

# **Data Mining:**

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## **Concepts and Techniques**

### **— Chapter 2 —**

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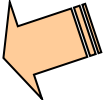
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# Chapter 2: Getting to Know Your Data

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- Data Objects and Attribute Types 
- Basic Statistical Descriptions of Data
- Data Visualization
- Measuring Data Similarity and Dissimilarity
- Summary

# Types of Data Sets

- Record

- Relational records
- Data matrix, e.g., numerical matrix, crosstabs
- Document data: text documents: term-frequency vector
- Transaction data

	team	coach	play	ball	score	game	win	lost	timeout	season
Document 1	3	0	5	0	2	6	0	2	0	2
Document 2	0	7	0	2	1	0	0	3	0	0
Document 3	0	1	0	0	1	2	2	0	3	0

- Graph and network

- World Wide Web
- Social or information networks
- Molecular Structures

- Ordered

- Video data: sequence of images
- Temporal data: time-series
- Sequential Data: transaction sequences
- Genetic sequence data

- Spatial, image and multimedia:

- Spatial data: maps
- Image data:
- Video data:

<i><b>TID</b></i>	<i><b>Items</b></i>
<b>1</b>	<b>Bread, Coke, Milk</b>
<b>2</b>	<b>Beer, Bread</b>
<b>3</b>	<b>Beer, Coke, Diaper, Milk</b>
<b>4</b>	<b>Beer, Bread, Diaper, Milk</b>
<b>5</b>	<b>Coke, Diaper, Milk</b>

# Important Characteristics of Structured Data

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- Dimensionality
  - Curse of dimensionality
- Sparsity
  - Only presence counts
- Resolution
  - Patterns depend on the scale
- Distribution
  - Centrality and dispersion

# Data Objects

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- Data sets are made up of data objects.
- A **data object** represents an entity.
- Examples:
  - sales database: customers, store items, sales
  - medical database: patients, treatments
  - university database: students, professors, courses
- Also called *samples*, *examples*, *instances*, *data points*, *objects*, *tuples*.
- Data objects are described by **attributes**.
- Database rows -> data objects; columns -> attributes.

# Attributes

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- **Attribute (or dimensions, features, variables):**  
a data field, representing a characteristic or feature of a data object.
  - *E.g., customer\_ID, name, address*
- Types:
  - Nominal
  - Binary
  - Numeric: quantitative
    - Interval-scaled
    - Ratio-scaled

# Attribute Types

- **Nominal:** categories, states, or “names of things”
  - *Hair\_color* = {*auburn, black, blond, brown, grey, red, white*}
  - marital status, occupation, ID numbers, zip codes
- **Binary**
  - Nominal attribute with only 2 states (0 and 1)
  - Symmetric binary: both outcomes equally important
    - e.g., gender
  - Asymmetric binary: outcomes not equally important.
    - e.g., medical test (positive vs. negative)
    - Convention: assign 1 to most important outcome (e.g., HIV positive)
- **Ordinal**
  - Values have a meaningful order (ranking) but magnitude between successive values is not known.
  - *Size* = {*small, medium, large*}, grades, army rankings

# Numeric Attribute Types

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- Quantity (integer or real-valued)
- **Interval**
  - Measured on a scale of **equal-sized units**
  - Values have order
    - E.g., *temperature in  $C^{\circ}$  or  $F^{\circ}$ , calendar dates*
  - No true zero-point
- **Ratio**
  - Inherent **zero-point**
  - We can speak of values as being an order of magnitude larger than the unit of measurement ( $10\text{ K}^{\circ}$  is twice as high as  $5\text{ K}^{\circ}$ ).
    - e.g., *temperature in Kelvin, length, counts, monetary quantities*



# Discrete vs. Continuous Attributes

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## ■ Discrete Attribute

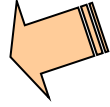
- Has only a finite or countably infinite set of values
  - E.g., zip codes, profession, or the set of words in a collection of documents
- Sometimes, represented as integer variables
- Note: Binary attributes are a special case of discrete attributes

## ■ Continuous Attribute

- Has real numbers as attribute values
  - E.g., temperature, height, or weight
- Practically, real values can only be measured and represented using a finite number of digits
- Continuous attributes are typically represented as floating-point variables

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# Basic Statistical Descriptions of Data

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- Motivation
  - To better understand the data: central tendency, variation and spread
- Data dispersion characteristics
  - median, max, min, quantiles, outliers, variance, etc.
- Numerical dimensions correspond to sorted intervals
  - Data dispersion: analyzed with multiple granularities of precision
  - Boxplot or quantile analysis on sorted intervals
- Dispersion analysis on computed measures
  - Folding measures into numerical dimensions
  - Boxplot or quantile analysis on the transformed cube

# Measuring the Central Tendency

- Mean (algebraic measure) (sample vs. population):

Note:  $n$  is sample size and  $N$  is population size.

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i \quad \mu = \frac{\sum x}{N}$$

- Weighted arithmetic mean:
- Trimmed mean: chopping extreme values

$$\bar{x} = \frac{\sum_{i=1}^n w_i x_i}{\sum_{i=1}^n w_i}$$

- Median:

- Middle value if odd number of values, or average of the middle two values otherwise
- Estimated by interpolation (for *grouped data*):

$$median = L_1 + \left( \frac{n/2 - (\sum freq)l}{freq_{median}} \right) width$$

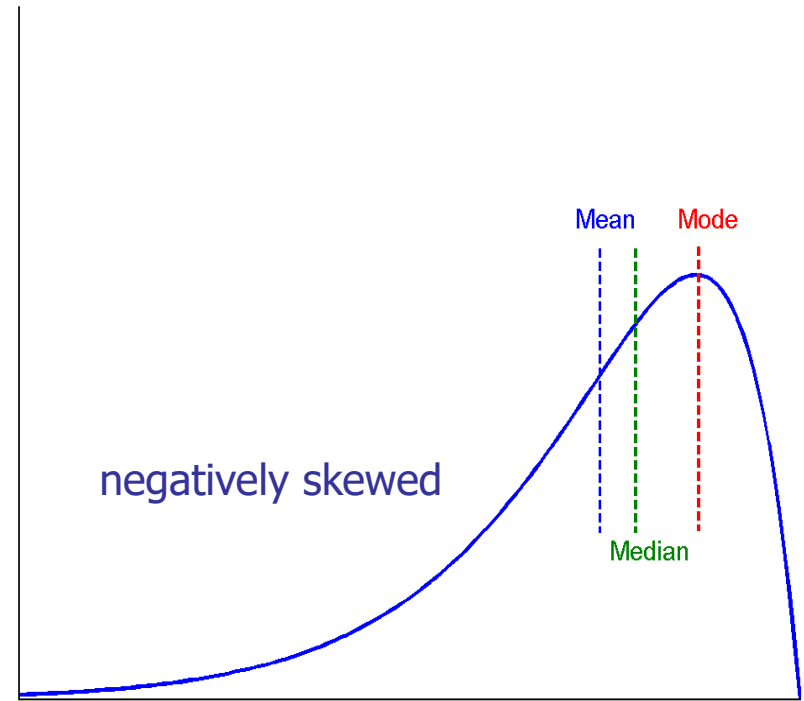
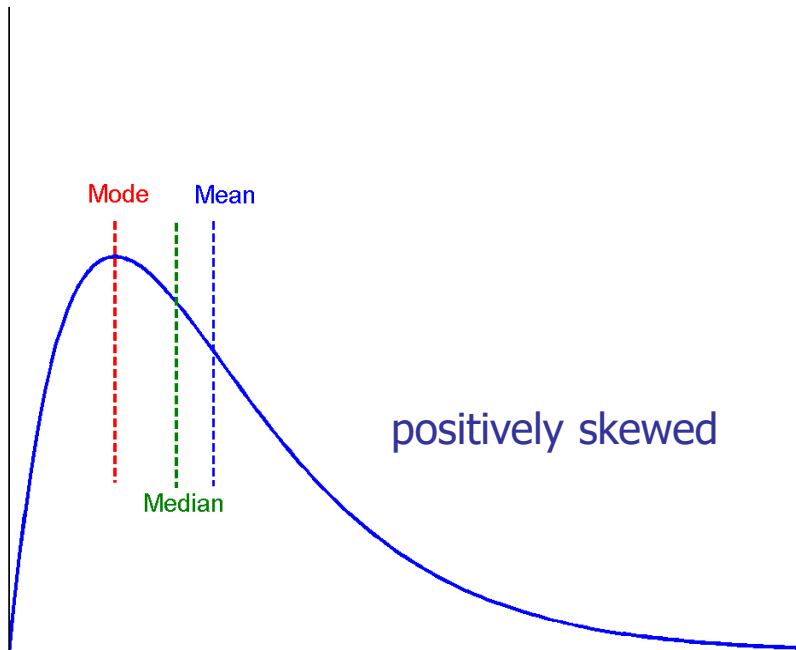
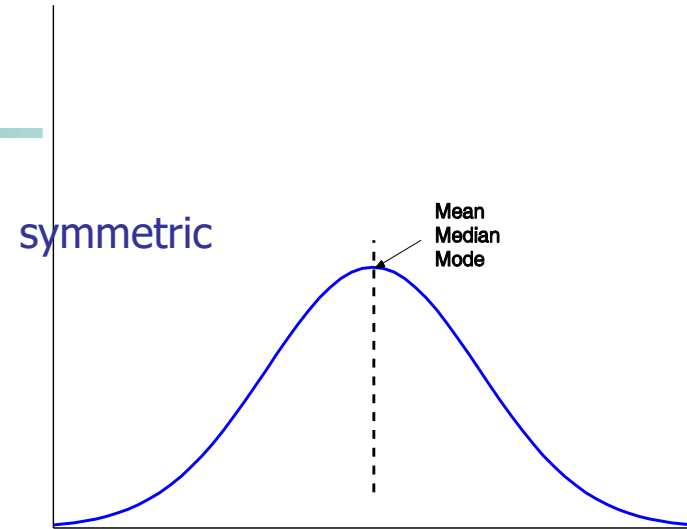
- Mode

- Value that occurs most frequently in the data
- Unimodal, bimodal, trimodal
- Empirical formula:  $mean - mode = 3 \times (mean - median)$

age	frequency
1-5	200
6-15	450
16-20	300
21-50	1500
51-80	700
81-110	44

# Symmetric vs. Skewed Data

- Median, mean and mode of symmetric, positively and negatively skewed data



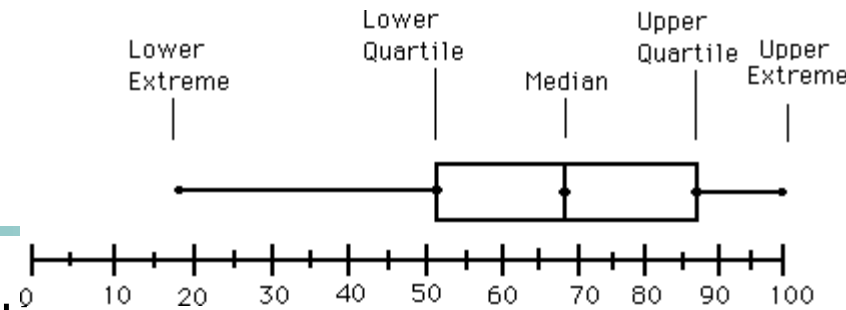
# Measuring the Dispersion of Data

- Quartiles, outliers and boxplots
  - **Quartiles:**  $Q_1$  (25<sup>th</sup> percentile),  $Q_3$  (75<sup>th</sup> percentile)
  - **Inter-quartile range:**  $IQR = Q_3 - Q_1$
  - **Five number summary:** min,  $Q_1$ , median,  $Q_3$ , max
  - **Boxplot:** ends of the box are the quartiles; median is marked; add whiskers, and plot outliers individually
  - **Outlier:** usually, a value higher/lower than  $1.5 \times IQR$
- Variance and standard deviation (*sample:  $s$ , population:  $\sigma$* )
  - **Variance:** (algebraic, scalable computation)

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2 = \frac{1}{n-1} \left[ \sum_{i=1}^n x_i^2 - \frac{1}{n} \left( \sum_{i=1}^n x_i \right)^2 \right] \quad \sigma^2 = \frac{1}{N} \sum_{i=1}^n (x_i - \mu)^2 = \frac{1}{N} \sum_{i=1}^n x_i^2 - \mu^2$$

- **Standard deviation**  $s$  (*or*  $\sigma$ ) is the square root of variance  $s^2$  (*or*  $\sigma^2$ )

# Boxplot Analysis

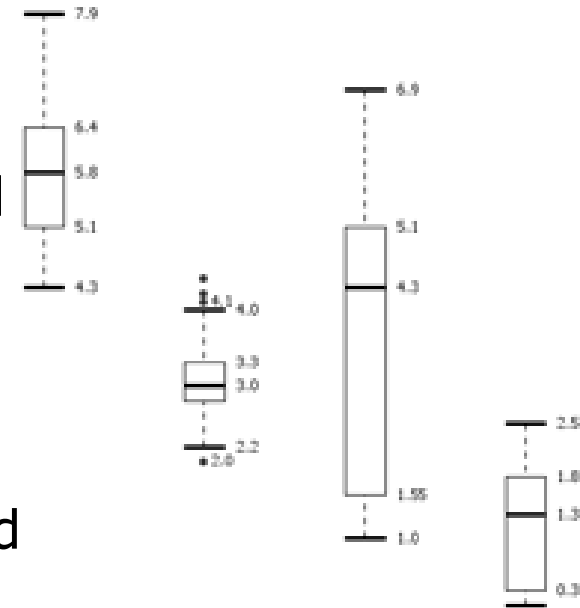


- **Five-number summary** of a distribution

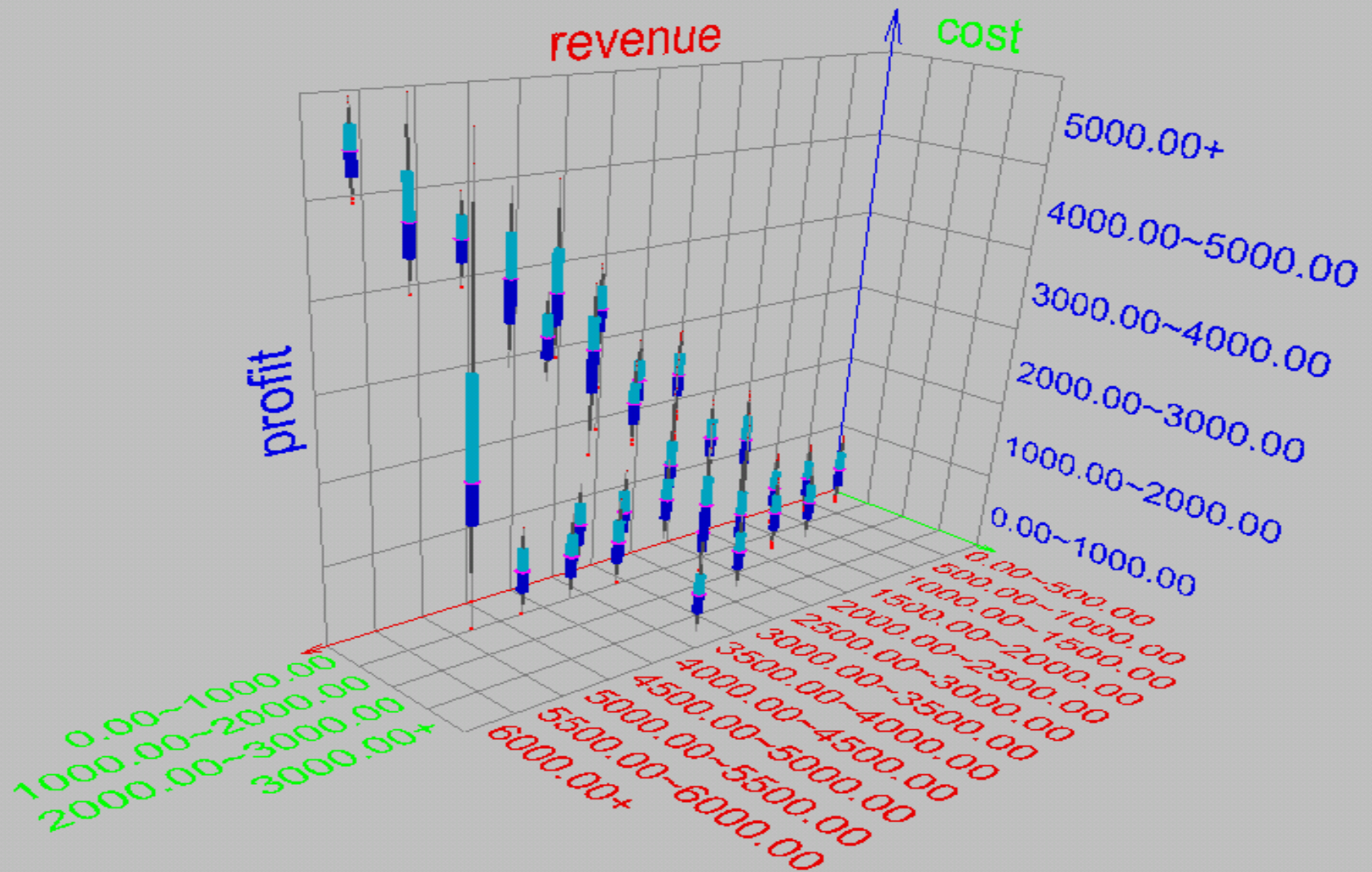
- Minimum, Q1, Median, Q3, Maximum

- **Boxplot**

- Data is represented with a box
- The ends of the box are at the first and third quartiles, i.e., the height of the box is IQR
- The median is marked by a line within the box
- Whiskers: two lines outside the box extended to Minimum and Maximum
- Outliers: points beyond a specified outlier threshold, plotted individually



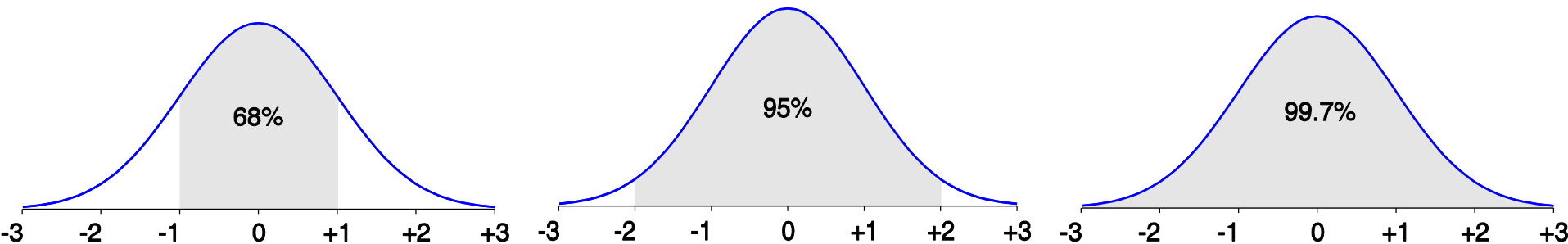
# Visualization of Data Dispersion: 3-D Boxplots





# Properties of Normal Distribution Curve

- The normal (distribution) curve
  - From  $\mu - \sigma$  to  $\mu + \sigma$ : contains about 68% of the measurements ( $\mu$ : mean,  $\sigma$ : standard deviation)
  - From  $\mu - 2\sigma$  to  $\mu + 2\sigma$ : contains about 95% of it
  - From  $\mu - 3\sigma$  to  $\mu + 3\sigma$ : contains about 99.7% of it



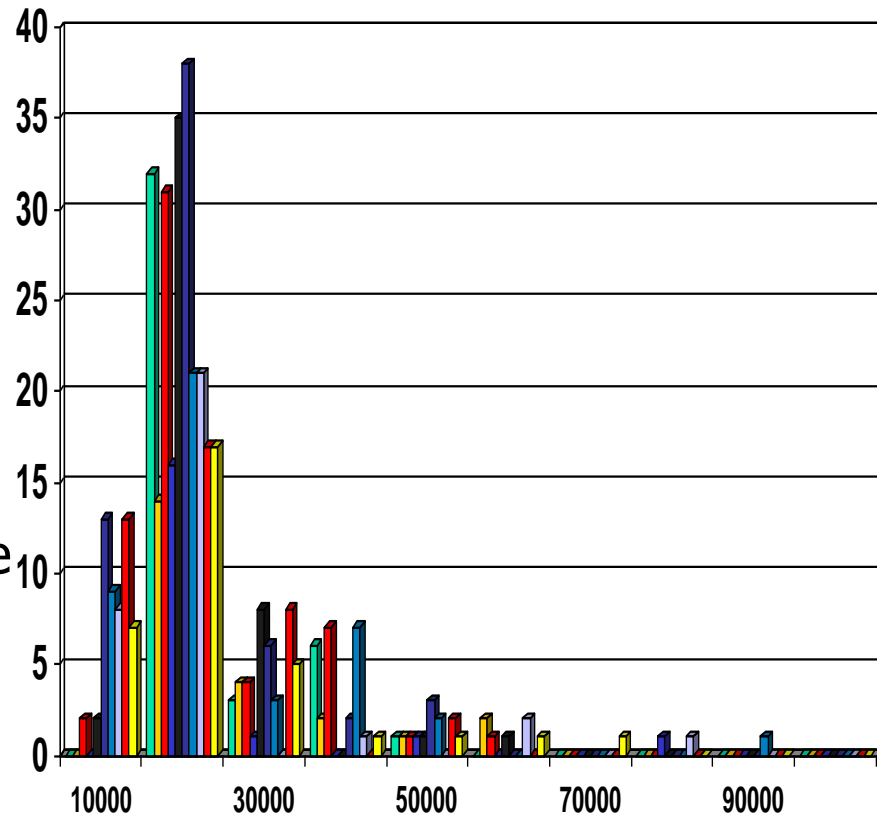
# Graphic Displays of Basic Statistical Descriptions

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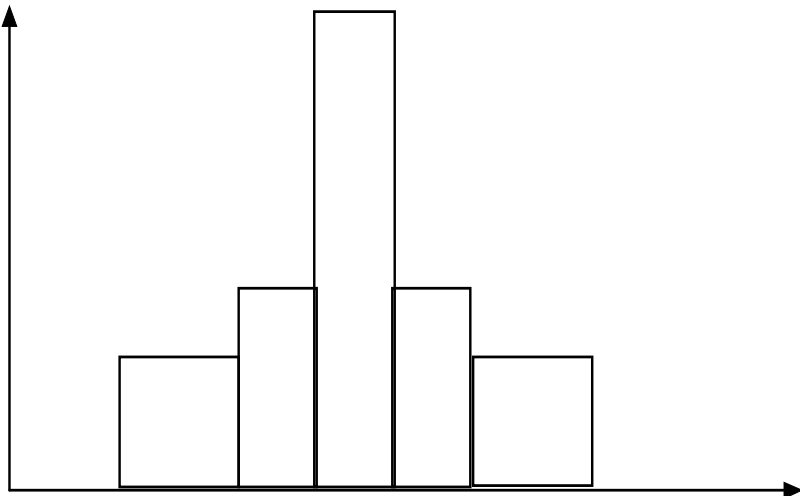
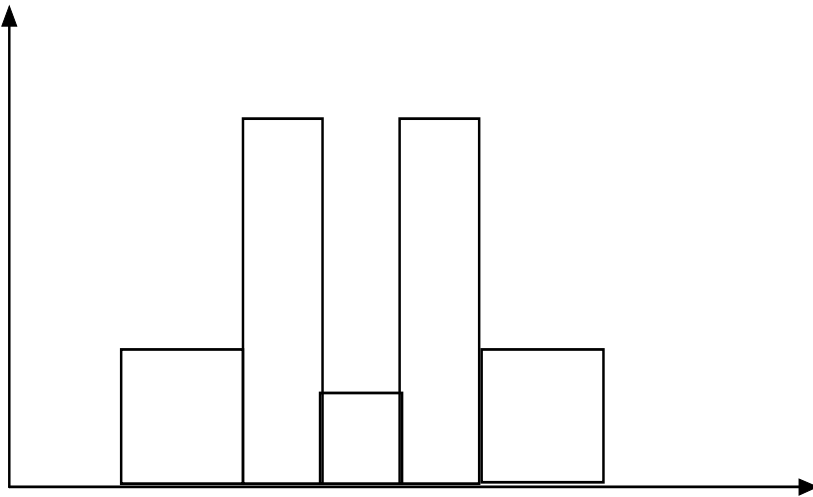
- **Boxplot:** graphic display of five-number summary
- **Histogram:** x-axis are values, y-axis repres. frequencies
- **Quantile plot:** each value  $x_i$  is paired with  $f_i$  indicating that approximately 100  $f_i$ % of data are  $\leq x_i$
- **Quantile-quantile (q-q) plot:** graphs the quantiles of one univariant distribution against the corresponding quantiles of another
- **Scatter plot:** each pair of values is a pair of coordinates and plotted as points in the plane

# Histogram Analysis

- Histogram: Graph display of tabulated frequencies, shown as bars
- It shows what proportion of cases fall into each of several categories
- Differs from a bar chart in that it is the *area* of the bar that denotes the value, not the height as in bar charts, a crucial distinction when the categories are not of uniform width
- The categories are usually specified as non-overlapping intervals of some variable. The categories (bars) must be adjacent



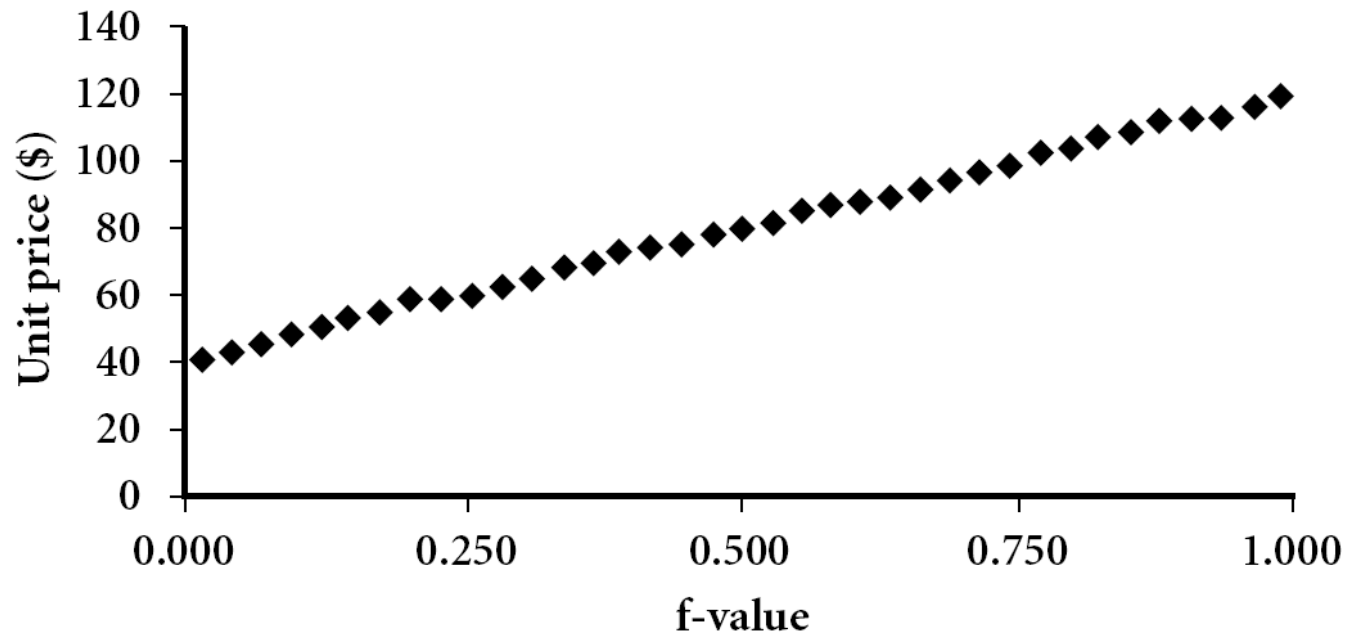
# Histograms Often Tell More than Boxplots



- The two histograms shown in the left may have the same boxplot representation
  - The same values for: min, Q1, median, Q3, max
- But they have rather different data distributions

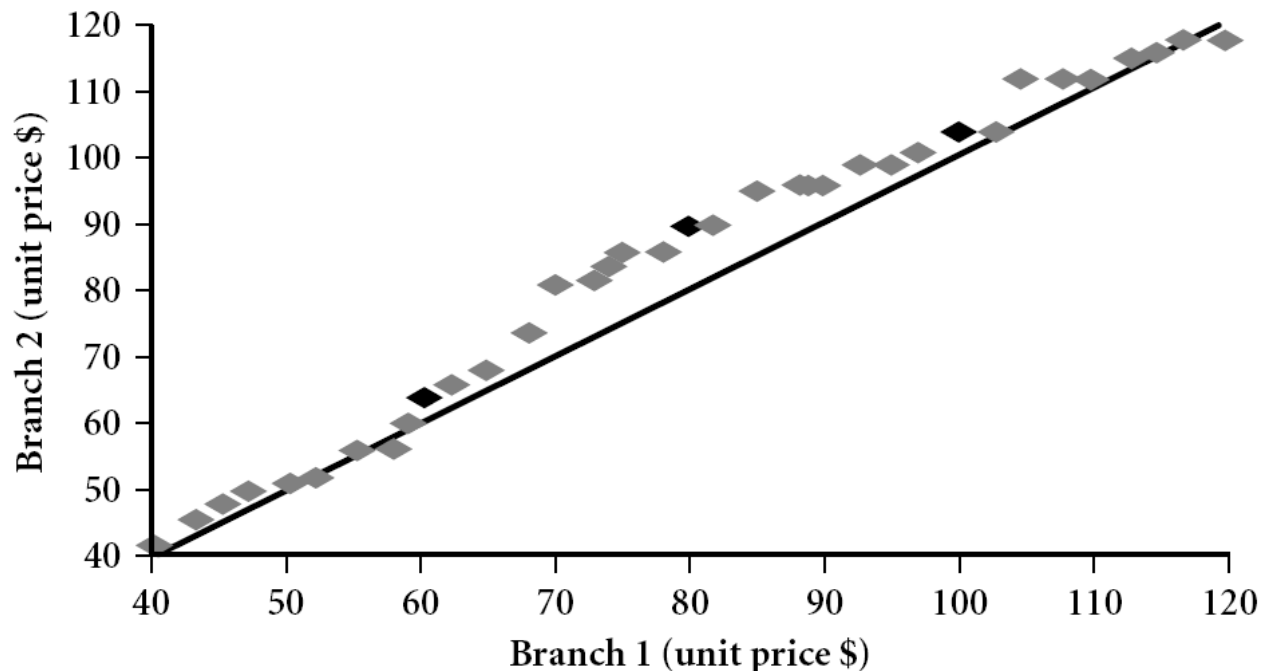
# Quantile Plot

- Displays all of the data (allowing the user to assess both the overall behavior and unusual occurrences)
- Plots **quantile** information
  - For a data  $x_i$  data sorted in increasing order,  $f_i$  indicates that approximately 100  $f_i$ % of the data are below or equal to the value  $x_i$



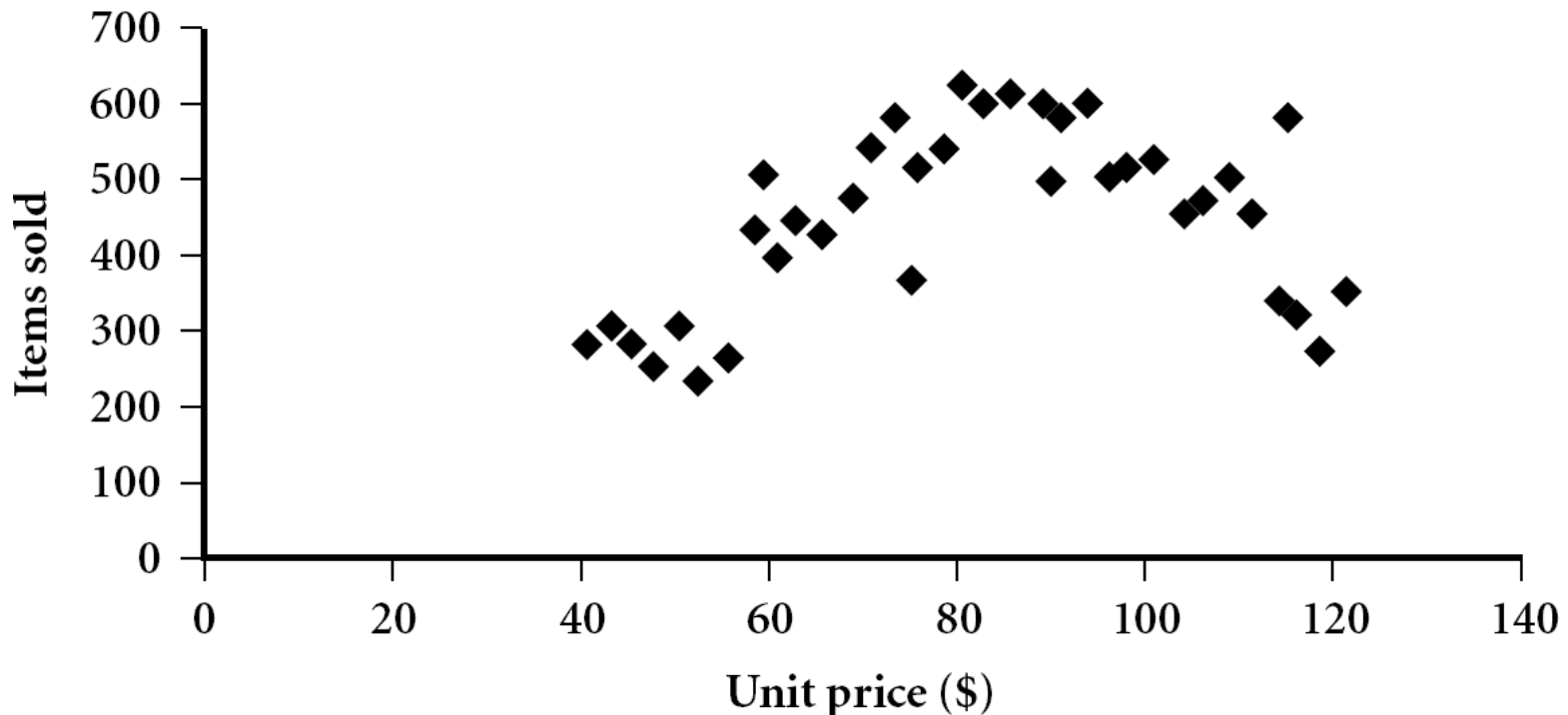
# Quantile-Quantile (Q-Q) Plot

- Graphs the quantiles of one univariate distribution against the corresponding quantiles of another
- View: Is there is a shift in going from one distribution to another?
- Example shows unit price of items sold at Branch 1 vs. Branch 2 for each quantile. Unit prices of items sold at Branch 1 tend to be lower than those at Branch 2.

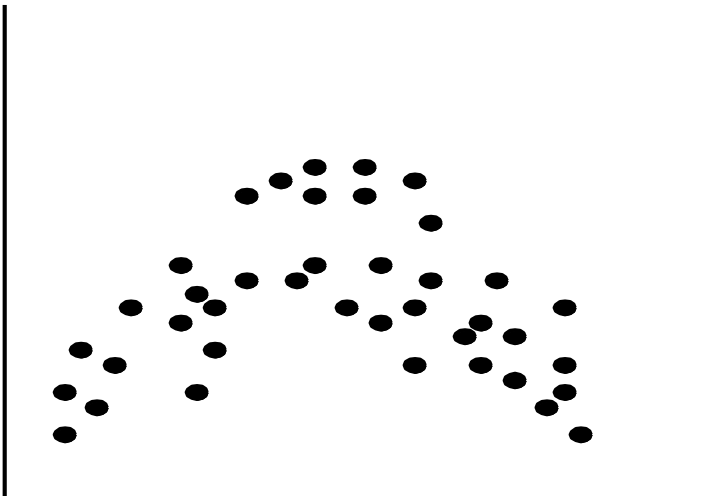
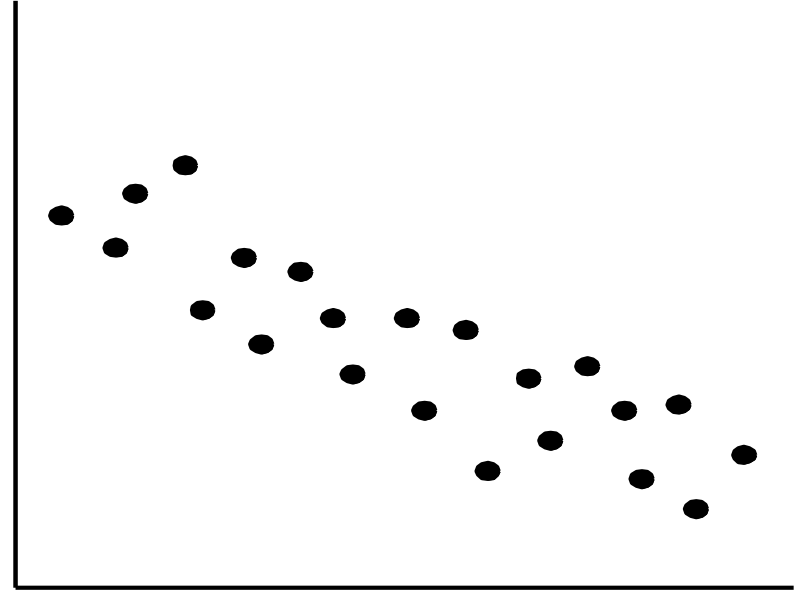
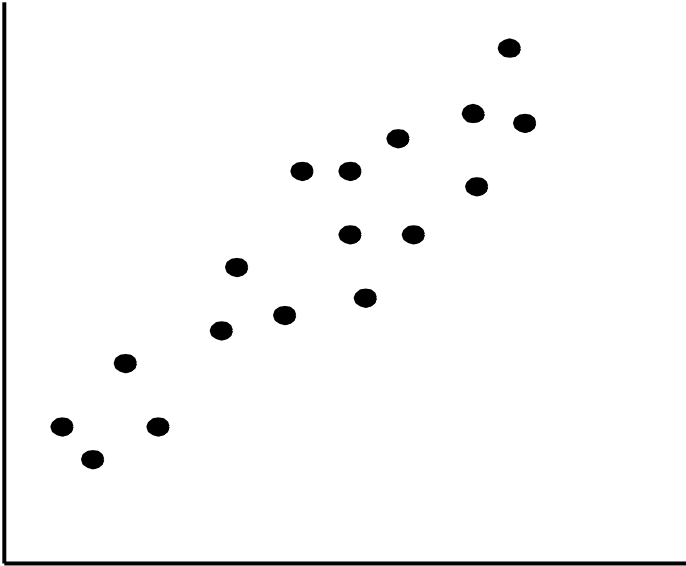


# Scatter plot

- Provides a first look at bivariate data to see clusters of points, outliers, etc
- Each pair of values is treated as a pair of coordinates and plotted as points in the plane



# Positively and Negatively Correlated Data



- The left half fragment is positively correlated
- The right half is negative correlated



# Uncorrelated Data

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