

# STATISTICAL LABORATORY

## Lecture 3: Exploratory Data Analysis (Continued)

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Applied Mathematics for Economics and Management  
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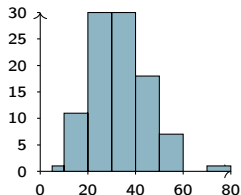
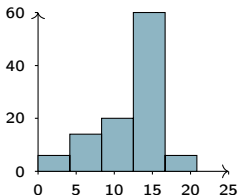
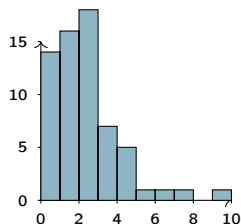
# Announcement: Venue change

- 1 Tomorrow (Wednesday, September 23rd), our practice session will be held in room 202 Francesinhas 2.
- 2 Starting next week, we will return to room 109 Francesinhas 2.
- 3 The written test will take place during the 12th week (the fourth week of November, likely on the 24th or 25th). I will confirm the exact date as soon as possible and let you know.

- 1 Fundamental Concepts of Statistics
- 2 Exploratory Data Analysis
- 3 Organizing and Summarizing Data
- 4 Association and Relationships Between Variables
- 5 Index Numbers
- 6 Time Series Analysis

# Skewness in Histograms

Is the histogram right skewed, left skewed, or symmetric?



- **Right-skewed (positively skewed) Distribution:**

- Tail extends to the right (higher values). Most data concentrated on the left. Skewness  $> 0$ .

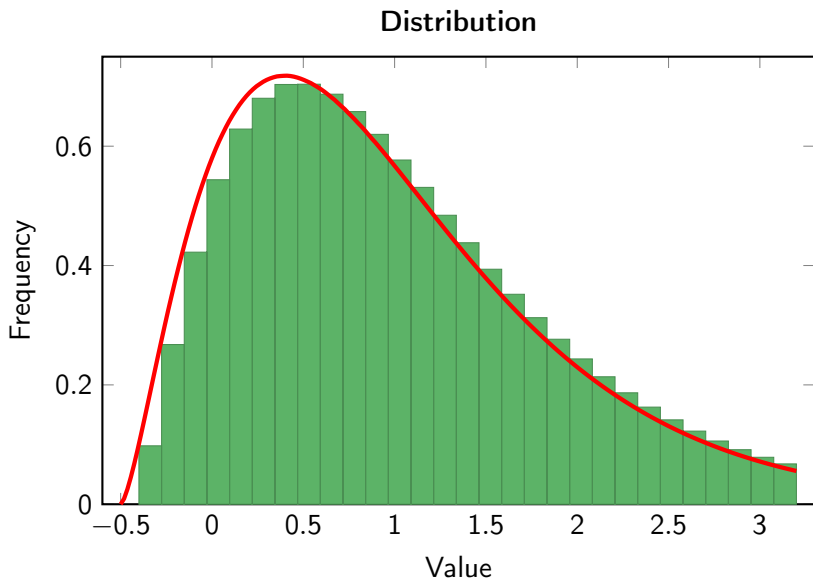
- **Left-skewed (negatively skewed) Distribution:**

- Tail extends to the left (lower values). Most data concentrated on the right. Skewness  $< 0$ .

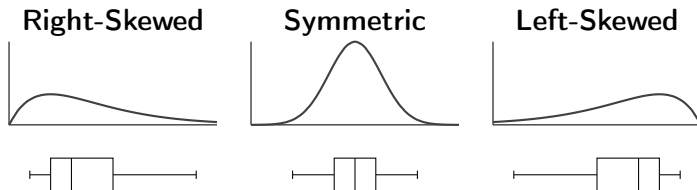
- **Symmetric Distribution:**

- Tails roughly equal on both sides. Data evenly distributed around the center. Skewness  $\approx 0$ .

# Skewness in Histograms

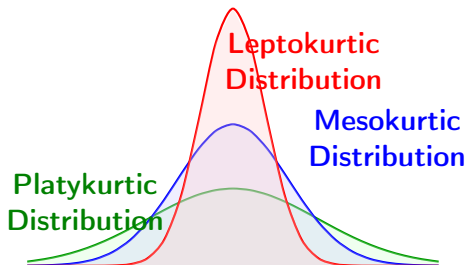


# Skewness in Boxplots



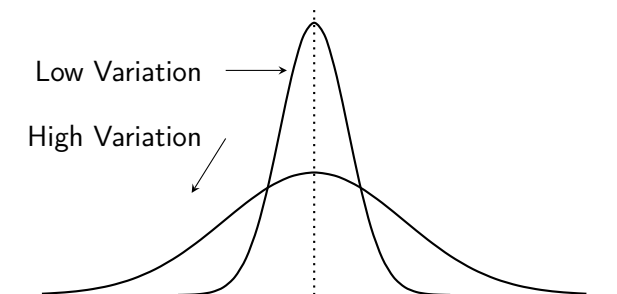
- Boxplots also clearly reflect distribution skewness (Right-skewed, Symmetric, Left-skewed).
- Note: The complete boxplot anatomy will be discussed in subsequent slides.

# Kurtosis Visualization



- **Leptokurtic Distribution** (high peak, heavy tails):
  - Most data concentrated near the mean.
  - More extreme values (outliers) likely.
- **Platykurtic Distribution** (low peak, light tails):
  - Data more evenly spread.
  - Fewer extreme values.
- **Mesokurtic Distribution** (moderate peak, normal tails):
  - Moderate concentration of data around the mean.

# Variation Visualization



- **Low Variation:** Narrow, tall curve. Data values cluster tightly around the center.
- **High Variation:** Wide, flat curve. Data values are widely dispersed.

# Stem-and-Leaf Diagram: Example

- A simple way to see distribution details in a data set.
- **Method:** Separate the sorted data series into leading digits (the stem) and the trailing digits (the leaves).

Data in ordered array:

21, 24, 24, 26, 27, 27, 30, 32, 38, 41

Completed stem-and-leaf diagram:

| Stem | Leaves      |
|------|-------------|
| 2    | 1 4 4 6 7 7 |
| 3    | 0 2 8       |
| 4    | 1           |

# Exercise 1.32

**1.32** Consider the following data:

17, 62, 15, 65, 28, 51, 24, 65, 39, 41, 35, 15, 39, 32, 36, 37, 40, 21, 44, 37, 59,  
13, 44, 56, 12

**Tasks:**

- 1 Construct a stem-and-leaf display.

## Exercise 1.32 d): Solution

### Steps for Stem-and-Leaf Solution:

- **Step 1 - Data:** Sorted raw values.
- **Step 2 - Determine stems & leaves:** Tens digit  $\rightarrow$  stem, Units digit  $\rightarrow$  leaf.
- **Step 3 & 4 - Assign and Order Leaves:**
  - Stem 1: 2, 3, 5, 5, 5, 7 (Freq: 6)
  - Stem 2: 1, 4, 8 (Freq: 3)
  - Stem 3: 2, 5, 6, 7, 7, 9, 9 (Freq: 7)
  - Stem 4: 0, 1, 4, 4 (Freq: 4)
  - Stem 5: 1, 4, 6, 9, 9 (Freq: 5)
  - Stem 6: 2, 4, 5, 5 (Freq: 4)

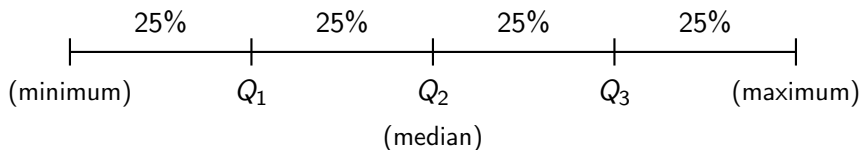
## Exercise 1.32 d): Solution

- Step 5 - Final stem-and-leaf plot with frequencies:

| Stem | Leaf          | Frequency |
|------|---------------|-----------|
| 1    | 2 3 5 5 5 7   | 6         |
| 2    | 1 4 8         | 3         |
| 3    | 2 5 6 7 7 9 9 | 7         |
| 4    | 0 1 4 4       | 4         |
| 5    | 1 4 6 9 9     | 5         |
| 6    | 2 4 5 5       | 4         |

- Step 6 - Interpretation: Most data are concentrated in the 30s. Frequencies highlight dominant stems.

# Quantiles: Definition



- **What are Quantiles?** Values that divide a dataset into equal parts.
- **Special cases:**
  - **Quartiles ( $Q_1, Q_2, Q_3$ ):** Divide data into 4 equal parts (Median =  $Q_2$ ).
  - **Deciles ( $D_1, \dots, D_{10}$ ):** Divide data into 10 equal parts (Median =  $D_5$ ).
  - **Percentiles ( $P_1, \dots, P_{100}$ ):** Divide data into 100 equal parts (Median =  $P_{50}$ ).
- **Order- $p$  quantile ( $q_p$ ):** With  $0 < p < 1$ , the observed value dividing sample data into values below and above  $q_p$ .
  - If  $p = 0.9$ : You are looking for the value ( $q_{0.9}$ ) where 90% of the data is below it, leaving only the top 10% above.

# Percentiles and Quartiles

- To find percentiles and quartiles, data must first be arranged in order from the smallest to the largest values.
- ***P*th percentile:** A value such that **approximately**  $P\%$  of the observations are at or below that number.
  - Percentiles separate large ordered data sets into 100 intervals.
  - The 50th percentile is the median.

- **Position formula:**

$P$ th percentile = value located in the  $\left(\frac{P}{100}\right)(n+1)$ th ordered position

- **Quartiles:** Descriptive measures that separate large data sets into four quarters.
  - **First quartile ( $Q_1$  / 25th percentile):** Separates approximately the smallest 25% of the data from the remainder.
  - **Second quartile ( $Q_2$  / 50th percentile):** The median.

# Percentiles and Quartiles

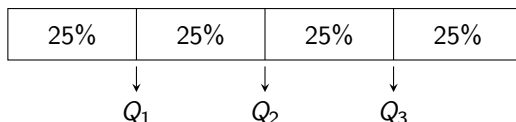
- Percentiles and Quartiles indicate the position of a value relative to the entire set of data.
- Generally used to describe large data sets.
- Example: An IQ score at the 90<sup>th</sup> percentile means that 10% of the population has a higher IQ score and 90% have a lower IQ score.

$P^{\text{th}}$  percentile = value located in the  $\left(\frac{P}{100}\right)(n+1)^{\text{th}}$  ordered position

Newbold et al. (2013)

# Quartiles

- Quartiles split the ranked data into 4 segments with an equal number of values per segment (note that the widths of the segments may be different).



- The first quartile,  $Q_1$ , is the value for which 25% of the observations are smaller and 75% are larger.
- $Q_2$  is the same as the median (50% are smaller, 50% are larger).
- Only 25% of the observations are greater than the third quartile.

Newbold et al. (2013)

# Quartile Formulas

Find a quartile by determining the value in the appropriate position in the ranked data, where  $(n + 1)p$ :

- First quartile position:

$$Q_1 = 0.25(n + 1)$$

- Second quartile position (the median position):

$$Q_2 = 0.50(n + 1)$$

- Third quartile position:

$$Q_3 = 0.75(n + 1)$$

where  $n$  is the number of observed values.

Newbold et al. (2013)

# Steps to Calculate an Order- $p$ Quantile ( $q_p$ )

- 1 **Order the sample:** Arrange data in ascending order:  
 $x_{(1)}, x_{(2)}, \dots, x_{(n)}$ .
- 2 **Determine the quantile position:** Compute  $(n+1)p$ . Let  $(n+1)p = r + a$ , where  $r$  is the integer part and  $a$  is the fractional part.
- 3 **Calculate the quantile:**
  - If  $(n+1)p$  is an integer:  $q_p = x_{(r)}$
  - If  $(n+1)p$  is not an integer:  $q_p = (1-a)x_{(r)} + ax_{(r+1)}$  (set it to a value between  $x_{(r)}$  and  $x_{(r+1)}$ )

*Note: Different software packages (SPSS, Excel) may use alternative formulas yielding minor discrepancies. Choose one method and apply consistently.*

# Compare the two formulas

- Suppose that  $p = 0.9$ .  $q_{0.9}$  and 90th percentile share the same meaning ( $p = P/100$ ).
- So the percentile formula is

$$\frac{90}{100} \times (n + 1) = 0.9 \times (n + 1). \quad (1)$$

- If  $0.9 \times (n + 1)$  is an integer and say it equals to  $r = 50$ , the 90th percentile is  $x_{(50)}$ .
- Note that the notation  $x_{(r)}$  denotes the  $r$ -th value value in the ascending-ordered dataset. (So in general  $x_{50} \neq x_{(50)}$ ).
- However, if the position  $r = 50.1$  (not an integer), we take an interpolated value between  $x_{(50)}$  and  $x_{(51)}$ .

# Steps to Calculate an Order- $p$ Quantile ( $q_p$ ) with Example

## ● Step 1 – Order the sample

- Data: 12, 15, 7, 20, 18
- Arrange in ascending order:

$$x_{(1)} = 7, x_{(2)} = 12, x_{(3)} = 15, x_{(4)} = 18, x_{(5)} = 20$$

## ● Step 2 – Determine the quantile position

- Suppose we want the **0.4 quantile** ( $p = 0.4$ , i.e., the 40th percentile).
- Compute  $(n + 1)p = (5 + 1) \cdot 0.4 = 2.4$
- Here,  $r = 2$  (integer part),  $a = 0.4$  (fractional part)

## ● Step 3 – Calculate the quantile

- Since  $(n + 1)p$  is not an integer, interpolate:

$$q_{0.4} = (1 - 0.4)x_{(2)} + 0.4x_{(3)} = 0.6 \cdot 12 + 0.4 \cdot 15 = 13.2$$

# Percentiles and Quartiles

- Indicate the position of a value relative to the entire dataset.
- Generally used to describe large datasets.
- **Pth Percentile formula:** Value located in the position:

$$\left( \frac{P}{100} \right) (n + 1)$$

- **Quartiles:** Split ranked data into 4 segments ( $Q_1 = 25\text{th percentile}$ ,  $Q_2 = \text{Median}/50\text{th percentile}$ ,  $Q_3 = 75\text{th percentile}$ ).

# Interquartile Range (IQR)

- Measures the spread in the middle 50
- Eliminates extreme high- and low-valued observations.
- **Formula:**

$$\text{IQR} = Q_3 - Q_1$$

- **Note on terminology:** While IQR strictly denotes the numerical difference  $Q_3 - Q_1$ , the interval  $[Q_1, Q_3]$  is sometimes informally described as the interquartile interval containing the middle 50% of observations.

# Interquartile Range (IQR): Example

Imagine comparing the daily high temperatures in April for two cities in Portugal:

- Lisbon (Low IQR): Because of the moderating effect of the Atlantic Ocean, the middle 50% of April days in Lisbon might have highs between  $18^{\circ}\text{C}$  and  $21^{\circ}\text{C}$ . The IQR is just  $3^{\circ}\text{C}$ . This means the core spring weather is quite consistent.
- Évora (High IQR): The middle 50% of April days might range from  $14^{\circ}\text{C}$  all the way to  $26^{\circ}\text{C}$ . The IQR is  $12^{\circ}\text{C}$ . This higher IQR shows that the spring weather in Évora is much more variable and spread out.
- If a heat wave occurs in April for 1-2 days, it will not affect the value of IQR, since we only care about the middle 50% of the data.

# Five-Number Summary

The five-number summary consists of five descriptive measures arranged in order:

- 1 Minimum
- 2 First Quartile ( $Q_1$ )
- 3 Median ( $Q_2$ )
- 4 Third Quartile ( $Q_3$ )
- 5 Maximum

$$\text{minimum} \leq Q_1 \leq \text{median} \leq Q_3 \leq \text{maximum}$$

## Exercise 2.6

**2.6** During the last 3 years Consolidated Oil Company expanded stations into convenience food stores (CFSs). Daily sales (in hundreds of dollars) from a random sample of 10 weekdays are:

6, 8, 10, 12, 14, 9, 11, 7, 13, 11

### Tasks:

- 1 Find the five-number summary.

## Exercise 2.6 ): Solution

### Step 1 - Order the sample:

$$x_{(1)} = 6, x_{(2)} = 7, x_{(3)} = 8, x_{(4)} = 9, x_{(5)} = 10, \\ x_{(6)} = 11, x_{(7)} = 11, x_{(8)} = 12, x_{(9)} = 13, x_{(10)} = 14$$

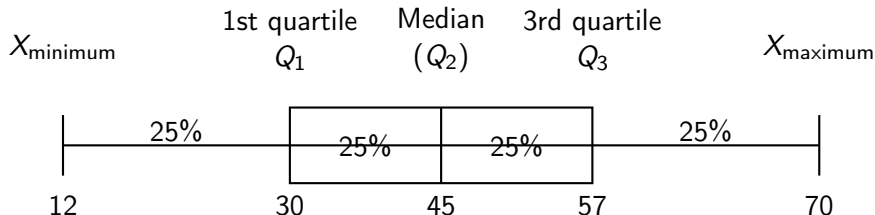
### Step 2 - Quantile positions ( $n = 10$ ):

- Minimum: 6
- Maximum: 14
- $Q_1$ :  $(10 + 1)(0.25) = 2.75 \implies Q_1 = 0.25(7) + 0.75(8) = 7.75$
- Median ( $Q_2$ ):  $(10 + 1)(0.5) = 5.5 \implies \text{Median} = 0.5(10) + 0.5(11) = 10.5$
- $Q_3$ :  $(10 + 1)(0.75) = 8.25 \implies Q_3 = 0.75(12) + 0.25(13) = 12.25$

### Step 3 - Five-number summary:

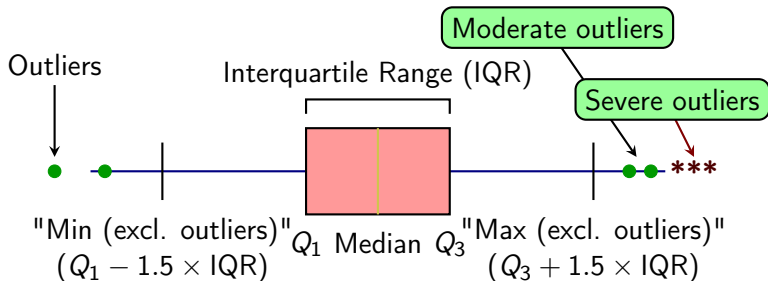
$$\text{Min} = 6, \quad Q_1 = 7.75, \quad \text{Median} = 10.5, \quad Q_3 = 12.25, \quad \text{Max} = 14$$

# Box-and-Whisker Plot / Boxplot



- Visualizes the five-number summary graphically.
- Box spans from  $Q_1$  to  $Q_3$  (containing the middle 50% of data), divided by the median line ( $Q_2$ ).
- Whiskers extend to minimum and maximum values (or specific fence boundaries).
- Can be oriented horizontally or vertically.

# Boxplots and Outliers



- **Interquartile Range:**  $IQR = Q_3 - Q_1$
- **Inner Fences (Moderate outlier thresholds):**

$$Q_1 - 1.5 \times IQR \quad \text{and} \quad Q_3 + 1.5 \times IQR$$

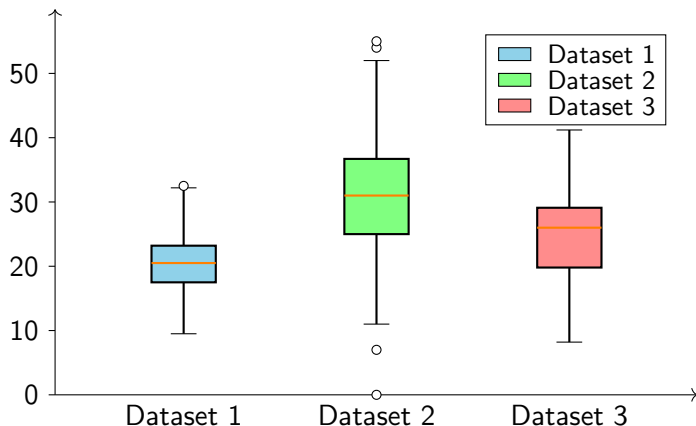
Points beyond these are **moderate outliers** (marked with circles).

- **Outer Fences (Severe outlier thresholds):**

$$Q_1 - 3 \times IQR \quad \text{and} \quad Q_3 + 3 \times IQR$$

Points beyond these are **severe outliers** (marked with asterisks).

# Comparing Data Sets



- Use measures of central tendency, dispersion, and shape.
- Visualize with histograms, boxplots, and stem-and-leaf plots.
- Identify patterns, similarities, and differences between data sets.

# Quantile-Quantile (Q-Q) Plots: Definition & Purpose

- **Definition:** A graphical tool used to compare the distribution of two datasets or check if a dataset follows a theoretical distribution.
- **Purpose:**
  - If data and reference distributions share the same shape, points form a straight line.
  - Deviations reveal structural differences.
- **Common Applications:**
  - Normality checks (comparing sample quantiles to a normal distribution).
  - Distribution comparisons.
  - Outlier detection.

# Q-Q Plots: Construction & Interpretation

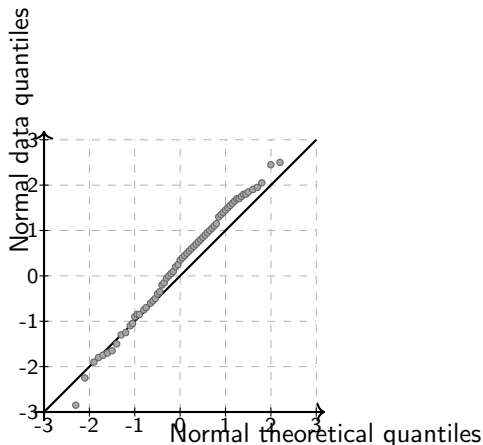
- **Construction Steps:**

- 1 Sort the observed data in ascending order.
- 2 Calculate corresponding theoretical quantiles.
- 3 Plot observed quantiles ( $y$ -axis) against theoretical quantiles ( $x$ -axis).

- **Interpretation:**

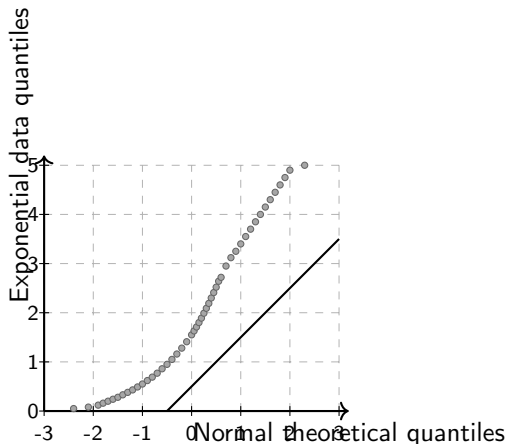
- Points close to a straight line  $\rightarrow$  data follows the reference distribution.
- Nonlinear patterns or shifts  $\rightarrow$  departures from normality, presence of skewness, or differences in mean/variance.

# Q-Q plots Examples



- A normal Q-Q plot comparing randomly generated, independent standard normal data on the vertical axis to a standard normal population on the horizontal axis.
- The linearity of the points suggests that the data are normally distributed.

# Normal Q-Q Plot for Exponential Data

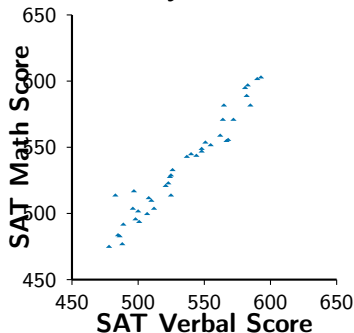


- A normal Q-Q plot of randomly generated, independent standard exponential data ( $X \sim \text{Exp}(1)$ ).
- The points follow a strongly nonlinear pattern, suggesting non-normality, while the offset indicates a non-zero mean.

# Scatter Diagram / Scatterplot

| Average SAT scores by state: 1998 |        |      |
|-----------------------------------|--------|------|
|                                   | Verbal | Math |
| Alabama                           | 562    | 558  |
| Alaska                            | 521    | 520  |
| Arizona                           | 525    | 528  |
| Arkansas                          | 568    | 555  |
| California                        | 497    | 516  |
| Colorado                          | 537    | 542  |
| Connecticut                       | 510    | 509  |
| Delaware                          | 501    | 493  |
| D.C.                              | 488    | 476  |
| Florida                           | 500    | 501  |
| Georgia                           | 486    | 482  |
| Hawaii                            | 483    | 513  |
| ...                               |        |      |
| W.Va.                             | 525    | 513  |
| Wis.                              | 581    | 594  |
| Wyo.                              | 548    | 546  |

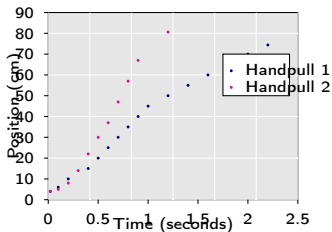
**Average SAT Math vs. Verbal Scores by State**



- Used for paired observations taken from two numerical variables.
- One variable is measured on the vertical axis and the other on the horizontal axis.
- Helps identify bivariate relationships, correlations, and clusters (e.g., SAT Math vs. Verbal scores by state).

- **Definition:** A graphical representation displaying the temporal sequence or ordering of events.
- **Purpose:**
  - Show data or events in chronological order.
  - Identify trends, patterns, or anomalies over time.
- **Key Features:**
  - Time/sequence on the  $x$ -axis, observations on the  $y$ -axis.
  - Often displayed as connected dots, especially in financial publications like *The Wall Street Journal*.

# Time Plot Examples



The following graph shows a physics-related time plot with the position vs. time for two spark tapes pulled through a spark timer at different constant speeds.



While a time plot can resemble a scatter plot, with a series of dots, you will often see these plots with the dots connected, especially in financial publications like The Wall Street Journal.

Frequency table for discrete and continuous variable

# Frequency table for discrete variable: example

## 1. Discrete Variable Example

Suppose we collected  $n = 15$  observations of a discrete variable  $x$ :

$$x = (2, 3, 3, 4, 2, 5, 3, 4, 2, 5, 4, 3, 2, 4, 3)$$

**Step 1:** Identify distinct values

$$x'_1 = 2, \quad x'_2 = 3, \quad x'_3 = 4, \quad x'_4 = 5$$

**Step 2:** Count frequencies

### Absolute Frequencies

$$n_1 = 4, \quad n_2 = 5, \quad n_3 = 4, \quad n_4 = 2$$

**Step 3:** Compute relative frequencies

### Relative Frequencies

$$f_j = \frac{n_j}{n} \implies f_1 = 0.267, \quad f_2 = 0.333, \quad f_3 = 0.267, \quad f_4 = 0.133$$

**Step 4:** Determine range

$$x_{\min} = 2, \quad x_{\max} = 5, \quad \text{Range} = 3$$

# Frequency table for discrete variable: example

## Step 5: Table (Frequency Distributions)

| $x'_j$       | Absolute Frequencies ( $n_j$ ) | Relative Frequencies ( $f_j$ ) |
|--------------|--------------------------------|--------------------------------|
| 2            | 4                              | 0.267                          |
| 3            | 5                              | 0.333                          |
| 4            | 4                              | 0.267                          |
| 5            | 2                              | 0.133                          |
| <b>Total</b> | <b>15</b>                      | <b>1.0</b>                     |

**Sample Size**  
 $n = 15$

$$\sum_{j=1}^m n_j = n \quad \sum_{j=1}^m f_j = 1$$

# Frequency table for discrete variable: example

Suppose we collected  $n = 12$  observations of a continuous variable  $x$ :

$$x = (1.2, 2.3, 1.8, 2.0, 3.1, 2.7, 3.5, 1.5, 2.8, 3.0, 1.9, 2.5)$$

**Step 1:** Determine range and class width

$$x_{\min} = 1.2, \quad x_{\max} = 3.5, \quad \text{Range} = 3.5 - 1.2 = 2.3$$

Using Sturges' rule,  $m = 4$  classes:

$$\text{Class width} = \frac{\text{Range}}{m} = \frac{2.3}{4} \approx 0.575 \approx 0.6$$

# Frequency table for discrete variable: example

**Step 2:** Construct classes ( $I_j = [l_{j-1}, l_j], j = 1, 2, \dots, m$ )

Start at 1.2:

①  $[1.2, 1.8]$

②  $(1.8, 2.4]$

③  $(2.4, 3.0]$

④  $(3.0, 3.6]$

$$I_j \cap I_k = \emptyset \text{ (non-overlap) and}$$

$$D \subset \bigcup_{j=1}^m I_j \text{ (inclusive)}$$

$$D = [x_{\min}, x_{\max}]$$

**Step 3:** Find midpoints

$$x'_1 = \frac{1.2 + 1.8}{2} = 1.5, \quad x'_2 = \frac{1.8 + 2.4}{2} = 2.1, \quad x'_3 = \frac{2.4 + 3.0}{2} = 2.7,$$

$$x'_4 = \frac{3.0 + 3.6}{2} = 3.3$$

# Frequency table for discrete variable: example

## Step 4: Count frequencies

- Class  $[1.2, 1.8]$ :  $1.2, 1.5, 1.8 \implies n_1 = 3$
- Class  $]1.8, 2.4]$ :  $1.9, 2.0, 2.3 \implies n_2 = 3$
- Class  $]2.4, 3.0]$ :  $2.5, 2.7, 3.0 \implies n_3 = 3$
- Class  $]3.0, 3.6]$ :  $3.1, 3.5 \implies n_4 = 2$

## Step 5: Compute relative frequencies

$$f_1 = \frac{3}{12} = 0.25, \quad f_2 = 0.25, \quad f_3 = 0.25, \quad f_4 = \frac{2}{12} \approx 0.167$$

## Step 6: Table (Frequency Distribution with Totals)

| Class $I_j$  | Midpoint $x'_j$ | $n_j$     | $f_j$      |
|--------------|-----------------|-----------|------------|
| $[1.2, 1.8]$ | 1.5             | 3         | 0.25       |
| $]1.8, 2.4]$ | 2.1             | 3         | 0.25       |
| $]2.4, 3.0]$ | 2.7             | 3         | 0.25       |
| $]3.0, 3.6]$ | 3.3             | 2         | 0.167      |
| <b>Total</b> | –               | <b>12</b> | <b>1.0</b> |

## Cumulative frequency function

# Cumulative Absolute Frequency Function

## Cumulative Absolute Frequency Function

$$N(x) = \begin{cases} 0 & x < x'_1 \\ n_1 & x'_1 \leq x < x'_2 \\ n_1 + n_2 & x'_2 \leq x < x'_3 \\ \dots & \dots \\ n_1 + n_2 + \dots + n_{m-1} & x'_{m-1} \leq x < x'_m \\ n_1 + n_2 + \dots + n_m = n & x \geq x'_m \end{cases}$$

- $N(x)$  is a function whose domain is the real line, with  $N(-\infty) = 0$ ,  $N(+\infty) = n$  and  $0 \leq N(x) \leq n$ .

- It is a **non-decreasing function and right-continuous**.

- It is a **step function** where each jump corresponds to an absolute frequency. For example:  $N(x'_3) - N(x'_2) = (n_1 + n_2 + n_3) - (n_1 + n_2) = n_3$ .

# Cumulative Relative Frequency Function

## Cumulative Relative Frequency Function

$$F^*(x) = \begin{cases} 0 & x < x'_1 \\ f_1 & x'_1 \leq x < x'_2 \\ f_1 + f_2 & x'_2 \leq x < x'_3 \\ \dots & \dots \\ f_1 + f_2 + \dots + f_{m-1} & x'_{m-1} \leq x < x'_m \\ f_1 + f_2 + \dots + f_m = 1 & x \geq x'_m \end{cases}$$

- $F^*(x)$  has properties similar to  $N(x)$ , except with respect to the codomain:  
 $0 \leq F^*(x) \leq 1$ .
- $F^*(-\infty) = 0$  and  $F^*(+\infty) = 1$ .
- It represents the proportion of observations less than or equal to  $x$ .

# Cumulative frequency function for discrete variable: Example (Data)

## Sample information:

- Observations:  $x = (2, 3, 3, 4, 2, 5, 3, 4, 2, 5, 4, 3, 2, 4, 3)$
- Number of observations:  $n = 15$
- Sample range: Sample range  $= x_{\max} - x_{\min} = 5 - 2 = 3$

## Frequency Distribution Table (with cumulative frequencies)

| $x'_j$       | $n_j$     | $f_j$      | $N(x'_j)$ | $F^*(x'_j)$ |
|--------------|-----------|------------|-----------|-------------|
| 2            | 4         | 0.267      | 4         | 0.267       |
| 3            | 5         | 0.333      | 9         | 0.600       |
| 4            | 4         | 0.267      | 13        | 0.867       |
| 5            | 2         | 0.133      | 15        | 1.0         |
| <b>Total</b> | <b>15</b> | <b>1.0</b> | –         | –           |

# Cumulative frequency function for discrete variable: Example (Functions)

## Cumulative Absolute Frequency Function

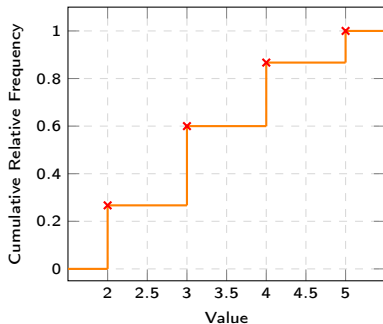
$$N(x) = \begin{cases} 0, & x < 2 \\ 4, & 2 \leq x < 3 \\ 9, & 3 \leq x < 4 \\ 13, & 4 \leq x < 5 \\ 15, & x \geq 5 \end{cases}$$

## Cumulative Relative Frequency Function

$$F^*(x) = \begin{cases} 0, & x < 2 \\ 0.267, & 2 \leq x < 3 \\ 0.600, & 3 \leq x < 4 \\ 0.867, & 4 \leq x < 5 \\ 1, & x \geq 5 \end{cases}$$

# Cumulative frequency graph for a discrete variable: example

- Here is the graph of the cumulative relative frequency function (ogive) for your sample: Values: 2, 3, 4, 5, Frequencies: 4, 5, 4, 5.
- The red marks indicate the cumulative proportions at each value.
- This shows how the cumulative relative frequency increases step by step until it reaches 1 (100%).



# Cumulative frequency function for continuous variable

## Cumulative Relative Frequency Function

$$F^*(l_0) = 0$$

$$F^*(l_1) = f_1$$

$$F^*(l_2) = f_1 + f_2$$

...

$$F^*(l_{m-1}) = f_1 + f_2 + \cdots + f_{m-1}$$

$$F^*(l_m) = f_1 + f_2 + \cdots + f_m = 1$$

It is assumed that the frequencies are **uniformly distributed within each class** and that the cumulative frequencies refer to the **class boundaries**.

# Frequency cumulative function for continuous variable: Example (Data)

## Sample information:

- Observations:  $x = (1.2, 2.3, 1.8, 2.0, 3.1, 2.7, 3.5, 1.5, 2.8, 3.0, 1.9, 2.5)$
- Number of observations:  $n = 12$
- Sample range: Sample range =  $x_{\max} - x_{\min} = 3.5 - 1.2 = 2.3$
- Classes (Sturges, width  $\approx 0.6$ ):  $[1.2, 1.8], [1.8, 2.4], [2.4, 3.0], [3.0, 3.6]$
- Midpoints: 1.5, 2.1, 2.7, 3.3

## Frequency Distribution Table (with cumulative frequencies)

| Class        | $l_j$ | $n_j$ | $f_j$ | $N(x_j)$ | $F^*(x_j)$ |
|--------------|-------|-------|-------|----------|------------|
| $[1.2, 1.8]$ | 1.2   | 3     | 0.25  | 3        | 0.25       |
| $]1.8, 2.4]$ | 1.8   | 3     | 0.25  | 6        | 0.50       |
| $]2.4, 3.0]$ | 2.4   | 3     | 0.25  | 9        | 0.75       |
| $]3.0, 3.6]$ | 3.0   | 3     | 0.25  | 12       | 1.00       |
| Total        | –     | 12    | 1.00  | –        | –          |

# Frequency cumulative function for continuous variable: Example (Functions)

## Cumulative Absolute Frequency Function

$$N(x) = \begin{cases} 0, & x < 1.2 \\ 3, & 1.2 \leq x < 1.8 \\ 6, & 1.8 \leq x < 2.4 \\ 9, & 2.4 \leq x < 3.0 \\ 12, & x \geq 3.0 \end{cases}$$

## Cumulative Relative Frequency Function

$$F^*(x) = \begin{cases} 0, & x < 1.2 \\ 0.25, & 1.2 \leq x < 1.8 \\ 0.50, & 1.8 \leq x < 2.4 \\ 0.75, & 2.4 \leq x < 3.0 \\ 1.0, & x \geq 3.0 \end{cases}$$

# Cumulative Frequency Graph for a Continuous Variable: Properties

## Ogive (Cumulative Frequency Graph)

### Where does it start?

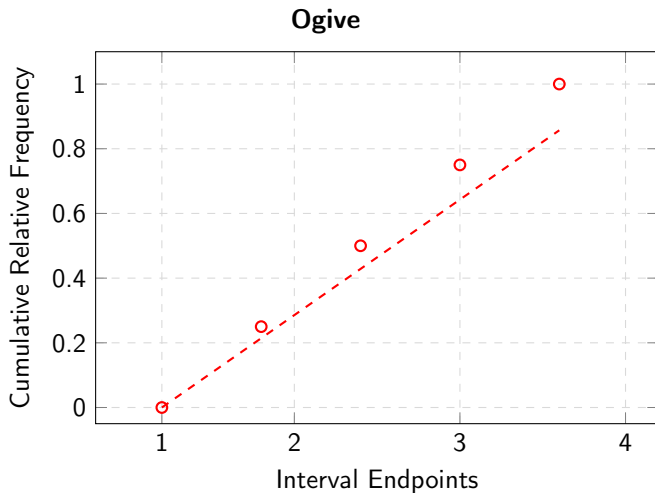
- It starts at the lower limit of the first class, with cumulative frequency = 0.
- The graph is built using the upper class limits on the x-axis and the cumulative frequencies (or percentages) on the y-axis.

### What is it used for?

- To show how data accumulate across classes.
- To identify the median, quartiles, deciles, and percentiles.
- To compare cumulative distributions.

*In this case, the ogive appears as a straight line rather than a curve, because the absolute and relative frequencies have the same values for all classes.*

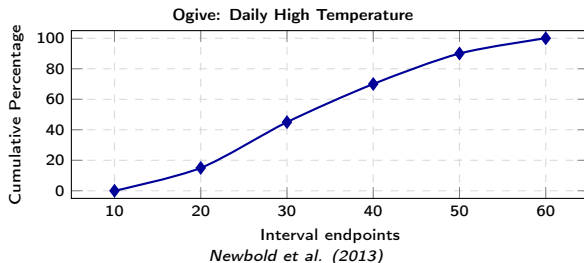
# Cumulative Frequency Graph for a Continuous Variable: Example



# The ogive graphing cumulative frequencies: other example

## Cumulative Relative Frequency Function

| Interval            | Upper endpoint | Cumulative Percentage |
|---------------------|----------------|-----------------------|
| Less than 10        | 10             | 0                     |
| 10 but less than 20 | 20             | 15                    |
| 20 but less than 30 | 30             | 45                    |
| 30 but less than 40 | 40             | 70                    |
| 40 but less than 50 | 50             | 90                    |
| 50 but less than 60 | 60             | 100                   |



# THANKS!

Questions?