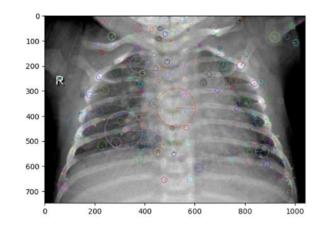


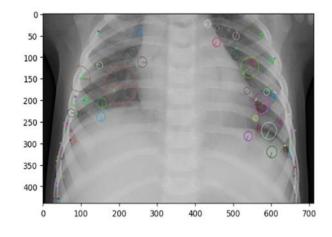




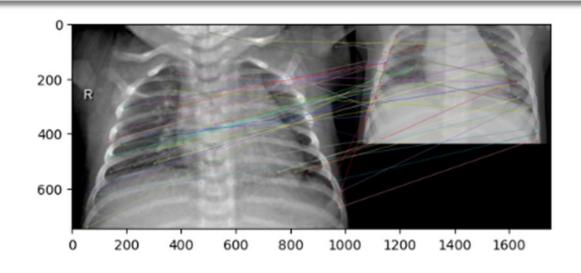


## SIFT Fllgorithm





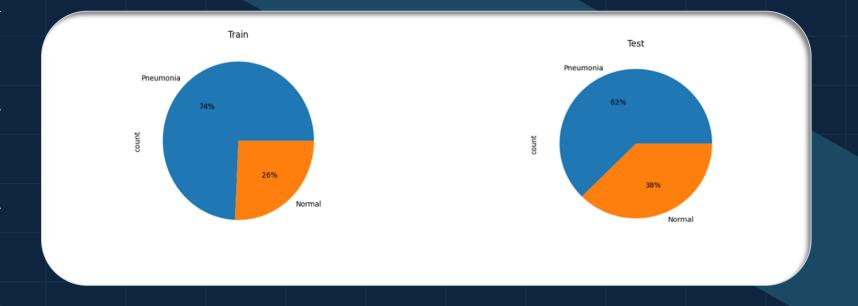
## SIFT Fllgorithm



#### Dataset

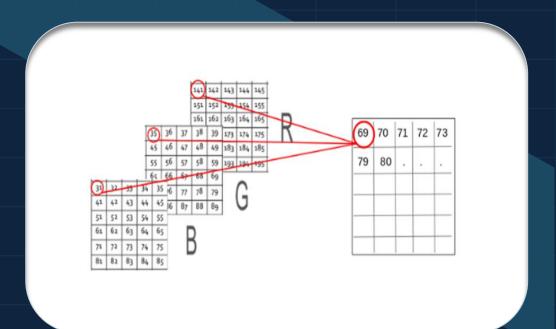
- Dataset consists of 5856 images of chest X-rays of which 4273 are pneumonia images and 1583 are normal chest X-ray images.
- A total of 80% of the data are used for training, producing 4642 images (3418 images of pneumonia and 1224 normal images)
- 15% of the data are used for testing, producing 919 images (641 cases of pneumonia and 278 normal images),
- the final 5% of the data are used for validation (214 cases of pneumonia and 81 non-pneumonia images).

## Dataset

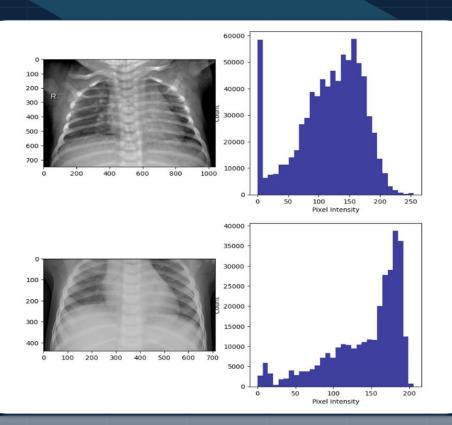


#### Feature Extraction

- Resize images to 64 x 64 pixels.
- We get three channels (r,g,b).
- Mean Pixel Value of Channels.
- Normalize data to have zero mean and unit variance.

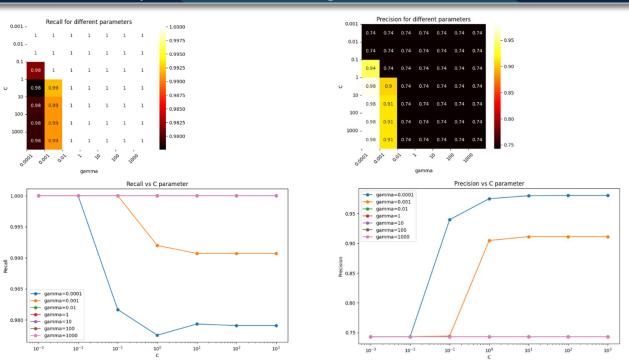


## Pixel intensity



#### Hyperparameter Tuning for svm with rbf kernel

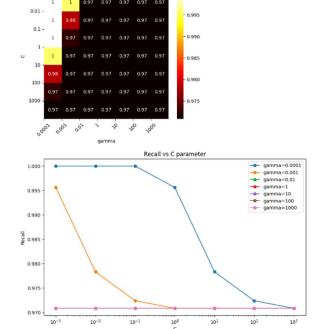
The hyperparameters we want to adjust in this case are C and gamma. As we can see the best option is C=100 and gamma=0.0001.



10cm

## Hyperparameter Tuning for svm with polynomial kernel

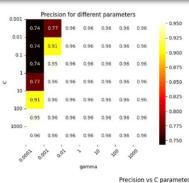
The hyperparameters we want to adjust in this case are C and gamma. As we can see the best option is C=1 and gamma=0.001.

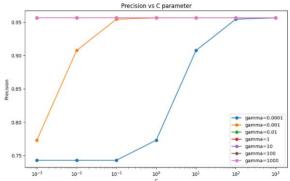


Recall for different parameters

0.001 -

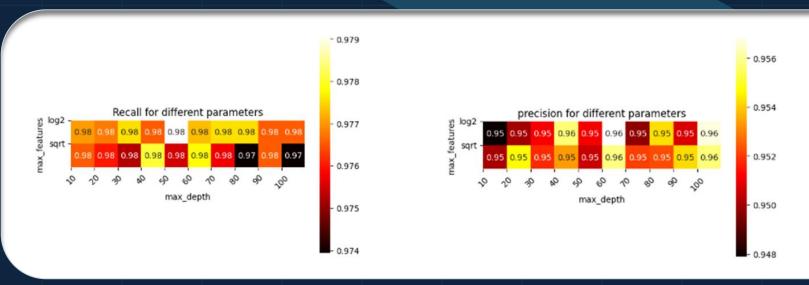
10cm





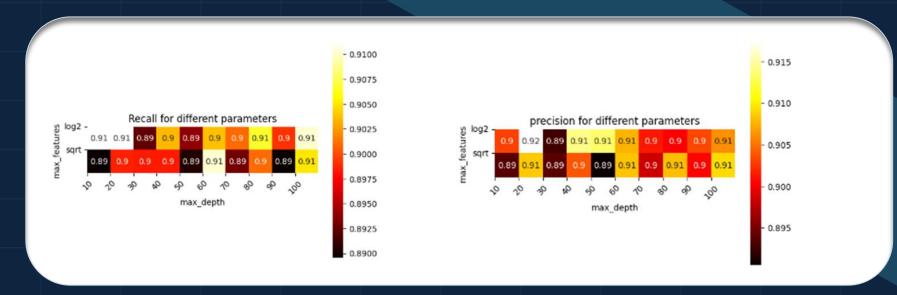
#### Hyperparameter Tuning for random forest

The hyperparameters we want to set in this case are max\_features and max\_depth. As we can see the best option is max\_features = log2 and max\_depth=50.



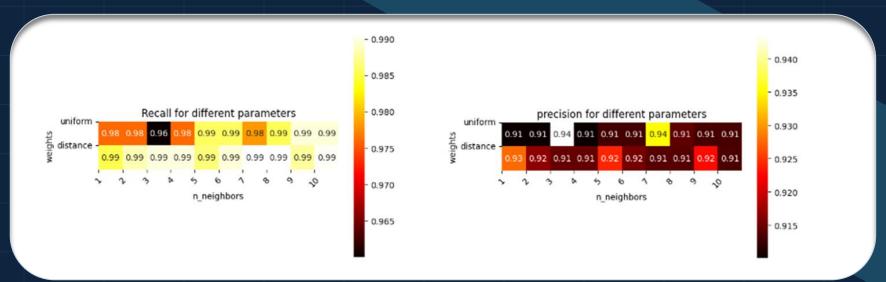
#### Hyperparameter Tuning for decision-tree

The hyperparameters we want to set in this case are max\_features and max\_depth. As we can see the best option is max\_features = log2 and max\_depth=20.

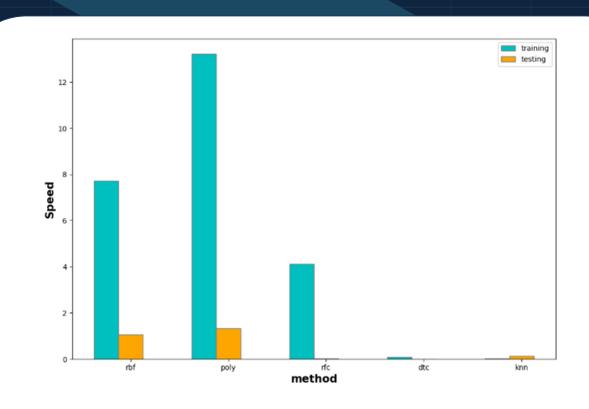


## Hyperparameter Tuning for KNN

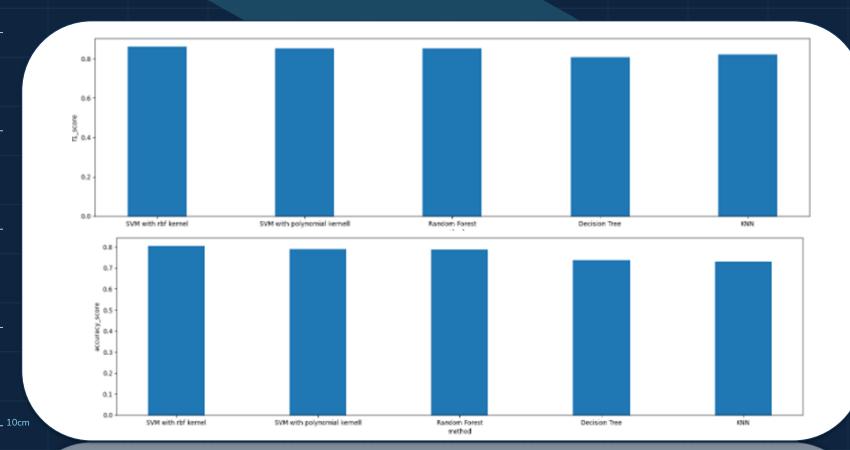
The hyperparameters we want to set in this case are n\_neighbors and weights. As we can see the best option is n\_neighbors=7 and weights= uniform.



## Model Selection



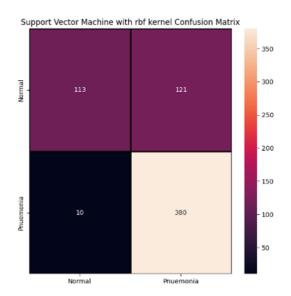
#### Model Selection



#### Model Selection

#### The model we choose is Svm with rbf kernel.

Support Vector Machine Classifier with rbf kernel accuracy score is:0.8044871794871795 recall score is:0.9743589743589743 precision score is:0.7723577235772358 f1 score is:0.8616780045351473 Support Vector Machine Classifier with polynomial kernell accuracy score is:0.7900641025641025 recall score is:0.9743589743589743 precision score is:0.7584830339321357 f1 score is:0.8529741863075196 Random Forest accuracy score 1s:0.7916666666666666 recall score is:0.9897435897435898 precision score is:0.75390625 f1 score is:0.8558758314855877 Decision Tree accuracy score is:0.6842948717948718 recall score is:0.8948717948717949 precision score is:0.691089108910891 f1 score 1s:0.7798882681564245 accuracy score is:0.7307692307692307 recall score is:0.9948717948717949 precision score is:0.7003610108303249 f1 score is:0.8220338983050848



#### User interface



#### User interface

