

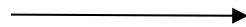
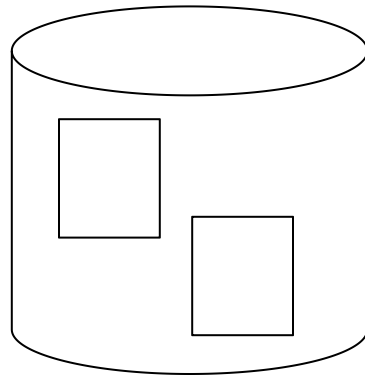
# Data Exploration



AnHai Doan

# Motivation

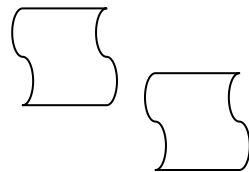
**X**



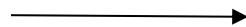
id	name	loc
$x_1$	Apple	CA
$x_2$	IBM	NY

**Y**

news  
articles



data  
extraction



id	cname	address	rev
$y_1$	IBM Corp	CA	25
$y_2$	Apple Inc	CA	51
$y_3$	GE	NY	351

**X**

id	name	loc
$x_1$	Apple	CA
$x_2$	IBM	NY

**Y**

id	cname	address	rev
$y_1$	IBM Corp	CA	25
$y_2$	Apple Inc	CA	51
$y_3$	GE	NY	351

data cleaning: GE revenue: 351  $\longrightarrow$  35.1

schema matching: name = cname  
loc = address

schema merging:  $\left. \begin{array}{l} X(\text{name, loc}) \\ Y(\text{cname, address, rev}) \end{array} \right\} Z(\text{name, loc, rev})$

data matching:

**M**

xid	yid
$x_1$	$y_2$
$x_2$	$y_1$

data merging: for name, return the longer string from X.name and Y.cname  
for loc, return X.loc

schema mapping:  $Z = \text{select merge\_name}(X.\text{name}, Y.\text{cname}), X.\text{loc}, Y.\text{rev}$   
from X, Y, M  
where  $X.\text{id} = M.\text{xid}$  and  $Y.\text{id} = M.\text{yid}$

**X**

<b>id</b>	<b>name</b>	<b>loc</b>
$x_1$	Apple	CA
$x_2$	IBM	NY

**Y**

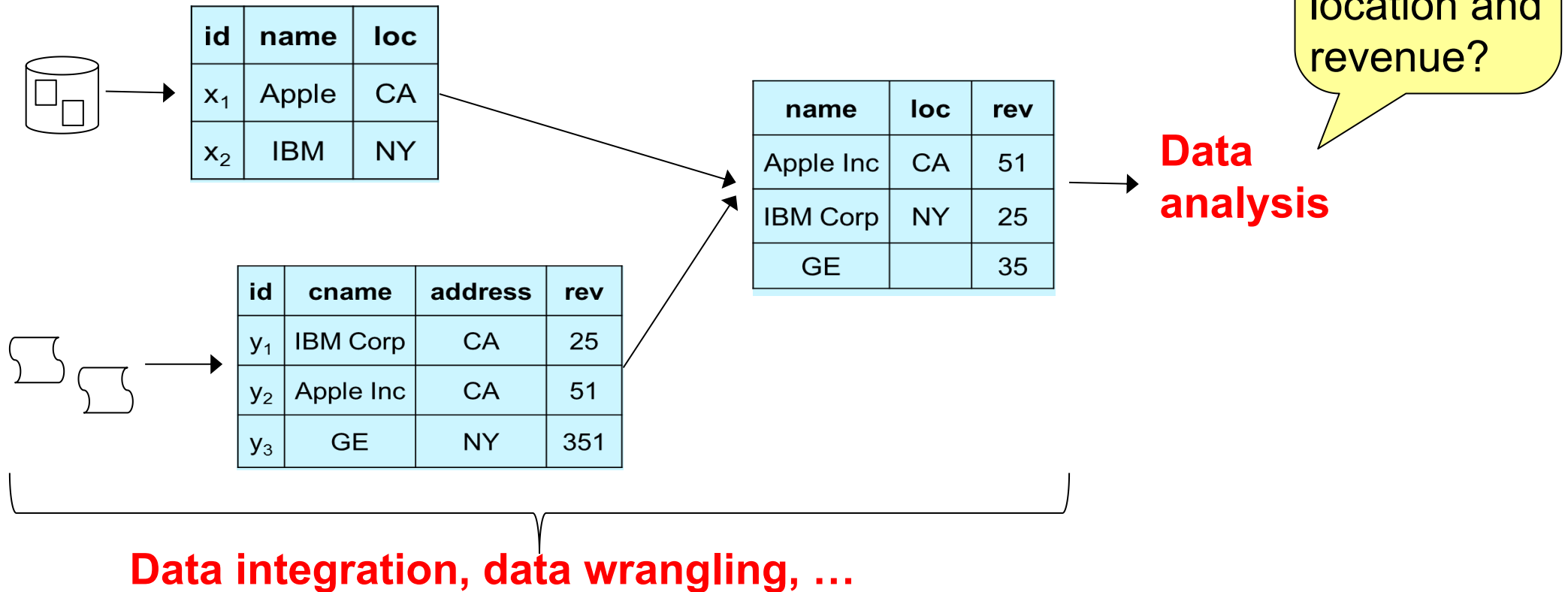
<b>id</b>	<b>cname</b>	<b>addresses</b>	<b>rev</b>
$y_1$	IBM Corp	CA	25
$y_2$	Apple Inc	CA	51
$y_3$	GE	NY	351

**Z**

<b>name</b>	<b>loc</b>	<b>rev</b>
Apple Inc	CA	51
IBM Corp	NY	25

# Another Example

- The raw data to insight pipeline



# Three Goals for Data Exploration

- **Understand the data**
  - basic characteristics, structure
- **Find interesting stuff**
- **Find quality issues**
  
- **These help decide what to do later, influence downstream actions**
  
- **Data exploration is also often known as exploratory data analysis (EDA)**

# Basic Operations

- **Browse and query**
- **Visualize**
- **Profile (mostly with automatic programs)**
  - compute statistics
  - detect more stuff (e.g., meta-data such as keys) about the data
  - find data quality problems
- **Two papers to read**
  - a book chapter called “data exploration”
  - a survey paper: “profiling relational data: a survey”
  - will provide them later

# Browse and Query

- **The simplest operations**
- **Yet difficult to do well today**
  - there is no good tool to browse a large table
  - no good tool to query large tables (e.g., 5G) quickly
  - no tool to allow users to ask NL queries (though they start appearing)
- **Today most folks in industry still use Excel**
  - show an example



# Visualize

- **Why?**

- can display a lot of information at once (dense information)
- can leverage human eyes, which are very good at detecting patterns and anomalies (thousands of years of evolution)

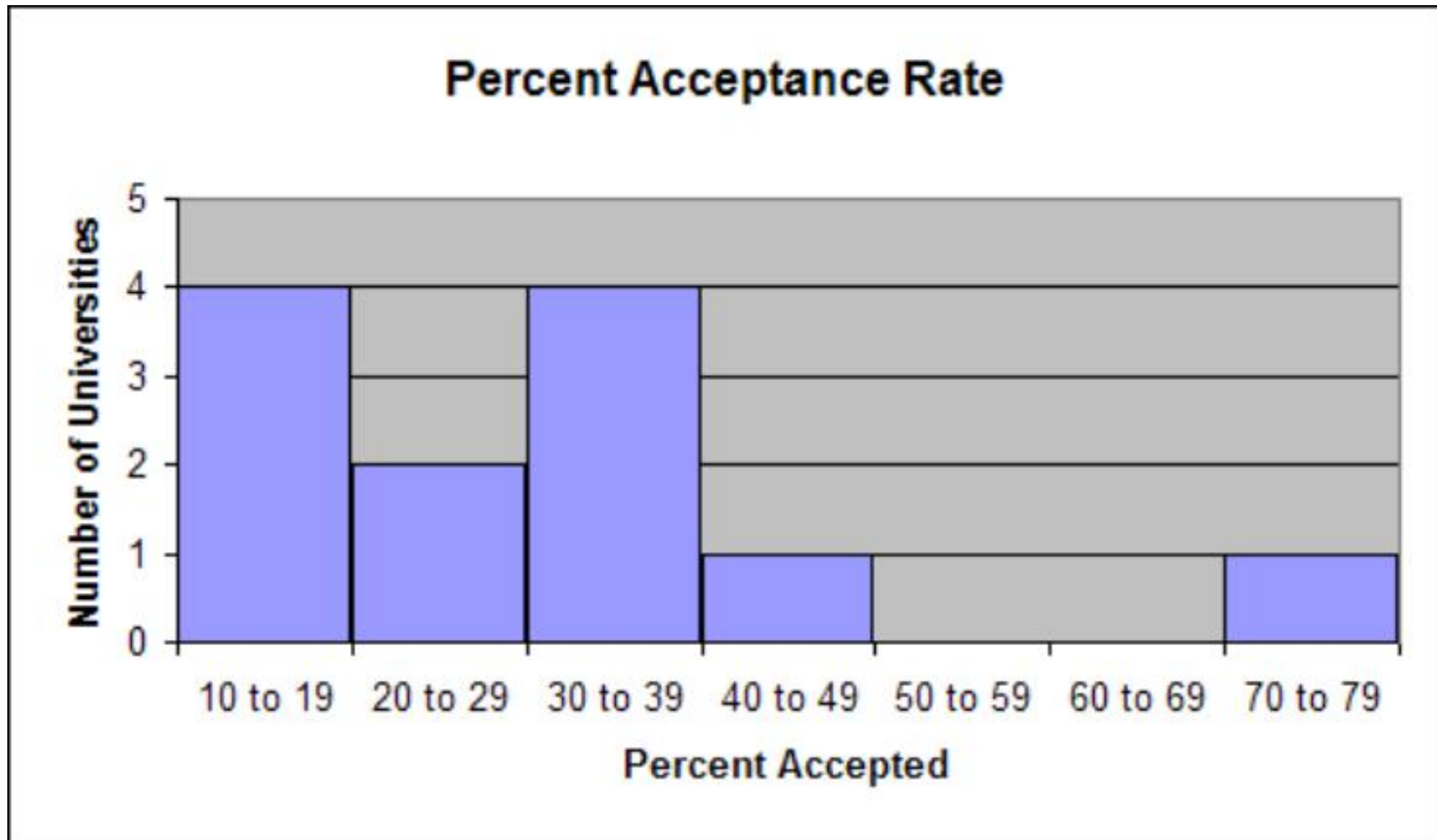
- **Basic types of visualization**

- visualizing a single attribute (aka column)
- visualizing several attributes
- visualizing a large number of attributes

- **Visualizing a single attribute**

- histogram and more (see chapter)

# Histogram Example



# Histogram Example

```
In [60]: profiler.profile_table(B, 'year')
```

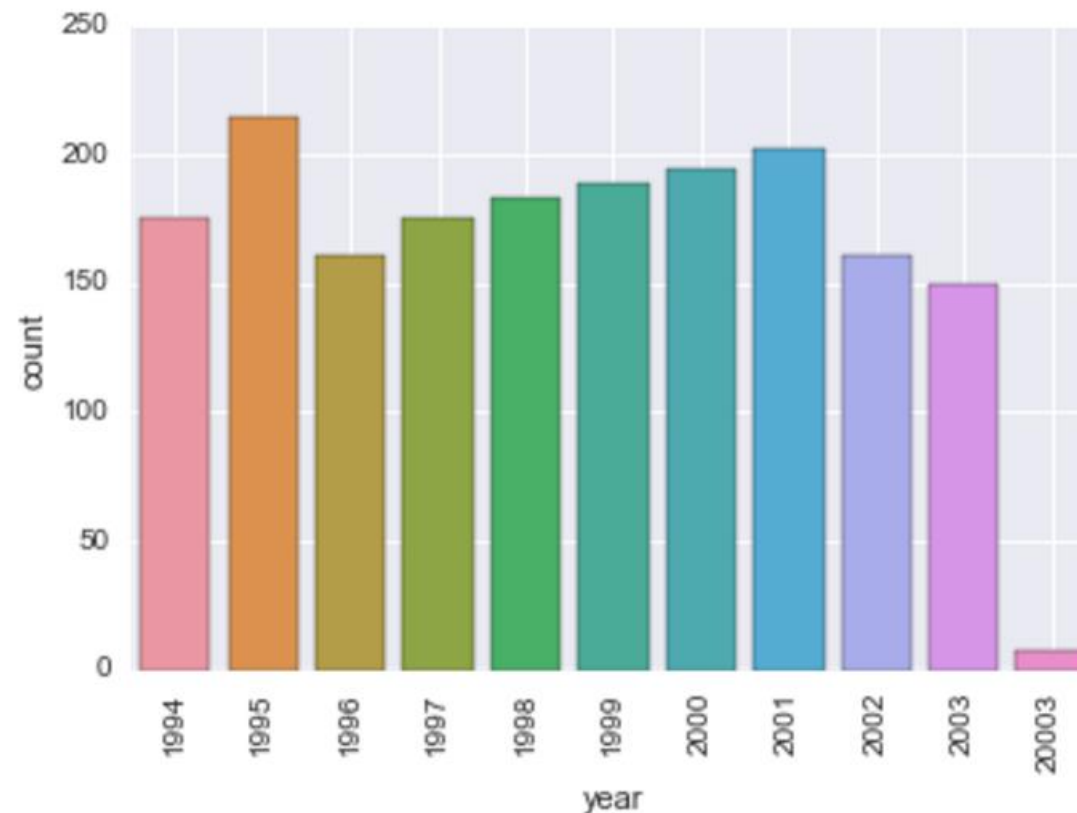
Number of unique values: 11

Number of missing values: 0

Unique values:

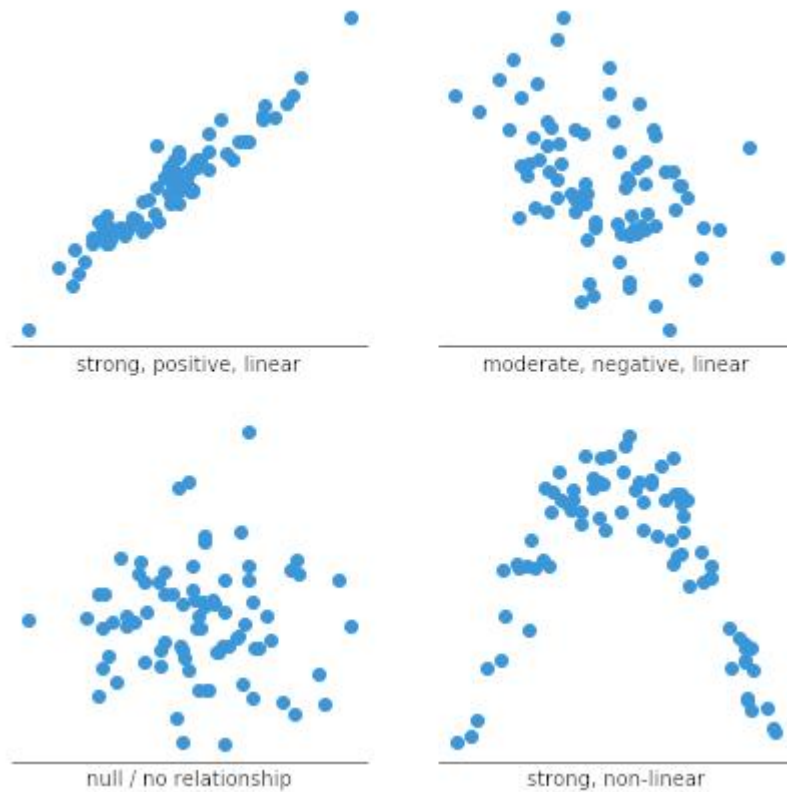
[1994, 1995, 1996, 1997, 1998, 1999, 2000, 2001, 2002, 2003, 20003]

Frequency plot:



# Visualizing Multiple Attributes

- Usually 2-4 attributes at the same time
- Scatterplot is a well-known technique



- More in the book chapter

# Basic Operations

- **Browse and query**
- **Visualize**
- **Profile (mostly with automatic programs)**
  - compute statistics
  - detect more stuff (e.g., meta-data such as keys) about the data
  - find data quality problems

# Types of Profiling

Inbox (3,360) x hpi dataprofil... x +

hpi.de/fileadmin/us... G +

WSJ E my-hp my-infa-zoom » Other bookmarks

4 / 27 - +

Ziawasch Abedjan et al.

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**Data profiling**

- Single column
  - Cardinalities
  - Patterns & data types
  - Value distributions
  - Domain Classification
- Multiple columns
  - Correlations & association rules
  - Clusters & outliers
  - Summaries & sketches
- Dependencies
  - Uniqueness
    - Key discovery
    - Conditional
    - (Approximate)
  - Inclusion dependencies
    - Foreign key discovery
    - Conditional
    - (Approximate)
  - Functional dependencies
    - Conditional
    - (Approximate)

**Fig. 1** A classification of traditional data profiling tasks.

dataprofil...main.pdf ^ Show all x

# More Types of Profiling

- **Can target specifically at detecting certain data quality problems**
- **See the google doc “data-profiling-cleaning-examples”**

# More Advanced Stuff

- **30% of values in Salary column are missing, can you explain why?**
  - 95% are missing for Temp workers in the state CA
  - suggest a systematic error here



# Summary of Data Exploration

- **Basic operations**

- Browse and query
- Visualize
- Profile (mostly with automatic programs)
  - compute statistics
  - detect more stuff (e.g., meta-data such as keys) about the data
  - find data quality problems

- **Guidance on how to use them**

- **Systems that incorporate the above tools**