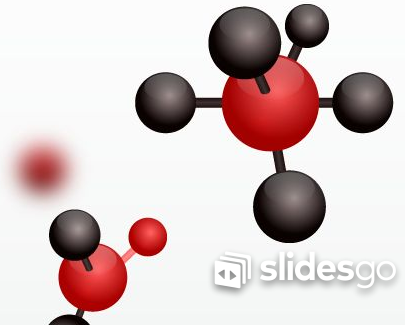




CLASSIFYING PNEUMONIA

Detecting pediatric pneumonia with
convolutional neural networks



OVERVIEW

01

BACKGROUND

Disease overview

02

METHODOLOGY

Presentation of methods
explored

03

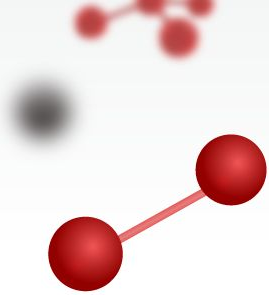
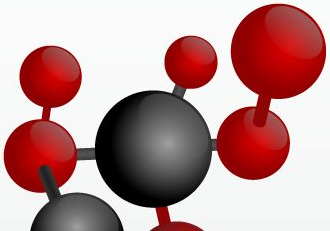
RESULTS ANALYSIS

Discussion of results

04

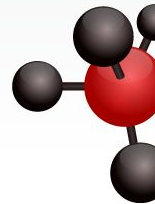
CONCLUSIONS

Next steps and
recommendations

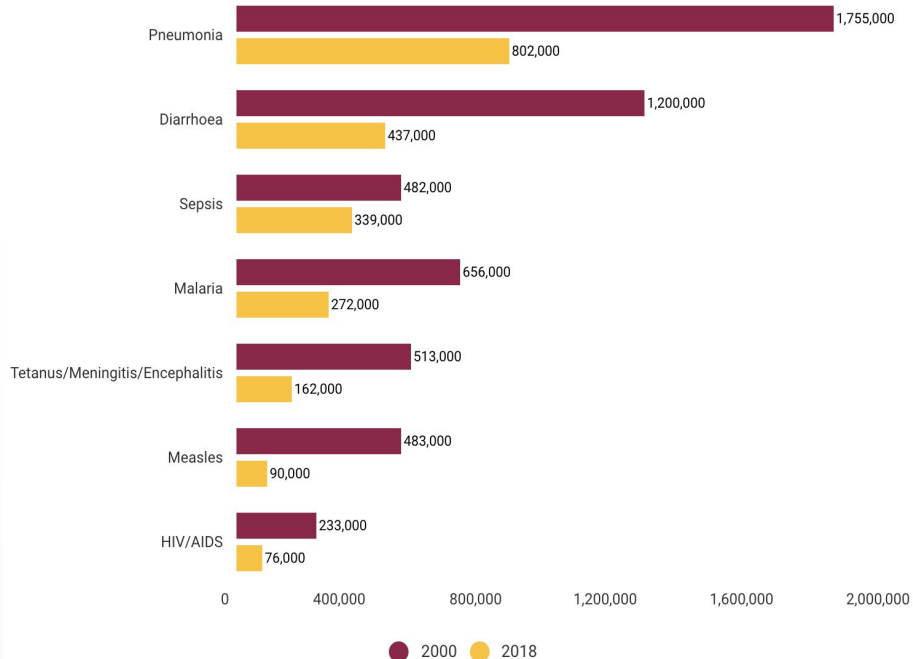




PNEUMONIA BACKGROUND



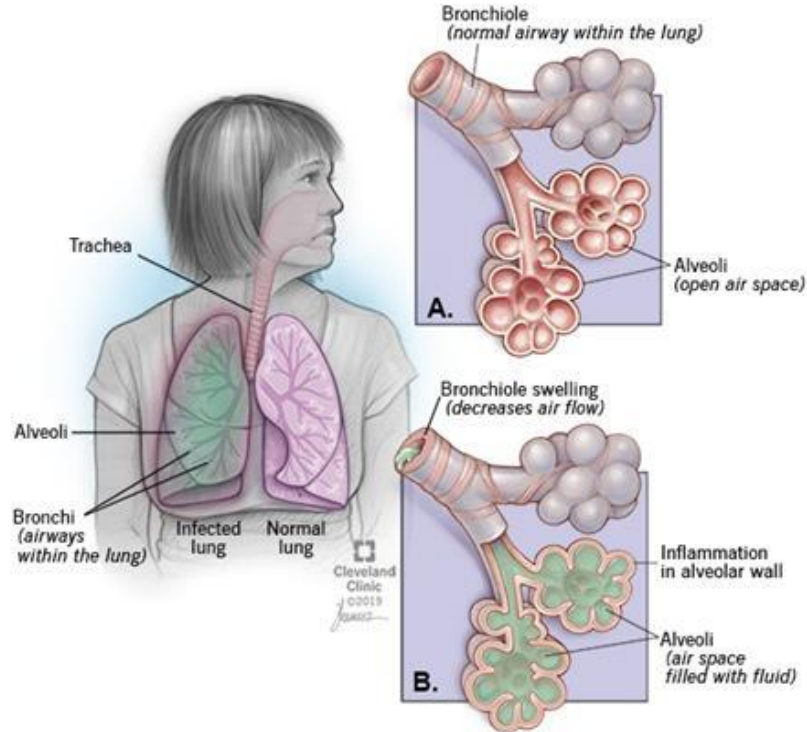
Deaths of children under five by infectious disease, 2000 vs 2018



Source: UNICEF analysis based on WHO and Maternal and Child Epidemiology Estimation Group interim estimates produced in September 2019, applying cause of deaths for the year 2017 to United Nations Inter-agency Group for Child Mortality Estimation estimates for the year 2018.

- A lung disease characterized by inflammation of the air sacs in the lungs
- Leading cause of death for children under the age of five
- Significantly slower progress for reducing deaths in children compared to any other infectious disease
- Correct diagnosis of pneumonia is essential

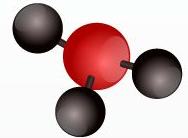
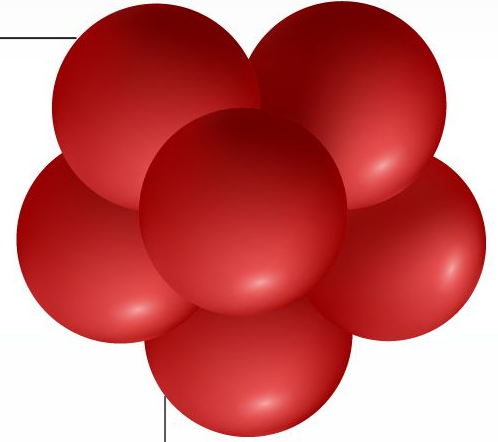
PNEUMONIA DIAGNOSIS

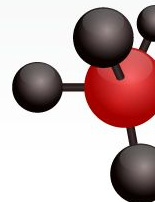
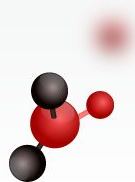


- Symptoms similar to flu or cold
- Chest X-ray is the typical method of diagnosis
- Diagnosis usually performed by expert physicians

PROBLEM STATEMENT

Can we build an automated system to accurately detect pneumonia based on chest X-rays?





DATA OVERVIEW

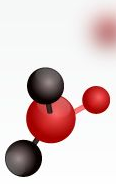


Pneumonia chest X-ray

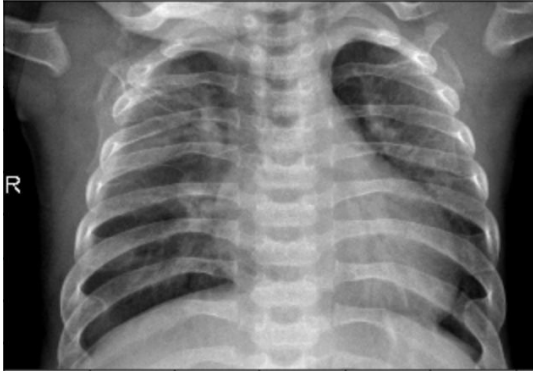


Normal chest X-ray

- 5,856 chest X-rays of pediatric patients ages one to five, from Guangzhou Women and Children's Medical Center
- Data obtained from Kaggle
- Varying image sizes
- 5,216 images in the training set
- Imbalanced data classes
 - Normal: 1,341
 - Pneumonia: 3,875



DATA AUGMENTATION



Original

- All Images rescaled
- Training images randomly flipped, rotated, zoomed, and/or skewed
- Increase training set



Horizontal Flip



Rotation Range (20)

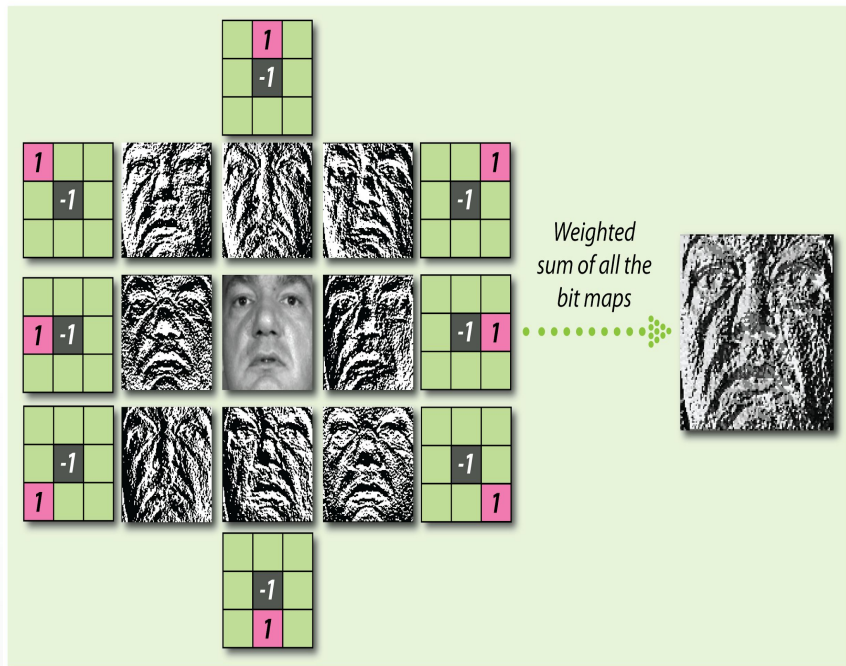


Zoom Range (0.2)



Shear Range (0.2)

TECHNIQUE OVERVIEW



- Neural Networks are a machine learning technique designed to recognize patterns
- Ability to improve with increased training and data size
- Convolutional Neural Networks are a type of NN which are especially effective at visual analysis
- Transfer learning
 - Use pretrained models (models trained on millions of other images) as the base layer of my model
 - Aid in image processing since all images have the same base operations such as edge detection

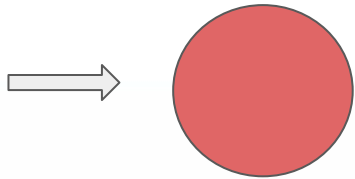
MODEL ARCHITECTURE

Input



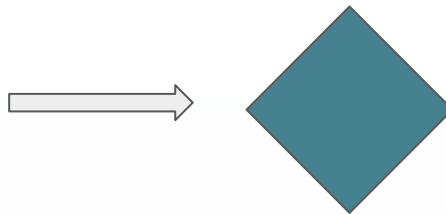
Inputs are processed images after data augmentation

Pretrained model



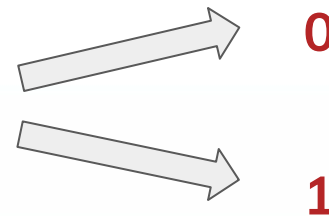
Pretrained models are effective at extracting base features from images

Custom Architecture



Custom architecture adapts the pretrained models to our specific problem

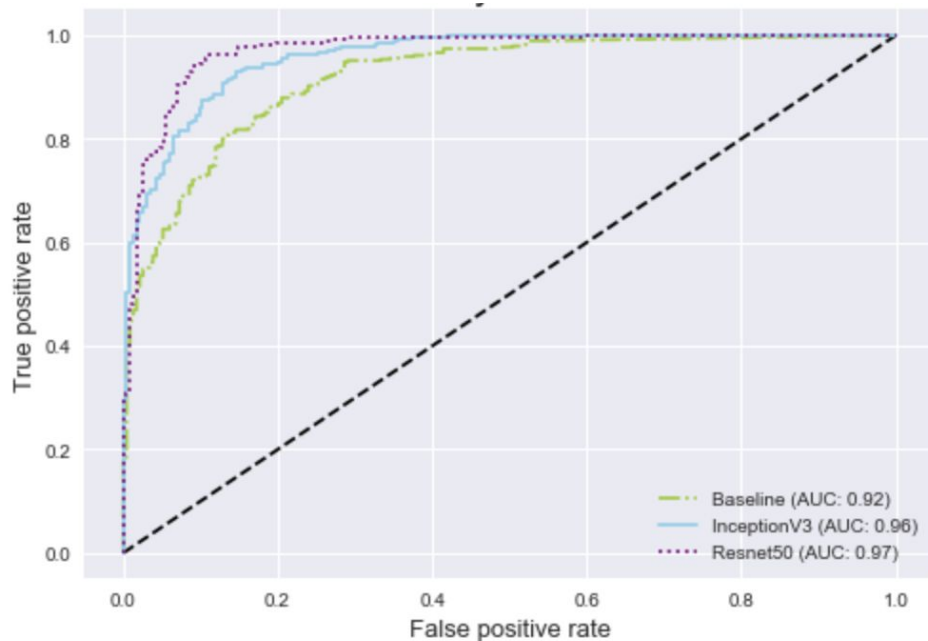
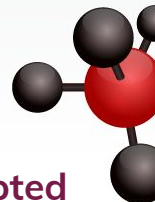
Classification



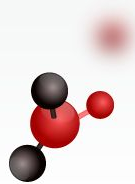
Output layer predicts binary class



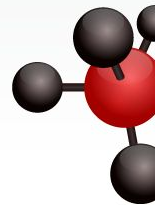
PRELIMINARY RESULTS



	Custom	Adapted InceptionV3	Adapted Resnet50
Accuracy	85.1%	89.1%	92.6%
AUROC	0.92	0.96	0.97
Recall	.93(P) .71(N)	.94(P) .80(N)	.94(P) .90(N)
Precision	.84(P) .87(N)	.89(P) .90(N)	.94(P) .91(N)



TYPES OF PNEUMONIA



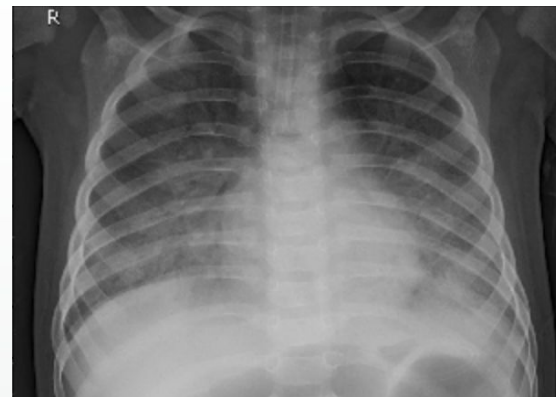
BACTERIAL

- Typically more localized
- Require antibiotic treatment
- 2533 images

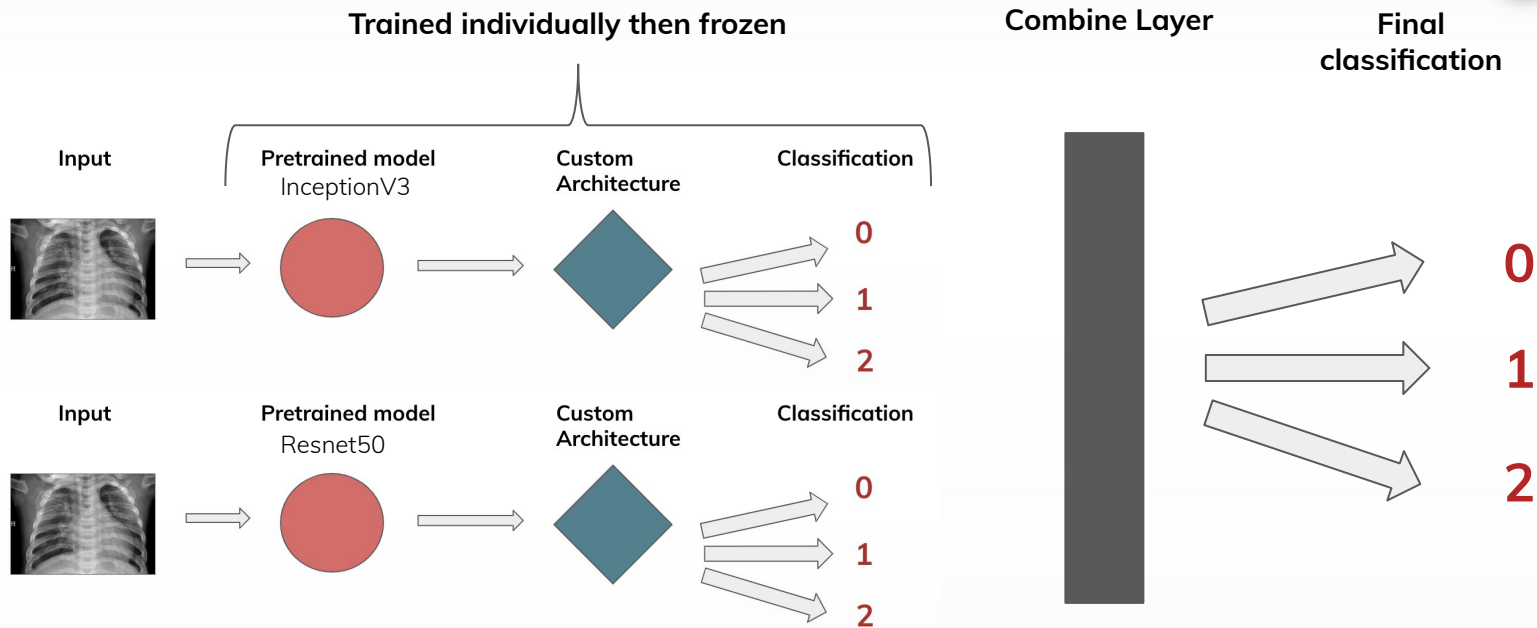


VIRAL

- Typically more diffused
- Predisposed to secondary bacterial pneumonia
- Contagious
- 1342 images

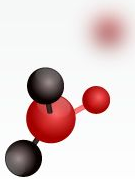


ENSEMBLE MODEL



Two stage process:

- Leverage two different pretrained network architectures
- Freeze weights learned in stage 1 and build an ensemble model to combine the predictions



RESULTS

	Accuracy	AUROC	Recall	Precision
InceptionV3	85.7%	0.98 (B) 0.97 (N) 0.94 (V)	0.95 (B) 0.82 (N) 0.77 (V)	0.85 (B) 0.95 (N) 0.75 (V)
Resnet50	81.7%	0.97 (B) 0.96 (N) 0.91 (V)	0.90 (B) 0.75 (N) 0.78 (V)	0.82 (B) 0.97 (N) 0.65 (V)
Ensemble	86.1%	0.98 (B) 0.97 (N) 0.94 (V)	0.95 (B) 0.81 (N) 0.78 (V)	0.85 (B) 0.97 (N) 0.75 (V)

CONFUSION MATRIX

ENSEMBLE

True Labels	BACTERIA	NORMAL	VIRUS
	231 (0.95)	5 (0.02)	6 (0.02)
	11 (0.05)	190 (0.81)	33 (0.14)
	31 (0.21)	1 (0.01)	116 (0.78)
	BACTERIA	NORMAL	VIRUS

Predicted Labels

INCEPTIONV3

True Labels	BACTERIA	NORMAL	VIRUS
	230 (0.95)	7 (0.03)	5 (0.02)
	9 (0.04)	191 (0.82)	34 (0.15)
	31 (0.21)	3 (0.02)	114 (0.77)
	BACTERIA	NORMAL	VIRUS

Predicted Labels



REVIEW

Conclusion

We built an automated system that is not only able to accurately detect pneumonia, but is also successful at identifying the type of the disease.

Benefits

Our work has several important benefits:

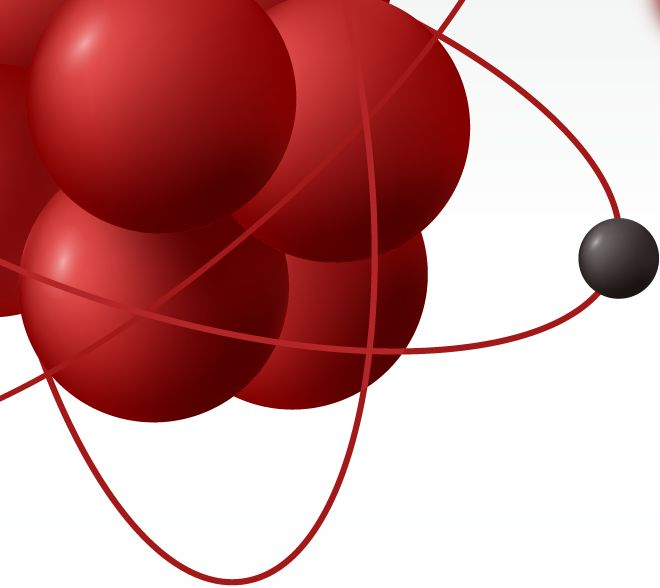
- Support physicians and expedite the diagnosis process
- Improve patient routing by prioritizing high likelihood cases
- Help design effective treatment plans

Next Steps

We propose several extensions to our work:

- Increase size of dataset
- Incorporate additional models in the ensemble
- Implement class weight function to address class imbalance
- Tune loss function to achieve desired levels of recall and precision
- Increase training time





THANK YOU

