

# AutoML Modeling Report



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Google Cloud Platform		Udacity AutoML Classifier		Search products and resources			
Vision		Datasets					
		NEW DATASET					
		Name	Type	Total images	Labeled images	Last updated	Status
		two_class_only_pneumonia ICN8523326727345143808	Single-Label Classification	235	235	Feb 17, 2021, 8:14:34 PM	Success: Training model
		clean_unbalanced_x_ray_pneumonia ICN956857140897644544	Single-Label Classification	400	400	Feb 16, 2021, 8:54:08 PM	Success: Training model
		3_class_balanced_x_ray_pneumonia ICN7274281518191607808	Single-Label Classification	300	300	Feb 16, 2021, 8:51:13 PM	Success: Training model
		dirty_balanced_x_ray_pneumonia ICN6166818222323531776	Single-Label Classification	200	200	Feb 16, 2021, 8:46:56 PM	Success: Training model
		chest_x_ray_images_pneumonia ICN182154876035770776	Single-Label Classification	200	200	Feb 16, 2021, 4:20:24 PM	Success: Exporting data

Figure: AutoML five trained models

## Binary Classifier with Clean/Balanced Data

### Train/Test Split

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

Label Stats				
Labels	Images	Train	Validation	Test
normal	100	80	10	10
pneumonia	100	81	10	9

### Confusion Matrix

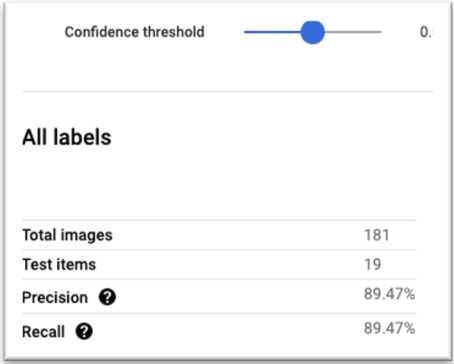

What do each of the cells in the confusion matrix describe? What values did you observe (include a screenshot)? What is the true positive rate for the “pneumonia” class? What is the false positive rate for the “normal” class?

Confusion matrix

True Label	Predicted Label	
	pneumonia	normal
pneumonia	8	1
normal	1	9

True Label	Predicted Label	
	pneumonia	normal
pneumonia	89%	11%
normal	10%	90%


The blue (diagonal) values refer to true predictions (positive or negative) while off-diagonal (grey) refers to false predictions (positive or negative). For example, in

	<p>the first row, a pneumonia (positive) case was predicted correctly (TP) 8 times (89% of the time) and was only once falsely predicted as normal (negative) or 11% of the time. With regard to the normal (negative) case, it was predicted correctly (TN) 8 times (90% of the time) and was only once falsely predicted (FP) as pneumonia 10% of the time.</p>										
<p><b>Precision and Recall</b>  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)?</p>	<p>Precision measures true predictions over total predication. While recall measures true predictions over total ground truth.</p>  <p>The screenshot shows a 'Confidence threshold' slider set to 0.5. Below it, a table titled 'All labels' displays the following metrics:</p> <table border="1"> <thead> <tr> <th>Metric</th> <th>Value</th> </tr> </thead> <tbody> <tr> <td>Total images</td> <td>181</td> </tr> <tr> <td>Test items</td> <td>19</td> </tr> <tr> <td>Precision ?</td> <td>89.47%</td> </tr> <tr> <td>Recall ?</td> <td>89.47%</td> </tr> </tbody> </table>	Metric	Value	Total images	181	Test items	19	Precision ?	89.47%	Recall ?	89.47%
Metric	Value										
Total images	181										
Test items	19										
Precision ?	89.47%										
Recall ?	89.47%										
<p><b>Score Threshold</b>  When you increase the threshold what happens to precision? What happens to recall? Why?</p>	 <p>The graph shows a Precision-Recall curve. The x-axis is labeled 'Confidence' and ranges from 0.0 to 1.0. The y-axis is labeled 'UU%' and ranges from 0% to 100%. A blue line represents 'Recall' and a red line represents 'Precision'. The Recall curve starts at (0.0, 100%) and drops to (1.0, 0%). The Precision curve starts at (0.0, 0%), rises sharply to about 89% at a confidence of 0.1, and then remains relatively flat until a confidence of 0.9, where it drops to 0%.</p> <p>Precision score is proportional to confidence and vice versa with recall case. For example, when the model predicts an image as normal with 60% confidence and 40% confidence, then in this case the outcome prediction would be normal and so on.</p>										

## Binary Classifier with Clean/Unbalanced Data

### Train/Test Split

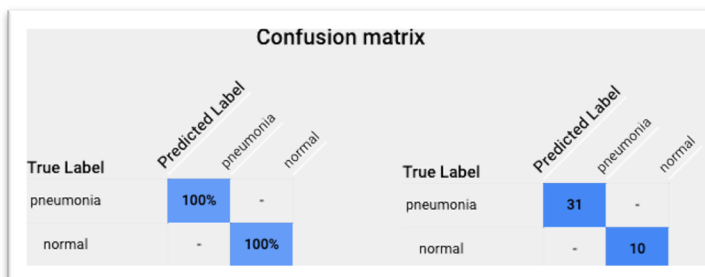
How much data was used for training? How much data was used for testing?



Label Stats					
Labels	Images	Train	Validation	Test	
normal	<div><div></div></div> 100	80	10	10	
pneumonia	<div><div></div></div> 300	239	30	31	

### Confusion Matrix

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



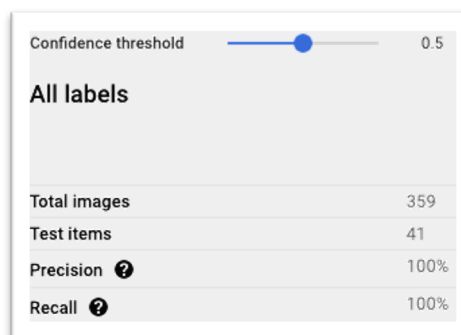
Confusion matrix		
True Label	Predicted Label	
	pneumonia	normal
pneumonia	100%	-
normal	-	100%

Confusion matrix		
True Label	Predicted Label	
	pneumonia	normal
pneumonia	31	-
normal	-	10

Yes, it improved to 100%.

### 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)?



Confidence threshold	
<div><div></div></div> 0.5	
All labels	
Total images	359
Test items	41
Precision ?	100%
Recall ?	100%

It (unexpectedly) increased to 100%.

### Unbalanced Classes

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

It weirdly gets higher scores probably due to higher exposure to training. Normally, such unbalance could create a bias or tendency in the model towards one prediction over another.

## 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.

Confusion matrix

True Label	Predicted Label	
	pneumonia	normal
pneumonia	80%	20%
normal	30%	70%

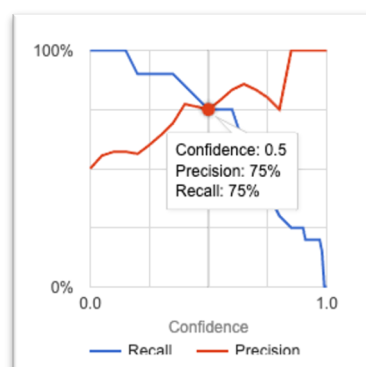
  

True Label	Predicted Label	
	pneumonia	normal
pneumonia	8	2
normal	3	7

It gets worse.

### 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?



It became very sensitive to confidence level with overall decrease in the performance by 15% to 75%.

### Dirty Data

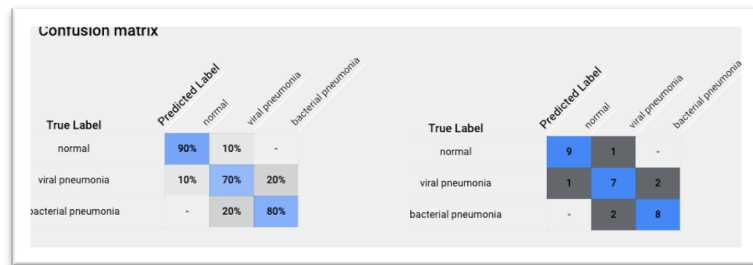
From what you have observed, how does dirty data affect a machine learning model?

It impacts the results negatively.

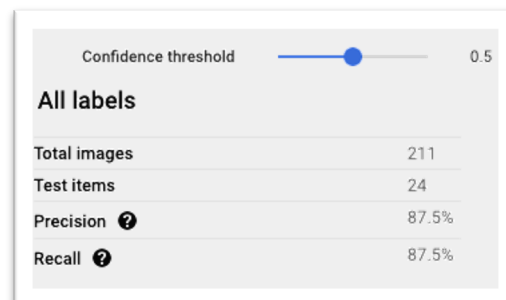
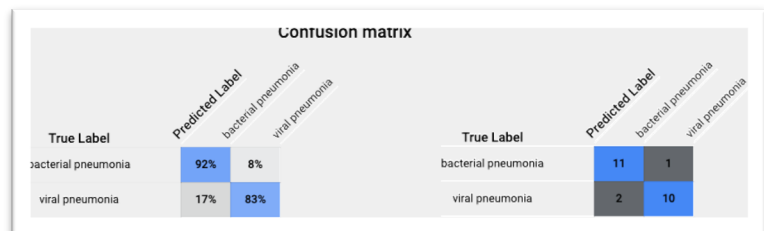
## 3-Class Model

### 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.



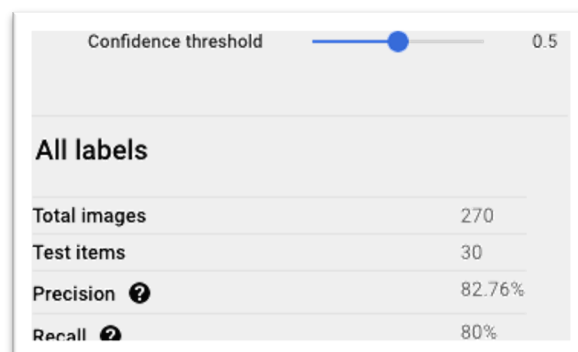
Viral pneumonia is most likely to confuse with least true prediction score (70%). While normal is most likely to get right with the highest prediction score (90%). To remedy that, I tried another model with a two-class pneumonia dataset (viral or bacterial) and got the following result:



So, the overall precision and recall scores were both enhanced, e.g. increased by 7.5%.

### 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)?



$$precision = \frac{\frac{9}{10} + \frac{7}{10} + \frac{8}{10}}{3} = 0.8$$

	$recall = \frac{\frac{9}{10} + \frac{7}{10} + \frac{8}{10}}{3} = 0.8$
<b>F1 Score</b> What is this model's F1 score?	$F1 = 2 \times \frac{82.76\% * 80\%}{82.76\% + 80\%} = 81.35\%$