measures Confusion Matrix Additional info Alternative (Classifier Uncertainty Information Cost ratio, skew) **Deterministic Classifiers** Continuous and Scoring Classifiers Prob. Classifiers (Reliability metrics) Multiclass Single-Class Focus Focus Information Distance/ Graphical Summary Error Theoretic measures Statistics measures Measures ROC Curves No KL divergence Chance PR Curves l measure RMSE K&B IR DET Curves Chance Correction BIR Correction Lift Charts Cost Curves Interestingness Comprehensibility Cohen's Kappa Precision-Recal Multicriteria Fleiss's Kappa Sens-Spec F measure Geom. Mean Dice

Figure 3.1. An ontology of performance metrics. (KL, Kullback–Leibler; BIR, Bayesian information reward; K & B IR, Kononenko and Bratko information reward.)

RESPONSE AND MODEL PERFORMANCE

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DATA 607

3/18/20

PRESENTATION GOALS

- Definitions
- Practical example: plot in R caret package
- Strengths and weaknesses of this measure

CUMULATIVE RESPONSE: DEFINITION

- Proportional measure of model performance
- X-axis: percentage of population tested
- Y-axis: proportion of true positives
- Related measure: lift and lift curve
- Visuals may be confused with each other

CUMULATIVE RESPONSE VISUALLY

Comparison of several classifiers

Vertical axis:

percent true positive, or

hit rate

Blue line y = x
Random guess

More efficient = Steeper slope

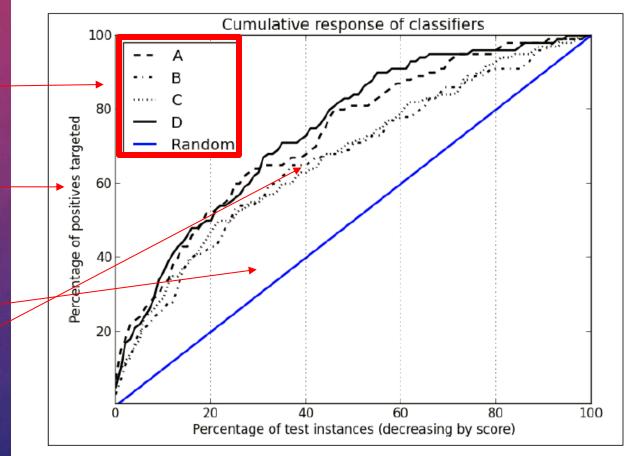


Figure 8-6. Four example classifiers (A-D) and their cumulative response curves.

COMPARISON: LIFT CURVE

Y axis of cumulative response divided by x-axis

Blue line: y = x, Same as before

"y times more lift" =
"y better than random"

Effective classifiers get more positives relative to sample

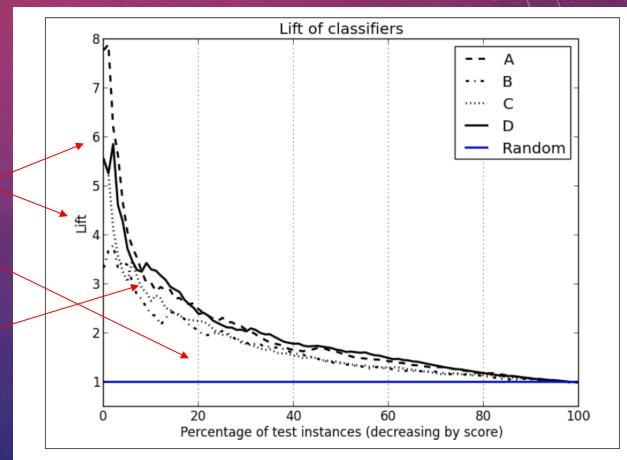


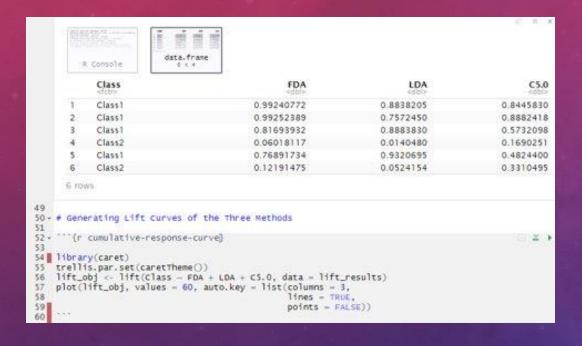
Figure 8-7. The four classifiers (A-D) of Figure 8-6 and their lift curves.

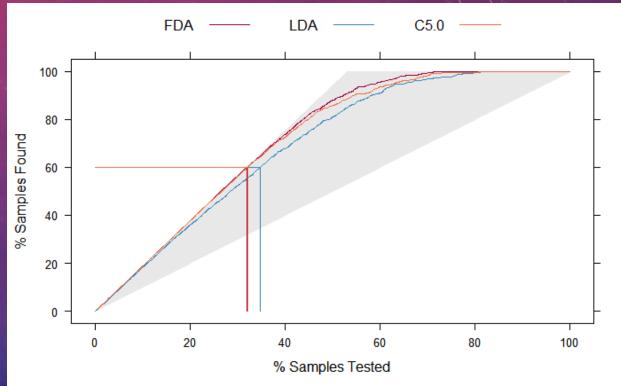
EXAMPLE: CARET PACKAGE

Classification And REgression Training

```
`{r caret-example-data}
    set.seed(2)
   lift_training <- twoClassSim(1000)
   lift_testing <- twoClassSim(1000)
20
   ctrl <- trainControl(method = "cv", classProbs = TRUE,
                         summaryFunction = twoClassSummary)
23
   set.seed(1045)
   fda_lift <- train(Class ~ ., data = lift_training,
                      method = "fda", metric = "ROC",
27
                      tuneLength = 20.
28
                      trControl = ctrl)
   set.seed(1045)
   lda_lift <- train(Class ~ ., data = lift_training,</pre>
                      method = "lda", metric = "ROC",
32
                      trControl = ctrl)
33
   library(C50)
    set.seed(1045)
   c5_lift <- train(Class ~ ., data = lift_training,
                     method = "C5.0", metric = "ROC",
37
38
                     tuneLength = 10,
                     trControl = ctrl.
                     control = C5.0Control(earlyStopping = FALSE))
41
   ## Generate the test set results
   lift_results <- data.frame(Class = lift_testing$Class)</pre>
  lift_results$FDA <- predict(fda_lift, lift_testing, type = "prob")[,"Class1"]
   lift_results$LDA <- predict(lda_lift, lift_testing, type = "prob")[,"Class1"]
   lift_results$C5.0 <- predict(c5_lift, lift_testing, type = "prob")[,"class1"]
   head(lift_results)
```

EXAMPLE: CARET VISUAL





STRENGTHS AND LIMITATIONS

- Strength: Intuitive representation
- Weakness: hard to generalize to scenarios with different class proportions
- Strength: can be run without known operations costs
- Weakness: cannot measure the profitability of a model
- Strength: considers beyond the accuracy of a model
- Weakness: No direct comparison between all tp and fp rates (ROC space)

