

SciKit Learn



Mass Street
University
PER EDUCATIONEM PROGRESSUS

Accuracy

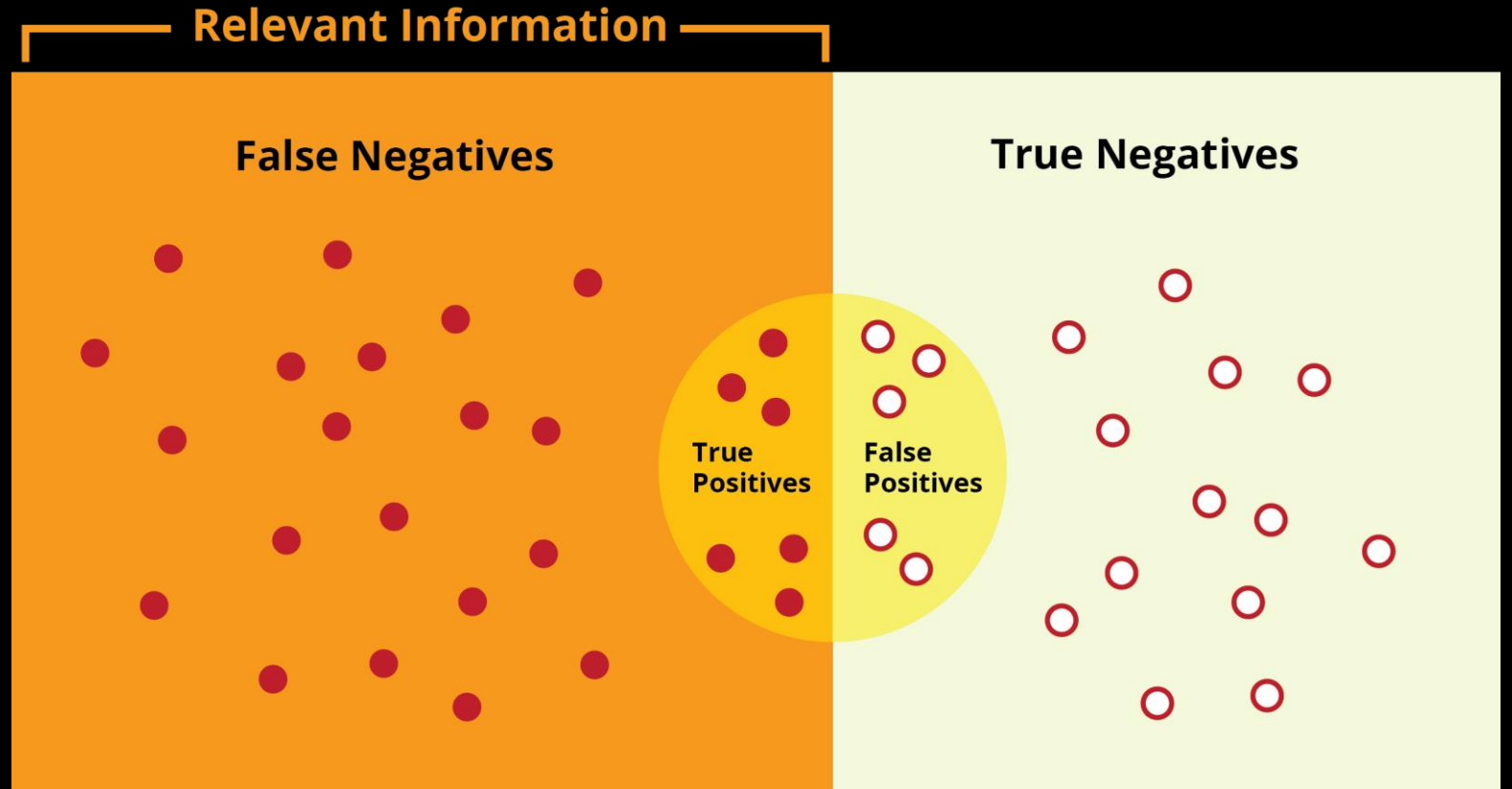
- Y true (left hand) equals the number of true cases we have in our data.
- Y pred (right hand) equals the number of predicted cases we have in our data.
- This will always be a value between 0 and 1.

$$A = \frac{y}{\hat{y}}$$



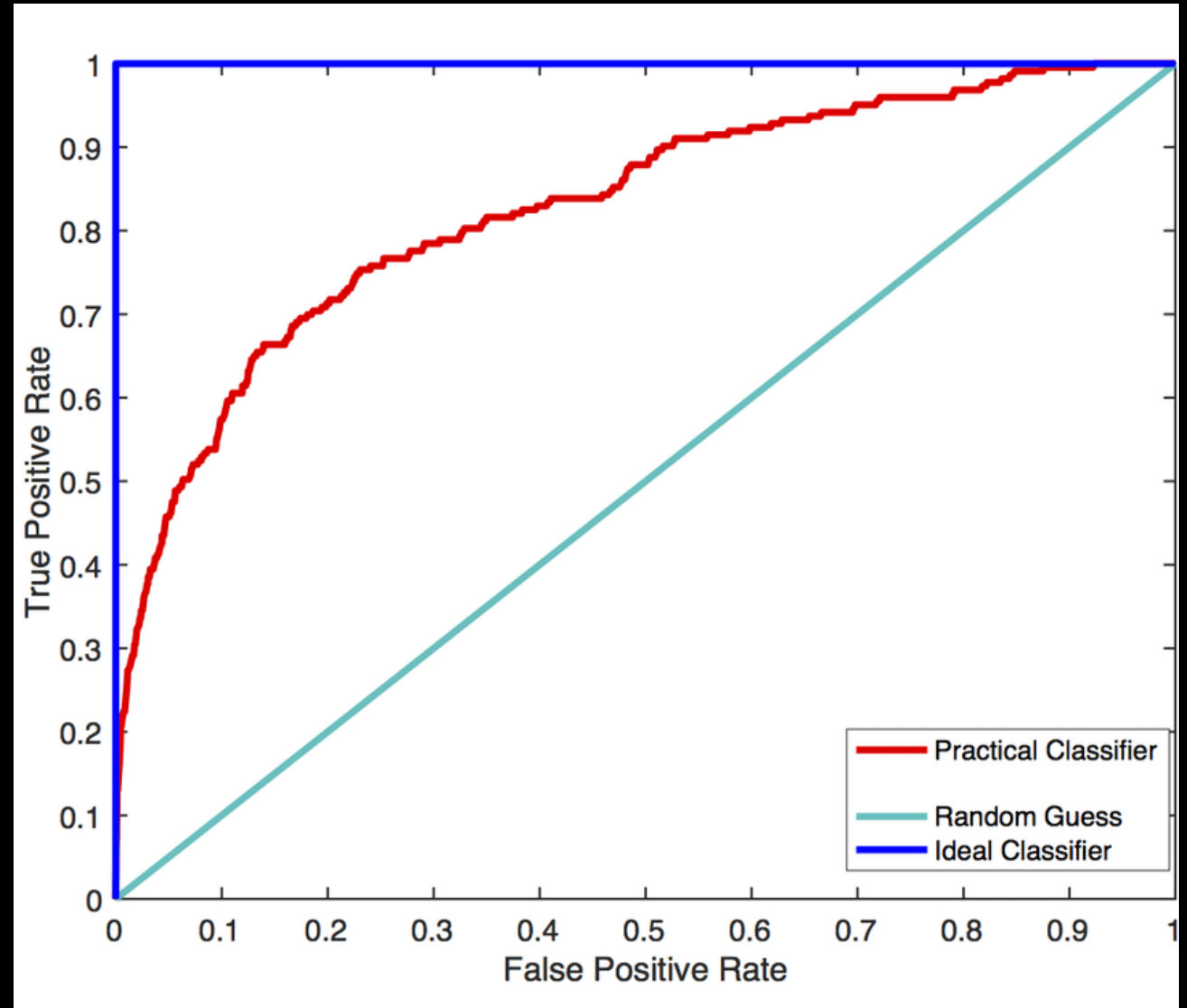
True and False

- The circle is the positive predictions the square outside the circle is the negative predictions.
- The orange side are actual positives. The white side are actual negatives.



ROC curve

- Perfect classifications are at $(0, 1)$.
- The worst classifications are at $(1, 0)$.



Confusion matrix

	Predicted (0)	Predicted (1)
Actual (0)	TN	FP
Actual (1)	FN	TP

True Positive (TP)

Patient has pneumonia
Model predicts: pneumonia

Number of occurrences: 1

False Positive (FP)

Patient is healthy
Model predicts: pneumonia

Number of occurrences: 1

False Negative (FN)

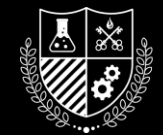
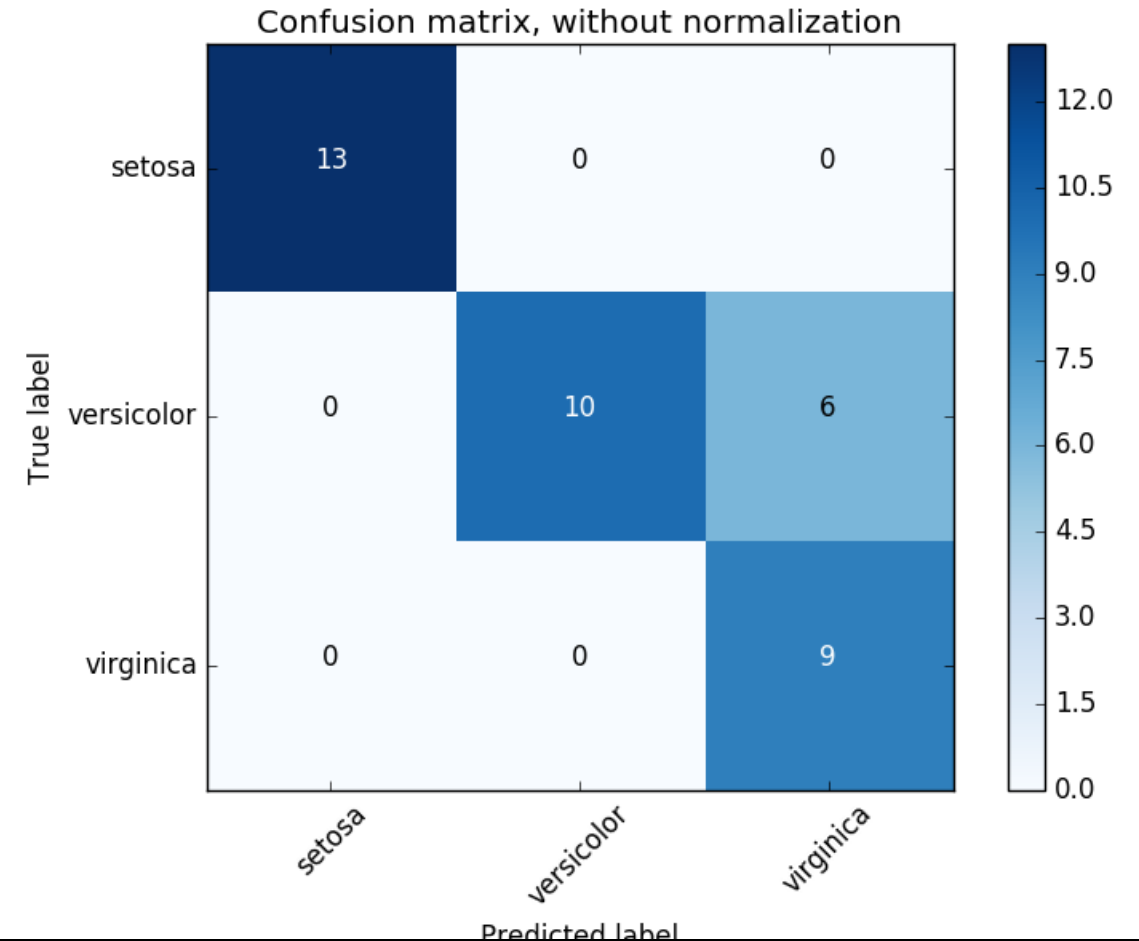
Patient has pneumonia
Model predicts: healthy

Number of occurrences: 8

True Negative (TN)

Patient is healthy
Model predicts: healthy

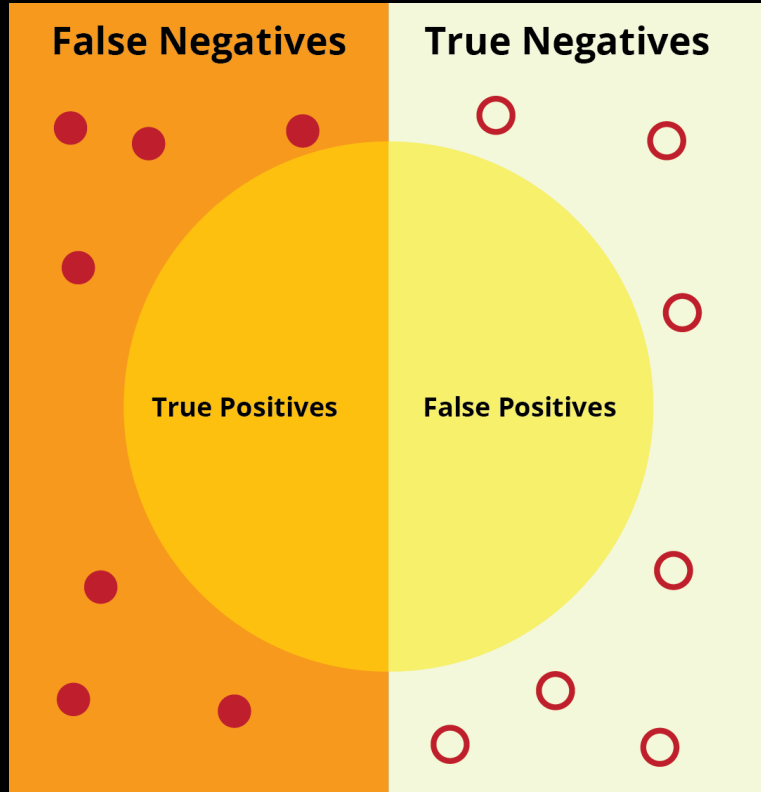
Number of occurrences: 90



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Precision vs Recall



How many selected items are relevant?

$$\text{Precision} = \frac{\text{True Positives}}{\text{True Positives} + \text{False Positives}}$$

How many relevant items are selected?

$$\text{Recall} = \frac{\text{True Positives}}{\text{True Positives} + \text{False Negatives}}$$



Mean Squared Error

- Very sensitive to outliers.
- Better alternative is mean squared error.

$$MSE = \frac{1}{n} \sum \left(y - \hat{y} \right)^2$$

The square of the difference
between actual and
predicted



Cross Entropy (Log Loss)

$$H_p(q) = -\frac{1}{N} \sum_{i=1}^N y_i \cdot \log(p(y_i)) + (1 - y_i) \cdot \log(1 - p(y_i))$$

Binary Cross-Entropy / Log Loss



Cross Validation

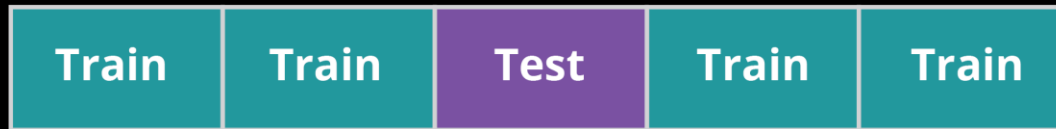
Iteration 1



Iteration 2



Iteration 3



Iteration 4



Iteration 5



Holdout Testing

