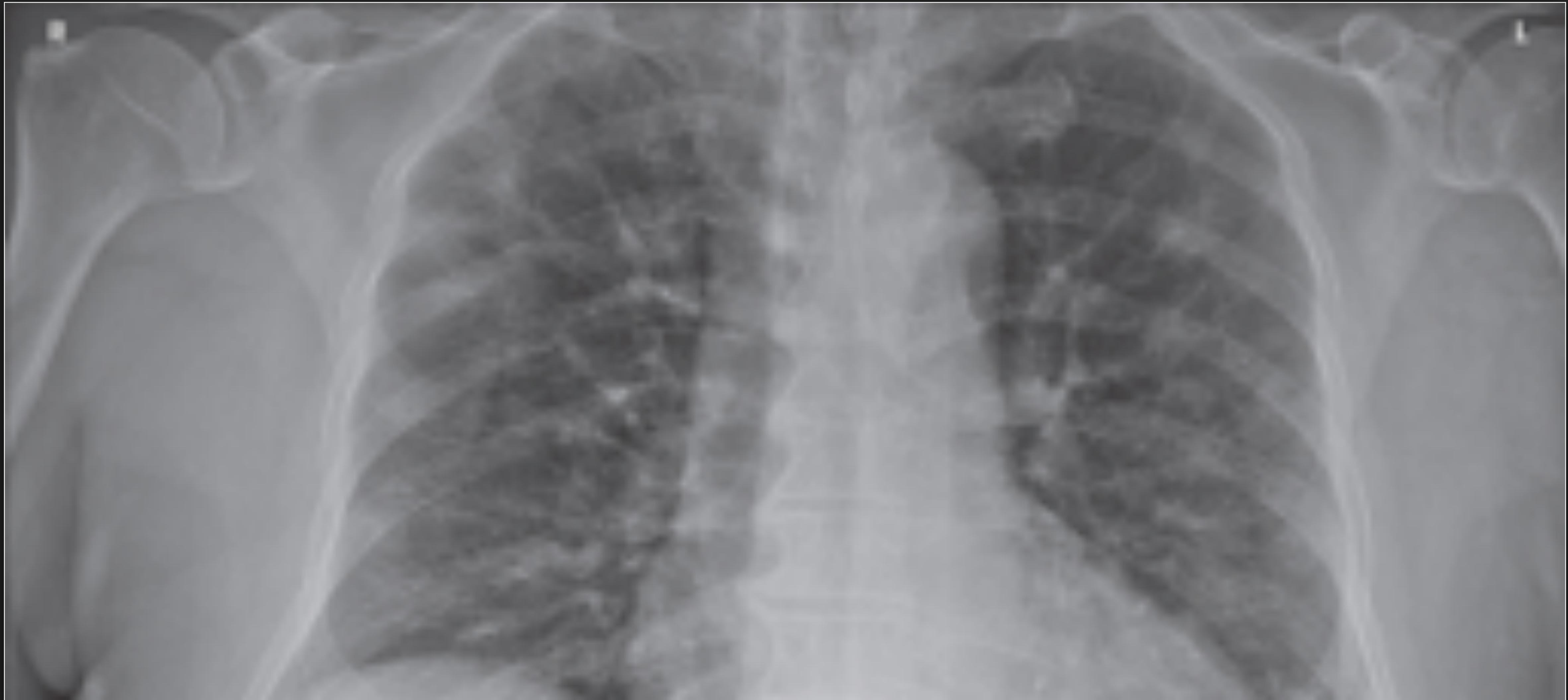
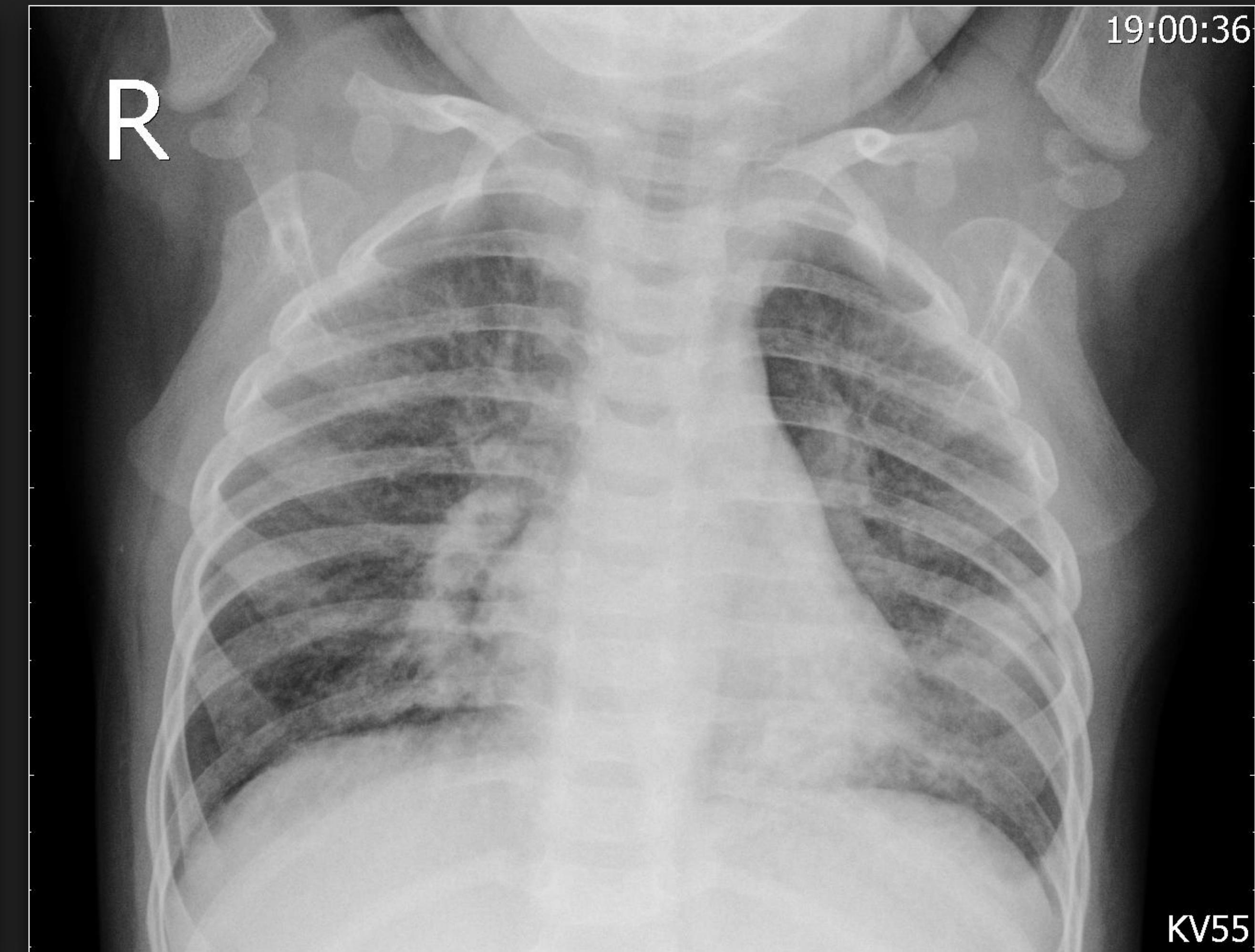
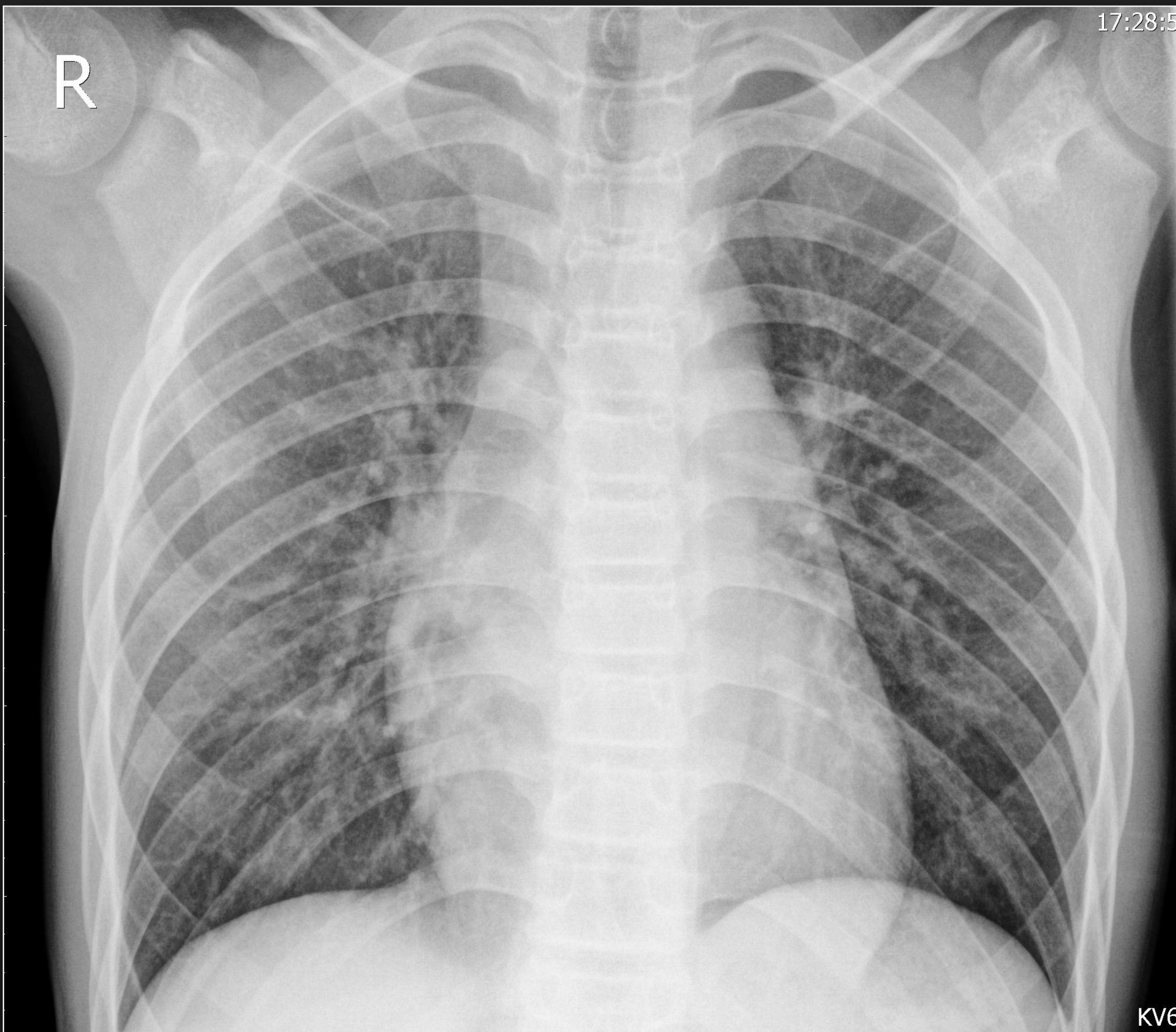


Chest X-Ray Classification

Brent Thayer

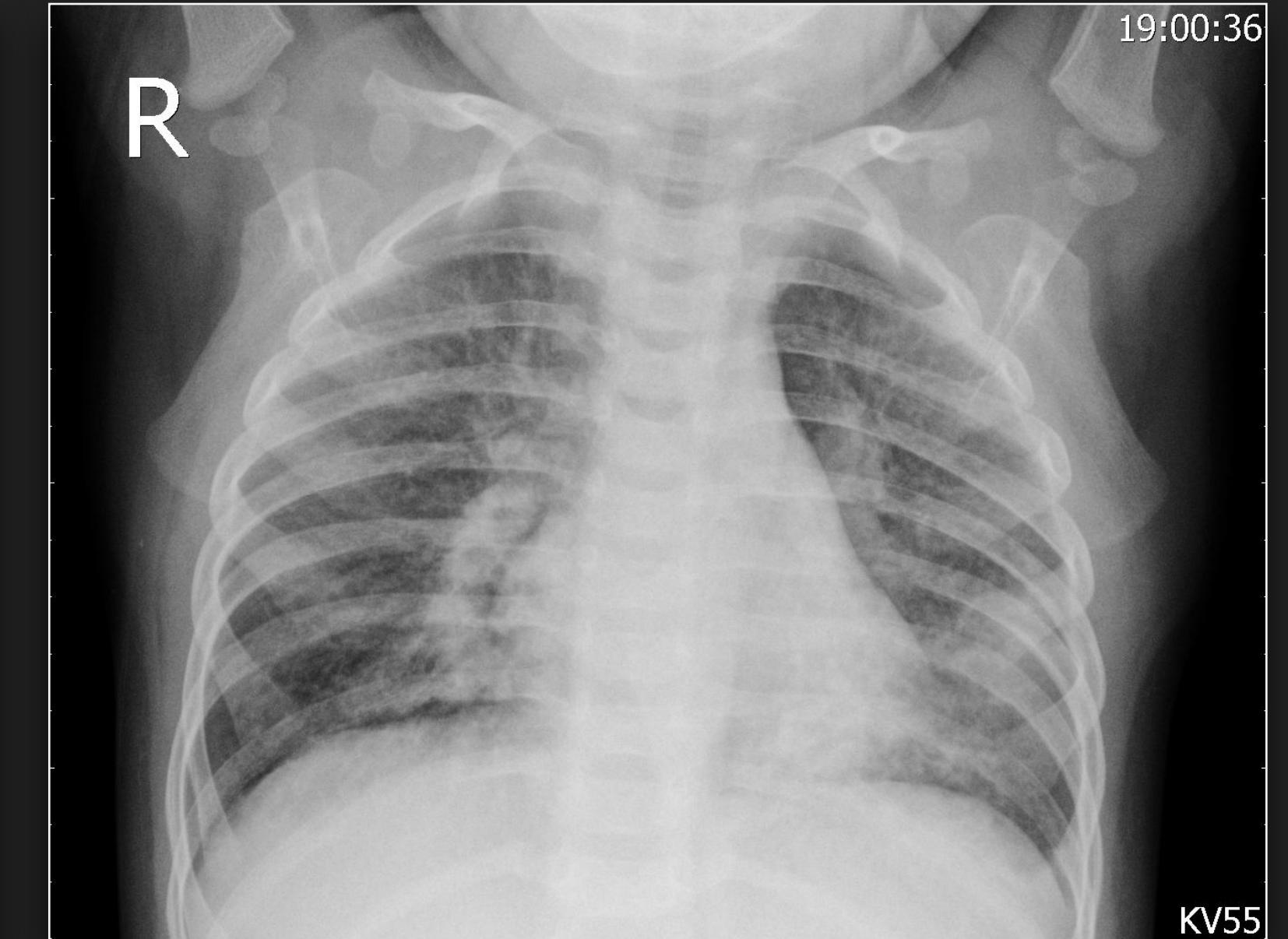
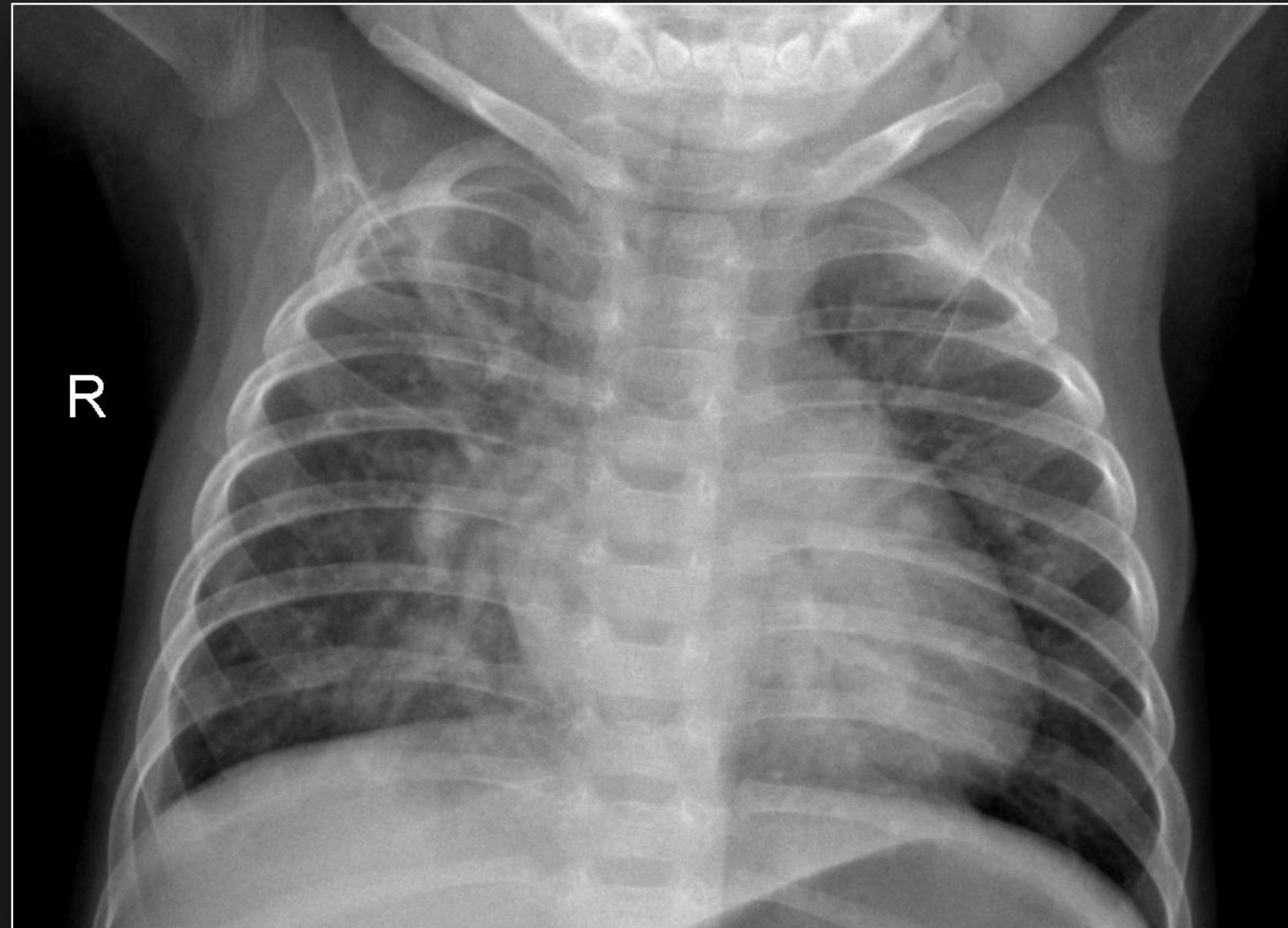
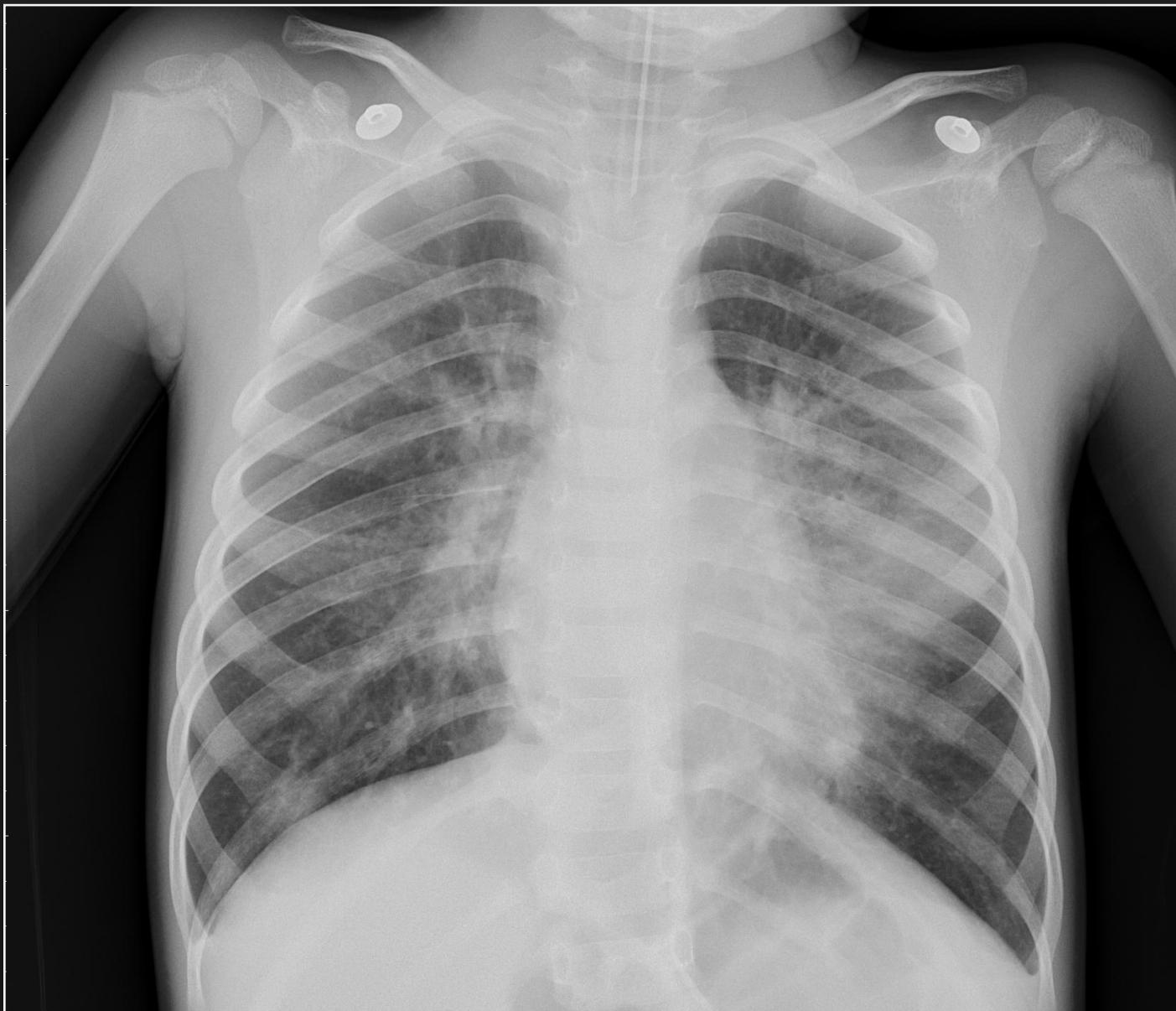




Is it possible to train a model to classify pneumonia from a chest X-ray?

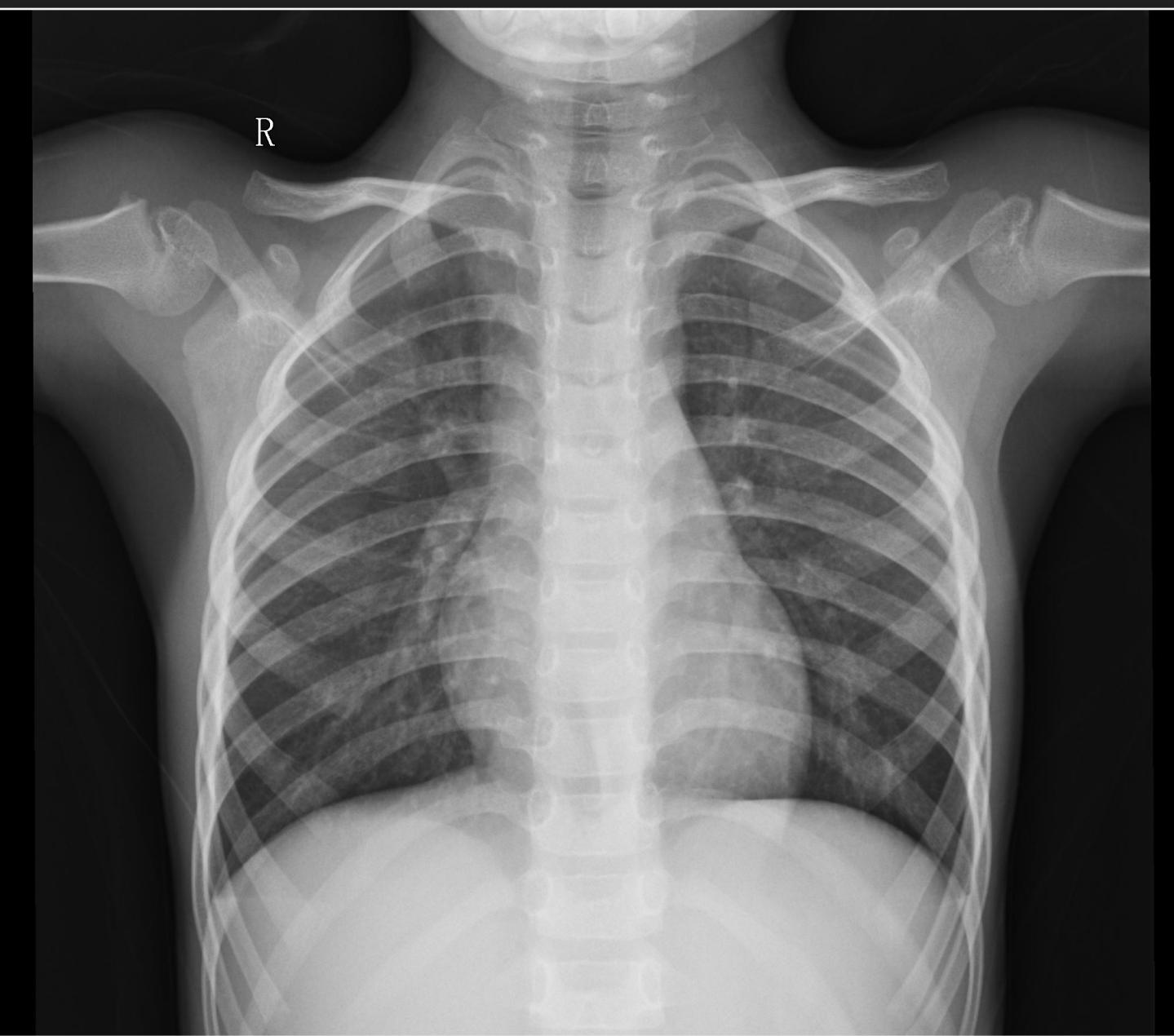
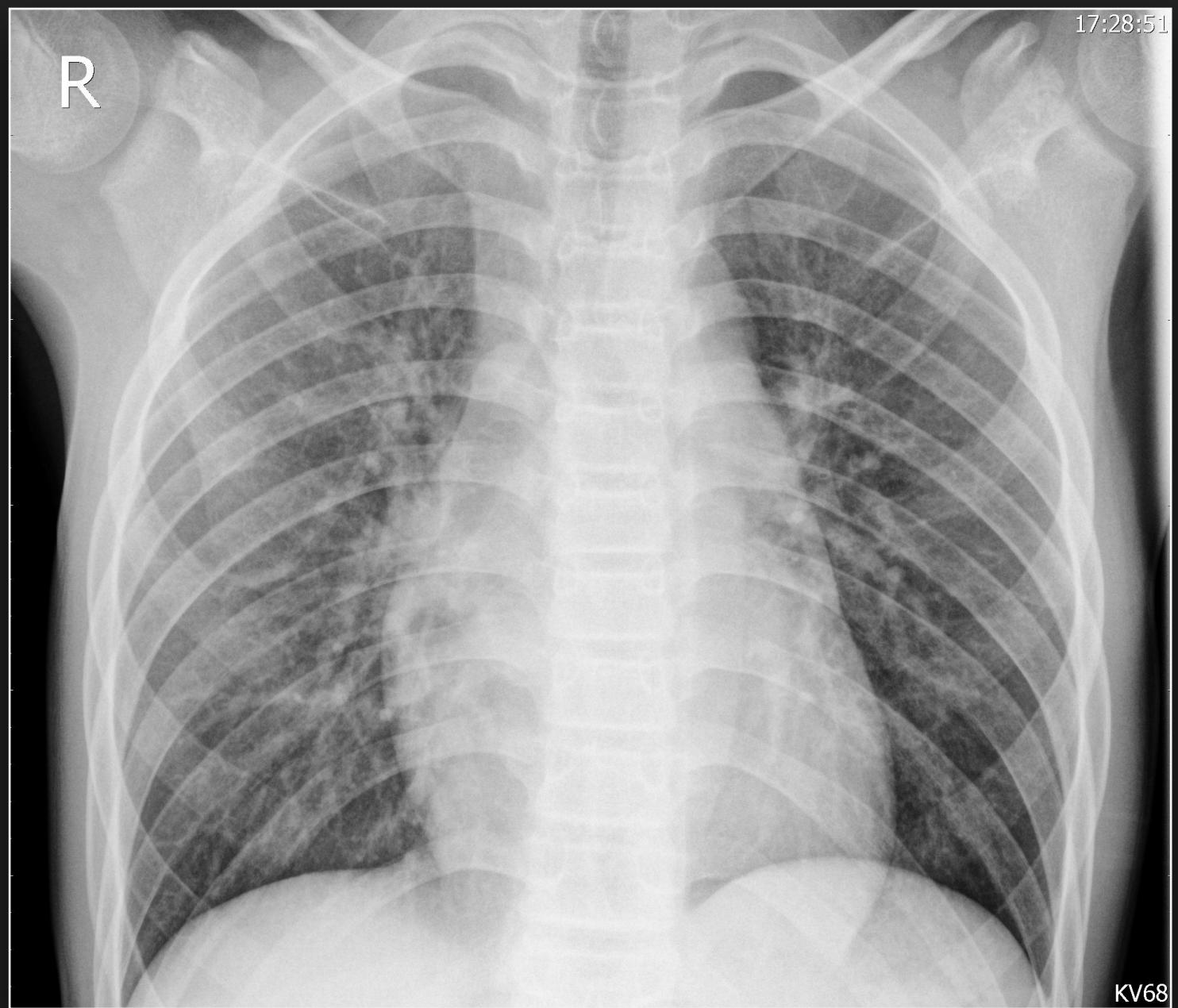
An X-ray allows a radiologist to see the lungs, heart and blood vessels. When interpreting the X-ray, the radiologist will look for white spots in the lungs (called infiltrates) that identify an infection.

Pneumonia

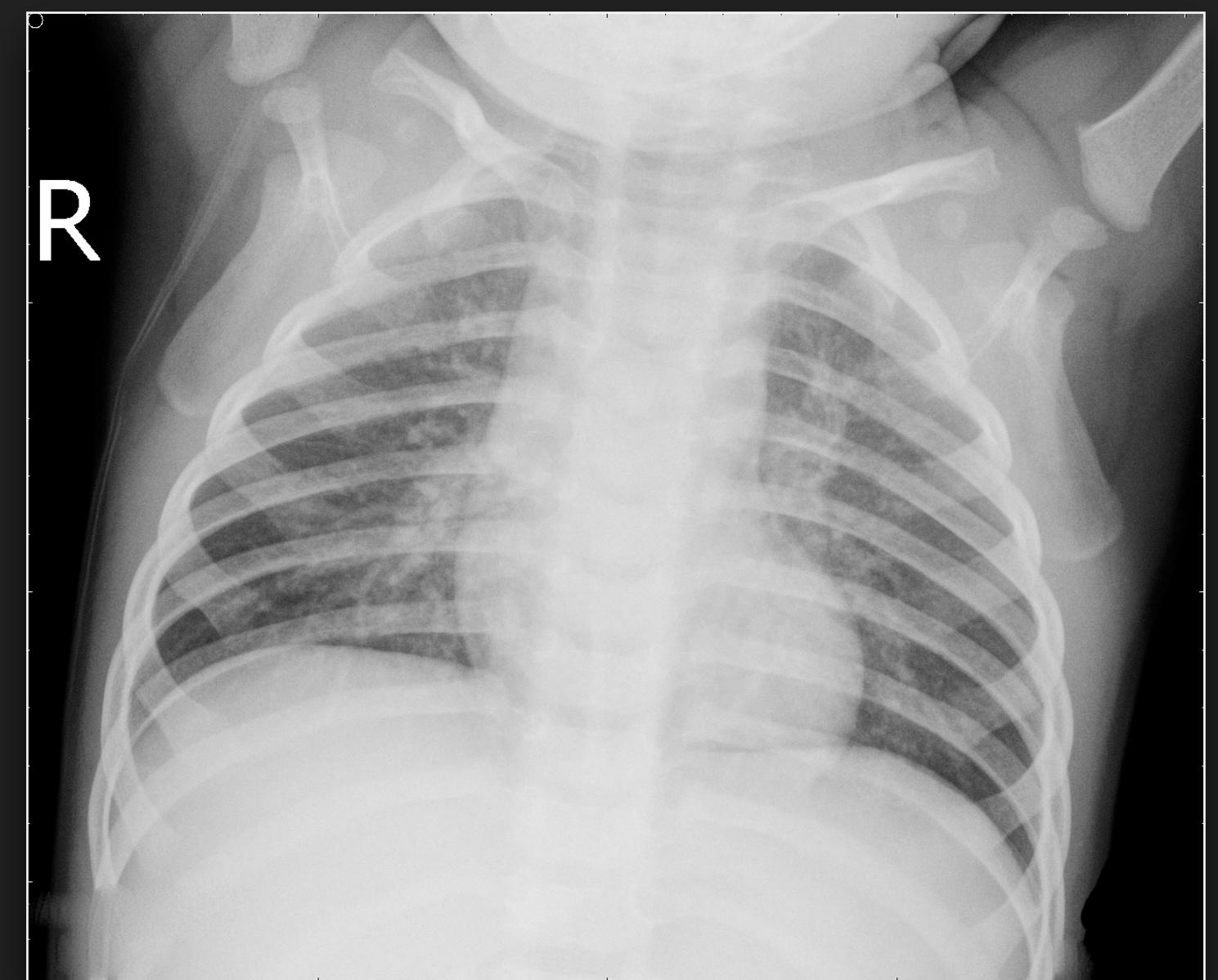
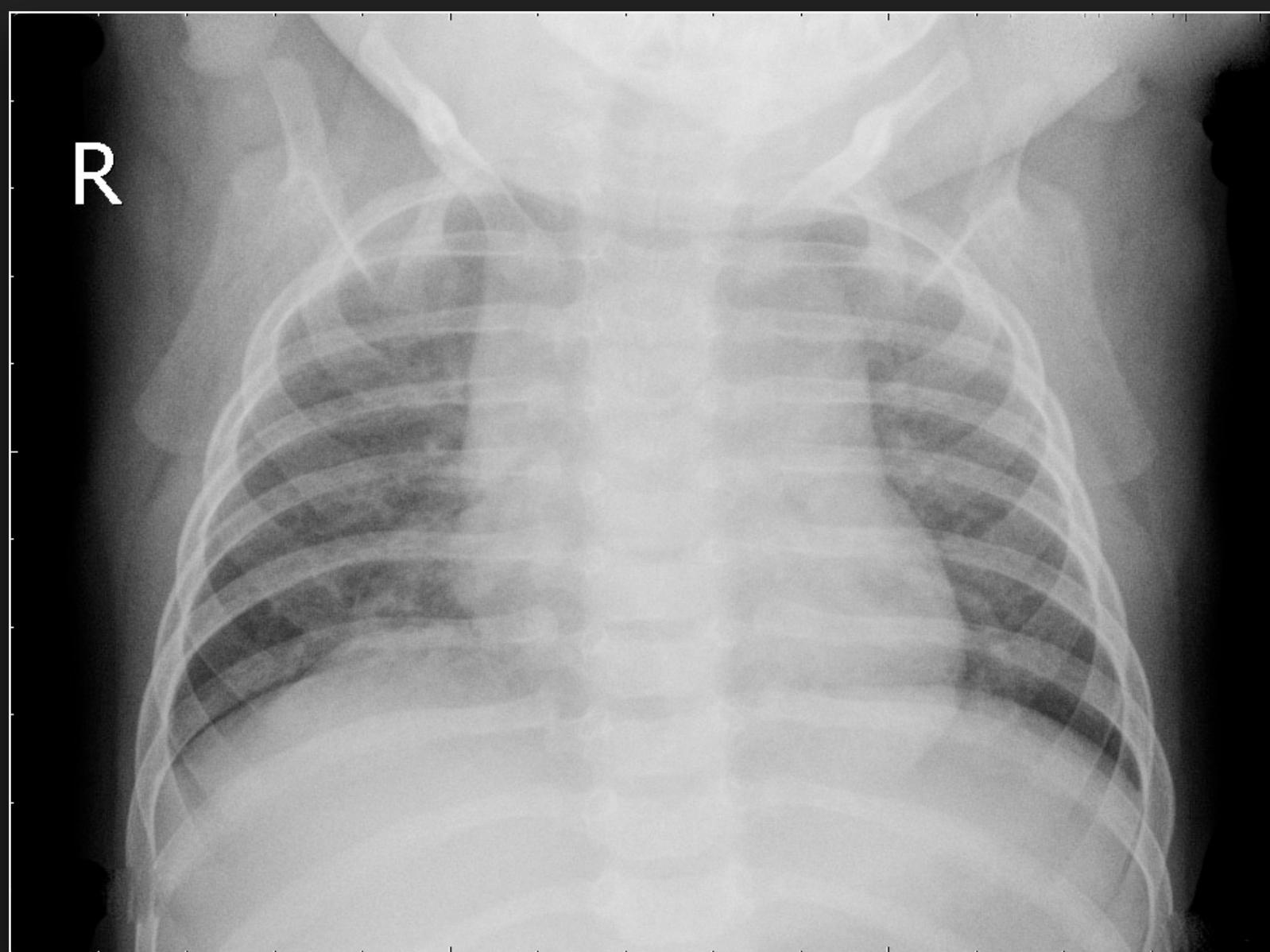


Pneumonia is an infection that inflames the air sacs in one or both lungs. The air sacs may fill with fluid or pus (purulent material), causing cough with phlegm or pus, fever, chills, and difficulty breathing. A variety of organisms, including bacteria, viruses and fungi, can cause pneumonia.

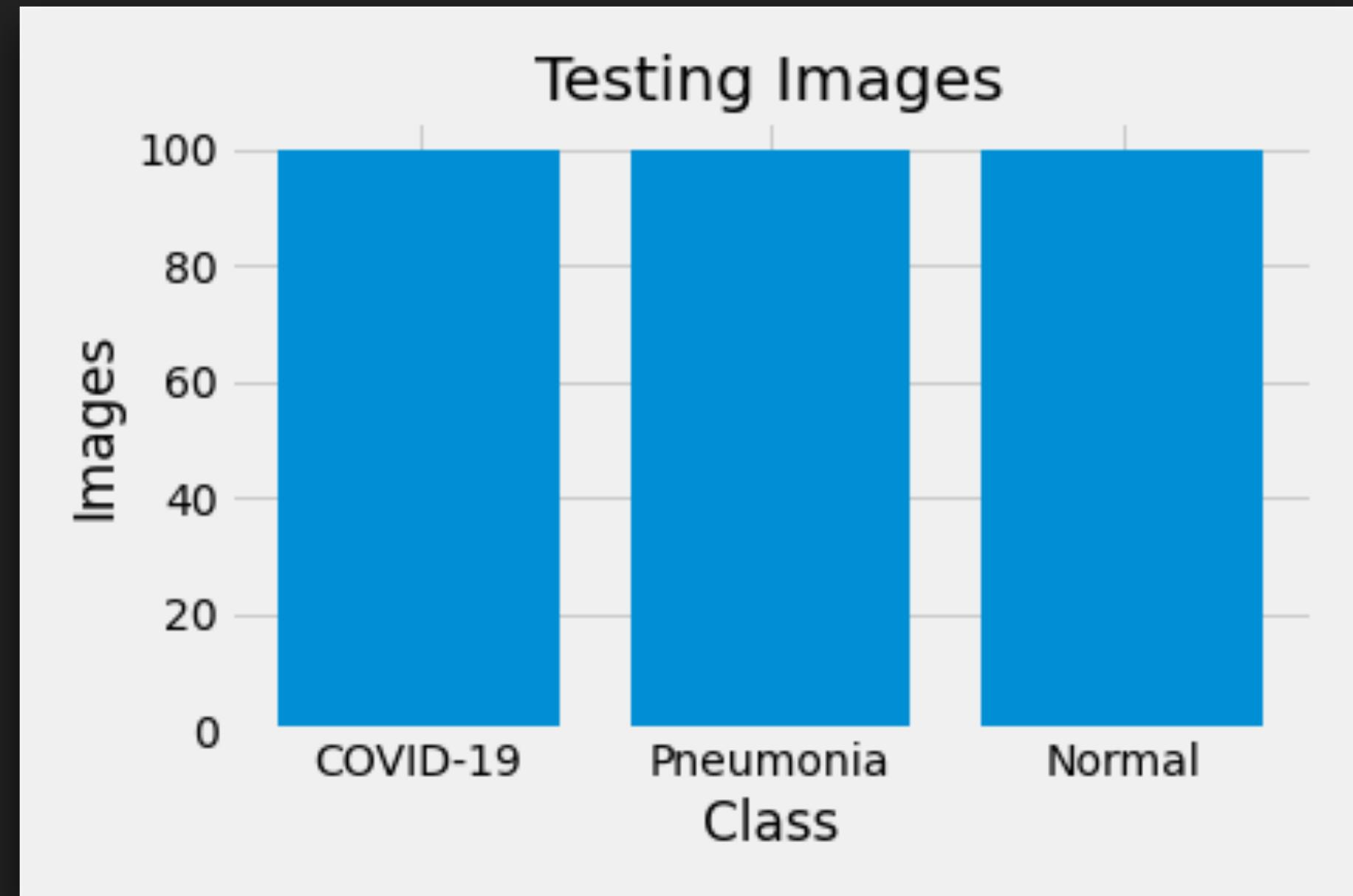
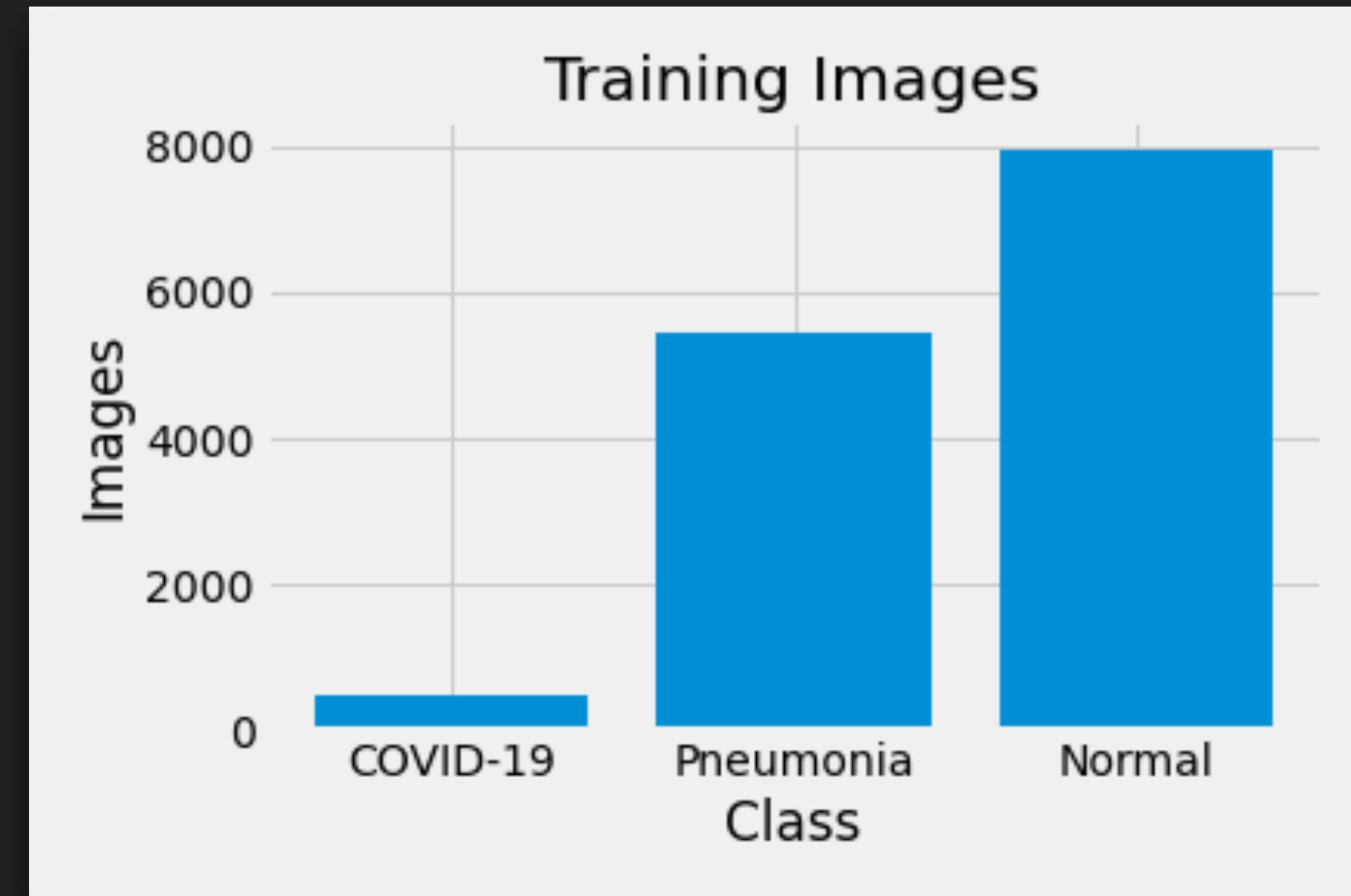
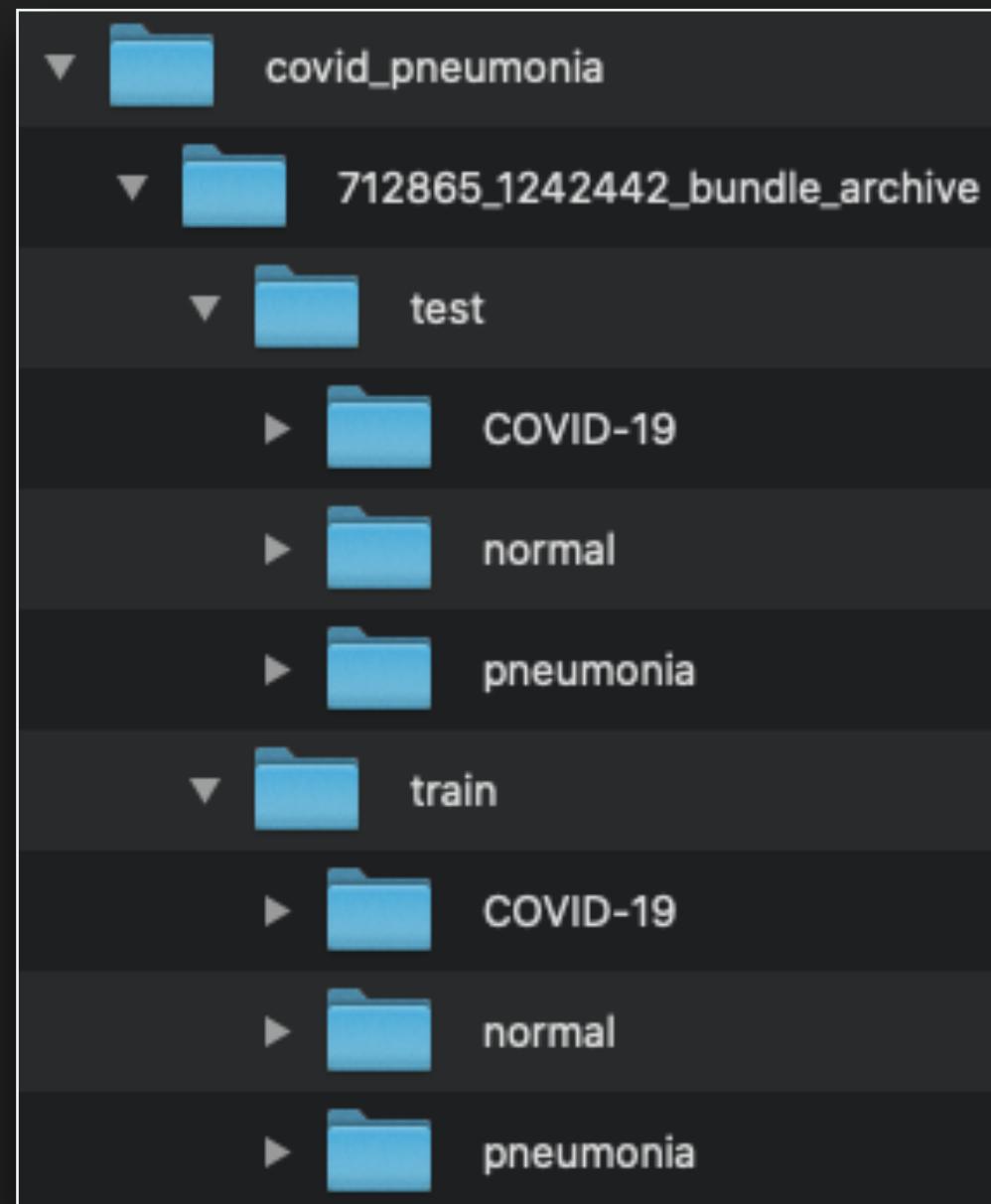
Normal



Anterior X-rays



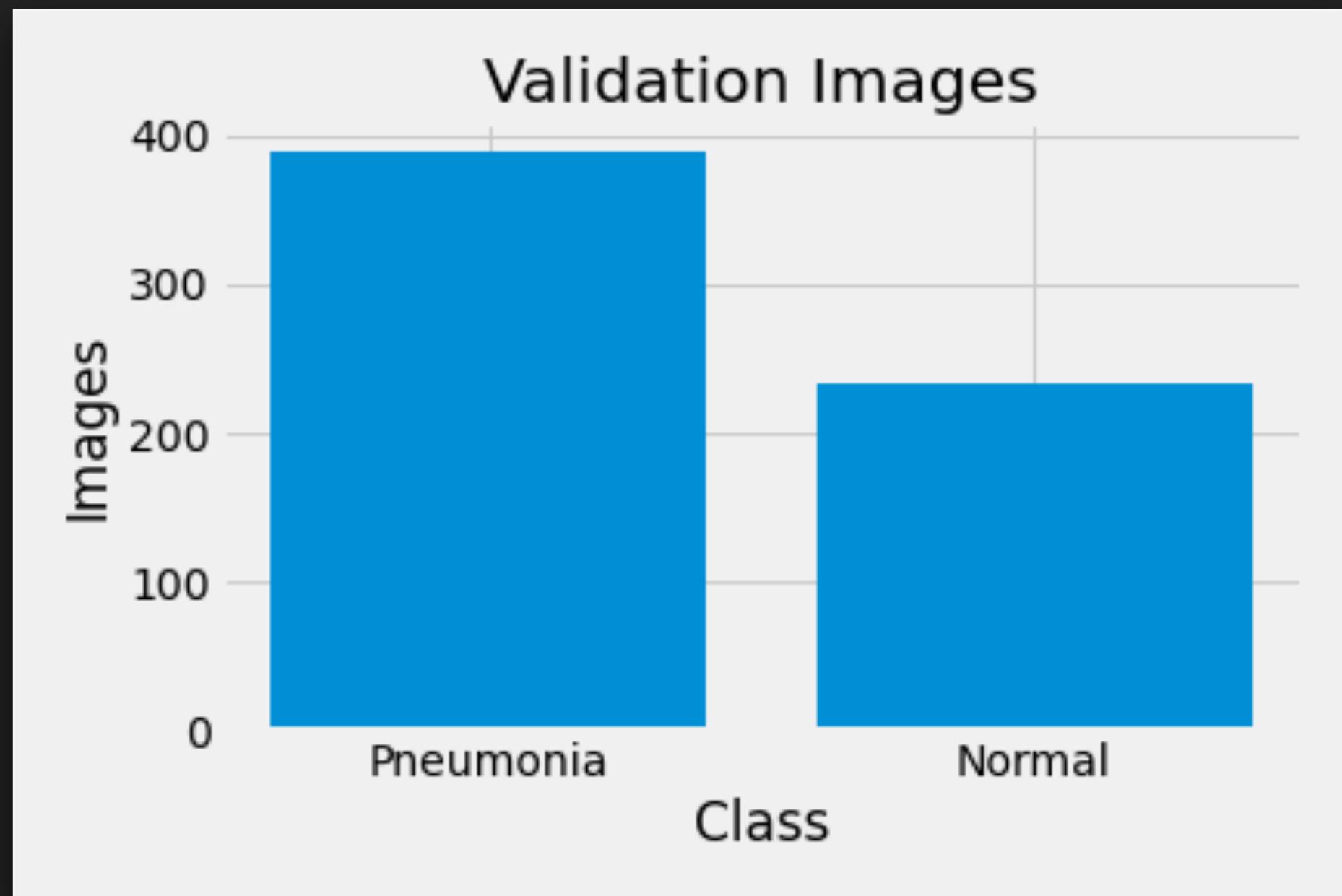
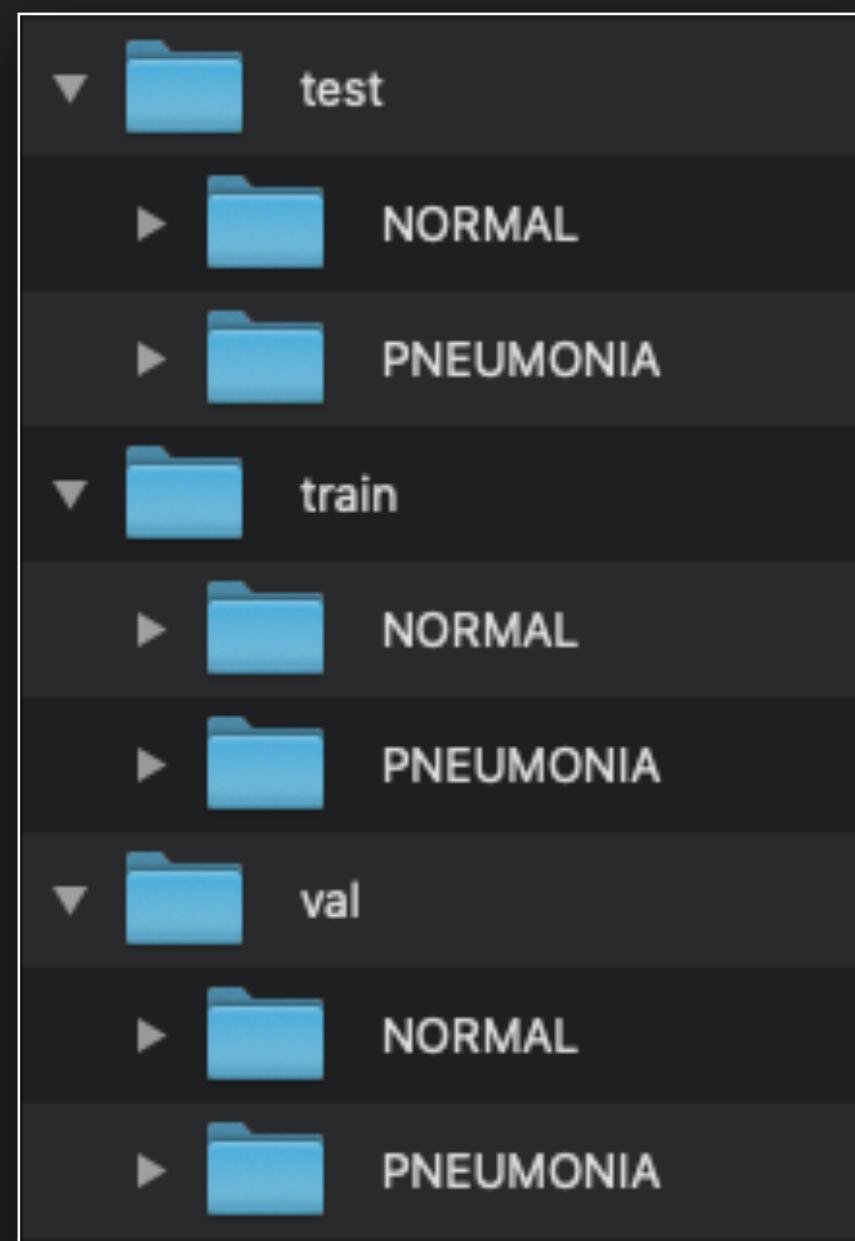
First Dataset



COVID-19 was removed to create a binary classification problem.

The model was learning to always predict normal.

New Dataset

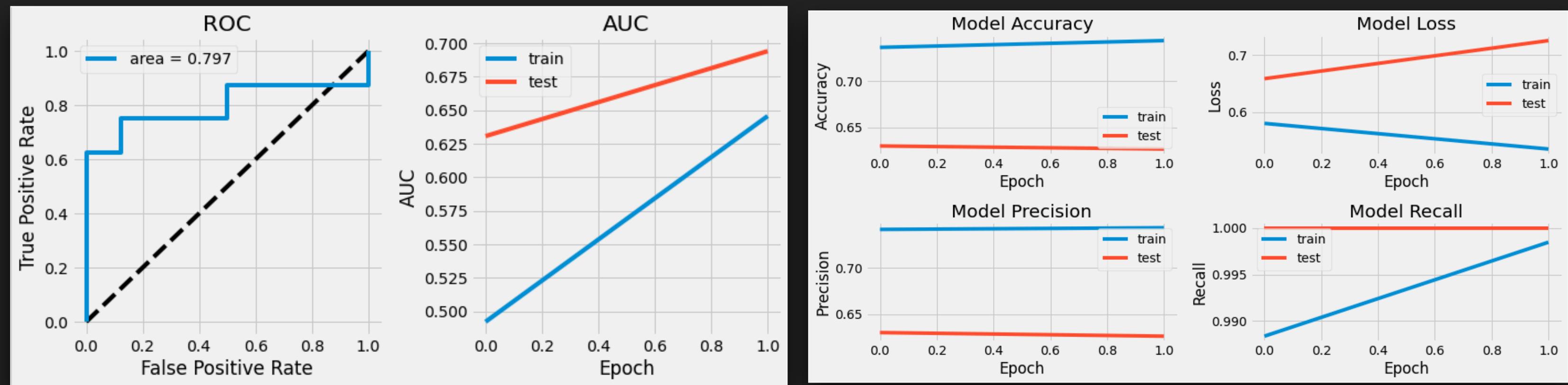


Interestingly enough, the model was now learning to always predict pneumonia after only a couple batches had processed.

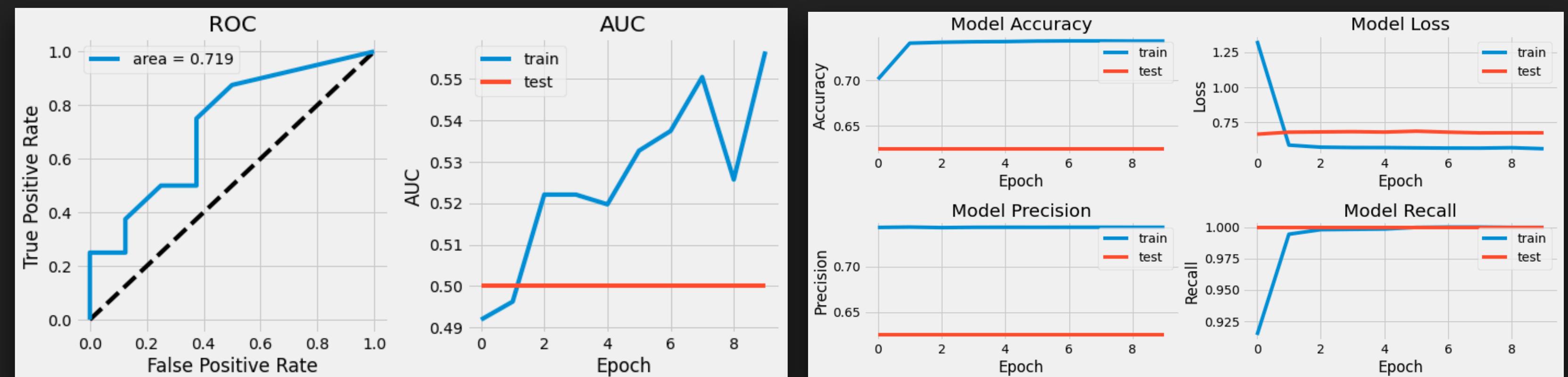
Model Progress

After Switching Datasets

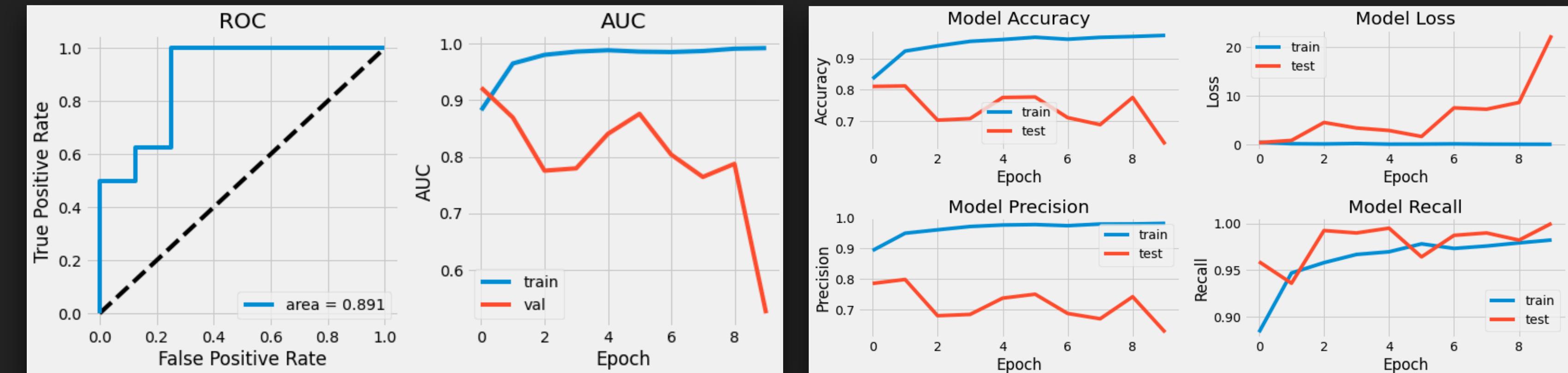
Initial



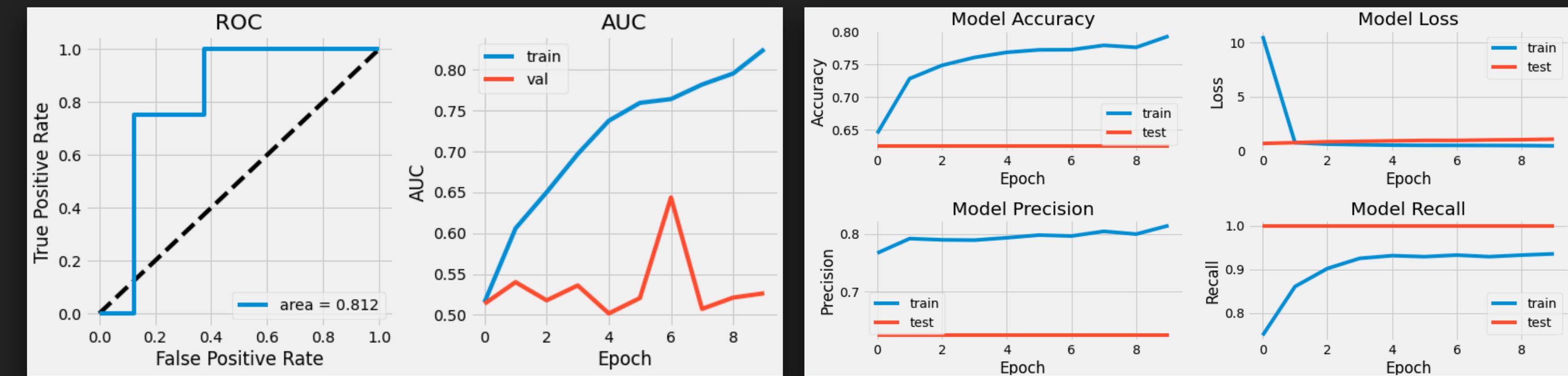
Removed
Grayscale and
Augmentation



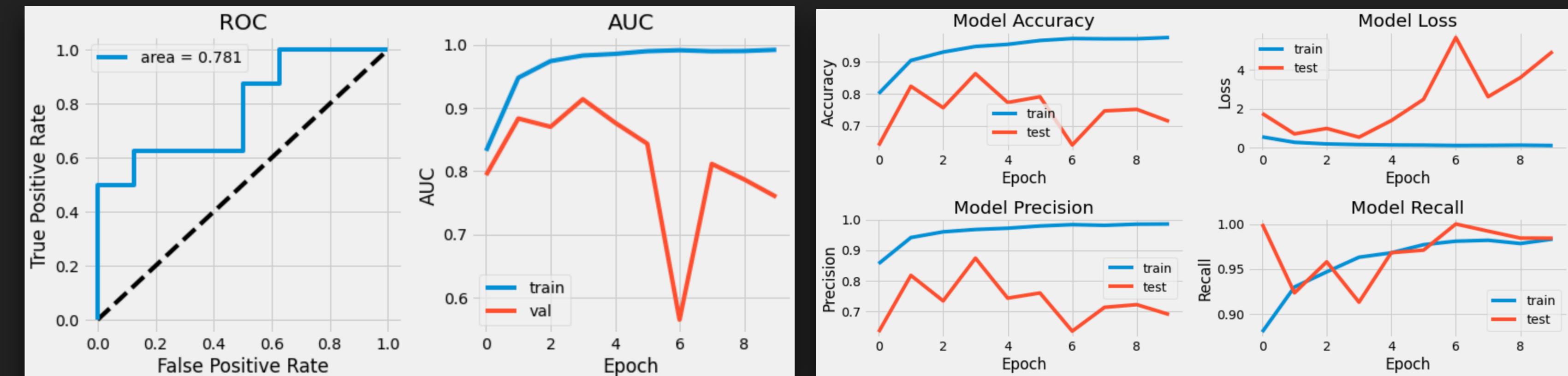
Unscaled Images / Removed Dropout



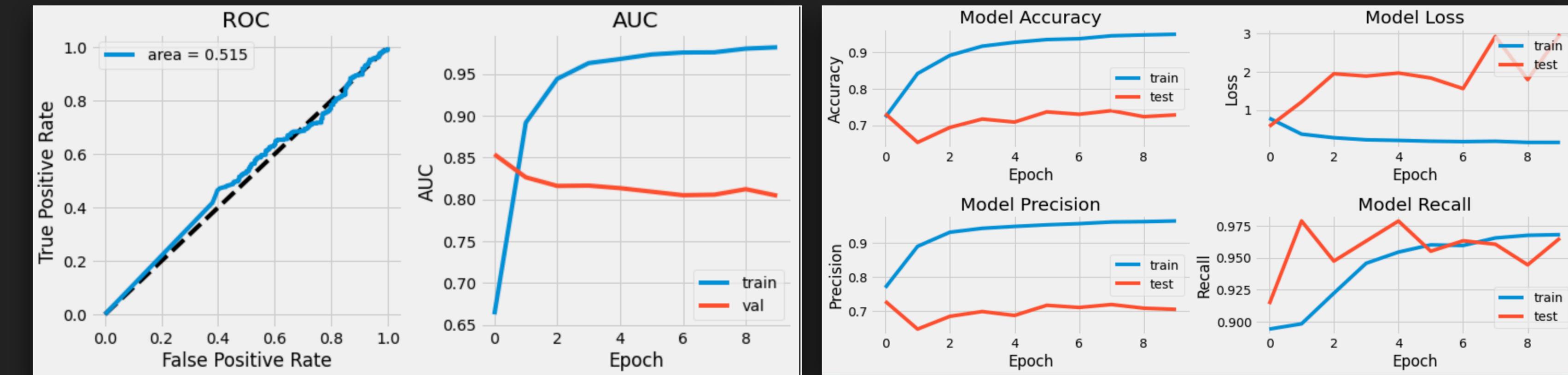
ReLU Activations / Added Dropout



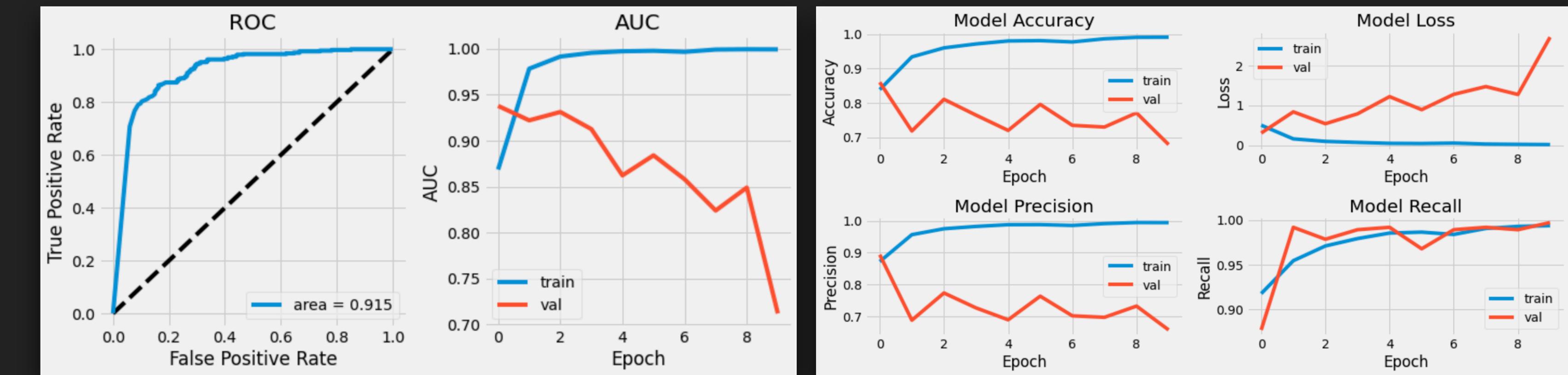
Weighted Normal to 2



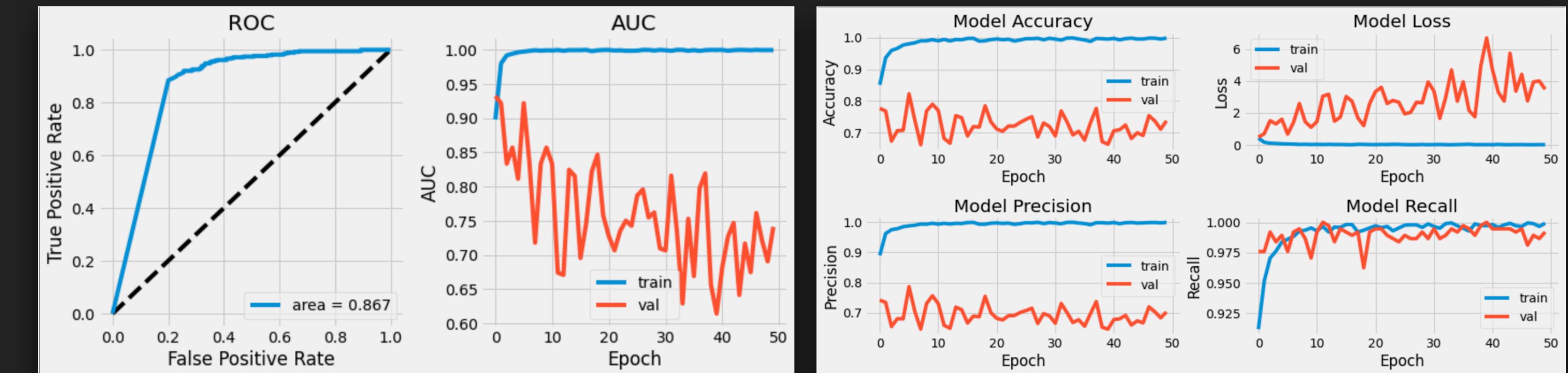
Changed Validation Images



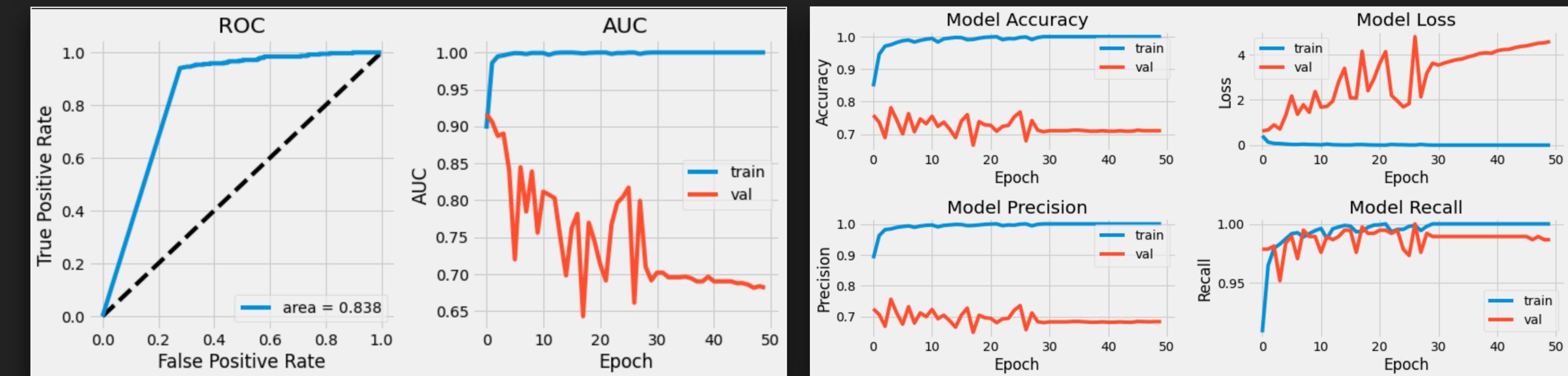
Removed Validation Shuffle



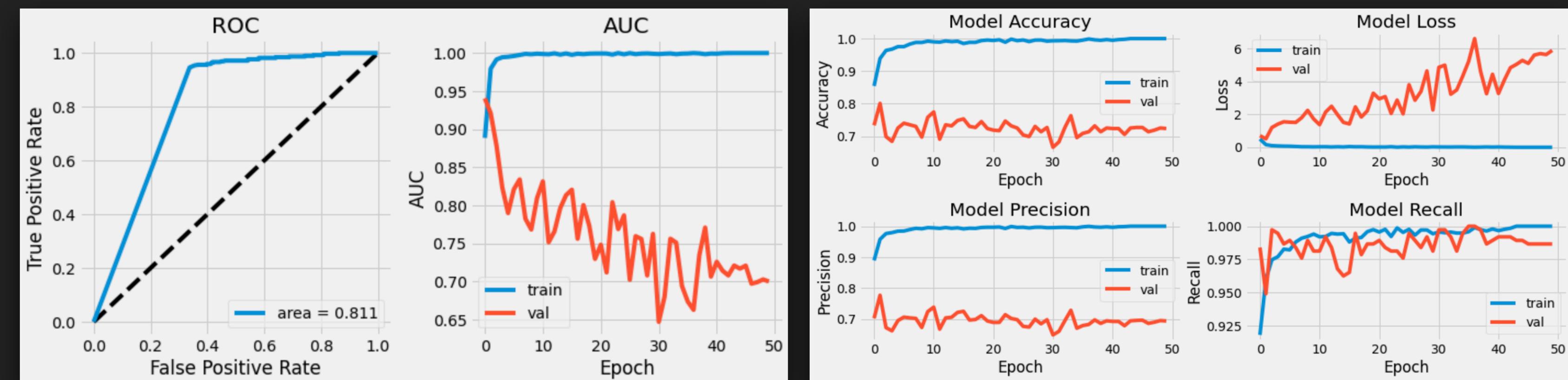
50 Epochs / 1 Dense Layer / Weighted Normal to 1.2



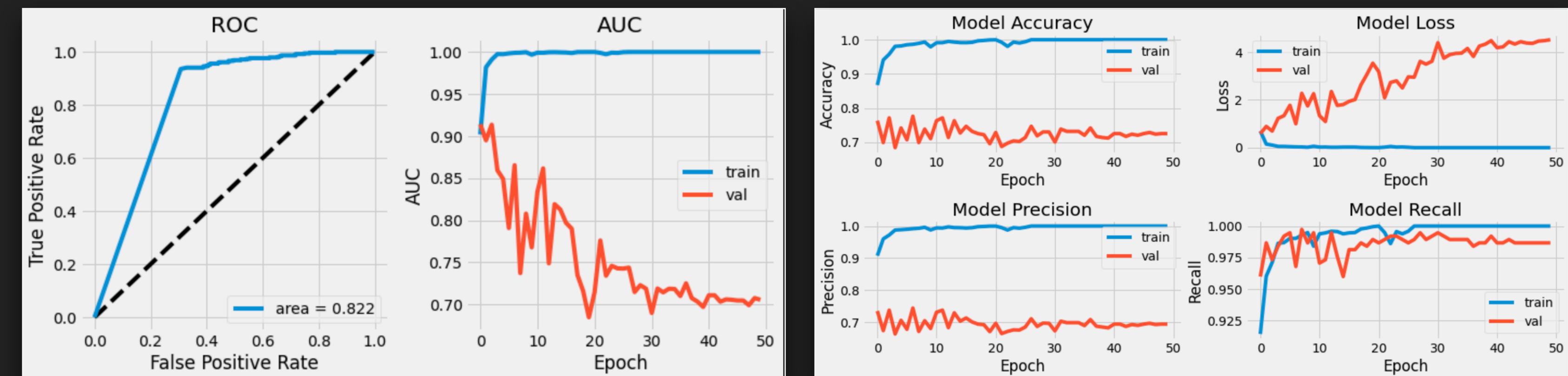
50 Epochs /
All Activations Swish



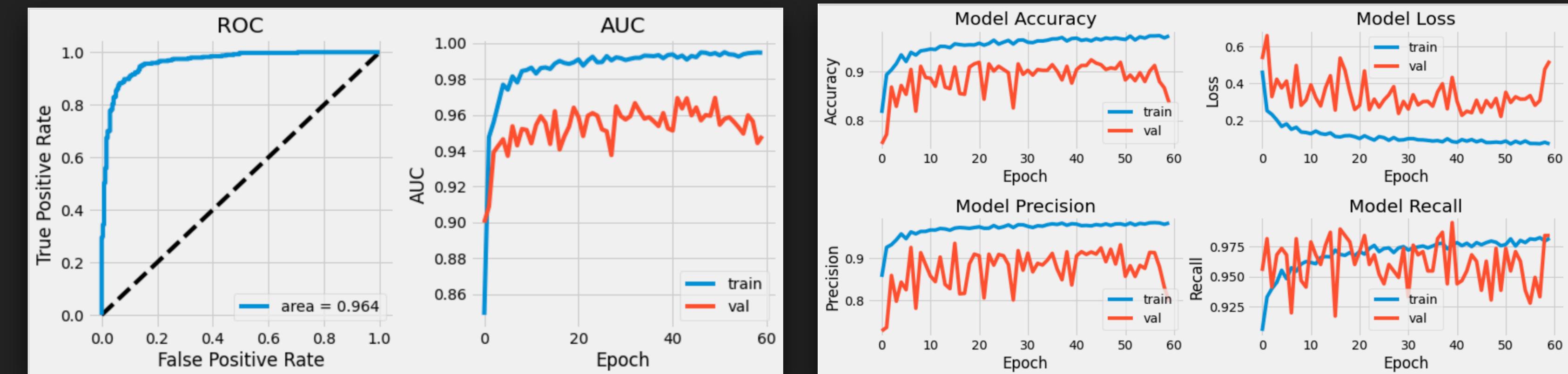
50 Epochs /
All Activations ReLU



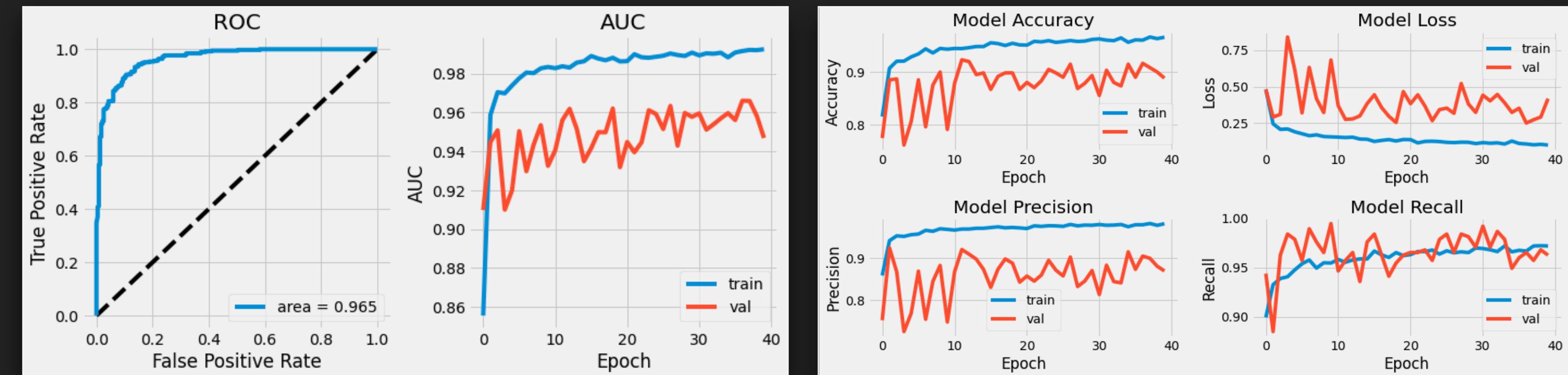
Removed last conv
layer. Dense units
512 to 256



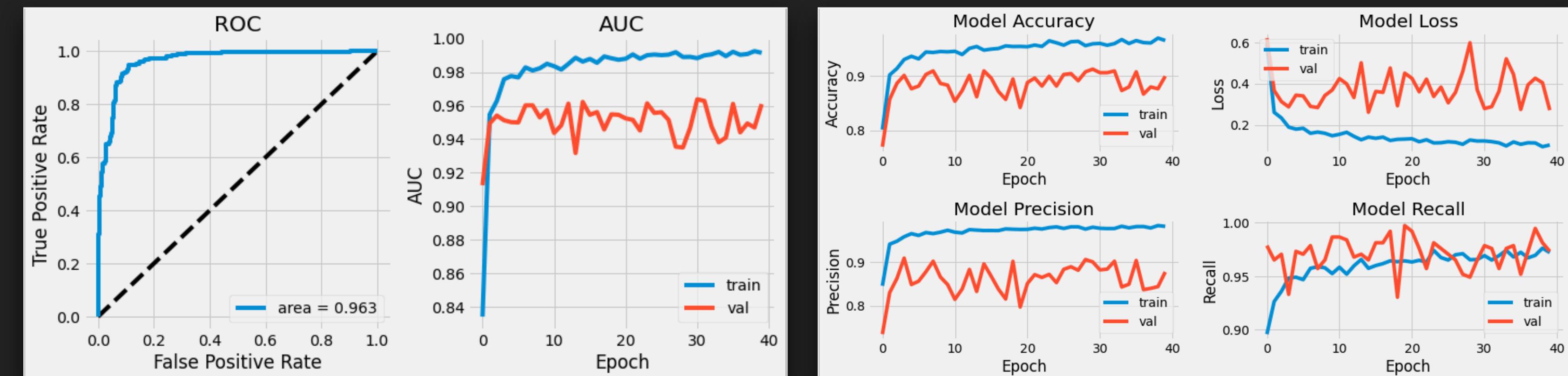
**Added
Augmentation /
60 Epochs /
Swish Activation**



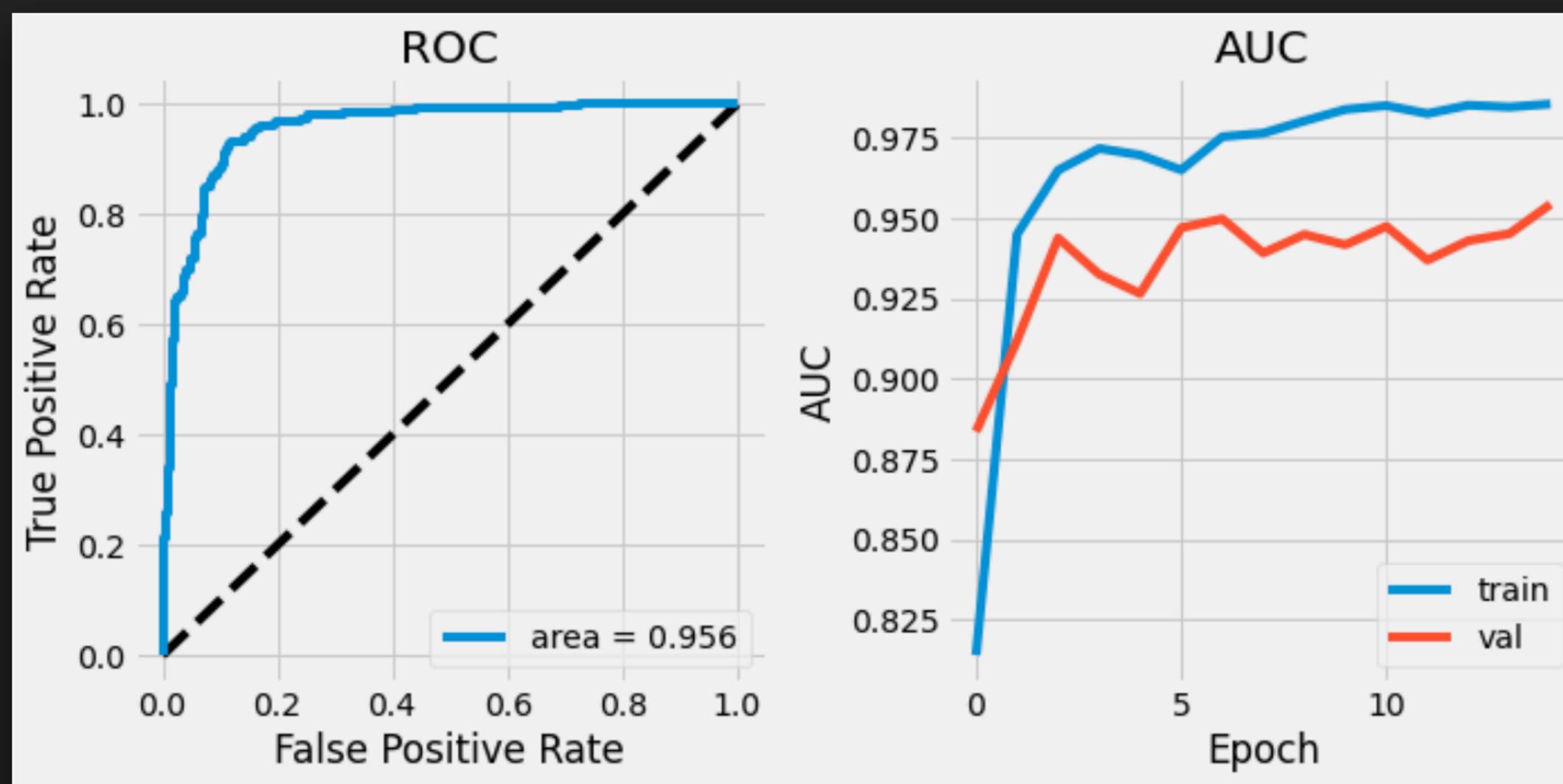
**Added Second
Dense Layer**



**Added Second
Dropout**

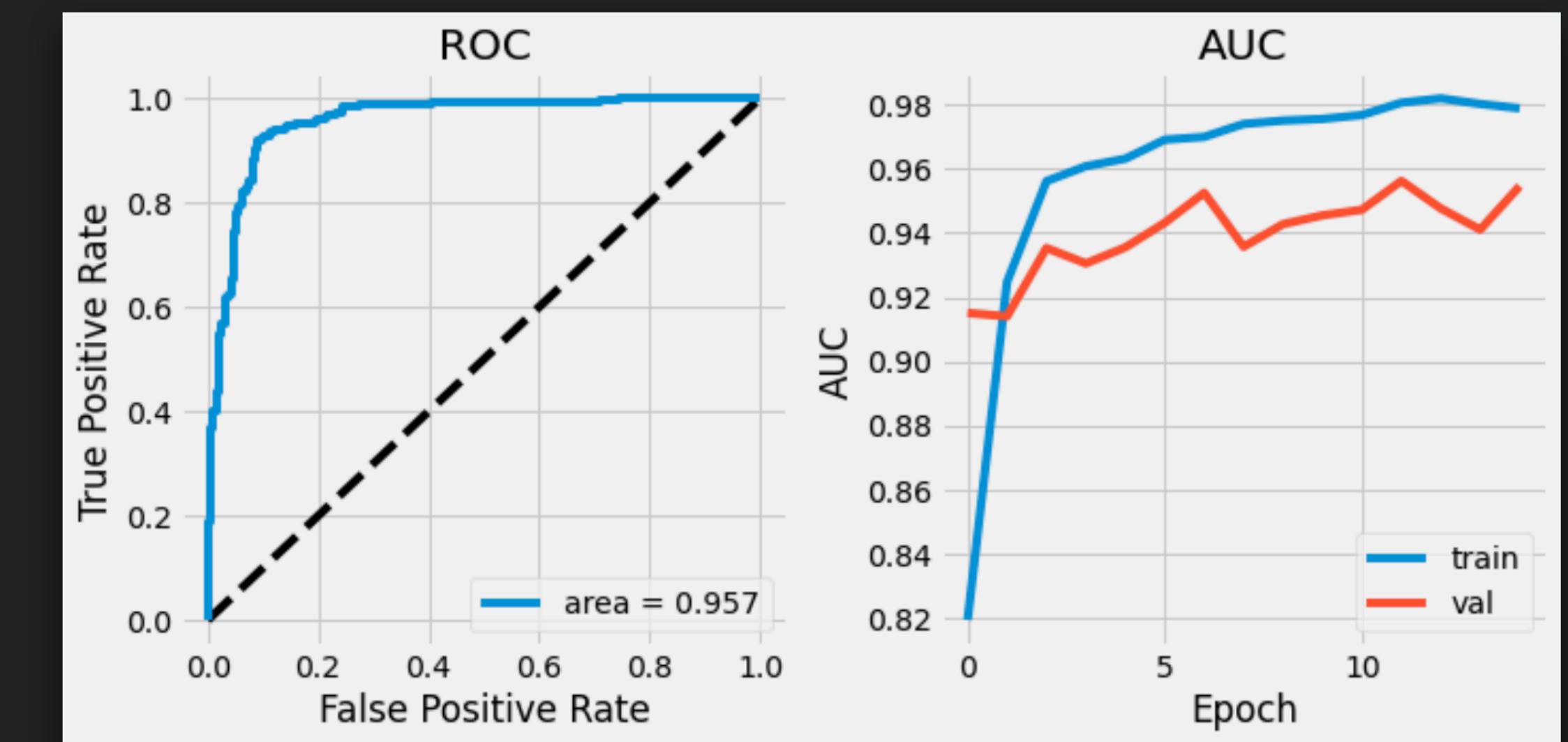


Swish



```
val_loss: 0.3222 - val_ACCURACY: 0.8832 - val_PRECISION: 0.8722  
val_loss: 0.5814 - val_ACCURACY: 0.7845 - val_PRECISION: 0.7435  
val_loss: 0.3392 - val_ACCURACY: 0.8882 - val_PRECISION: 0.8768  
val_loss: 0.3694 - val_ACCURACY: 0.8882 - val_PRECISION: 0.8660  
val_loss: 0.2897 - val_ACCURACY: 0.9030 - val_PRECISION: 0.8870
```

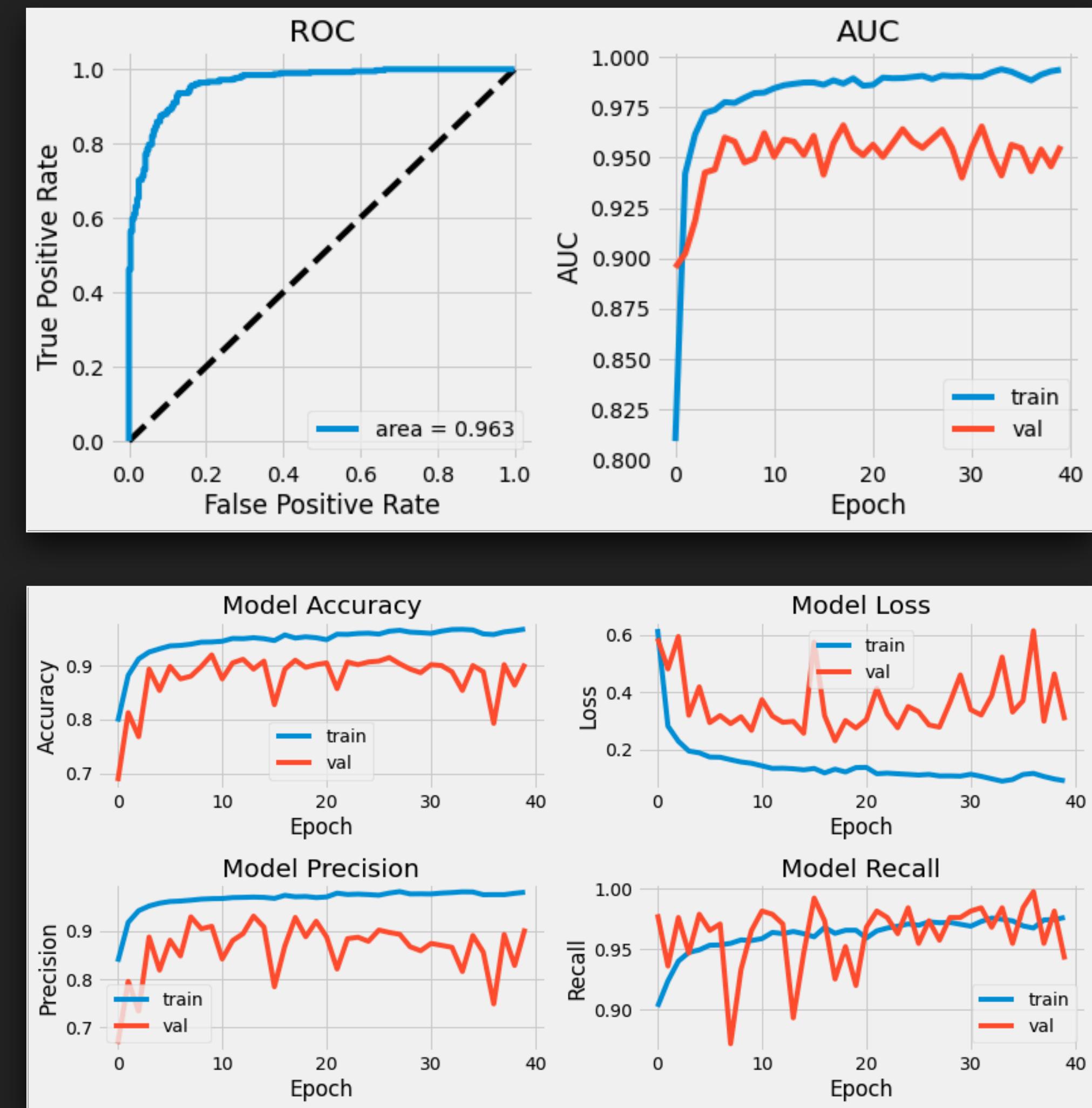
ReLU



```
val_loss: 0.3002 - val_ACCURACY: 0.9062 - val_PRECISION: 0.8914  
val_loss: 0.3494 - val_ACCURACY: 0.8684 - val_PRECISION: 0.8341  
val_loss: 0.3133 - val_ACCURACY: 0.8898 - val_PRECISION: 0.8828  
val_loss: 0.4338 - val_ACCURACY: 0.8618 - val_PRECISION: 0.8404  
val_loss: 0.2806 - val_ACCURACY: 0.9013 - val_PRECISION: 0.8925
```

Final Model

Layer (type)	Output Shape	Param #
conv2d_5 (Conv2D)	(None, 300, 300, 32)	896
max_pooling2d_5 (MaxPooling2)	(None, 150, 150, 32)	0
conv2d_6 (Conv2D)	(None, 150, 150, 64)	18496
max_pooling2d_6 (MaxPooling2)	(None, 75, 75, 64)	0
conv2d_7 (Conv2D)	(None, 38, 38, 64)	36928
max_pooling2d_7 (MaxPooling2)	(None, 19, 19, 64)	0
conv2d_8 (Conv2D)	(None, 10, 10, 128)	73856
max_pooling2d_8 (MaxPooling2)	(None, 5, 5, 128)	0
conv2d_9 (Conv2D)	(None, 3, 3, 128)	147584
max_pooling2d_9 (MaxPooling2)	(None, 1, 1, 128)	0
flatten_1 (Flatten)	(None, 128)	0
dense_2 (Dense)	(None, 256)	33024
dropout_1 (Dropout)	(None, 256)	0
dense_3 (Dense)	(None, 1)	257



Accuracy: 0.9030 AUC: 0.9558
 Precision: 0.9049 Recall: 0.9412

Conclusions

Even though I can not personally, differentiate between a chest X-ray showing pneumonia and one that does not, the network seemed to pick up on the difference fairly quickly.

Key Takeaways

Class Imbalance

True Positives | True Negatives

ROC Curves | Area Under the Curve

Model Complexity

Overfitting / Image Augmentation

Cloud Computing

Next Steps

Apply this model to X-rays from the local hospital

Go back and try COVID again

