Inductive Learning Algorithms and Representations for Text Categorization

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Text Categorization

real-time sorting emails/files

topic identification

structured search and/or browsing

finding documents that match long-term standing interests

Old School

Dewey Decimal

MeSH(Medical Subject Headings)

Yahoo!'s topic hierarchy

CyberPatrol

Evaluation

Results & Others

Data:

a collection of hand-tagged financial newswire stories from Reuters.

http://www.research.att.com/~lewis/reuters21578.html (no longer available)

Classifiers

Inductive Learning of Classifiers

Classifiers

Classifiers

$$\rightarrow -x = (x1,x2,x3...xn)$$

$$f(\rightarrow x) = confidence(class)$$

eg. class- interest

if (interest AND rate) OR (quarterly), then confidence(cat interest) = 0.9

confidence(interest cat) = 0.3*interest + 0.4*rate + 0.7*quarterly

Inductive Learning of Classifiers

Find Similar (a variant of Rocchio's method for relevance feedback)

Decision Tree

Naive Bayes

Naive Nets

SVM

*All methods require only on a small amount of labeled training data The effectiveness of the model is tested on previously unseen instances.

Inductive Learning of Classifiers

Find Similar (a variant of Rocchio's method for relevance feedback)

-tf*idf

-all features used

$$x_j = \alpha \cdot x_{q,j} + \beta \cdot \frac{\sum\limits_{i \in rel} x_{i,j}}{n_r} + \gamma \cdot \frac{\sum\limits_{i \in non-rel} x_{i,j}}{N-n_r}$$

$$= \sum_{\beta \cdot \frac{i \in rel}{n_r}} x_{i,j}$$

*no error minimization is applied

Inductive Learning of Classifiers

Feature selection

$$MI(x_i,c) = \sum_{x_i \in [0,1]} \sum_{c \in [0,1]} P(x_i,c) \log \frac{P(x_i,c)}{P(x_i)P(c)}$$

SVM: K = 300

The remaining: K = 50

Only binary feature values are used

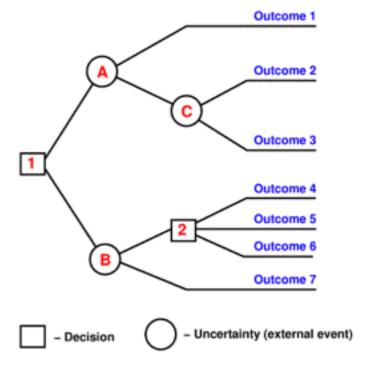
Inductive Learning of Classifiers

Decision Tree

Recursive greedy splitting

Bayesian posterior probability

Node → class probability



Inductive Learning of Classifiers

Naive Bayes

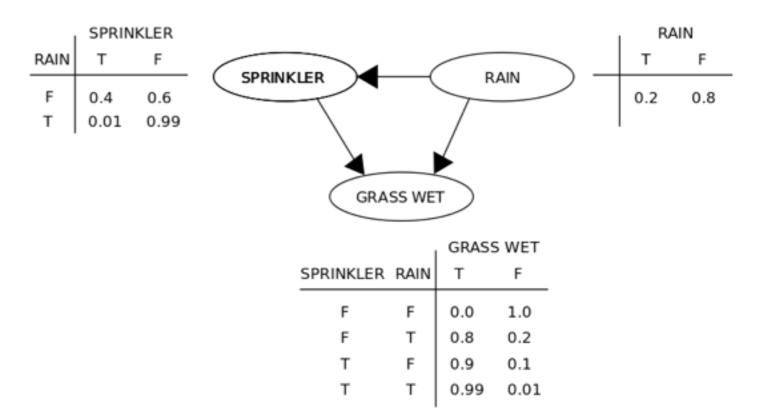
$$P(C = c_k \mid \vec{x}) = \frac{P(\vec{x} \mid C = c_k)P(C = c_k)}{P(\vec{x})}$$

Assume the features X1,....Xn are conditionally independent

$$P(\vec{x} \mid C = c_k) = \prod_i P(x_i \mid C = c_k)$$

Inductive Learning of Classifiers

Bayes Nets

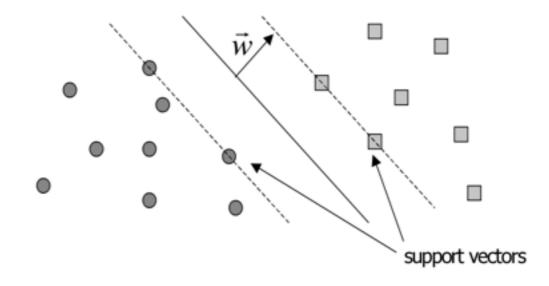


2-dependence Bayesian classifier

Inductive Learning of Classifiers

SVM

Simplest linear version



$$y_i(\vec{w}\cdot\vec{x}_i-b)\geq 1$$

Evaluation

Reuters-21578

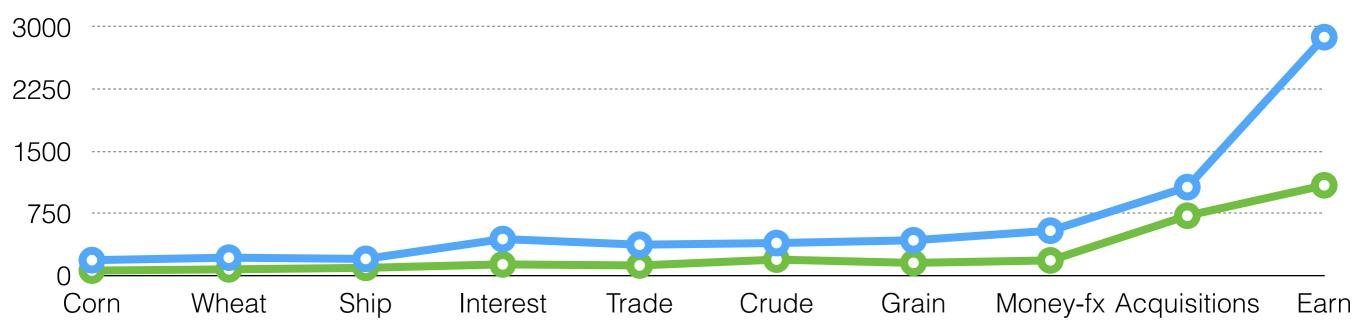
Summary of Inductive Learning Process

Reuters-21578

21578 collection, 200 words in length

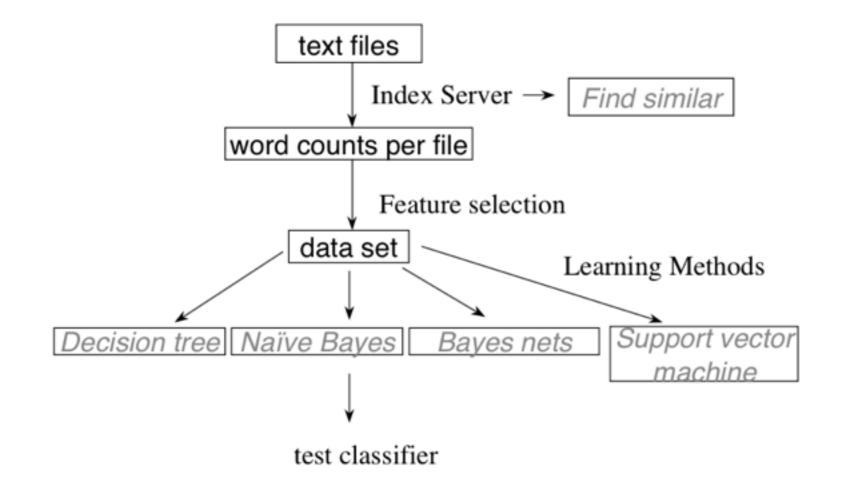
118 categories

75% train, 25% test



Category Name	Num Train	Num Test	
Earn	2877	1087	
Acquisitions	1650	719	
Money-fx	538	179	
Grain	433	149	
Crude	389	189	
Trade	369	118	
Interest	347	131	
Ship	197	89	
Wheat	212	71	
Corn	182	56	

Summary of Inductive Learning Process



Average of precision and recall(F measure?)
Train/test dataset not optimized

Results & Others

Training Time

Classification Speed for New Instances

Classification Accuracy

Other Experiments

Training Time

266 MHz Pentium II running Windows NT.

Fastest: Find Similar (<1 CUP sec/cat)

SVM (<2 CUP sec/cat)

Naive Bayes(8 CPU sec/cat)

Decision Trees (~70 CUP sec/cat)

Slowest: Bayes Nets(~145 CUP sec/cat)

	Findsim	NBayes	BayesNets	Trees	LinearSVM
earn	92.9%	95.9%	95.8%	97.8%	98.0%
acq	64.7%	87.8%	88.3%	89.7%	93.6%
money-fx	46.7%	56.6%	58.8%	66.2%	74.5%
grain	67.5%	78.8%	81.4%	85.0%	94.6%
crude	70.1%	79.5%	79.6%	85.0%	88.9%
trade	65.1%	63.9%	69.0%	72.5%	75.9%
interest	63.4%	64.9%	71.3%	67.1%	77.7%
ship	49.2%	85.4%	84.4%	74.2%	85.6%
wheat	68.9%	69.7%	82.7%	92.5%	91.8%
corn	48.2%	65.3%	76.4%	91.8%	90.3%
Avg Top 10	64.6%	81.5%	85.0%	88.4%	92.0%
Avg All Cat	61.7%	75.2%	80.0%	N/A	87.0%

Table 2 – Breakeven Performance for 10 Largest Categories, and over all 118 Categories.

New Instances?

All less than 2 sec

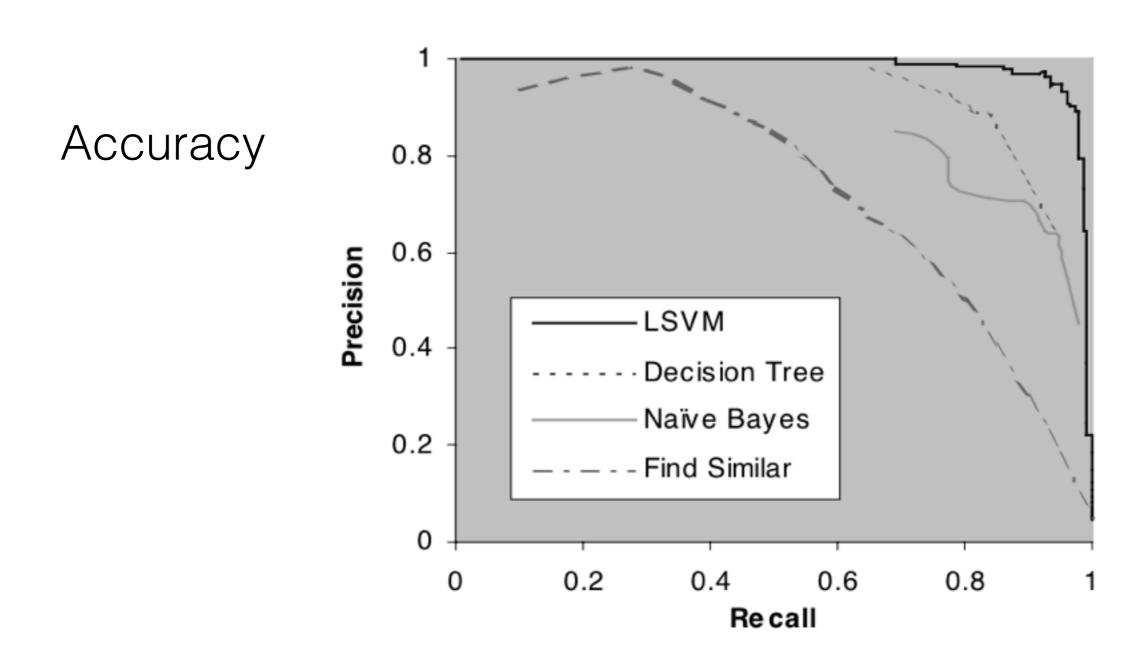


Figure 3 – Precision-Recall Curve for Category "grain"

Others?

Sample Size

N-gram

Binary vs. 0/1/2 features

