Evaluation: From precision, recall, and F-measure to ROC, informedness, markedness and correlation

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Co-authors View all... Trent Wilson Lewis Martin Luerssen Richard Leibbrandt Adham Atvabi Donagiana Yana Sean Patrick Fitzgibbon

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Introduction

► Contingency table



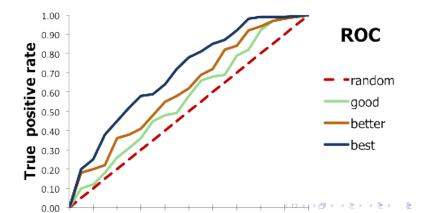


- Measure to evaluate Machine Learning System
- 1. Precision
- 2. Recall
- 3. F Measure
- 4. Issues with them
- Alternate Techniques
- 1. Accuracy
- 2. Cohen Kappa



ROC Analysis

- ► ROC analysis give geometric insights into the nature of the measures and their sensitivity to skew
- Compare Classifiers
- Choose parameters based on maximization of AUC



Informedness and Markedness

Markedness

Markedness quantifies how marked a condition is for the specified predictor, and specifies the probability that a condition is marked by the predictor (versus chance).

Markedness is a deep measure of how consistently the outcome has the Predictor as a Marker by combining surface measures about what proportion of Predictions are correct

Precision + Inverse Precision - 1

Informedness

Informedness quantifies how informed a predictor is for the specified condition, and specifies the probability that a prediction is informed in relation to the condition (versus chance).

Informedness is a deep measure of how consistently the predictor predicts the outcome by combining surface measures about what proportion of outcomes are correctly predicted.

Regression

- ► Linear Regression
- Estimating Coefficients
- 1. Rp
- 2. Rr
- 3. Rg

Conclusion

Infromedness usually is a better evaluation measure in binary classification.

Further work to research into the multiclass application of the technique.

Explore the relationship between Infromedness and Markedness

We have a better and a more intuitive understanding of the mearues that we could use for evaluating our ML system.

- Accuracy
- Cohen Kappa
- Precision
- Recall
- ► F1-Score
- ► Bookmaker Informedness / Delta P'
- Markedness

Skipped

- Significance
- Montecarlo Simulation
- ► Evenness (Used in multiclass problems)