

# EXPLORING DATA



# Data Exploration

- A preliminary exploration of the data to better understand its characteristics.
- Helping to select the right tool for preprocessing or analysis
- In our discussion of data exploration, we focus on
  - ▣ Summary statistics
  - ▣ Visualization

# Iris Sample Data Set

Many of the exploratory data techniques are illustrated with the Iris Plant data set.

Can be obtained from the UCI Machine Learning Repository

<http://www.ics.uci.edu/~mlearn/MLRepository.html>

- ❖ Three flower types (classes):
  - ❖ Setosa
  - ❖ Virginica
  - ❖ Versicolour
- ❖ Four (non-class) attributes
  - ❖ Sepal width and length
  - ❖ Petal width and length



# Summary Statistics

**Summary statistics** are quantities, such as the mean and standard deviation, that capture various characteristics of a potentially large set of values with a single number or a small set of numbers.

Summarized properties include frequency, location and spread

Examples:      location - mean  
                 spread - standard deviation

# Frequencies and the Mode

The frequency of an attribute value is the percentage of time the value occurs in the data set

- ❖ For example, given the attribute 'gender' and a representative population of people, the gender 'female' occurs about 50% of the time.

The mode of an attribute is the most frequent attribute value

- ✓ The notions of frequency and mode are typically used with nominal data

# Percentiles

For continuous data, the notion of a percentile is more useful.

Given an ordinal or continuous attribute  $x$  and a number  $p$  between 0 and 100, the  $p$ th percentile is a value  $x_p$  of  $x$  such that  $p\%$  of the observed values of  $x$  are less than  $x_p$ .

For instance, the 50th percentile is the value  $x_{50\%}$  such that 50% of all values of  $x$  are less than  $x_{50\%}$ .

$$\min(x) = x_{0\%} \text{ and } \max(x) = x_{100\%}$$

# Example

Percentile	Sepal Length	Sepal Width	Petal Length	Petal Width
0	4.3	2.0	1.0	0.1
10	4.8	2.5	1.4	0.2
20	5.0	2.7	1.5	0.2
30	5.2	2.8	1.7	0.4
40	5.6	3.0	3.9	1.2
50	5.8	3.0	4.4	1.3
60	6.1	3.1	4.6	1.5
70	6.3	3.2	5.0	1.8
80	6.6	3.4	5.4	1.9
90	6.9	3.6	5.8	2.2
100	7.9	4.4	6.9	2.5

# Measures of Location: Mean and Median

## Mean

**mean** is the most common measure of the location of a set of points.

$$\{x_1, \dots, x_m\}$$



attribute values of  $x$  for  
 $m$  objects.

$$\text{mean}(x) = \bar{x} = \frac{1}{m} \sum_{i=1}^m x_i$$

mean is very sensitive to outliers

## Median

$$\{x_{(1)}, \dots, x_{(m)}\}$$

$$\text{median}(x) = \begin{cases} x_{(r+1)} & \text{if } m \text{ is odd, i.e., } m = 2r + 1 \\ \frac{1}{2}(x_{(r)} + x_{(r+1)}) & \text{if } m \text{ is even, i.e., } m = 2r \end{cases}$$



# trimmed mean

## trimmed mean

- ✓ A percentage  $p$  between 0 and 100 is specified
- ✓ top and bottom  $(p/2)\%$  of the data is thrown out
- ✓ mean is then calculated in the normal way.

Measure	Sepal Length	Sepal Width	Petal Length	Petal Width
mean	5.84	3.05	3.76	1.20
median	5.80	3.00	4.35	1.30
trimmed mean (20%)	5.79	3.02	3.72	1.12

# Measures of Spread: Range and Variance

- ✓ **Range** is the difference between the max and min
- ✓ **Variance** or standard deviation is the most common measure of the spread of a set of points.

$$\text{variance}(x) = s_x^2 = \frac{1}{m-1} \sum_{i=1}^m (x_i - \bar{x})^2$$

sensitive to outliers

**absolute average deviation (AAD)**

$$\text{AAD}(x) = \frac{1}{m} \sum_{i=1}^m |x_i - \bar{x}|$$

**median absolute deviation (MAD)**

$$\text{MAD}(x) = \text{median}\left(\{|x_1 - \bar{x}|, \dots, |x_m - \bar{x}|\}\right)$$

$$\text{interquartile range}(x) = x_{75\%} - x_{25\%}$$

# Multivariate Summary Statistics

if  $x_i$  and  $x_j$  are the  $i$ th and  $j$ th attributes, then

$$s_{ij} = \text{covariance}(x_i, x_j)$$

$$\text{covariance}(x_i, x_j) = \frac{1}{m-1} \sum_{k=1}^m (x_{ki} - \bar{x}_i)(x_{kj} - \bar{x}_j)$$

$$r_{ij} = \text{correlation}(x_i, x_j) = \frac{\text{covariance}(x_i, x_j)}{s_i s_j}$$

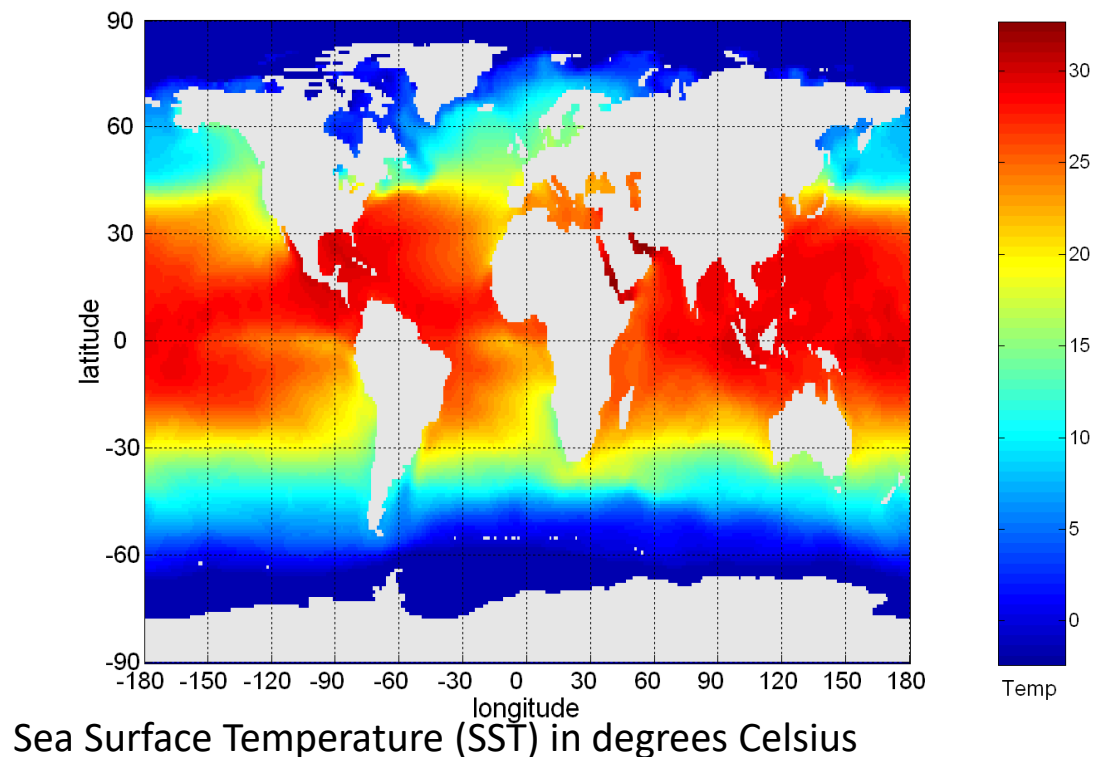
**correlation matrix R**



# Visualization

# Visualization

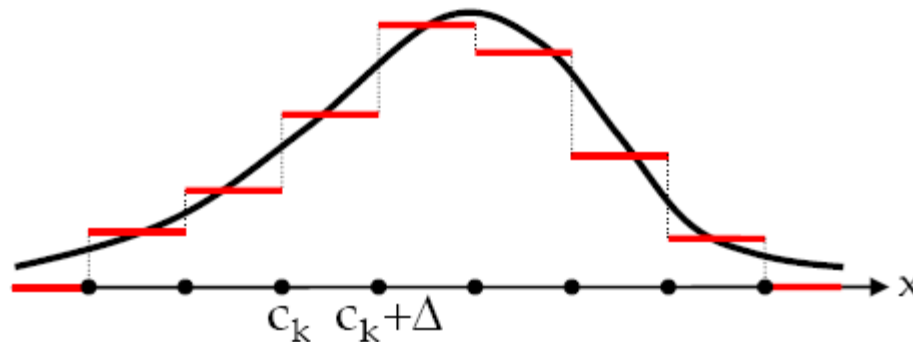
- ✓ Visualization is the conversion of data into a visual format so that the characteristics of the data and the relationships among data items or attributes can be analyzed or reported
- ✓ people can quickly absorb large amounts of visual information



# Visualization Techniques: Histograms

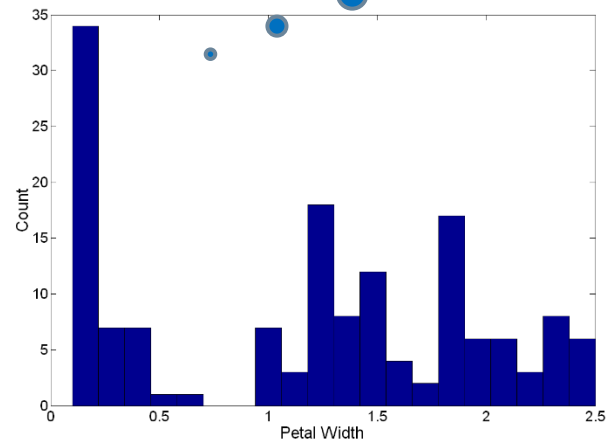
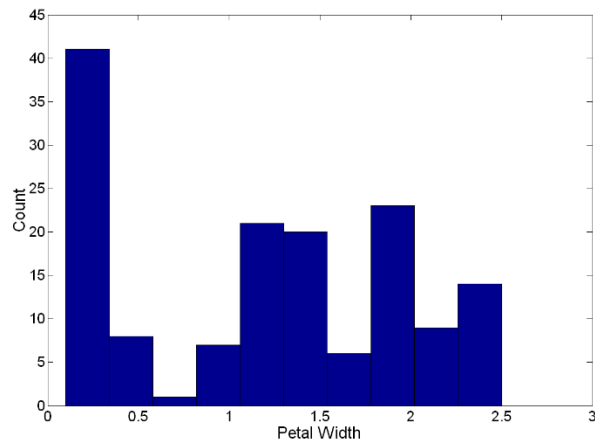
## Histogram

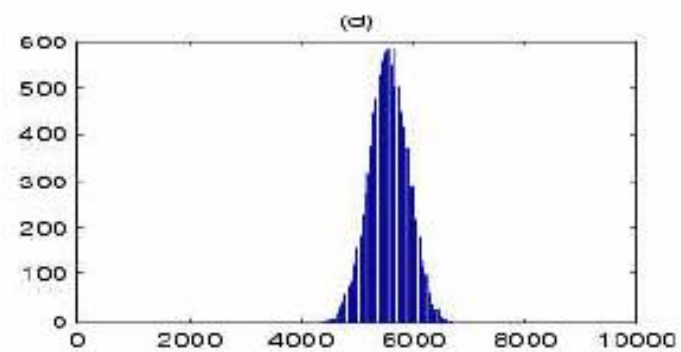
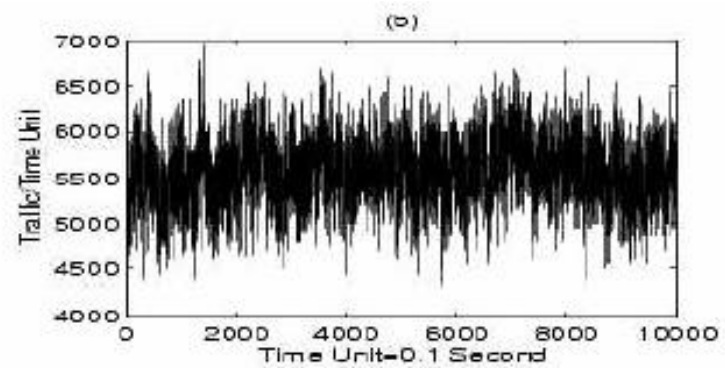
- ❖ Usually shows the distribution of values of a single variable
- ❖ Divide the values into bins and show a bar plot of the number of objects in each bin.
- ❖ The height of each bar indicates the number of objects
- ❖ Shape of histogram depends on the number of bins



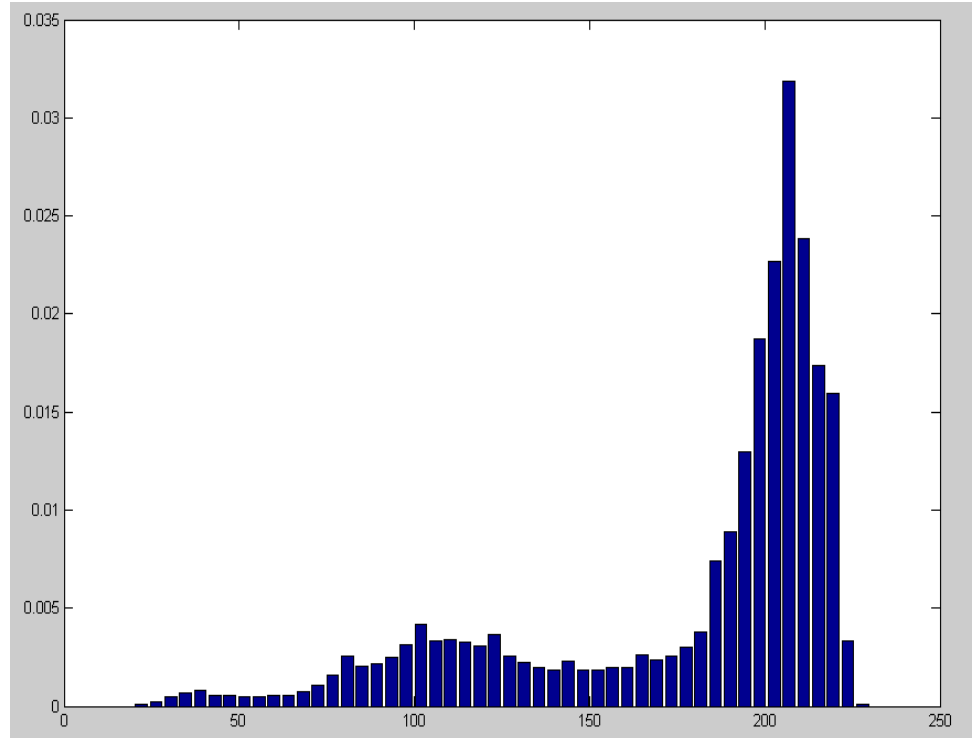
**Example:** Petal Width (10 and 20 bins, respectively)

relative frequency  
histogram





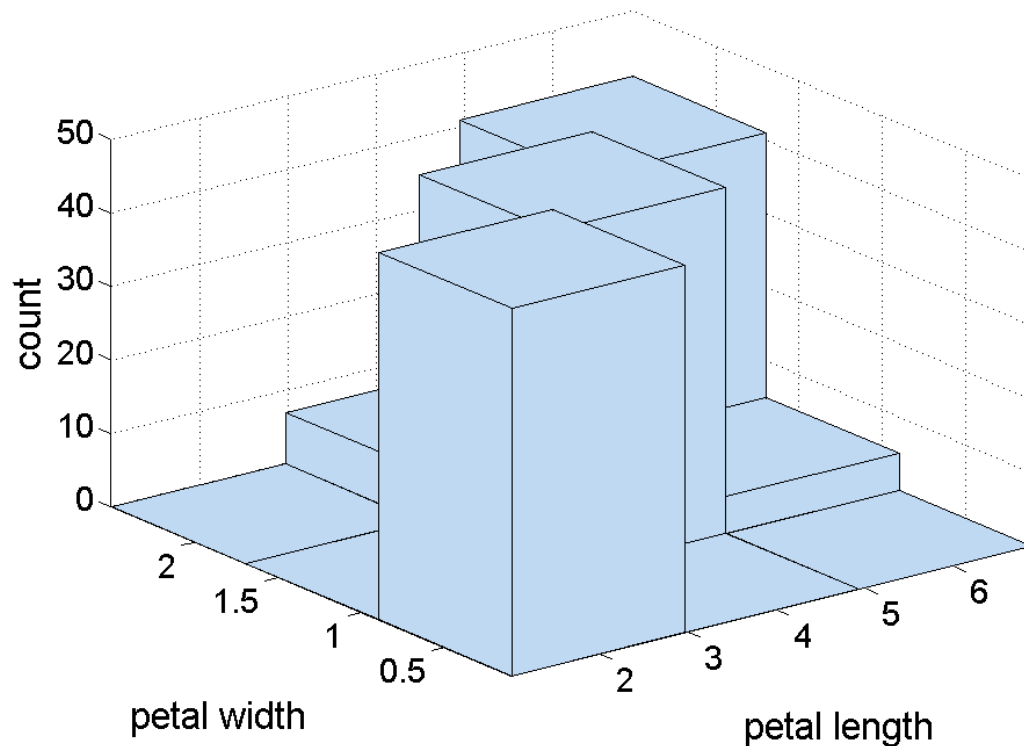




# Two-Dimensional Histograms

Show the joint distribution of the values of two attributes

**Example:** petal width and petal length

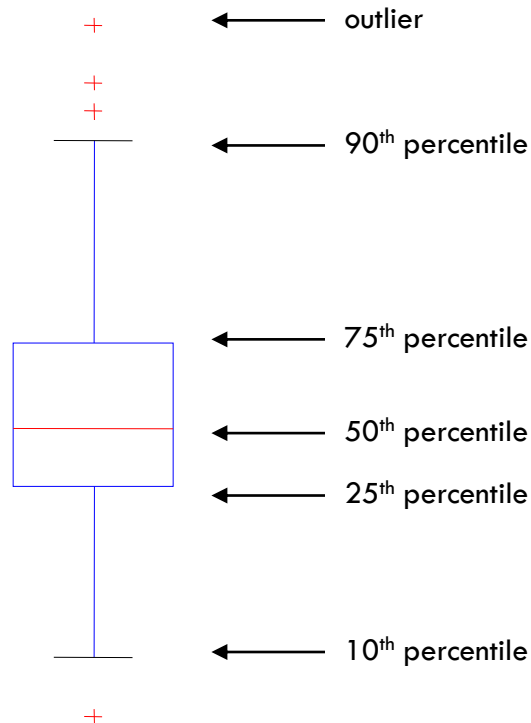


# Visualization Techniques: Box Plots

## Box Plots

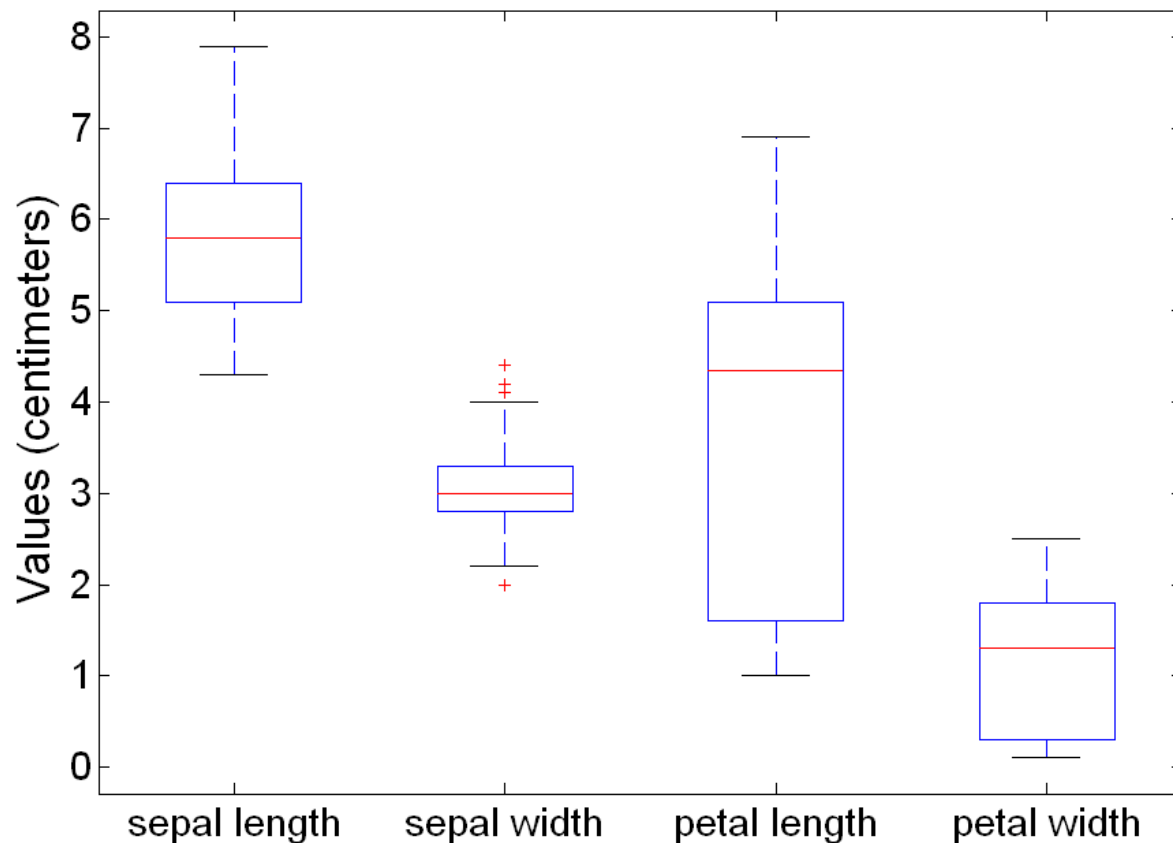
Another way of displaying the distribution of data

Following figure shows the basic part of a box plot



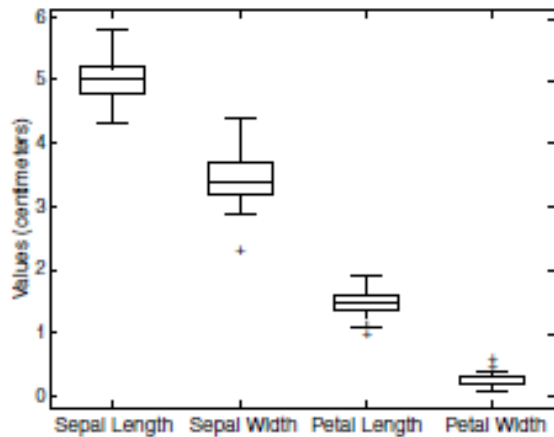
# Visualization Techniques: Box Plots

Box plots can be used to compare attributes

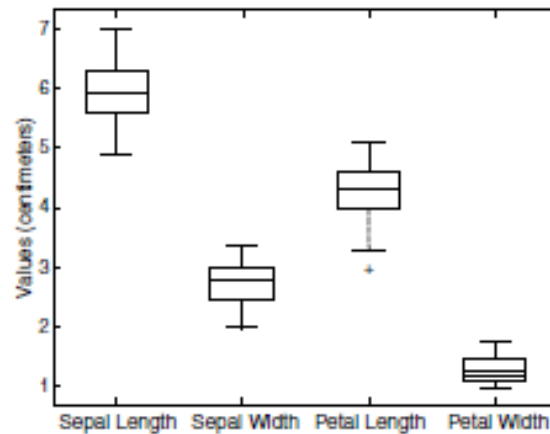


# Visualization Techniques: Box Plots

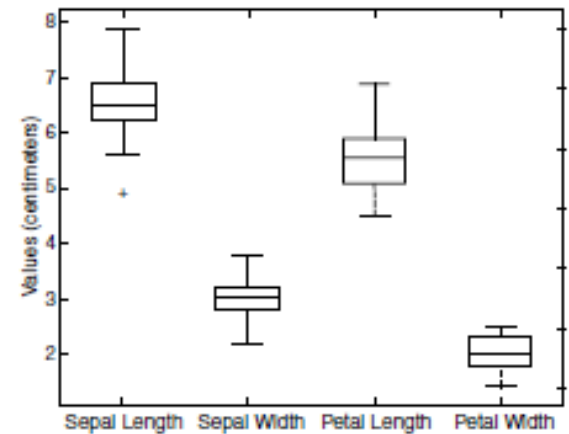
Box plots can also be used to compare how attributes vary between different classes of objects



(a) Setosa.



(b) Versicolour.



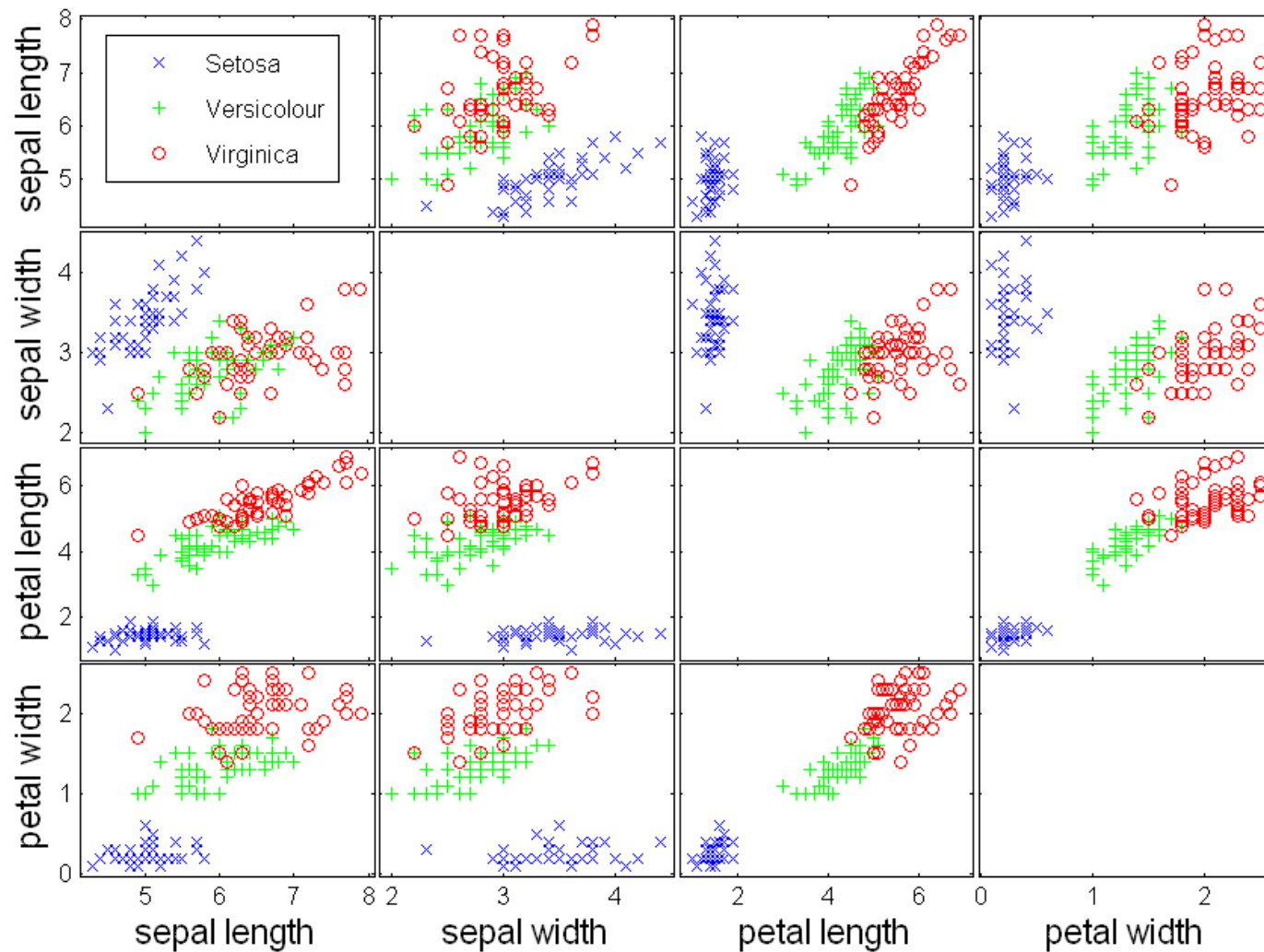
(c) Virginica.

# Visualization Techniques: Scatter Plots

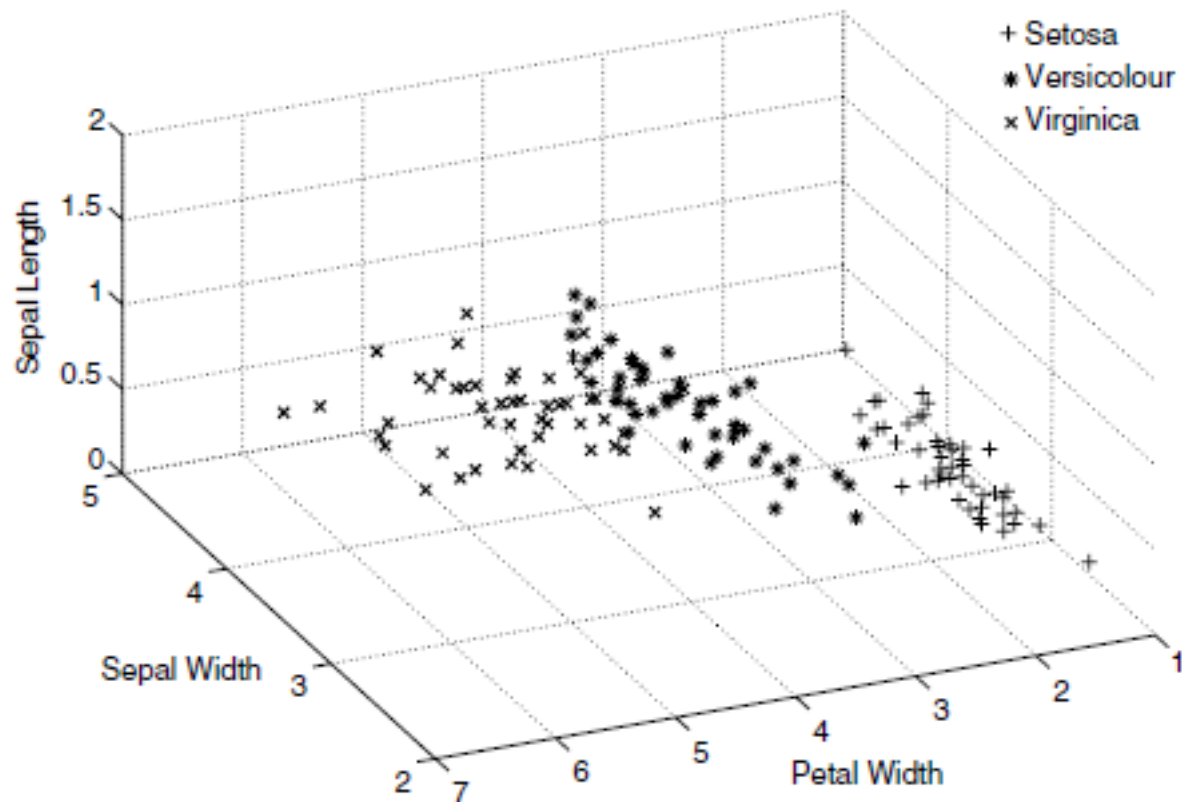
## Scatter plots

- ✓ Attributes values determine the position
- ✓ Two-dimensional scatter plots most common, but can have three-dimensional scatter plots
- ✓ Often additional attributes can be displayed by using the size, shape, and color of the markers that represent the objects
- ✓ It is useful to have arrays of scatter plots can compactly summarize the relationships of several pairs of attributes

# Visualization Techniques: Scatter Plots



# Visualization Techniques: Scatter Plots



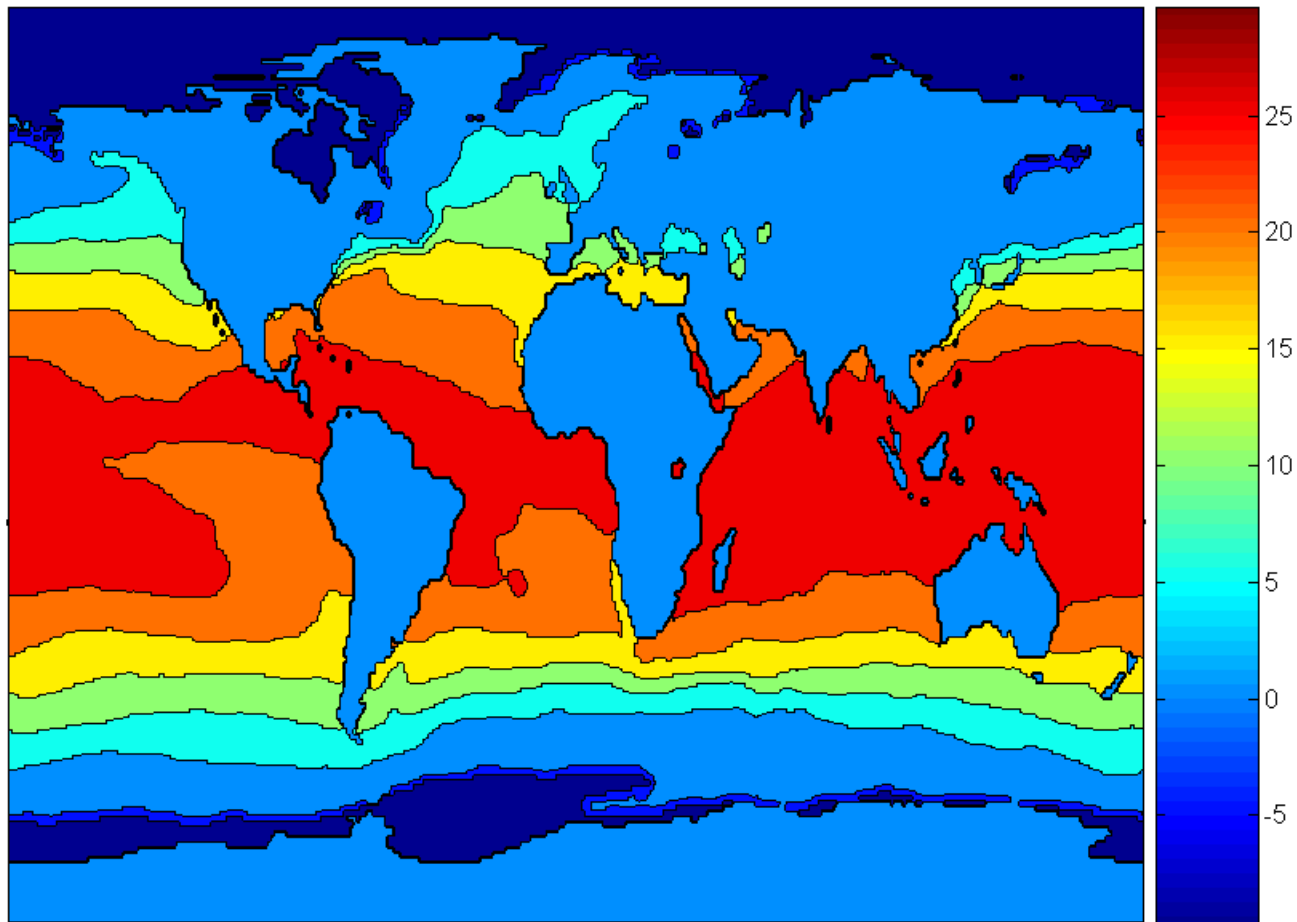


# Visualization Techniques: Contour Plots

## Contour plots

- ✓ Useful when a continuous attribute is measured on a spatial grid
- ✓ They partition the plane into regions of similar values
- ✓ The contour lines that form the boundaries of these regions connect points with equal values
- ✓ The most common example is contour maps of elevation
- ✓ Can also display temperature, rainfall, air pressure, etc.

# Visualization Techniques: Contour Plots



Celsius

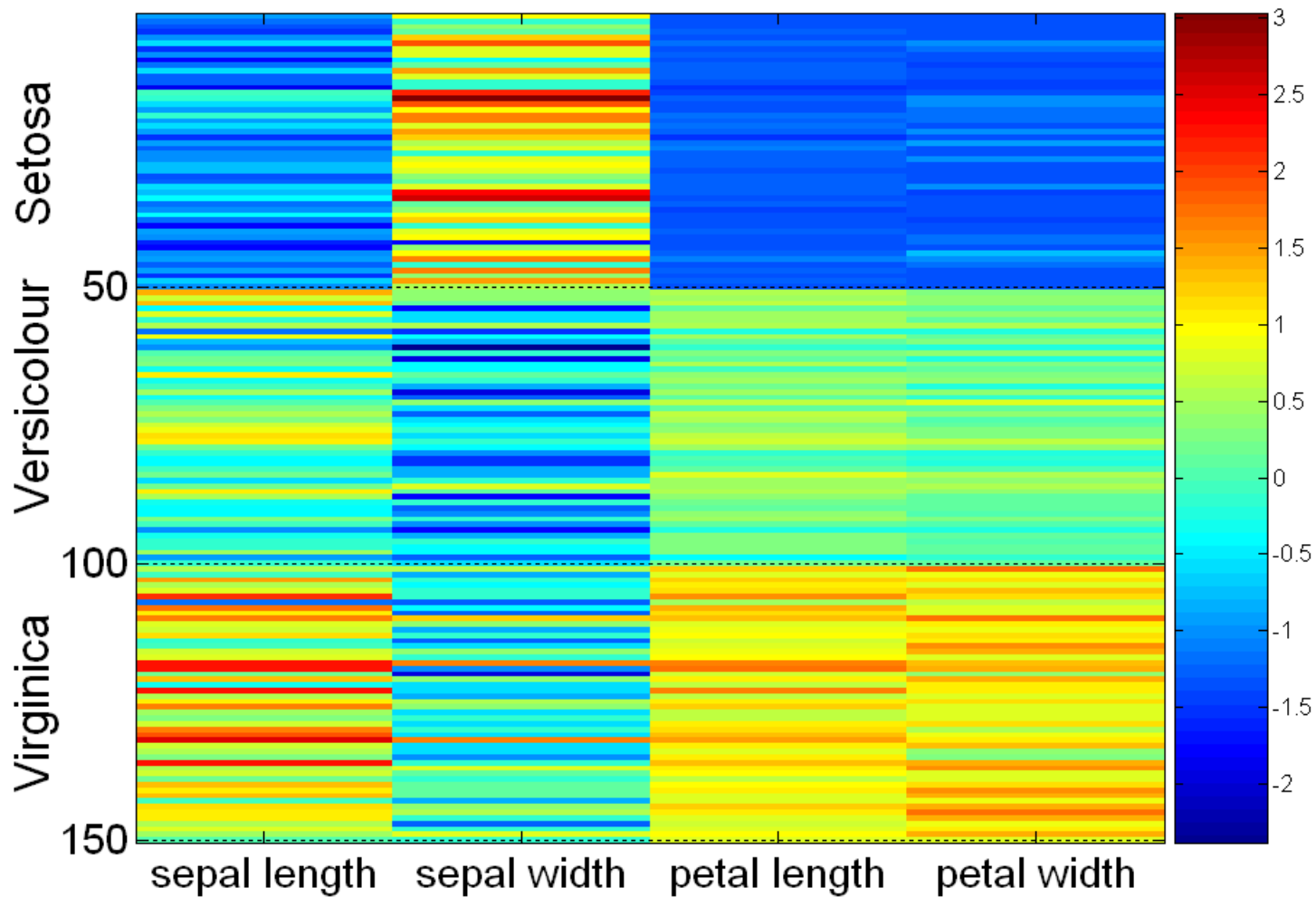
# Visualization Techniques: Matrix Plots



## Matrix plots

- ✓ Can plot the data matrix
- ✓ This can be useful when objects are sorted according to class
- ✓ Typically, the attributes are normalized to prevent one attribute from dominating the plot
- ✓ Plots of similarity or distance matrices can also be useful for visualizing the relationships between objects

# Visualization Techniques: Matrix Plots



# Visualization Techniques: Matrix Plots

