

TIM 245 Lecture 6 (4/19/17)

Agenda

1) Review Homework 1

2) Feature Selection using  
Information Gain

3) Wrapper based methods  
and general comments on  
feature selection

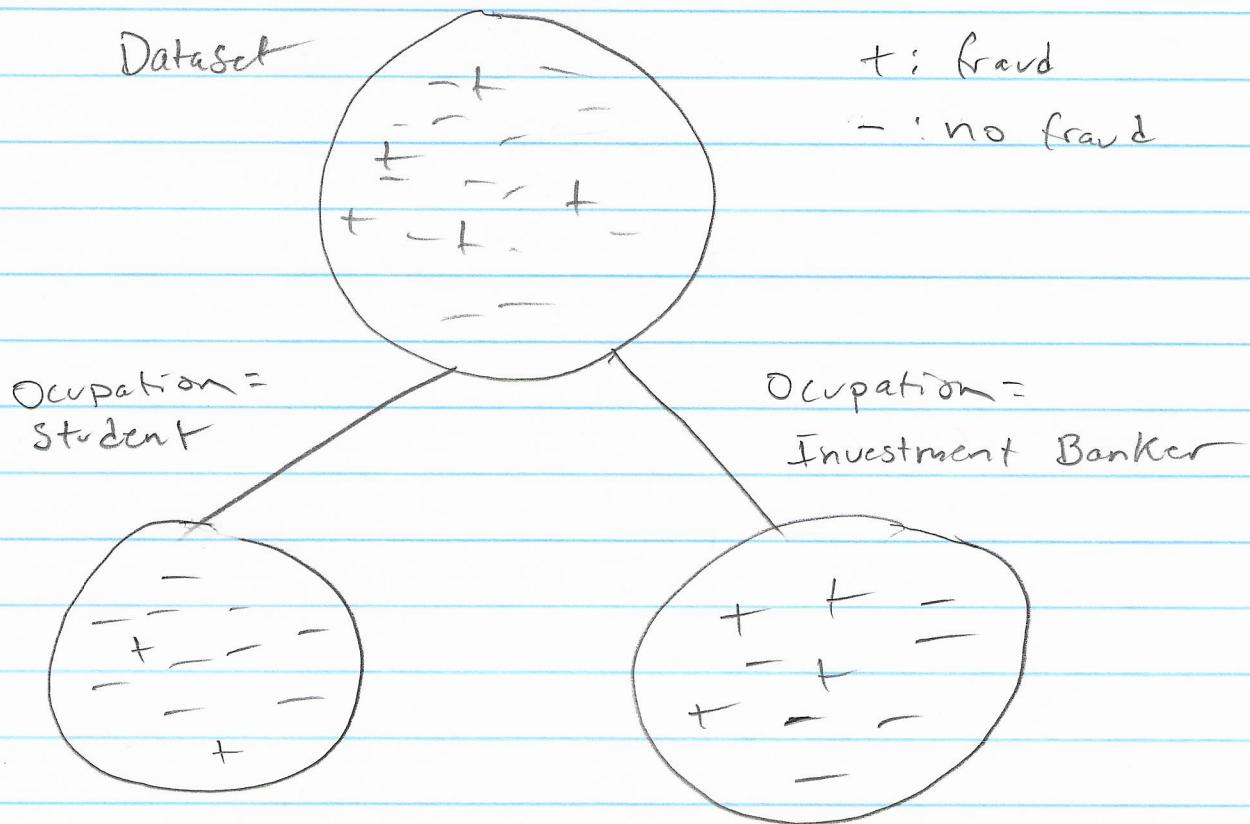
4) Dimensionality Reduction: PCA

5) Project Phase II and roadmap  
for the course

## ② Feature Selection using Information Gain

Each attribute gives us information on the target  $y$ .

Example : Fraud Detection



How can we select the attributes that provide the most information on  $y$ ?

Let :

$c_i^y \triangleq$  i<sup>th</sup> ( $i=1, 2, \dots, r$ ) of  
the target  $y$

$c_k^{A_j} \triangleq$  k<sup>th</sup> ( $k=1, 2, \dots, s$ ) of  $A_j$

$p(c_i^y) \triangleq$  probability of  $c_i^y$  estimated  
as  $|c_i^y| / |D|$

Process for computing information gain  
of attribute  $A_j$ :

- 1) Compute the entropy of the complete dataset  $D$

$$\text{Entropy}(D) = -\sum_{i=1}^r p(c_i^y) \log_2 (p(c_i^y))$$

- 2) Partition  $D$  into  $s$  subsets where partition  $D_K$  contains only instances where  $A_j = c_k^{A_j}$

$$\text{Entropy}_{A_j}(D) = \sum_{K=1}^s \frac{|D_K|}{|D|} \times \text{Entropy}(D_K)$$

- 3) Compute the information gain

$$\text{Gain}(A_j) = \text{Entropy}(D) - \text{Entropy}_{A_j}(D)$$

Information gain can be biased towards attributes with a large number of values.

Gain ratio is a normalized version of information gain.

$$\text{SplitInfo}_{A_j}(D) = -\sum_{k=1}^s \frac{|D_k|}{|D|} \times \log_2\left(\frac{|D_k|}{|D|}\right)$$

$$\text{Gain Ratio}(A_j) = \frac{\text{Gain}(A_j)}{\text{SplitInfo}_{A_j}(D)}$$

### ③ Wrapper Based Methods

Given a particular learning algorithm, e.g. kNN, we can search for the subset of attributes that provide the best performance.

This is called a wrapper based method.

Search can be forwards or backwards through the attributes:

$$\{\emptyset\} \rightarrow \{A_1\} \rightarrow \{A_1, A_2\}$$

$$\{A_1, A_2, A_3\} \rightarrow \{A_1, A_2\}$$

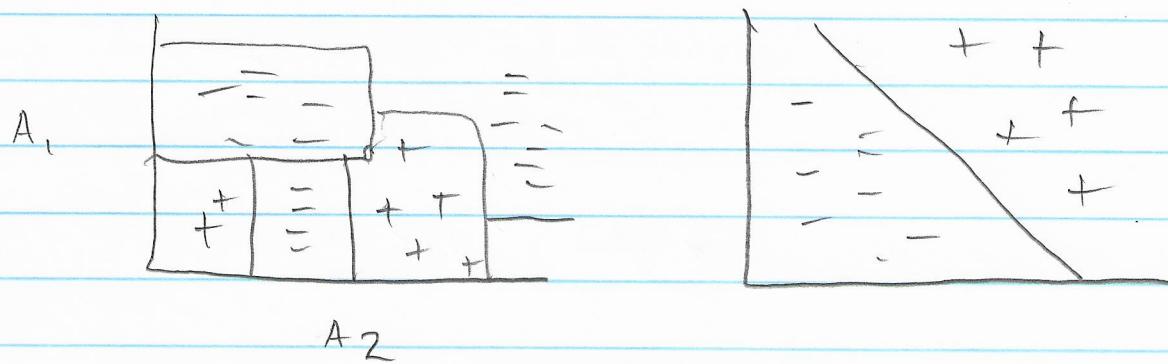
Solved using hill climb or greedy algorithm

Result is the best set of attributes for the specific algorithm.

## General Comments on Feature Selection

- 1) Start with fast filter based methods to come up with a preliminary set of attributes

Correlation and information gain  
find different kinds of relationships



Better suited to  
information gain

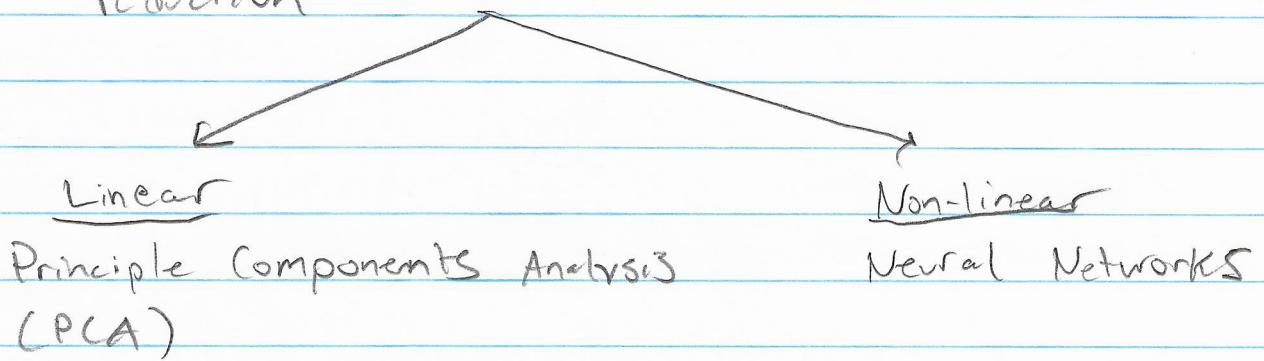
Better suited to  
correlation

- 2) Experiment with a variety of different learning algorithms

- 3) Apply wrapper based feature selection using the best learning algorithm

## ④ Dimensionality Reduction

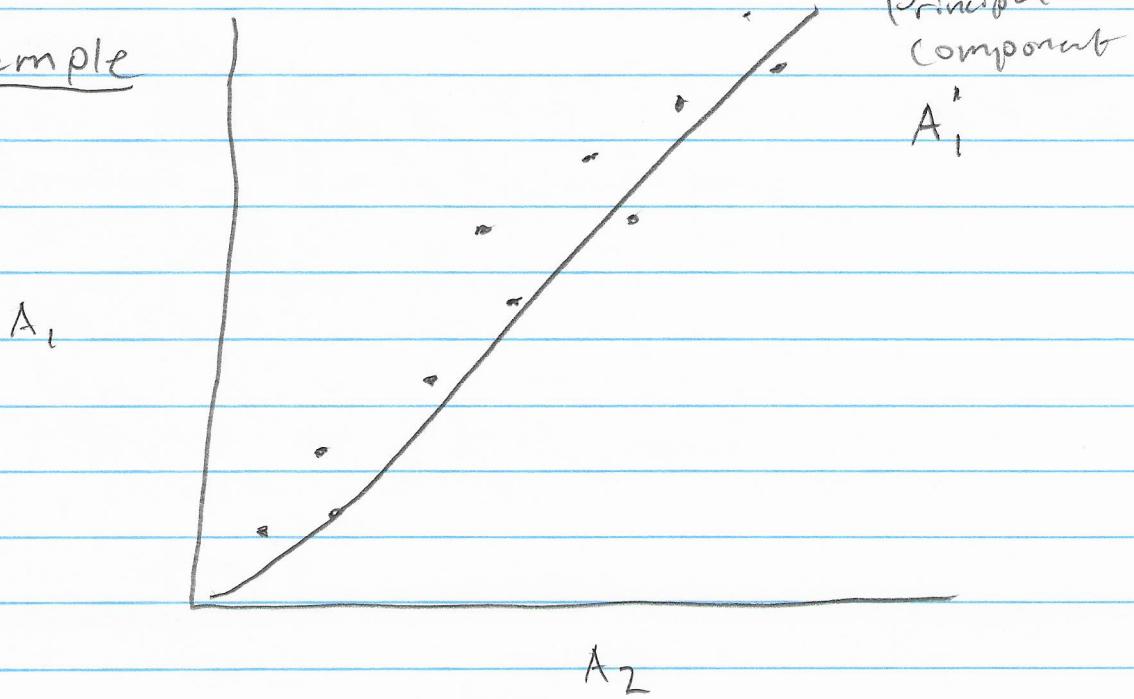
Two general approaches to dimensionality reduction



### Principle Components Analysis

Find the internal "axes" of the data set. Each "axes" becomes a new high-level feature

#### Example



### Assumptions :

- 1) Relationship between the Variables  
is linear
- 2) Mean and Covariance is important
- 3) Large Variances have important dynamics