
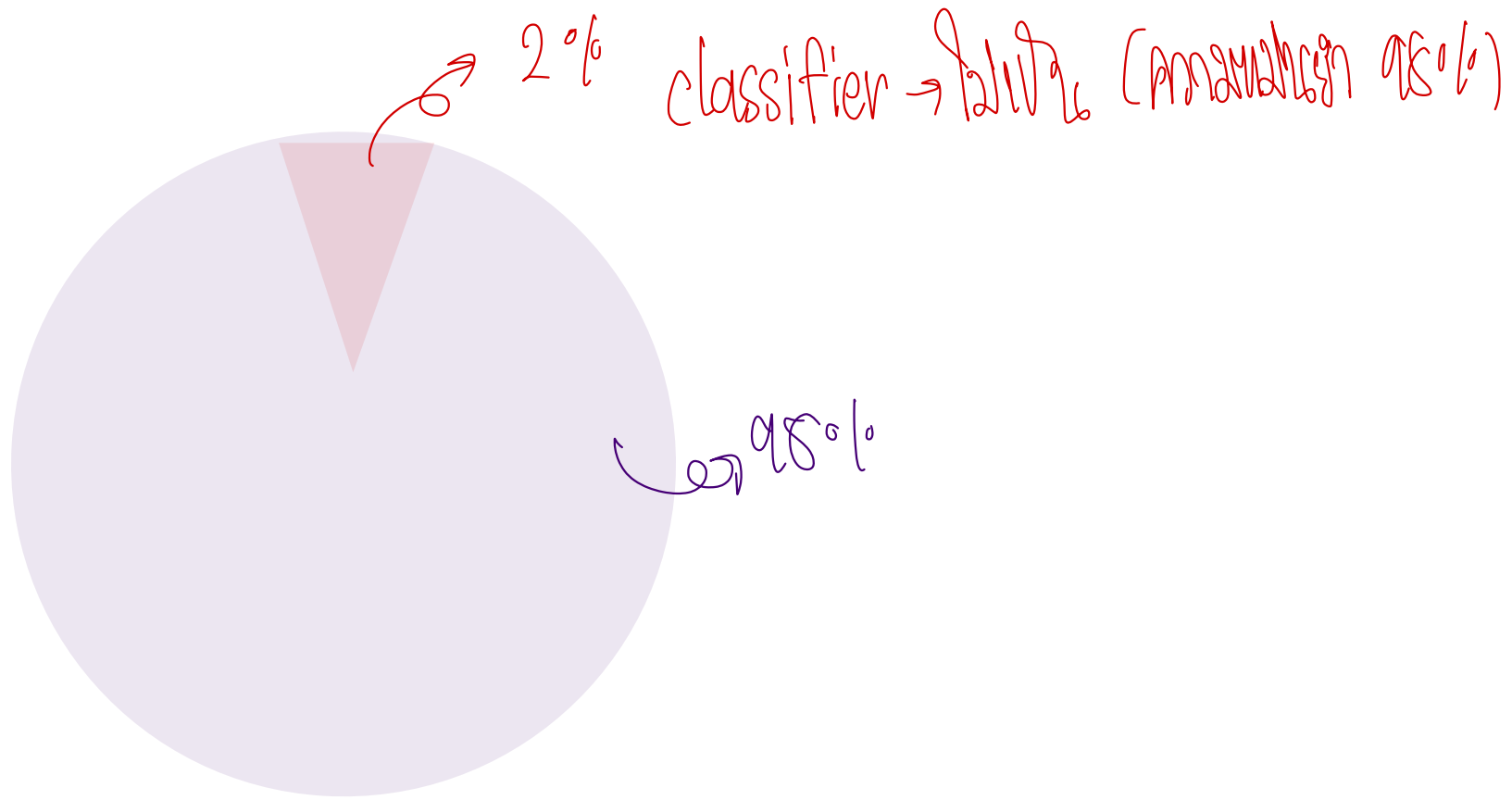


Chapter 8. Classification: Basic Concepts

- ❑ Classification: Basic Concepts
- ❑ Decision Tree Induction
- ❑ Bayes Classification Methods
- ❑ Linear Classifier
- ❑ Model Evaluation and Selection 
- ❑ Techniques to Improve Classification Accuracy: Ensemble Methods
- ❑ Additional Concepts on Classification
- ❑ Summary

Model Evaluation and Selection

- ❑ Evaluation metrics
 - ❑ How can we measure accuracy?
 - ❑ Other metrics to consider?
- ❑ Use **validation test set** of class-labeled tuples instead of training set when assessing accuracy
- ❑ Methods for estimating a classifier's accuracy
 - ❑ Holdout method
 - ❑ Cross-validation
 - ❑ Bootstrap
- ❑ Comparing classifiers:
 - ❑ ROC Curves



Classifier Evaluation Metrics: Confusion Matrix

Confusion Matrix: *Amou nintisa*

Actual class \ Predicted class	C_1 <i>yes</i>	$\neg C_1$ <i>no</i>
C_1	True Positives (TP)	False Negatives (FN) <i>Recall</i>
$\neg C_1$	False Positives (FP)	True Negatives (TN)

Handwritten notes:
 - Above TP: *precision*
 - Above FN: *recall*
 - Above FP: *no yes no*
 - Above TN: *yes no no*

In a confusion matrix w. m classes, $CM_{i,j}$ indicates # of tuples in class i that were labeled by the classifier as class j

May have extra rows/columns to provide totals

Example of Confusion Matrix:

Actual class \ Predicted class	buy_computer = yes <i>post</i>	buy_computer = no <i>neg</i>	Total
buy_computer = yes <i>positive</i>	6954	46	7000
buy_computer = no <i>negative</i>	412	2588	3000
Total	7366	2634	10000

Classifier Evaluation Metrics: Accuracy, Error Rate, Sensitivity and Specificity

A\P	C	¬C	
C	TP	FN	P
¬C	FP	TN	N
	P'	N'	All

- **Classifier accuracy**, or recognition rate

- Percentage of test set tuples that are correctly classified

$$\text{Accuracy} = (TP + TN) / \text{All}$$

- **Error rate**: $1 - \text{accuracy}$, or
 $\text{Error rate} = (FP + FN) / \text{All}$

- **Class imbalance problem**

- One class may be *rare*
 - E.g., fraud, or HIV-positive
- Significant *majority of the negative class* and minority of the positive class
- Measures handle the class imbalance problem
 - **Sensitivity** (recall): True positive recognition rate
 - $\text{Sensitivity} = TP / P$
 - **Specificity**: True negative recognition rate
 - $\text{Specificity} = TN / N$

Classifier Evaluation Metrics: Precision and Recall, and F-measures

- ❑ **Precision:** Exactness: what % of tuples that the classifier labeled as positive are actually positive?

$$P = \text{Precision} = \frac{TP}{TP + FP}$$

which model marked pos not actually pos

- ❑ **Recall:** Completeness: what % of positive tuples did the classifier label as positive?

$$R = \text{Recall} = \frac{TP}{TP + FN}$$

which pos truly marked as pos actually

- ❑ Range: [0, 1]
- ❑ The “inverse” relationship between precision & recall
- ❑ **F measure (or F-score):** harmonic mean of precision and recall
 - ❑ In general, it is the weighted measure of precision & recall

$$F_{\beta} = \frac{1}{\alpha \cdot \frac{1}{P} + (1 - \alpha) \cdot \frac{1}{R}} = \frac{(\beta^2 + 1)PR}{\beta^2 P + R}$$

Assigning β times as much weight to recall as to precision)

- ❑ **F1-measure (balanced F-measure)**

- ❑ That is, when $\beta = 1$,

$$F_1 = \frac{2PR}{P + R}$$

F score is } average of precision and recall