

AutoML Modeling Report



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Binary Classifier with Clean/Balanced Data

Train/Test Split

How much data was used for training? How much data was used for testing?

Training: 100 images labeled as normal and 100 images labeled as pneumonia

Testing \ validation: 10 images were used for testing and 10 images for validation

Confusion Matrix

What do each of the cells in the confusion matrix describe? What values did you observe (include a screenshot)? What is the true positive rate for the "pneumonia" class? What is the false positive rate for the "normal" class?

Each cell describes the number of times the model classified an image in each category shown in percentage

In the first row, the matrix shows that 100% of pneumonia images was correctly classified as "pneumonia. this section includes the correctly predicted actual positive so that TPR is 100%.

On the same row a none of pneumonia images was predicted as normal is shown in the second cell (no FN for pneumonia class)

In the second row, 10% of normal images was classified as "pneumonia". which tells that this number of normal images were misclassified and that FPR is 10%

in the next cell, 90% of normal images were classified correctly as "Normal"

In the pneumonia actual, TPR for this class is 90%

True Label	Predicted Label	
	pneumonia	normal
pneumonia	100%	-
normal	10%	90%

Precision and Recall

What does precision measure? What does recall measure? What precision and recall did the model achieve (report the values for a score threshold of 0.5)?

The **precision** is the ratio of true positives to predicted

positives. It measures the number of images the model predicted correctly in its class compared to all predictions

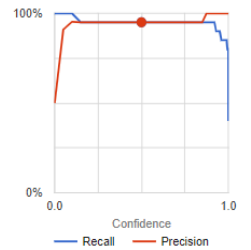
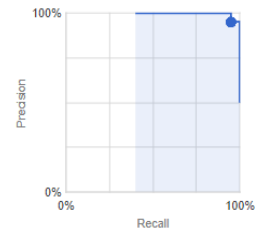
Recall is the ratio of true positives predictions to actual positives.

The Model achieved a precision and recall of **95%**

All labels

Total images	179
Test items	20
Precision ?	95%
Recall ?	95%

Use the slider to see which confidence threshold works best for your model on the precision-recall tradeoff curve.
[Learn more about these metrics and graphs.](#)



Score Threshold

When you increase the threshold what happens to precision? What happens to recall? Why?

When the threshold is increased above 0.8 the precision value will keep increasing but the recall value will decrease. Because precision takes in consideration the number of images predicted as positive and when we increase the model confidence the precision value will increase. While, for recall value is affected by the presence false negative prediction which decreases its value.

Binary Classifier with Clean/Unbalanced Data

Train/Test Split

How much data was used for training? How much data was used for testing?

Training: 100 images labelled as normal
300 images labelled as pneumonia

Testing \ Validation: 10 images for testing \ 10 validation

Confusion Matrix

How has the confusion matrix been affected by the unbalanced data? Include a screenshot of the new confusion matrix.

The matrix shows that the model has predicted **83%** of pneumonia images and **90%** of normal images correctly in their respective classes.

Also, **10%** of normal images was classified as “pneumonia” while **17%** of pneumonia images were classified as “Normal”

True Label	Predicted Label	
	pneumonia	normal
pneumonia	83%	17%
normal	10%	90%

Precision and Recall

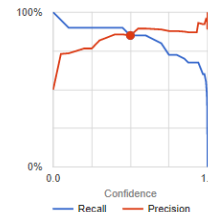
How have the model's precision and recall been affected by the unbalanced data (report the values for a score threshold of 0.5)?

The model achieved 85% in both metrics (85% precision, 85 % recall) at 0.5 confidence threshold.

All labels

Total images	359
Test items	40
Precision ②	85%
Recall ②	85%

Use the slider to see which confidence threshold works best for your model on the precision-recall tradeoff curve.
[Learn more about these metrics and graphs.](#)



Unbalanced Classes

From what you have observed, how do unbalanced classes affect a machine learning model?

Increased the number of false positive on normal class 5 images were incorrectly predicted as normal

Increased the number of False negatives on pneumonia class the model predicted 5 images with pneumonia as normal.

False negatives

Your model should have predicted pneumonia on these images



Score(s): 0.67493474



Score(s): 0.4561833



Score(s): 0.6164537



Score(s): 0.7443263



Score(s): 0.72207546



Score(s): 0.47874397

Binary Classifier with Dirty/Balanced Data

Confusion Matrix

How has the confusion matrix been affected by the dirty data? Include a screenshot of the new confusion matrix.

Dirty data has affected the accuracy of the model
The number of False Negatives has increased in each class.
40% of pneumonia images were classified as normal
30% of normal images classified as pneumonia

True Label	Predicted Label	
	pneumonia	normal
pneumonia	60%	40%
normal	30%	70%

Precision and Recall

How have the model's precision and recall been affected by the dirty data (report the values for a score threshold of 0.5)? Of the binary classifiers, which has the highest precision? Which has the highest recall?

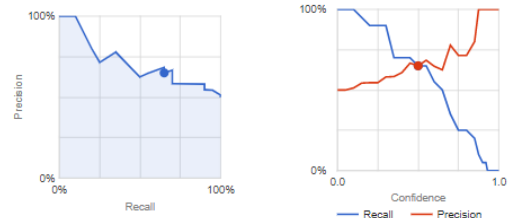
The model precision and recall have dropped to 65% at 0.5 threshold compared to the model metric when clean data was used for training.

By comparing all 3 classifiers, Clean Balanced classifier has performed well among all three with the highest precision and recall value of 95%
Which indicates the importance of training the model with balanced and clean dataset.

All labels

Total Images	180
Test items	20
Precision	65%
Recall	65%

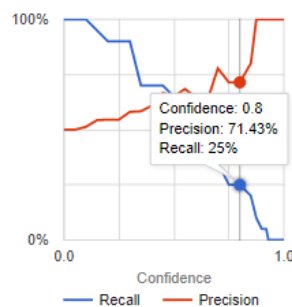
Use the slider to see which confidence threshold works best for your model on the precision-recall tradeoff curve.
[Learn more about these metrics and graphs.](#)



Dirty Data

From what you have observed, how does dirty data affect a machine learning model?

Training the model with dirty data has affected the accuracy of the model it can be seen when we increase our confidence threshold to 0.8 the recall drops to 25% which indicates that the Ratio of true positives to actual positives is low. The model is misclassifying actual positives and predicting them as negative.



3-Class Model

Confusion Matrix

Summarize the 3-class confusion matrix. Which classes is the model most likely to confuse? Which class(es) is the model most likely to get right? Why might you do to try to remedy the model's "confusion"? Include a screenshot of the new confusion matrix.

if we look at the values in the diagonal shape, we see that they mark the highest numbers in the matrix which indicates that the model has correctly predicted most of the data in its correct class.

The model predicted all images of normal class correctly.


Confusion appears in bacterial pneumonia class in predicting 10% of the class images as viral pneumonia

True Label	Predicted Label		
	normal	viral pneumonia	bacterial pneumonia
normal	100%	-	-
viral pneumonia	10%	80%	10%
bacterial pneumonia	-	10%	90%

Another area of confusion is when the model has classified 10% of the viral pneumonia as normal and 10% as bacteria pneumonia

A sample of the false negative is shown below

False negatives
Your model should have predicted viral pneumonia on these images



Score(s): 0.032637544

Score(s): 0.13728897

To lessen the confusion, I might train the model on more images of viral pneumonia and select images with better quality.

Precision and Recall

What are the model's precision and recall?
How are these values calculated (report the values for a score threshold of 0.5)?

At 0.5 score of thresholds

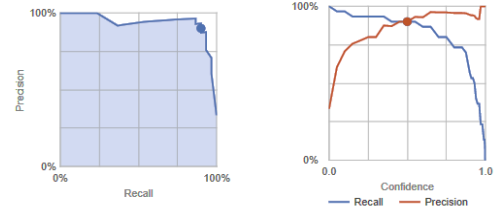
Precision: 90%

Recall: 90%

All labels

Total images	269
Test items	30
Precision ?	90%
Recall ?	90%

Use the slider to see which confidence threshold works best for your model on the precision-recall tradeoff curve.
[Learn more about these metrics and graphs.](#)



To calculate the model precision, we need 1st to calculate the **precision** to each class

$$P_{\text{Normal}} = \frac{TP}{TP + FP} = \frac{100}{100 + 10} = \frac{100}{110} = 0.909 = 0.91$$

$$P_{\text{Viral}} = \frac{TP}{TP + FP} = \frac{80}{80 + 10} = 0.888 = 0.89$$

$$P_{\text{Bacterial}} = \frac{TP}{TP + FP} = \frac{90}{90 + 10} = 0.9$$

$$P_{\text{Model}} = (P_{\text{Normal}} + P_{\text{Viral}} + P_{\text{Bacterial}}) / 3 = (0.91 + 0.89 + 0.9) / 3 = \mathbf{0.9}$$

We will calculate the **Recall** in the same manner

$$R_{\text{Normal}} = \frac{TP}{TP + FN} = \frac{100}{100 + 0} = 1$$

$$R_{\text{Viral}} = \frac{TP}{TP + FN} = \frac{80}{80 + 10 + 10} = \frac{80}{100} = 0.8$$

$$R_{\text{Bacterial}} = \frac{TP}{TP + FN} = \frac{90}{90 + 10} = \frac{90}{100} = 0.9$$

$$R_{\text{Model}} = (R_{\text{Normal}} + R_{\text{Viral}} + R_{\text{Bacterial}}) / 3 = (1 + 0.8 + 0.9) / 3 = \mathbf{0.9}$$

F1 Score

What is this model's F1 score?

$$F1 = (2 * \text{precision} * \text{recall}) / (\text{precision} + \text{recall})$$

$$F1 = (2 * 0.9 * 0.9) / (0.9 + 0.9)$$

$$F1 = 1.62 / 1.8 = \mathbf{0.9}$$

Thank you

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