# **AutoML Modeling Report**



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## Binary Classifier with Clean/Balanced Data

#### **Train/Test Split** Initially, 198 images were used. Out of which 99 to How much data was used for Normal class and remaining 99 belongs to pneumonia training? How much data was used class to create a Clean/Balanced data. for testing? August 10, 2024 at 13:33:42 2 training labels, 158 training PneumoniaDetection images, 2 test labels, 40 test In which the 80% of the images are used for Training and 20% are used for Testing. **Confusion Matrix** A **confusion matrix** is a table that is often used to What do each of the sections in describe the performance of a classification model (or the confusion matrix describe? "classifier") on a set of test data for which the true What values did you observe values are known. (include a screenshot)? What is True positives (TP): These are cases in which we the true positive rate for the predicted yes for the data "pneumonia" class? What is the True negatives (TN): We predicted no for the data false positive rate for the "normal" False positives (FP): We predicted yes, but it's not class? (Also known as a "Type I error.") False negatives (FN): We predicted no, but it's true (Also known as a "Type II error.")

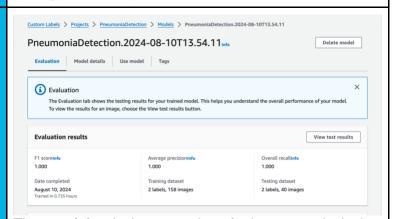
```
(4) EvaluationResultSummary-PneumoniaDetection-PneumoniaDetection.2024-08-10T13.54.11 ) No Selection
         "AggregatedEvaluationResults": {
            "ConfusionMatrix": [
                "GroundTruthLabel": "normal",
                "PredictedLabel": "normal", "Value": 0.95
                "GroundTruthLabel": "normal",
  11
12
13
                "PredictedLabel": "pneumonia",
                "Value": 0.05
  14
15
16
17
18
19
                "GroundTruthLabel": "pneumonia",
               "PredictedLabel": "normal",
               "Value": 0.0
                "GroundTruthLabel": "pneumonia",
"PredictedLabel": "pneumonia",
  20
21
22
  23
           "F1Score": 1.0,
           "Precision": 1.0.
           "Recall": 1.0
```

The true positive rate for the "pneumonia" class is 1.0 (the model has a 100% success rate in detecting pneumonia cases)

The false positive rate for the "normal" class is 0.05 (the model mistakenly identified 5% of the normal cases as having pneumonia)

#### **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 100% and recall of 100%.

### Binary Classifier with Clean/Unbalanced Data

#### **Train/Test Split**

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

298 images were used, in which 99 belongs to Normal class and 199 belongs to Pneumonia class

Project details

Project name
unbalancedBinaryClassifier
August 10, 2024 at 15:02:24
(UTC+03:00)

Dataset
2 training labels, 238 training images, 2 test labels, 60 test images

80% of the images are used for Training, 20% used for Testing.

#### **Confusion Matrix**

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

```
(4) EvaluationResultSummary-unbalancedBinaryClassifier-unbalancedBinaryClassifier.2024-08-10T15.37.41 ) No Selection
        "AggregatedEvaluationResults": {
           "ConfusionMatrix": [
               "GroundTruthLabel": "normal",
               "PredictedLabel": "normal",
               "Value": 0.95
  10
               "GroundTruthLabel": "normal",
               "PredictedLabel": "pneumonia",
  12
13
               "Value": 0.05
  15
16
17
18
19
               "GroundTruthLabel": "pneumonia",
               "PredictedLabel": "normal",
              "Value": 0.0
  20
21
22
              "GroundTruthLabel": "pneumonia",
               "PredictedLabel": "pneumonia".
               "Value": 1.0
  23
          "F1Score": 1.0,
  26
          "Precision": 1.0.
          "Recall": 1.0
```

The confusion matrix shows that the model has a high true positive rate for the "pneumonia" class (1.0) but a relatively high false positive rate for the "normal" class (0.05). This indicates that the model is biased towards the majority "pneumonia" class due to the unbalanced dataset, where there are more pneumonia samples than normal samples in the training data.

#### **Precision and Recall**

How have the model's precision and recall been affected by the unbalanced data?

Evaluation results		View test results	
F1 scoreInfo	Average precisionInfo	Overall recallinfo	
1.000	1.000	1.000	
Date completed	Training dataset	Testing dataset	
August 10, 2024 Trained in 0.865 hours	2 labels, 238 images	2 labels, 60 images	

The model's precision and recall are both 1.0, indicating excellent overall performance. However, this is likely due to the unbalanced nature of the dataset, where the model has learned to perform well on the majority "pneumonia" class, while the performance on the

	minority "normal" class may be less reliable.
Unbalanced Classes From what you have observed, how do unbalanced classed affect a machine learning model?	Unbalanced classes in a dataset can significantly bias a machine learning model towards the majority class, leading to poor performance on the minority class. This affects evaluation metrics and makes it challenging for the model to converge during training.

## 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 information.

```
(4) EvaluationResultSummary-binaryClassifierWithDirtyData-binaryClassifierWithDirtyData.2024-08-10T16.13.21 ) No Selectio
         "AggregatedEvaluationResults": {
            "ConfusionMatrix": [
               "GroundTruthLabel": "normal",
               "PredictedLabel": "normal",
               "Value": 0.4
  10
11
12
               "GroundTruthLabel": "normal",
"PredictedLabel": "pneumonia",
  13
14
               "GroundTruthLabel": "pneumonia",
  16
17
18
19
20
21
22
23
               "PredictedLabel": "normal", "Value": 0.0
               "GroundTruthLabel": "pneumonia",
               "PredictedLabel": "pneumonia",
               "Value": 1.0
           "F1Score": 0.8742928975487115,
           "Precision": 0.8836317135549872,
  27 28 },
          "Recall": 0.875
with a high false positive rate for the "normal" class
```

with a high false positive rate for the "normal" class (60% of normal instances misclassified as pneumonia), while perfectly classifying the "pneumonia" class. This indicates the model has become biased towards the majority "pneumonia" class, likely due to noisy or mislabeled data in the training set.

#### **Precision and Recall**

How have the model's precision and recall been affected by the dirty data. Of the binary classifiers, which has the highest precision? Which has the highest recall?

Evaluation results		View test result:	
F1 scoreInfo	Average precisionInfo	Overall recallinfo	
0.874	0.884	0.875	
Date completed	Training dataset	Testing dataset	
August 10, 2024 Trained in 0.822 hours	2 labels, 158 images	2 labels, 40 images	

The Precision and Recall of the model went down to 0.884 and 0.875. Of the binary classifiers, the balanced, unbalanced data has the highest Precision and recall of 1.00.

#### **Dirty Data**

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

Dirty data leads to:

- Trade-offs between precision and recall for different classes
- Unbalanced performance across classes
- Overall degradation in model metrics like F1score

### 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 information.

The model is most likely to confuse "bacterial pneumonia" and "viral pneumonia" classes, with 5% of "bacterial pneumonia" predicted as "viral pneumonia" and 40% of "viral pneumonia" predicted as "bacterial pneumonia".

The model performs best on the "normal" class, correctly predicting 100% of the samples.

To improve the model's performance, you could:

- Collect more training data for the pneumonia classes
- 2. Explore better feature engineering or model architectures to capture differences between

	pneumonia types 3. Use class-weighted loss functions to focus the model on the more challenging pneumonia classes				
Precision and Recall	Evaluation results		View test results		
What are the model's precision and recall? How are these values	F1 scoreInfo 0.864	Average precisionInfo 0.867	Overall recallinfo 0.867		
calculated?	Date completed August 10, 2024 Trained in 1.006 hours	Training dataset 3 labels, 237 images	Testing dataset 3 labels, 60 images		
	The model's overall precision is 0.867 and recall is 0.867.  Precision measures the fraction of true positive predictions out of all positive predictions.  Recall measures the fraction of true positive predictions out of all actual positive instances				
F1 Score What is this model's F1 score?	The F1 score of the model is 0.864.				