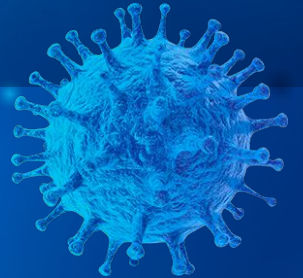
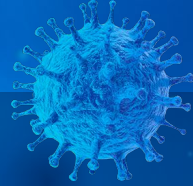


Pneumonia detection on chest X-ray images



Meet our team



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1. Problem Description

The background is a solid blue color with several stylized, spiky virus-like particles scattered across it. These particles are rendered in a lighter blue or white color, giving a sense of depth and movement. The overall aesthetic is clean and modern, typical of data visualization or public health communication.

2,560,000

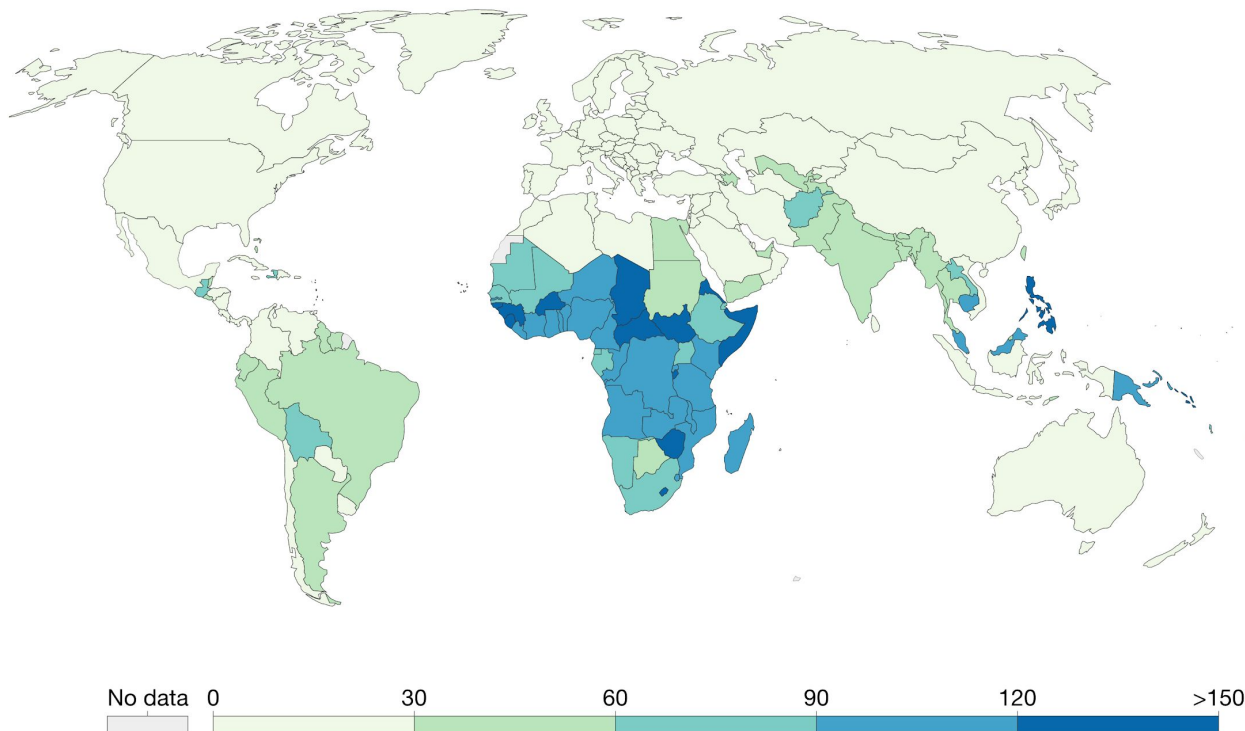
people died from pneumonia in 2017

<https://ourworldindata.org/pneumonia>

Death rate from pneumonia, 2017

The annual number of deaths from pneumonia per 100,000 people.


Our World
in Data



Source: Global Burden of Disease Study, IHME (2018)

OurWorldInData.org/pneumonia • CC BY

Note: To allow comparisons between countries and over time this metric is age-standardized. Deaths from 'clinical pneumonia', which refers to a diagnosis based on disease symptoms such as coughing and difficulty breathing and may include other lower respiratory diseases.



**Pneumonia is one
of the most lethal
COVID-19
complications**

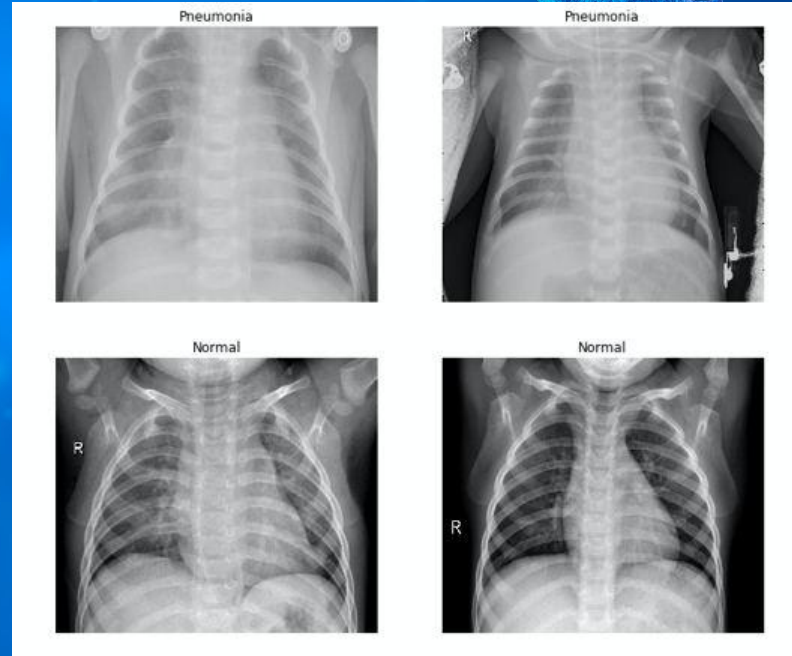
Expected value

- ▶ Rapid screen of all patients
- ▶ Aid in early detection of the disease
- ▶ Protection of health care workers



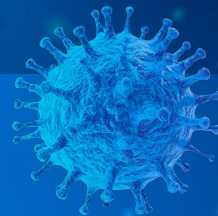
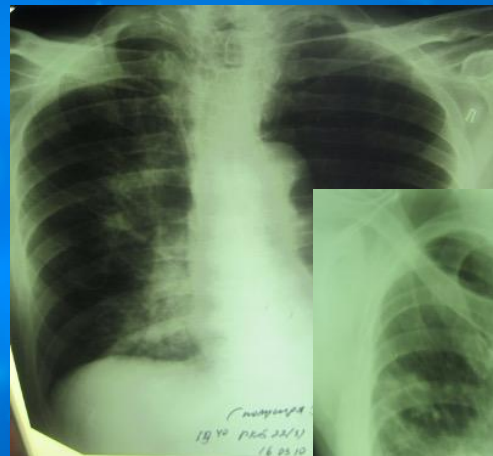
Dataset description

- ▶ Patients of 1-5 years old
- ▶ 5,863 X-Ray images
- ▶ 2 categories:
Pneumonia/Normal

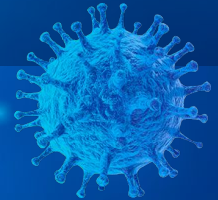
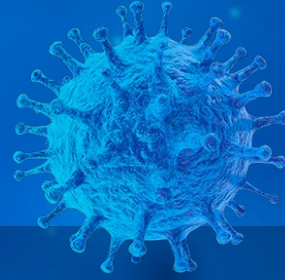


Additional dataset description

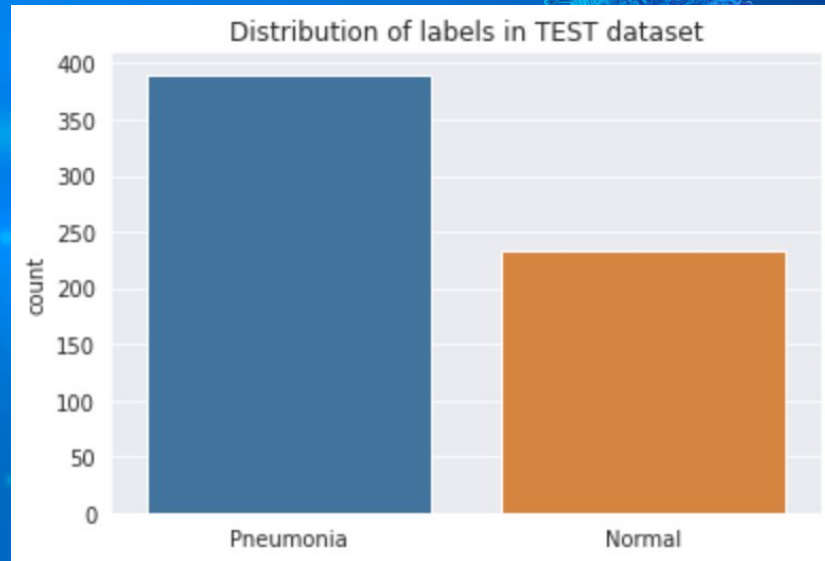
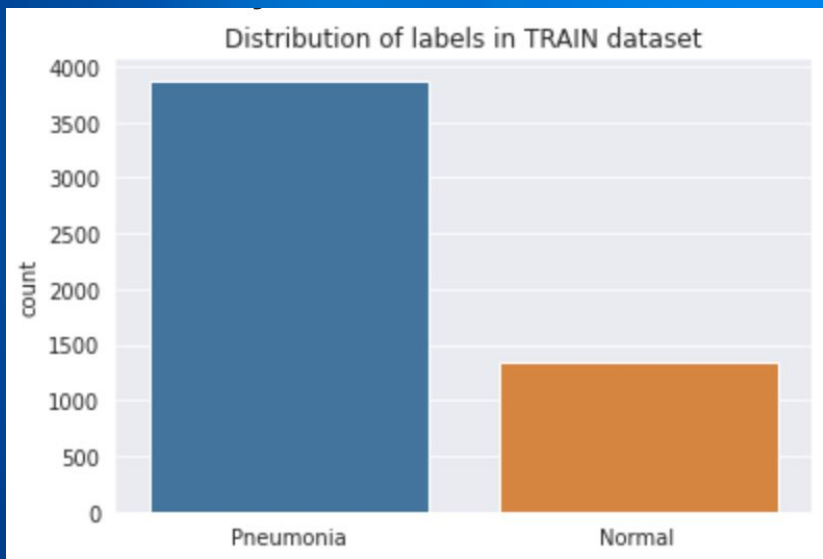
- ▶ From the personal archive by the doctor of Sechenov university, department of Pulmonology prof. Gainitdinova Viliya Vilevna
- ▶ 64 X-Ray images
- ▶ Only pneumonia

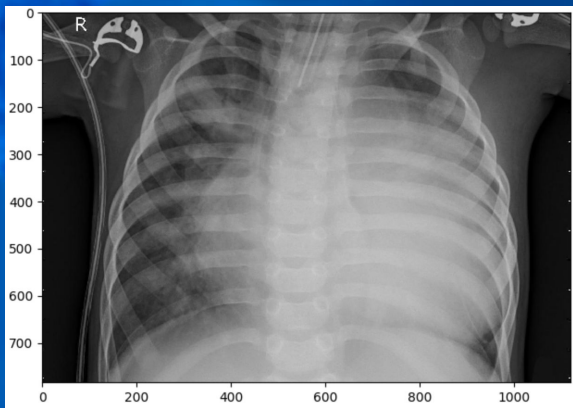
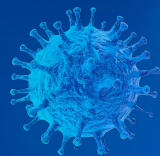


2. CV Algorithms

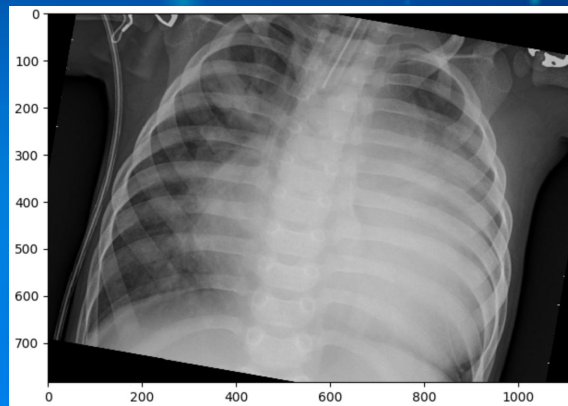


Data augmentation

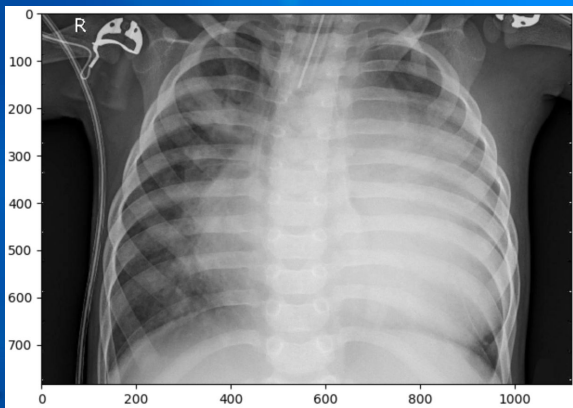




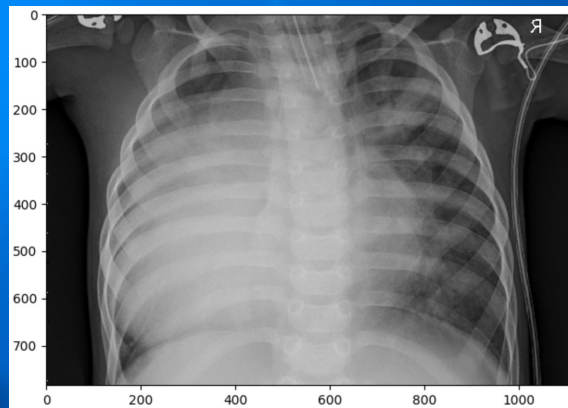
Original



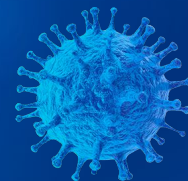
Random Rotation



Random Brightness

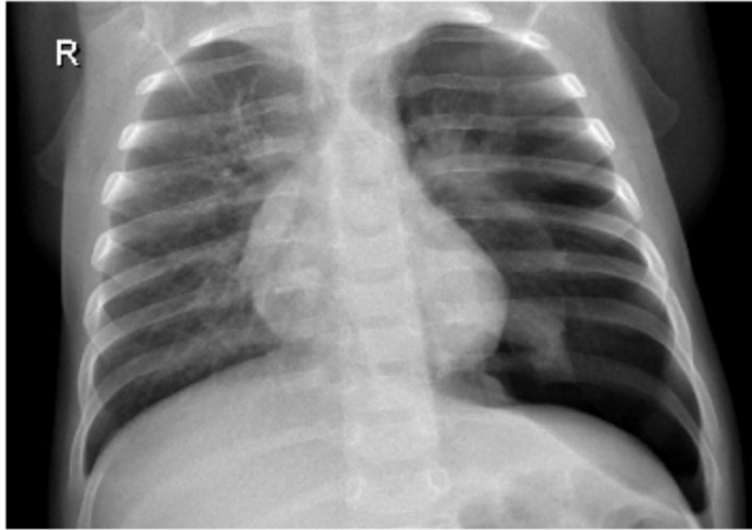


Horizontal Flip

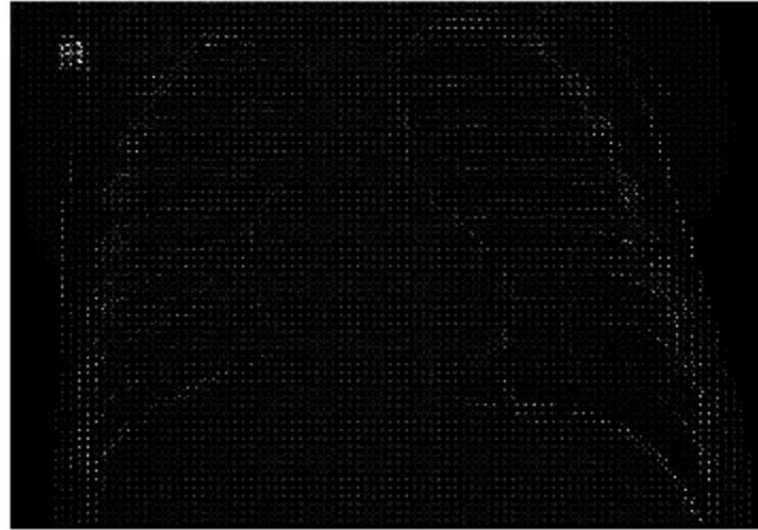


Feature extraction: HOG

Input image

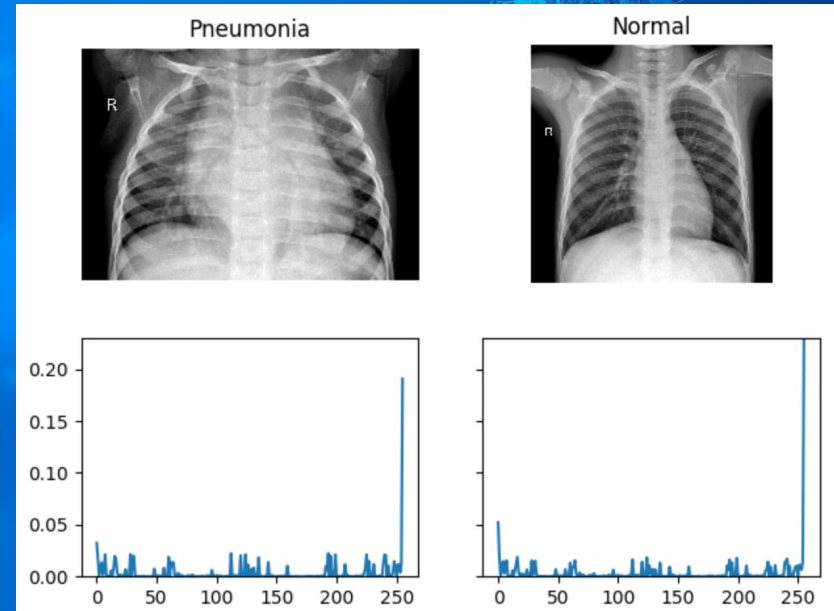


Histogram of Oriented Gradients

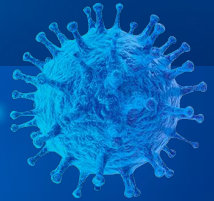
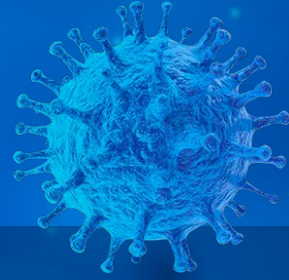


Feature extraction: LBP

- ▶ Local Binary Patterns
- ▶ Texture descriptor
- ▶ Labels the pixels of an image by thresholding the neighborhood of each pixel and considers the result as a binary number



3. Results



HOG + SVC (with data augmentation)

1. GridSearch parameters (cv = 5, verbose = 10):

```
param_grid = {'C': [0.001, 0.01, 0.1, 1, 10, 100],  
'gamma': [0.001, 0.01, 0.1, 1, 10, 100]}
```

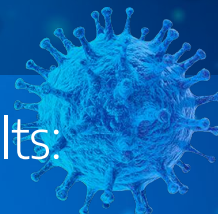


```
grid_search.best_params_  
{'C': 10, 'gamma': 0.1}
```

2. With the accuracy:

```
print(accuracy_score(labels_test, labels_hog_aug_pred))  
  
0.708974358974359
```

Results:



| | | |
|-----------|-----------|--------|
| PNEUMONIA | 387 | 3 |
| NORMAL | 224 | 166 |
| | PNEUMONIA | NORMAL |

HOG + SVC (without data augmentation)

1. GridSearch parameters (cv = 5):

```
param_grid = {'C': [0.01, 0.1, 1, 10, 100],  
              'gamma': [0.001, 0.01, 0.1, 1]}
```



```
grid_search.best_params_  
{'C': 0.01, 'gamma': 0.001}
```

2. With the accuracy:

```
print(accuracy_score(labels_test, labels_hog_pred))  
  
0.7532051282051282
```

Results:

| | PNEUMONIA | NORMAL |
|-----------|-----------|--------|
| PNEUMONIA | 388 | 2 |
| NORMAL | 152 | 82 |

LBP + SVC (with data augmentation)

1. GridSearch parameters (cv = 5):

```
C_vals = np.logspace(start=-1, stop=4, num=10, endpoint=True, dtype=float)
params = {'C': C_vals,
          'gamma': np.logspace(start=-1, stop=3, num=10)}
```



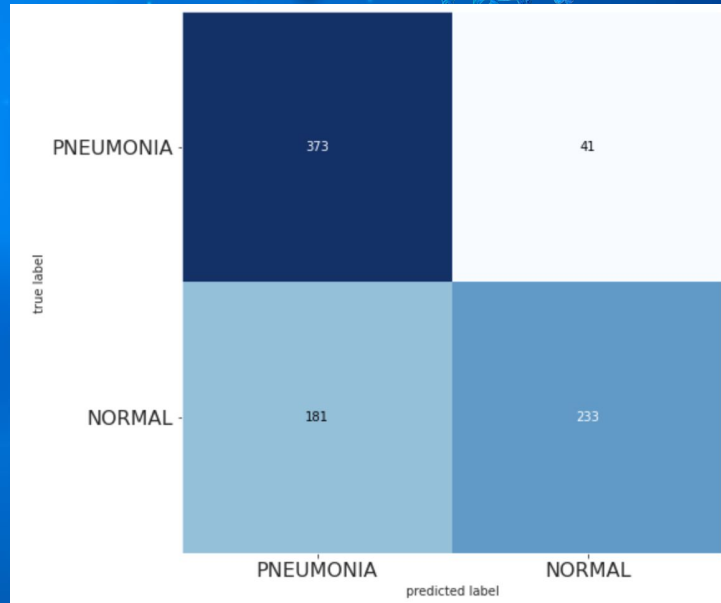
```
[51] grid_search_lbp.best_params_
      {'C': 215.44346900318845, 'gamma': 1000.0}
```

2. With the accuracy:

```
print(accuracy_score(y_aug_test, y_aug_test_pred))
print(grid_search_lbp.best_estimator_.score(X=X_aug_test, y=y_aug_test))
```

```
0.7318840579710145
0.7318840579710145
```

Results:



LBP + SVC (without data augmentation)

1. GridSearch parameters (cv = 5):

```
C_vals = np.logspace(start=-2, stop=4, num=15, endpoint=True, dtype=float)
params = {'C': C_vals,
          'gamma': np.logspace(start=-5, stop=5, num=15)}
```



```
grid_search_lbp.best_params_
{'C': 517.9474679231213, 'gamma': 26.826957952797272}
```

2. With the accuracy:

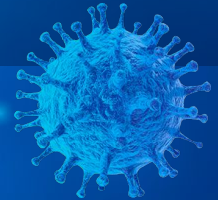
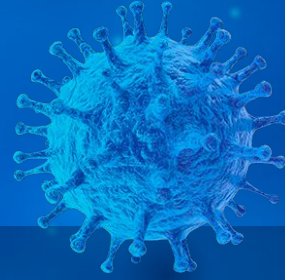
```
print(accuracy_score(y_test, y_test_pred))
print(grid_search_lbp.best_estimator_.score(X=X_test, y=y_test))
```

```
0.75625
0.75625
```

Results:

| | PNEUMONIA | NORMAL |
|-----------|-----------|--------|
| PNEUMONIA | 377 | 21 |
| NORMAL | 135 | 107 |

4. Conclusion



Results analysis

- ▶ Augmentation+LBP+SVC is the best according to confusion matrix
- ▶ Data augmentation may not help if imbalance is too dramatic
- ▶ Simple classical algorithms can still give decent results
- ▶ Results can be improved by:
 - ▷ Stronger classification algorithms (for example, Logistic Regression)
 - ▷ Neural Networks



THANKS!

Any questions?

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