


- 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

47

Classifier Evaluation Metrics: Confusion Matrix

□ Confusion Matrix:

Actual class \ Predicted class	Precision	
	C_1	$\neg C_1$
C_1	True Positives (TP)	False Negatives (FN)
$\neg C_1$	False Positives (FP)	True Negatives (TN)

- In a confusion matrix w. m classes, CM_{ij} 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	buy_computer = no	Total
buy_computer = yes	6954	46	7000
buy_computer = no	412	2588	3000
Total	7366	2634	10000

49

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

AVP	C	$\neg C$	
C	TP	FN	P
$\neg C$	FP	TN	N
	P'	N'	All

- **Classifier accuracy**, or recognition rate
 - Percentage of test set tuples that are correctly classified
 - **Accuracy** = $(TP + TN) / All$
- **Error rate**: $1 - accuracy$, or **Error rate** = $(FP + FN) / All$
- **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
 - **Sensitivity** = TP / P
- **Specificity**: True negative recognition rate
 - **Specificity** = TN / N

50

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

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

$$P = \text{Precision} = \frac{TP}{TP + FP}$$
- **Recall**: Completeness: what % of positive tuples did the classifier label as positive?

$$R = \text{Recall} = \frac{TP}{TP + FN}$$
 - 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 $\beta = 1$, $F_1 = \frac{2PR}{P + R}$

51

Chapter 9. Classification: Advanced Methods

ทฤษฎีการเรียนรู้ของเครื่องที่ใช้ในการตัดสินใจแบบอัตโนมัติ

- ❑ Bayesian Belief Networks
- ❑ Support Vector Machines
- ❑ Neural Networks and Deep Learning
- ❑ Pattern-Based Classification
- ❑ Lazy Learners and K-Nearest Neighbors
- ❑ Other Classification Methods
- ❑ Summary

3

ทฤษฎีการเรียนรู้ของเครื่องที่ใช้ในการตัดสินใจแบบอัตโนมัติ

ทฤษฎีการเรียนรู้ของเครื่องที่ใช้ในการตัดสินใจแบบอัตโนมัติ

From Naïve Bayes to Bayesian Networks

- ❑ Naïve Bayes classifiers assume that the value of a particular feature is independent of the value of any other feature, given the class variable
 - ❑ This assumption is often too simple to model the real world well
- ❑ Bayesian network (or Bayes network, belief network, Bayesian model or probabilistic directed acyclic graphical model) is a probabilistic **graphical model**
 - ❑ Represented by a set of *random variables* and *their conditional dependencies* via a *directed acyclic graph* (DAG)
 - ❑ Ex. Given symptoms, the network can be used to compute the probabilities of the presence of various diseases



4

Discussion on the k-NN Algorithm

1.1. ขั้นตอนการหา k-NN

1.2. ขั้นตอนการหา k-NN

1.3. ขั้นตอนการหา k-NN

- ❑ k-NN for real-valued prediction for a given unknown tuple
 - ❑ Returns the mean values of the **k nearest neighbors**
- ❑ Distance-weighted nearest neighbor algorithm
 - ❑ Weight the contribution of each of the k neighbors according to their distance to the query x_q
 - ❑ Give greater weight to closer neighbors
- ❑ Robust to noisy data by averaging **k-nearest neighbors**
- ❑ Curse of dimensionality: distance between neighbors could be dominated by irrelevant attributes
 - ❑ To overcome it, axes stretch or elimination of the least relevant attributes

$$w \equiv \frac{1}{d(x_q, x_i)^2}$$

67