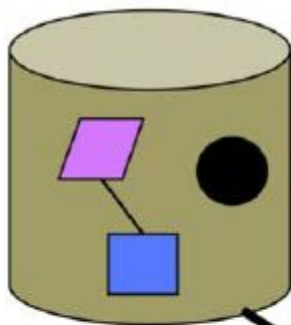


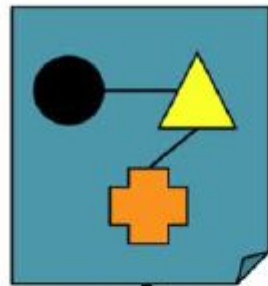
# Data Integration

## Lesson 3b

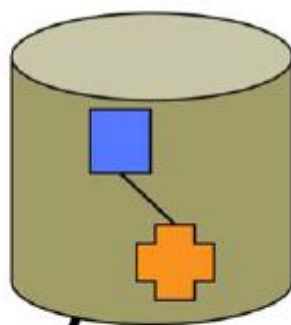
Database A



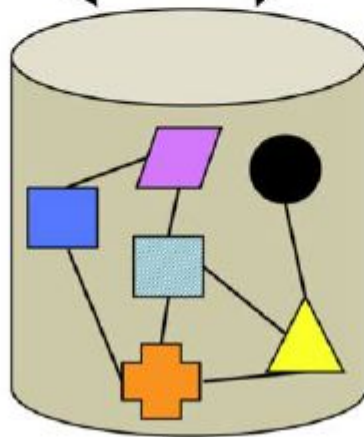
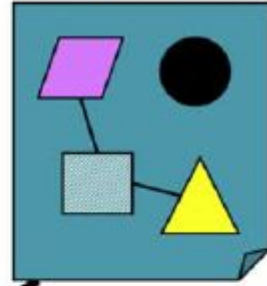
Data File B



Database C



Data File D



Integration

# Handling redundancy in data integration

- Redundant data occur often when integration of multiple databases
  - *Object identification*: The same attribute or object may have different names in different databases
  - *Derivable data*: One attribute may be a “derived” attribute in another table, e.g., annual revenue
- Redundant attributes may be able to be detected by *correlation analysis*
- Careful integration of the data from multiple sources may help reduce/avoid redundancies and inconsistencies and improve mining speed and quality

# Correlation Analysis

**Numerical Data :**

**Pearson's Correlation Coefficient**

**Categorical Data :**

**Chi Square Test**

## Correlation analysis : Numerical data (Pearson's Correlation Coefficient)

$$r = \frac{n(\sum xy) - (\sum x)(\sum y)}{\sqrt{[n\sum x^2 - (\sum x)^2][n\sum y^2 - (\sum y)^2]}}$$

❖  $> 0$  , A and B positively correlated

- ❖ values of A increase as values of B increase

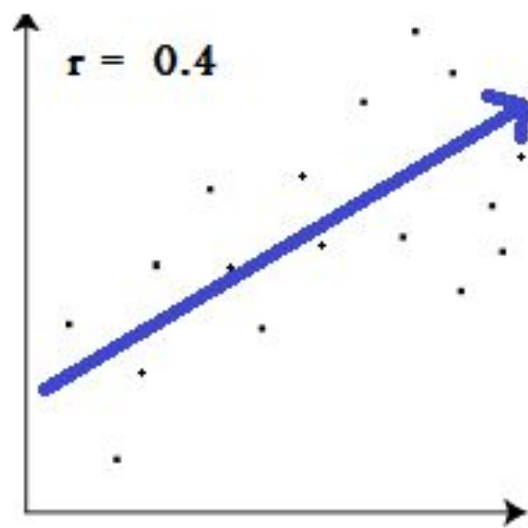
- ❖ The higher the value, the more each attribute implies the other

- ❖ High value indicate that A (or B) may be removed as a redundancy

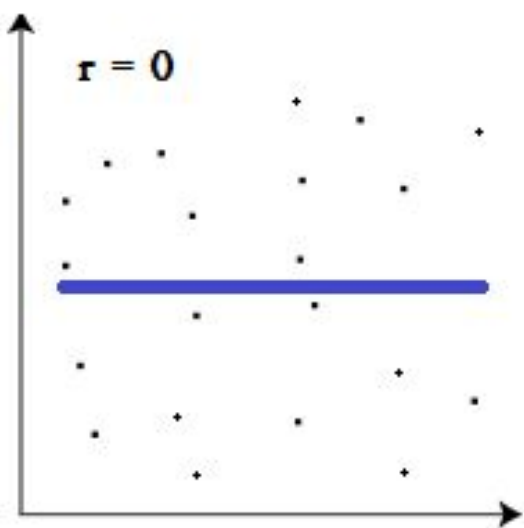
❖  $= 0$ , A and B independent (no correlation)

❖  $< 0$ , A and B negatively correlated

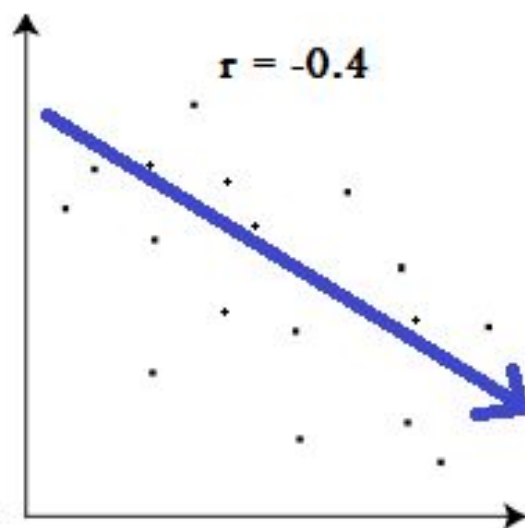
- ❖ Values of one attribute increase as the values of the other attribute decrease (discourages each other)



**Positive Correlation**



**No correlation**



**Negative**

- Find the value of the correlation coefficient from the table:

Subject	Age	Glucose level
1	43	99
2	21	65
3	25	79
4	42	75
5	57	87
6	59	81



# Solution

$$r = \frac{n(\sum xy) - (\sum x)(\sum y)}{\sqrt{[n\sum x^2 - (\sum x)^2][n\sum y^2 - (\sum y)^2]}}$$

SUBJECT	AGE X	GLUCOSE LEVEL Y	XY	X <sup>2</sup>	Y <sup>2</sup>
1	43	99	4257	1849	9801
2	21	65	1365	441	4225
3	25	79	1975	625	6241
4	42	75	3150	1764	5625
5	57	87	4959	3249	7569
6	59	81	4779	3481	6561
Σ	247	486	20485	11409	40022

$$r = \frac{n(\sum xy) - (\sum x)(\sum y)}{\sqrt{[n\sum x^2 - (\sum x)^2][n\sum y^2 - (\sum y)^2]}}$$

From our table:

- $\sum x = 247$
- $\sum y = 486$
- $\sum xy = 20,485$
- $\sum x^2 = 11,409$
- $\sum y^2 = 40,022$
- $n$  is the **sample size**, in our case = 6

The correlation coefficient =

$$\begin{aligned} & \bullet \quad 6(20,485) - (247 \times 486) / \sqrt{[6(11,409) - (247^2)] \times [6(40,022) - 486^2]} \\ & \quad = 0.5298 \end{aligned}$$

The range of the correlation coefficient is from -1 to 1. Our result is 0.5298 or 52.98%, which means the variables have a moderate positive correlation

# Correlation analysis (categorical data)

- $\chi^2$  (chi-square) test

$$\chi^2_{n-1} = \sum_{i=1}^n \frac{(\text{Observed}_i - \text{Expected}_i)^2}{\text{Expected}_i}$$

- $n$  is the number of possible values
- The larger the  $\chi^2$  value, the more likely the variables are related
- The cells that contribute the most to the  $\chi^2$  value are those whose actual count is very different from the expected count
- Correlation does not imply causality
  - # of hospitals and # of car-theft in a city are correlated
  - Both are causally linked to the third variable: population

## Observed Counts

Gender	Drank Alcohol in Last 2 Hours?		Total
	Yes	No	
Male	77	404	481
Female	16	122	138
Total	93	526	619

## Expected Counts

Gender	Drank Alcohol in Last 2 Hours?		Total
	Yes	No	
Male	$(93 \cdot 481) / 619 = 72.3$	$(526 \cdot 481) / 619 = 408.7$	481
Female	$(93 \cdot 138) / 619 = 20.7$	$(526 \cdot 138) / 619 = 117.3$	138
Total	93	526	619

# Example:

	Play chess	Not play chess	Sum (row)
Like science fiction	250	200	450
Not like science fiction	50	1000	1050
Sum (col.)	300	1200	1500

**Probability to play chess:  $P(\text{chess}) = 300/1500 = 0.2$**

**Probability to like science fiction:  $P(\text{SciFi}) = 450/1500 = 0.3$**

**If science fiction and chess playing are independent attributes, then the probability to like SciFi AND play chess is**

$$P(\text{SciFi, chess}) = P(\text{SciFi}) \cdot P(\text{chess}) = 0.06$$

**That means, we expect  $0.06 \cdot 1500 = 90$  such cases (if they are independent)**

	Play chess	Not play chess	Sum (row)
Like science fiction	250 (90)	200 (360)	450
Not like science fiction	50 (210)	1000 (840)	1050
Sum (col.)	300	1200	1500

- $\chi^2$  (chi-square) calculation (numbers in parenthesis are expected counts calculated based on the data distribution in the two categories)

$$\chi^2 = \frac{(250-90)^2}{90} + \frac{(50-210)^2}{210} + \frac{(200-360)^2}{360} + \frac{(1000-840)^2}{840} = 507.93$$

- It shows that like\_science\_fiction and play\_chess are correlated in the group

Next Topic: Association Rule Mining