# **AutoML Modeling Report**

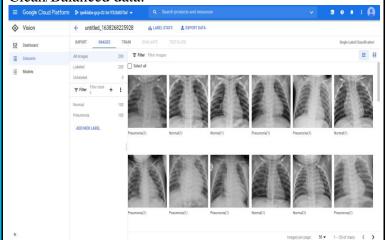


Etendra Verma

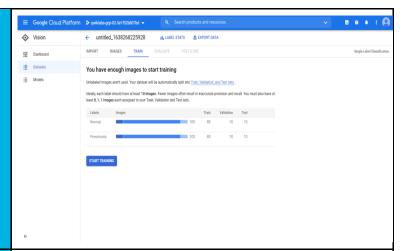
# Binary Classifier with Clean/Balanced Data

# Train/Test Split How much data was used for training? How much data was used for testing?

Initially, 200 images were used for first case 1. Out of which 100 belongs to Normal class and remaining 100 belongs to Pneumonia class in order to create a Clean/Balanced data.

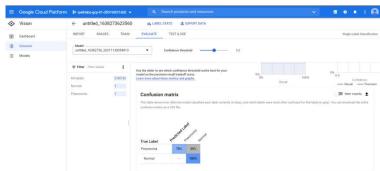


In which the 80% of the images are used for Training and 10 percent are used for Testing, remaining for Validation.

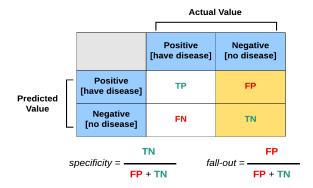


#### **Confusion Matrix**

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?



A confusion matrix is a summarized table of the number of correct and incorrect predictions (or actual and predicted values) yielded by a classifier (or classification model) for binary classification tasks.



**True Positives (TP):** These are cases in which we predicted yes for the data

**True Negatives (TN):** We predicted no for the data **False Positives (FP):** We predicted yes, but it's not true (Also known as a "Type I error.")

**False Negatives (FN):** We predicted no, but it's true (Also known as a "Type II error.")

True positive rate for Pneumonia class is 70%(all the cases are predicted accurately), False positive rate for Normal class is 0(not even a single case was predicted incorrectly.)

#### **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 proportion of relevant results in the list of all returned search results. The **Recall** is the ratio of the relevant results returned by the search engine to the total number of the relevant results that could have been returned.

The model achieved a precision of 85% and recall of 85%.

#### **Score Threshold**

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

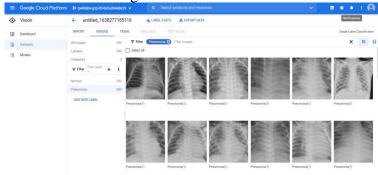
When you increase the score threshold, the precision goes up and recall seems to decrease. This is because when you increase the score threshold you want to be more confident when you make a prediction. Hence by increasing the score threshold, your will classify fewer images but it will have lower risk of misclassifying the images.

# Binary Classifier with Clean/Unbalanced Data



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

399 images were used, in which 100 belongs to Normal class and 299 belongs to Pneumonia class.

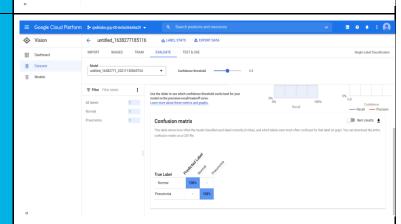


79.75% of the images are used for Training, 10% used for Testing and the remaining for Validation.

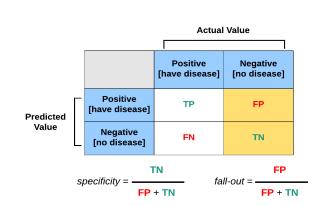


#### **Confusion Matrix**

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



A confusion matrix is a summarized table of the number of correct and incorrect predictions (or actual and predicted values) yielded by a classifier (or classification model) for binary classification tasks.



**True Positives (TP):** These are cases in which we predicted yes for the data

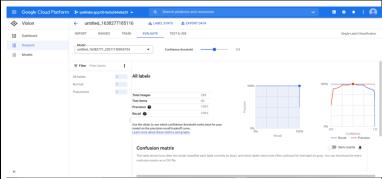
**True Negatives (TN):** We predicted no for the data **False Positives (FP):** We predicted yes, but it's not true (Also known as a "Type I error.")

**False Negatives (FN):** We predicted no, but it's true (Also known as a "Type II error.")

True positive rate for Normal class is 100% (all the cases are predicted accurately), True Negative positive rate for Pneumonia class is 0% (not even a single case was predicted wrongly). There is not much difference between the Balanced and unbalanced data.

#### **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 **Precision** is the proportion of relevant results in the list of all returned search results. The **Recall** is the ratio of the relevant results returned by the search engine to the total number of the relevant results that could have been returned.

#### **Unbalanced Classes**

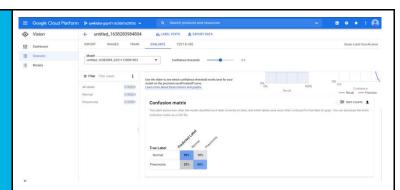
From what you have observed, how do unbalanced classed affect

Unbalanced data didn't introduce any bias in this case, the True positive and False negative rates were accurate and even the Precision, Recall of the model.

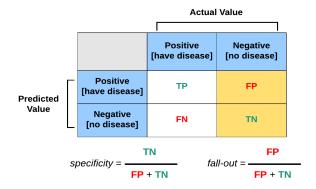
# Binary Classifier with Dirty/Balanced Data

#### **Confusion Matrix**

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



A confusion matrix is a summarized table of the number of correct and incorrect predictions (or actual and predicted values) yielded by a classifier (or classification model) for binary classification tasks.



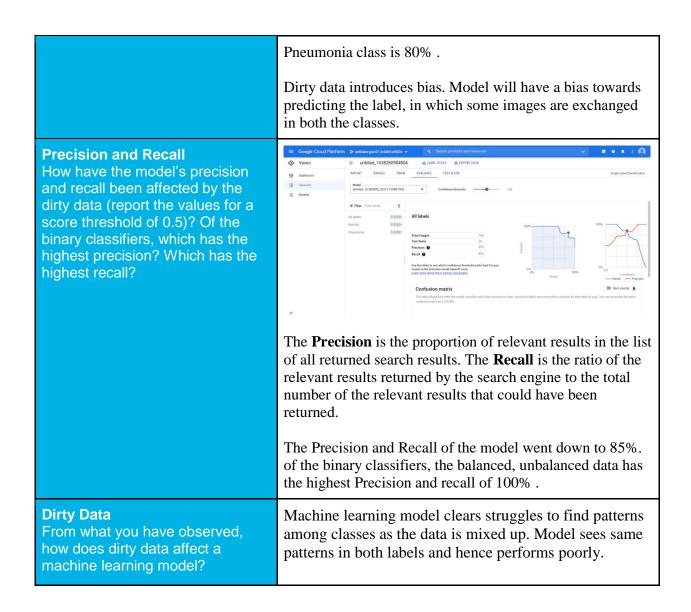
True Positives (TP): These are cases in which we predicted yes for the data

True Negatives (TN): We predicted no for the data False Positives (FP): We predicted yes, but it's not true (Also known as a "Type I error.")

False Negatives (FN): We predicted no, but it's true

(Also known as a "Type II error.")

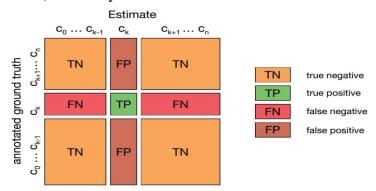
The True positive rate of Normal is 90%, False positive rate of Pneumonia class is 10%. Similarly, the False negative rate of Normal Class is 20% and True negative rate of



### 3-Class Model

# Confusion Matrix Summarize the 3-class confusion matrix. Which classes is the model most likely to confuse? Which classes(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.

A confusion matrix is a summarized table of the number of correct and incorrect predictions (or actual and predicted values) yielded by a classifier (or classification model) for binary classification tasks.



**True positives (TP):** These are cases in which we predicted yes for the data.

**True negatives (TN):** We predicted no for the data. **False positives (FP):** We predicted yes, but it's not true (Also known as a "Type I error.")

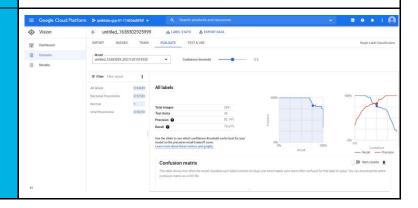
False negatives (FN): We predicted no, but it's true (Also known as a "Type II error.")

The True Negative rate of Viral Pneumonia class is 80%, True Negative rate of Normal class is 10%, the True Positive rate of Normal class is 100%, the True Negative rate of Viral Pneumonia class is 40%, True Negative rate of Bacterial Pneumonia rate is 60%.

The model which is like to confuse is Bacterial Pneumonia, Viral Pneumonia class. Normal class is likely to get right (True Positive rate -100%). We can add more images to each class as there only 100 images for each class now.

#### **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)?



	The Precision and Recall for the model is 82.14%, 76.67%. Precision is calculated with true positives by sum of true positives, false positives whereas Recall is calculated with true positives by sum of true positives and false negatives.
F1 Score What is this model's F1 score?	F1-score = $2 \times (0.8\% \times 0.6\%) / (0.8\% + 0.6\%) = 0.685714286\%$
	The F1 score of the model is 0.68%