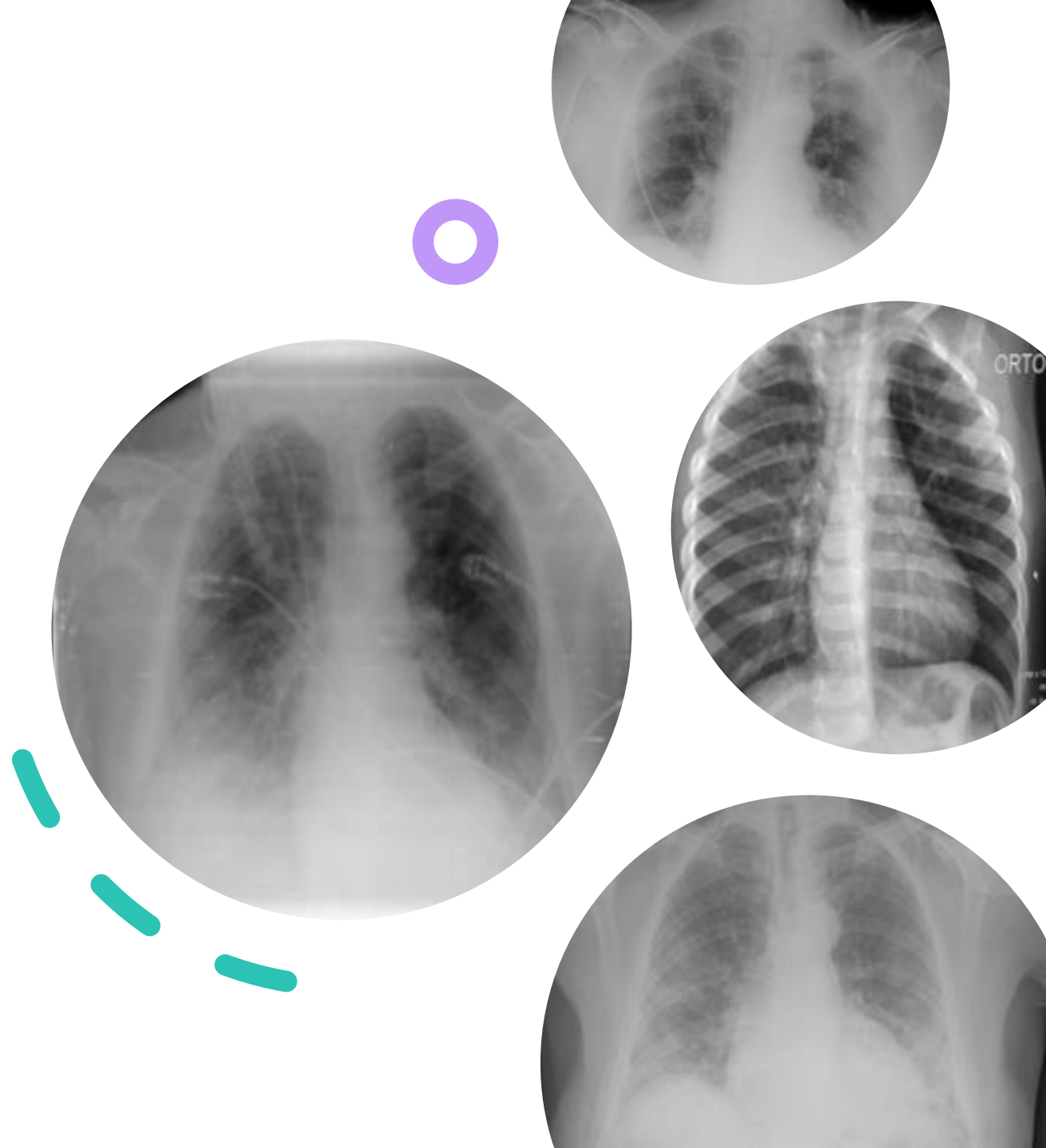


# Imaging's Potential to Assist in COVID-19 Crisis

- CSIC 5011 / MATH 5473 Project 2 -

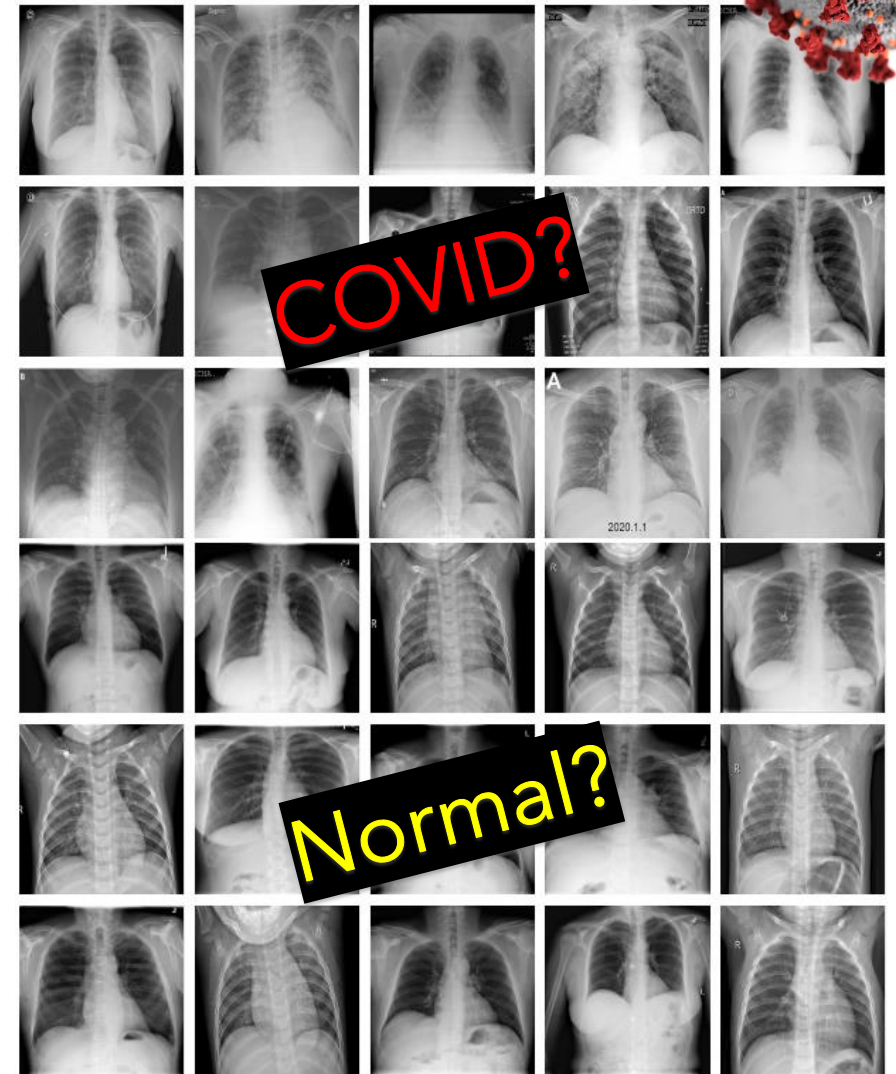
Yingshu CHEN<sup>1</sup>, Wing Hei SUM<sup>2</sup>, Ho Pan IP<sup>3</sup>, Zipeng WU<sup>4</sup>

Department of Computer Science and Engineering<sup>1</sup>, Industrial Engineering and Decision Analytics<sup>2</sup>, Mathematics<sup>3</sup>, Physics<sup>4</sup>



# Motivation

- Severe COVID-19 pandemic
  - How to detect COVID without detection kit?
  - How to speed up detection?
- Chest X-ray Image Classification!
  - Issue: lack sufficient data
- Data augmentation via synthesis!



# Outline

- Data and observation
- Data augmentation using generative model
- Data distribution
- Classification results

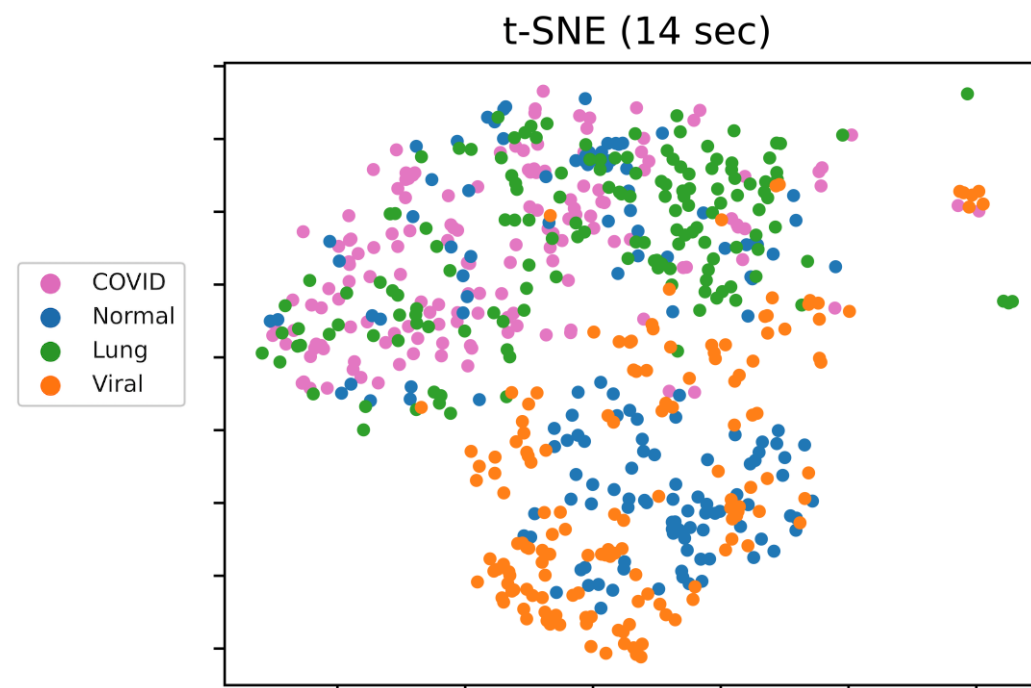
# Data Observation

- Data\* Statistics
  - Small original dataset
  - 20k images
  - Viral PNA only 1k
- Real Data visualization
  - Normal is sparse
  - Normal and Viral are close
  - COVID and Lung are close

Table 1: Data Statistics

Class	Normal	COVID-19	Lung Opacity <sup>a</sup>	Viral Pneumonia <sup>b</sup>
Train	9,942	3,366	5,762	1,095
Test	250	250	250	250
Total	10,192	3,616	6,012	1,345
Blended Train	9,942	7,000	5,762	7,000

<sup>a,b</sup> Non-COVID lung infection.



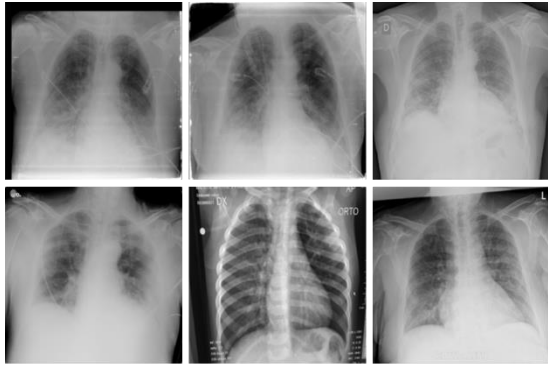
\*Data source: <https://www.kaggle.com/tawsifurrahman/covid19-radiography-database>

# Data Augmentation

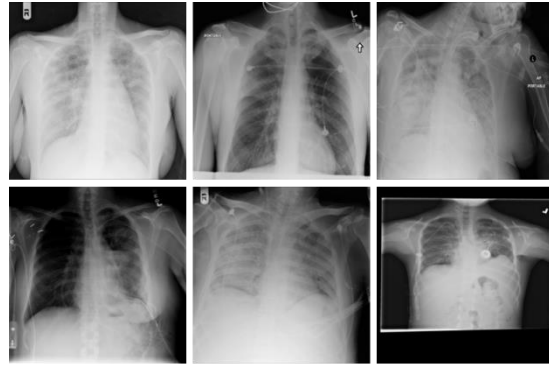
- Generative Model - SOTA conditional GAN
  - *stylegan2-ada* model with limited data (thousand scale)
  - Photorealistic synthesis
  - Unconditional vs Conditional GAN

# Data Augmentation

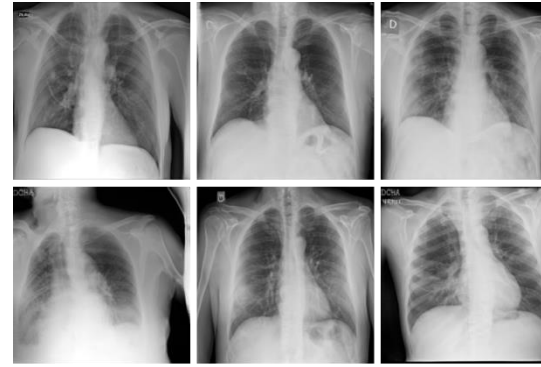
## - Photorealism of Synthesis



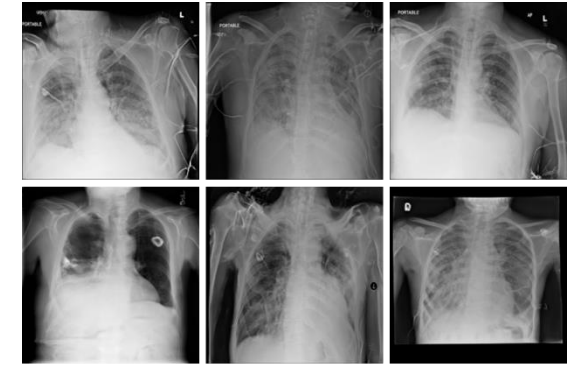
Real COVID-19 CXRs



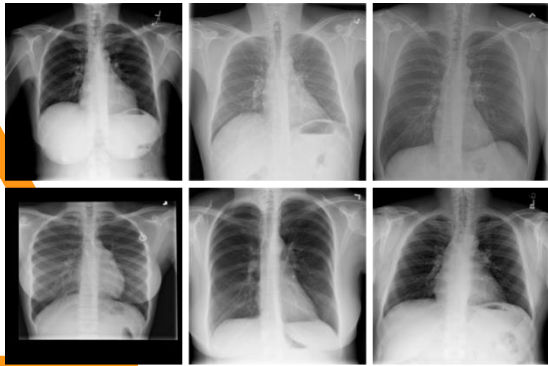
Real Lung Opacity CXRs



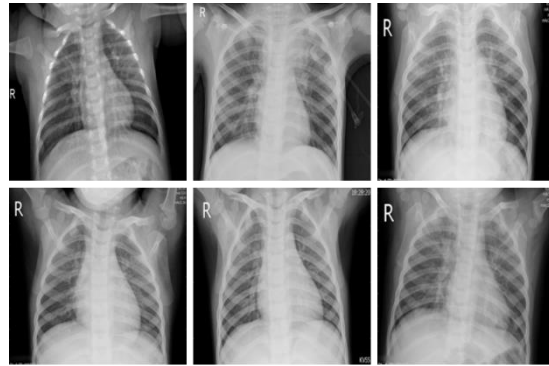
Synthetic COVID-19 CXRs



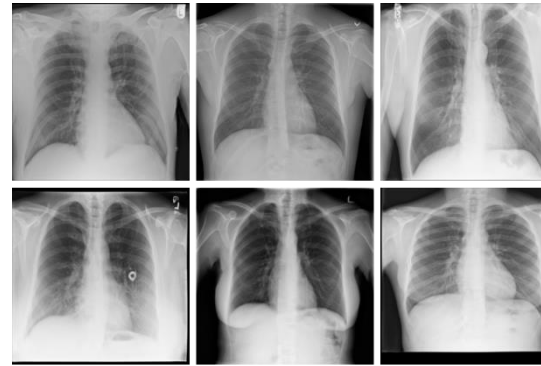
Synthetic Lung Opacity CXRs



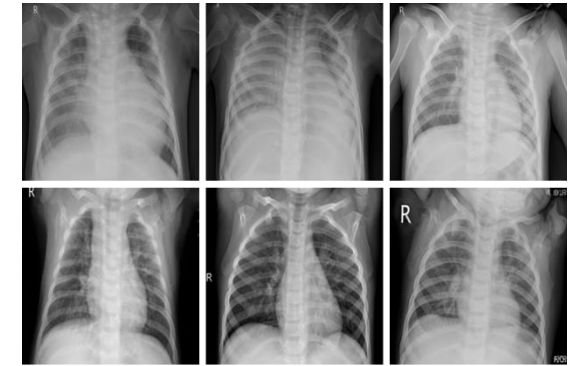
Real Normal CXRs



Real Viral PNA CXRs



Synthetic Normal CXRs



Synthetic Viral PNA CXRs



# Data Augmentation

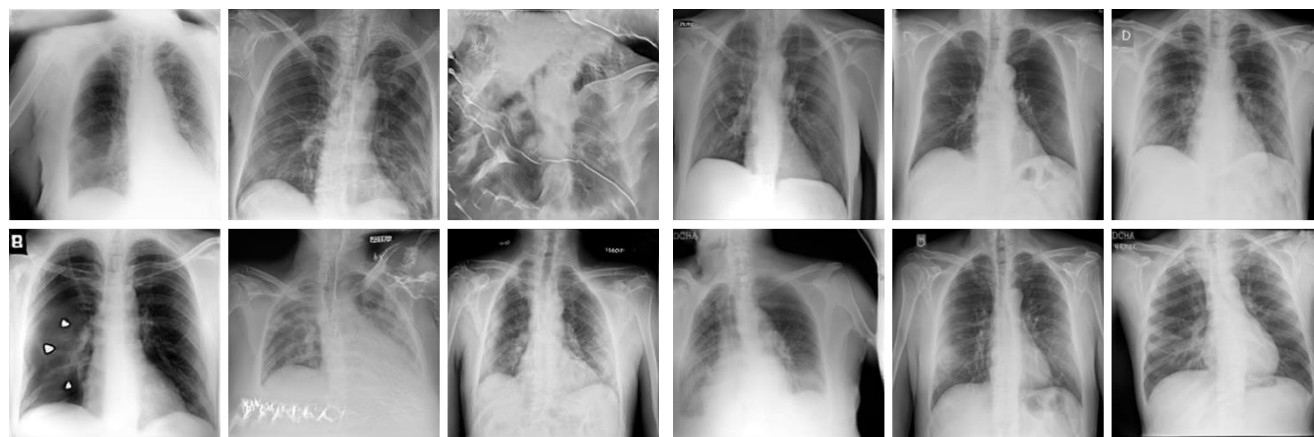
- Generative Model
  - SOTA conditional GAN
    - *stylegan2-ada* model with limited data (thousand scale)
    - Photorealistic synthesis
    - Unconditional vs Conditional GAN
  - Expected to assist in data augmentation

Table 3: Unconditional vs. Conditional Generative Model

Model*	Unconditional				Conditional			
	Normal	COVID-19	Lung Opacity	Viral PNA	Normal	COVID-19	Lung Opacity	Viral PNA
FID↓	18.28	33.48	28.74	28.25	<b>16.39</b>	<b>25.24</b>	<b>24.22</b>	<b>26.85</b>

\*Four unconditional models and one conditional model with 4 labels

Lower FID means closer to real data distribution

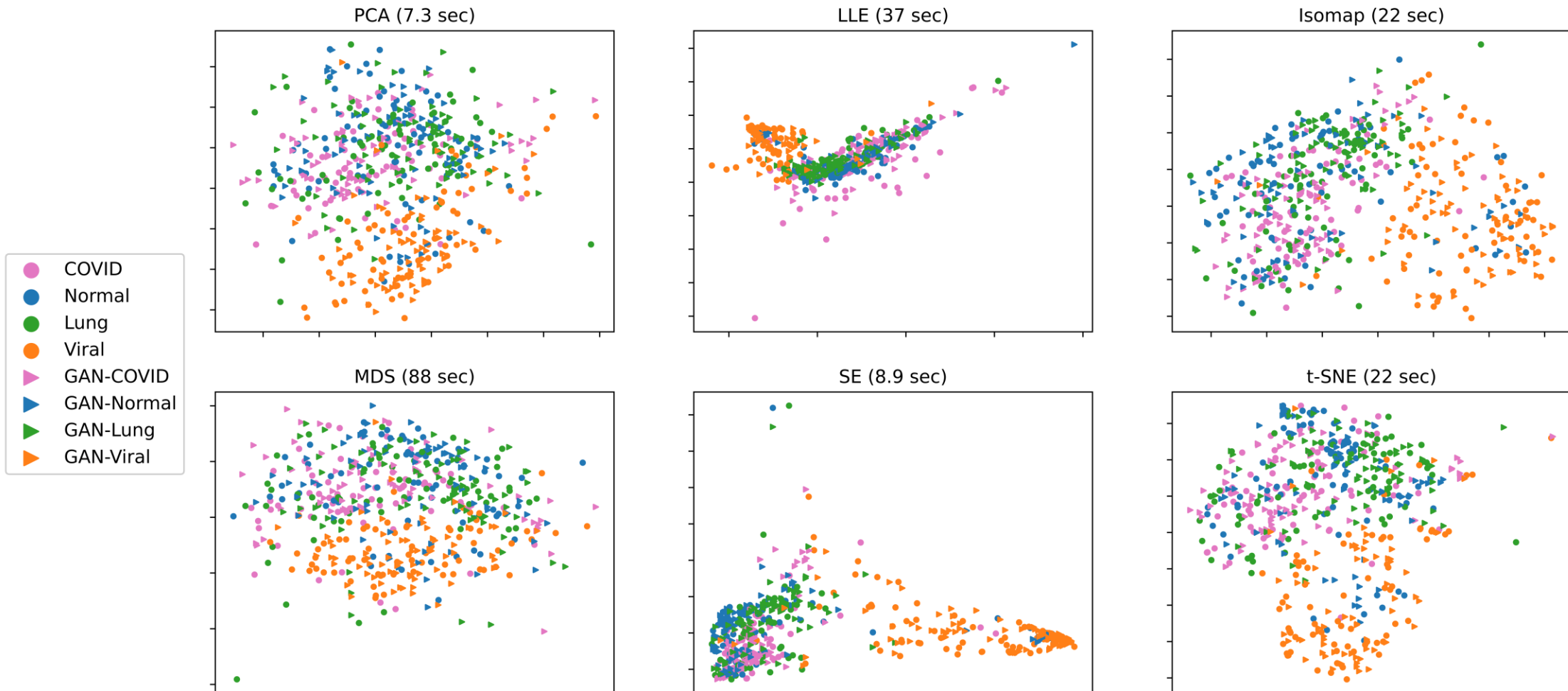


Unconditional - COVID-19

Conditional - COVID-19

# Blended Data Distribution Visualization

- Generative model to some extent reproduce real data distribution





# Classification

- Method description
  - Convolutional Neural Network (CNN)
  - Layers: 13
  - Architecture: Conv2D, Dense, Dropout, MaxPooling2D, Flatten
  - Activation functions: ReLU and Softmax
  - Early stopping to avoid over-fitting

Table 2: Hyper-parameter settings

Optimizer	Loss	Metrics	Batch Size	Epochs
Adam	sparse_categorical_crossentropy	Accuracy	32	20

## • Blended Data

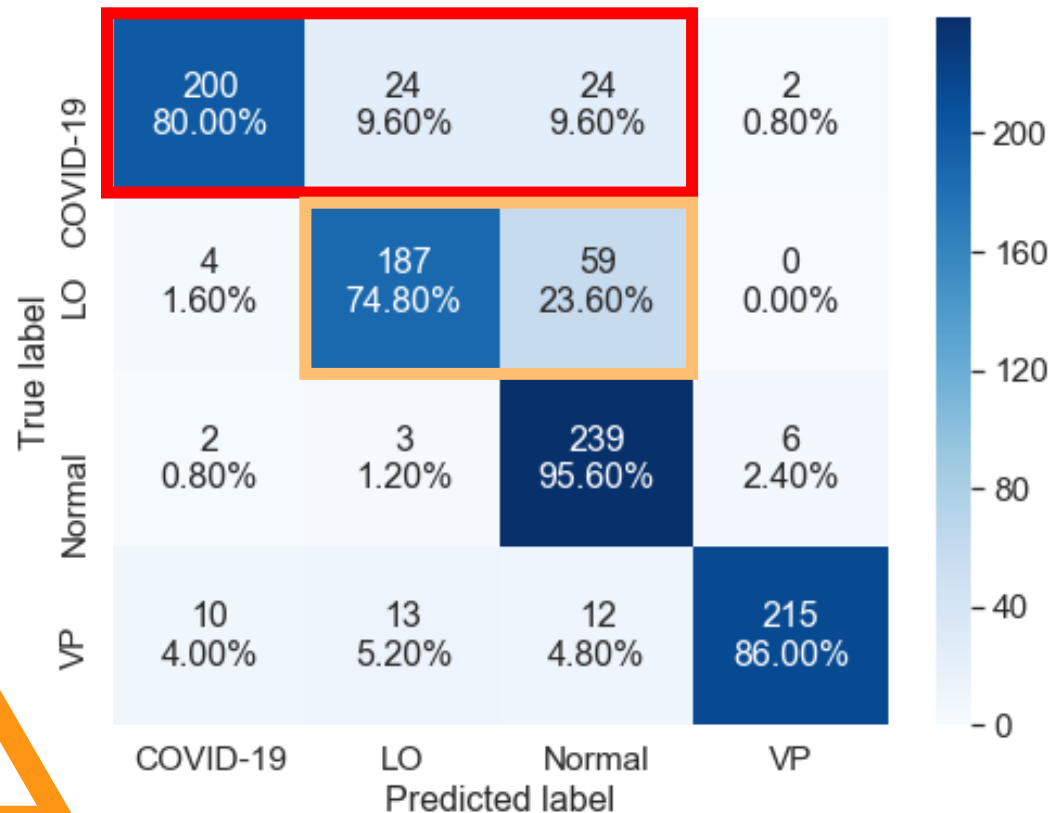
Table 5: Experiments on Data Combination

Data	Validation Acc <sup>a</sup>	Prediction Acc per Class <sup>b</sup>			
		COVID	Lung	Normal	Viral
Real only	86.41%	92.4%	75.2%	82.4%	92.0%
Real + 3,634 COVID synthesis	88.04%	90.4%	<b>82.4%</b>	89.2%	95.2%
Real + 1,238 Lung synthesis	86.05%	88.8%	81.2%	88.8%	90.4%
Real + 5,905 Viral synthesis	89.09%	94.4%	80.04%	87.6%	94.8%
Real + 3,634 COVID & 5,905 Viral synthesis	<b>90.35%</b>	<b>95.2%</b>	80.0%	<b>91.6%</b>	96.4%
Real + 3,634 COVID&1,238 Lung&5,905 Viral synthesis	89.35%	94.4%	81.2%	89.2%	<b>97.2%</b>

<sup>a, b</sup> Different from Sec.4.1, here we use  $\#correct/\#prediction$  in validation set<sup>a</sup> and testing set<sup>b</sup> as accuracy.

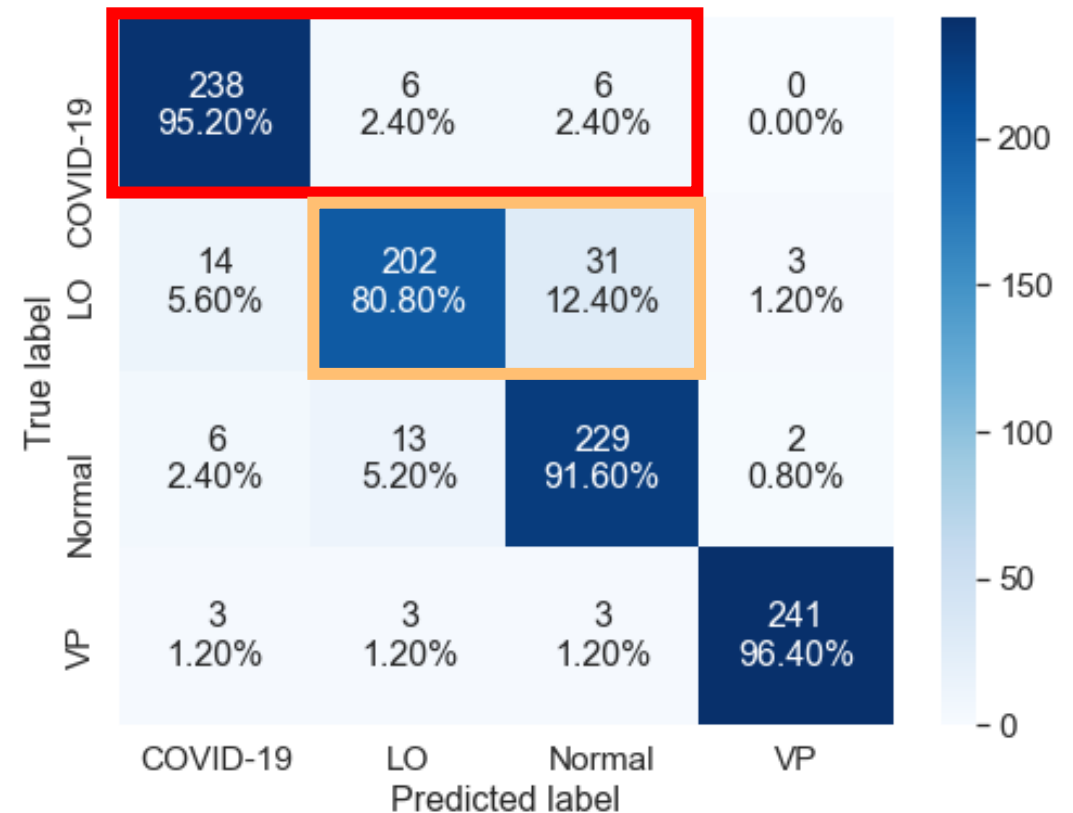
# Evaluation on the Classification

## Real Data



**Overall Accuracy: 84.1%**

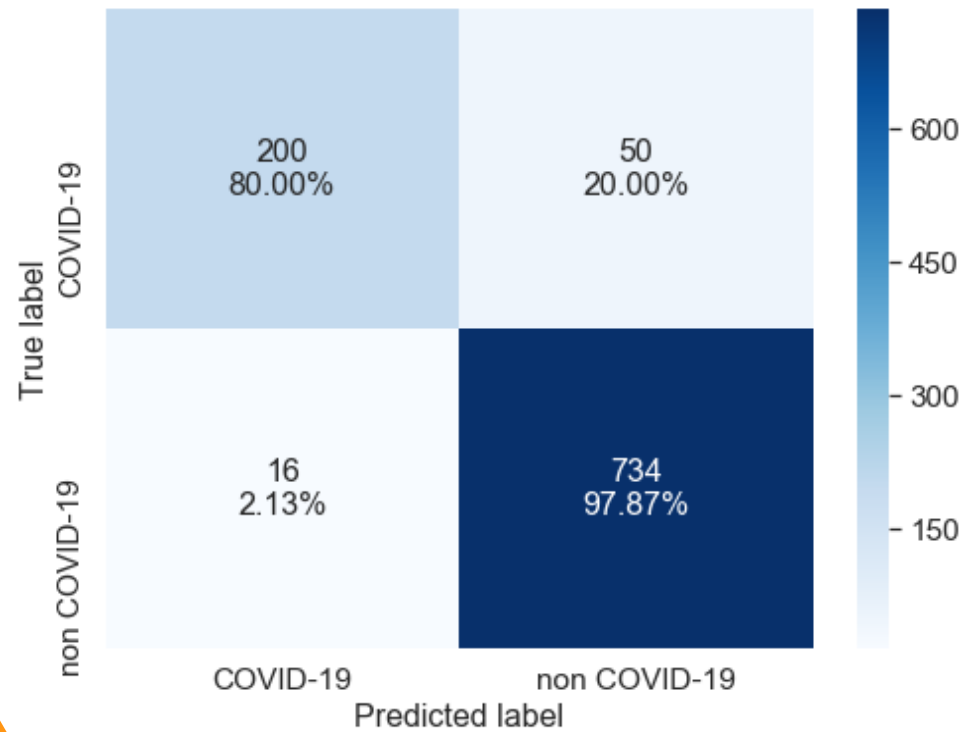
## Real Data + Synthetic Data (COVID-19 & Viral)



**Overall Accuracy: 91%**

# Evaluation on the Classification

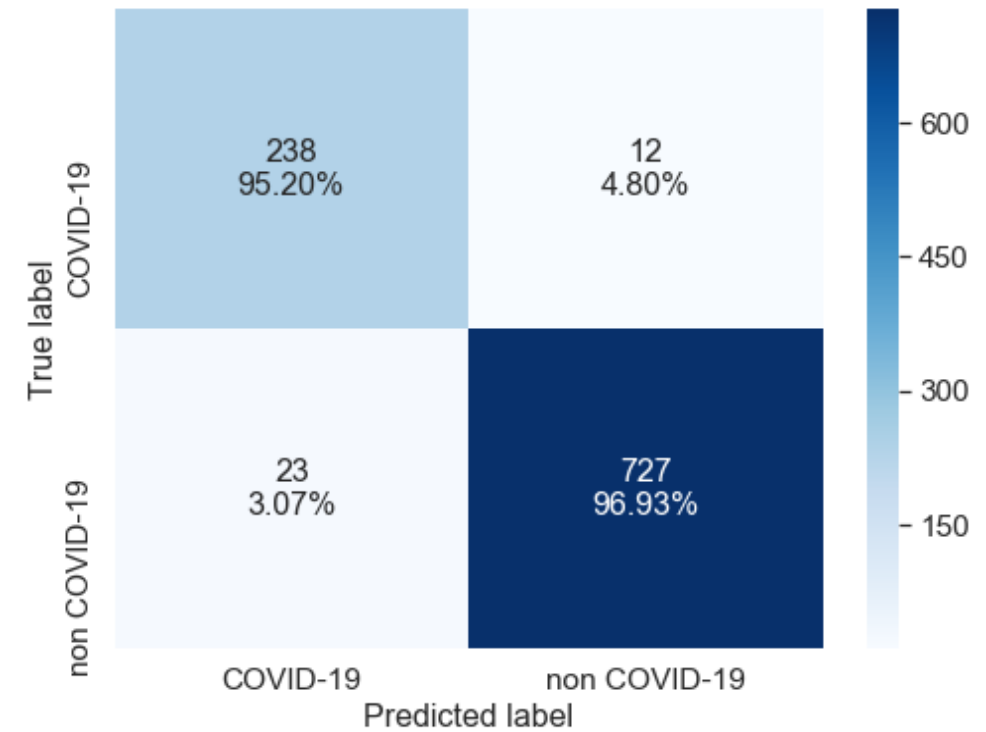
## Real Data



Accuracy=0.934  
Precision=0.936  
Recall=0.979  
F1 Score=0.957



## Real Data + Synthetic Data (COVID-19 & Viral)



Accuracy=0.965  
Precision=0.984  
Recall=0.969  
F1 Score=0.976

# Summary

- Dimensionality reduction methods can work as a visualization and evaluation means to visualize data distribution and get data distribution observation.
- High-quality chest X-ray images synthesis with limited data can be synthesized via a conditional GAN.
- We transfer the COVID-19 CXR detection problem to a 4-class classification problem considering other non-COVID lung infectious CXRs as well.
- We validate the effectiveness of synthetic images for medical data augmentation and visual classification enhancement.
- Qualitatively and quantitatively verify the potential of image generative model to assist in new respiratory infectious disease detection.