

MANE 3332.04

attendance
1 - E

Lecture 17, March 31

Agenda

- Midterm not graded: still contacting students who missed exam
- Continue working on Technical Report One Assignment
- Chapter Six
- Attendance
- Questions?

Grades

25% mid-term

$$\text{O mid-term} \rightarrow -100(.25) = \underline{-25}$$

↓
could make 75

Schedule

Monday Lecture	Wednesday Lecture
3/31: Chapter 6	4/2: Chapter 5
4/7: Chapter 7 & 8	4/9: Chapter 8, Case 1
4/14: Chapter 8: Case 2	4/16: Chapter 8: Case 3
4/21: Chapter 9, case 1	4/23: Chapter 9, Case 2
4/28: Chpater 9, Case 3	4/30: Chapter 11
5/5: Chapter 11	5/7: Review

12 classroom sessions plus Final Exam

Handouts

- [Chapter 6 Slides](#)
- Chapter 6 Slides marked

Data Analysis

- 1) location of Data \rightarrow mean, median, mode
- 2) Variability / Spread \rightarrow variance / standard dev.
- 3) Shape of data



Numerical Summaries

- Called Descriptive Statistics in Chapter 6
 - Descriptive statistics help us understand the location or central tendency of data and the scatter or variability in data
 - Included in all statistical software packages, R does a good job calculating descriptive statistics

Central Tendency

- Ostle, et. al. (1996) define central tendency as “the tendency of sample data to cluster about a particular numerical value”
- Population mean

Greek letter $\rightarrow \mu$
 N - population size

$$\mu = \frac{1}{N} \sum_{i=1}^N x_i$$

- Sample mean

n is now lower-case

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

hat notation

$\hat{\mu}$ - est
estimator
of μ

- Sample median - middle value $\rightarrow \tilde{x}$
- Sample mode - most commonly occurring number(s)

$$\text{Range} = X_{\max} - X_{\min}$$

Measures of Variability

- There are several statistics that measure the variability or spread present in data
- Population variance

$$\sigma^2 = \frac{\sum_{i=1}^N (x_i - \mu)^2}{N} \quad \text{calculator: } \frac{\sigma_N}{N} \text{ or } \sigma_n$$

- Sample variance

$$s^2 = \hat{\sigma}^2 = \frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n - 1} \quad \text{calculator: } \sigma_{n-1}$$

- Shortcut (Computational) Formula

$$s^2 = \frac{\sum_{i=1}^n x_i^2 - \frac{(\sum_{i=1}^n x_i)^2}{n}}{n - 1}$$

- Standard deviation is often used because it is measured in the original units

$$\sigma = \sqrt{\sigma^2}; \quad s = \sqrt{s^2}$$

all columns

midterm has 3 columns

R Function Su

- R code
summary
- Output is fi

```
26 {r}
27 summary(midterm)
28 ^
```

28:4 # Import Dataset

Console Terminal Render Background Jobs 6 25

R 4.3.1 · /Volumes/SAMSUNG T7/wfscsBackup/Teaching2/AY_2023_2024/MANE3332_spring2024/PartTwo/li

```
> knitr::opts_chunk$set(echo = TRUE)
> library(readxl)
> midterm <- read_excel("/Volumes/NO NAME/midterm.xlsx")
> View(midterm)
> summary(midterm)
```

Participation	QuizAverage	MidtermExam
Min. : 2.941	Min. : 0.00	Min. :28.00
1st Qu.: 67.500	1st Qu.: 60.00	1st Qu.:59.75
Median : 87.941	Median : 80.00	Median :65.00
Mean : 77.096	Mean : 74.42	Mean :66.07
3rd Qu.: 95.147	3rd Qu.: 92.00	3rd Qu.:74.75
Max. :100.000	Max. :100.00	Max. :92.00
		NA's :5

```
> |
```

Q1 →

$x = Q_1$

Descriptive Statistics

R Function Su

- R code
- Output is fr

```
30 {r}
31 summary(midterm$MidtermExam)
32 {r}
```

32:4 Import Dataset R Markdown

Console Terminal Render Background Jobs

R 4.3.1 · /Volumes/SAMSUNG T7/wfscsBackup/Teaching2/AY_2023_2024/MANE3332_spring2024/PartTwo/

```
> summary(midterm)
```

Participation	QuizAverage	MidtermExam
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1st Qu.: 67.500	1st Qu.: 60.00	1st Qu.:59.75
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3rd Qu.: 95.147	3rd Qu.: 92.00	3rd Qu.:74.75
Max. :100.000	Max. :100.00	Max. :92.00
		NA's :5

```
> View(midterm)
> View(midterm)
summary(midterm$MidtermExam)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
28.00	59.75	65.00	66.07	74.75	92.00	5

specific column

Descriptive Statistics

R Functions

- Summary
- Descriptive
- Descriptive
- R Console
- library
- des

```
34 > library(psych)
35 describe(midterm)
37
```

Description: df [3 × 13]

	vars <dbl>	n <dbl>	\bar{x} mean <dbl>	s sd <dbl>	\tilde{x} median <dbl>	trimmed <dbl>	mad <dbl>
Participation	1	33	77.10	25.65	87.94	81.35	16.13
QuizAverage	2	33	74.42	23.49	80.00	77.48	23.72
MidtermExam	3	28	66.07	13.73	65.00	66.62	13.34

3 rows | 1-8 of 13 columns

37:4 Import Dataset R Markdown

imported library

mean
absolute
deviation

forecastly

- Psych package output from Spring 2024

$$\sigma^2 \sim (x - \mu)^2 \rightarrow 2^{\text{nd}} \text{ order}$$

3rd order
4th order

Description: df [3 × 13]

median	trimmed	mad	min	Max m...	range	skew	kurtosis	se
<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
87.94	81.35	16.13	2.94	100	97.06	-1.31	0.79	4.47
80.00	77.48	23.72	0.00	100	100.00	-1.24	1.28	4.09
65.00	66.62	13.34	28.00	92	64.00	-0.46	0.39	2.59

3 rows | 6-14 of 13 columns

Describe Output

Standard error
Chapter 5 8 & 9
s
on

Calculating Quantiles

$$x_p = x_{(i)} + d[x_{(i+1)} - x_{(i)}]$$

$$p(n+1) = i + d$$

36

2.3.2 Sample Quantiles

In Example 2.8, we consider an ogive for the plated bracket data. The point (1.55, 0.567) is on that ogive, so we estimate that 56.7% of the sampled population of brackets weighed at most 1.55 ounces. Weights associated with other percentages can also be estimated by locating an appropriate point on the ogive. In general, if the point (x, p) is on the ogive, we can use x as an estimate of the weight with 100% of the population values at or below it. This estimate, called the 100% sample quantile, is denoted x_p .

If two persons (or computer programs) use different mappings to obtain an ogive, the resulting quantiles will differ. To remedy this deficiency, an algebraic procedure is required.

THE 100% SAMPLE QUANTILE

Several definitions of sample quantiles are in use. We use the one that agrees with the default values output by the UNIVARIATE procedure in SAS[®]. Also, the definition used here is consistent with our definition of the sample median.

Suppose a sample of size n is obtained from some population associated with a continuous variable. For $0 < p < 1$, let $p(n+1) = i + d$, with i the integer part of $p(n+1)$ and $0 < d < 1$ the decimal part. If $1 \leq i < n$, and n is the 100% sample quantile is $x_{(i)}$. If $1 \leq i < n$ and $0 < d < 1$, interpolate linearly between $x_{(i)}$ and $x_{(i+1)}$. In either case, the 100% sample quantile is

$$x_p = x_{(i)} + d[x_{(i+1)} - x_{(i)}] \quad (2.4)$$

when $1 \leq i < n$. If $n = 0$ or n , the 100% sample quantile does not exist. If 100% is an integer, the corresponding quantile is called a *percentile*.

EXAMPLE 2.18

Suppose we want to find the 43rd percentile of the sample of plated weights in Table 2.1. Since

there are $n = 75$ observations in the sample and $p = 0.43$, we find $p(n+1) = (0.43)(75+1) = 32.68$. Letting $i = 32$ and $d = 0.68$, we use Equation (2.4) to obtain $x_{0.43} = x_{(32)} + (0.68)(x_{(33)} - x_{(32)})$. The 32nd ordered value in Figure 2.1(b) is $x_{(32)} = 1.50$ and the 33rd ordered value is $x_{(33)} = 1.51$. Thus, the 43rd percentile for these data is $x_{0.43} = 1.50 + (0.68)(1.51 - 1.50) = 1.5068 \approx 1.507$. Using this as a point estimate of the population percentile, we can say that approximately 43% of the plated brackets produced on the day the data were collected had weights of 1.507 ounces or less.

The Sample Median is a Percentile

Suppose we want to find the 50th percentile and the data set contains n values. When n is even, $(0.50)(n+1) = (n/2) + (0.50)$, with $n/2$ a positive integer. Using Equation (2.4) with $i = n/2$ and $d = 0.50$, $x_{0.50} = x_{(i)} + (0.50)[x_{(i+1)} - x_{(i)}] = [x_{(i)} + x_{(i+1)}]/2$. When n is odd, $(0.50)(n+1) = (n+1)/2$, with $(n+1)/2$ a positive integer. Using Equation (2.4) with $i = (n+1)/2$ and $d = 0$, we find $x_{0.50} = x_{(i)}$. But, this is precisely how the sample median was defined. Thus, $x = x_{0.50}$.

SAMPLE QUANTILES

The percentiles $x_{0.25}$, $x_{0.50}$, and $x_{0.75}$ are known as the *first*, *second*, and *third sample quantiles*, respectively. These quantiles are often denoted q_1 , q_2 , and q_3 .

EXAMPLE 2.18

Consider the plated bracket weights in Table 2.1. Using the ordered stem-and-leaf display presented in Figure 2.1(b), we find the following:

- First Quartile:** Since $(0.25)(75+1) = 19$, $q_1 = x_{0.25} = x_{(19)} = 1.46$.
- Second Quartile (Median):** Since $(0.50)(75+1) = 38$, $q_2 = x = x_{0.50} = x_{(38)} = 1.53$.

reference for calculating quantiles

1) Sort data

2) find $p = .5$ (median)

$x_1 = 6, x_2 = 4, \dots, x_8 = 6 \rightarrow$ order in data

$x_{(1)} = 3, x_{(2)} = 4, x_{(3)} = 4, \dots, x_{(8)} = 7$

Qu

8 observations from binomial distribution with

[REDACTED]

sorted

6, 4, 5, 7, 3, 5, 4, 6 \rightarrow 3, 4, 4, 5, 5, 6, 6, 7

\uparrow \uparrow
 $x_{(4)}$ $x_{(5)}$

Quantile Example

$$p = .5, n = 8$$

$i \rightarrow$ integer

$$p(n+1) = i + d$$

$d \rightarrow$ decimal remainder

$$.5(8+1) = \underline{4.5}$$

$$\rightarrow i = 4, d = .5$$

$$X_{.5} = X_{(4)} + .5[X_{(5)} - X_{(4)}]$$

$$= 5 + .5[5 - 5]$$

$$= \underline{5}$$

What am I doing?

linear interpolation

② add to note card for final

Exploratory Data (Graphical) Analysis

- Exploratory data analysis (EDA) is the use of graphical procedures to analyze data.
- John Tukey was a pioneer in this field and invented several of the procedures
- Tools include stem-and-leaf diagrams, box plots, time series plots and digidot plots *dot plots*

look at graph and
get data values

Stem and Leaf Diagram

- Excellent tool that maintains data integrity
- The stem is the leading digit or digits
- The leaf is the remaining digit
- Make sure to include units
- R Code

```
stem(midterm$MidtermExam)
```

Stem and

The decimal point is 1 digit(s) to the right of the

units

• Round

2 | 8
3 |
4 | 4
5 | 11566
6 | 13334446679
7 | 2247
8 | 00147
9 | 2

Stem - 7
leaf - 4
units $\rightarrow 7(10)$
value $7(10) + 4 = 74$

Stem and Leaf Plot of Midterm Exam Scores



Histogram

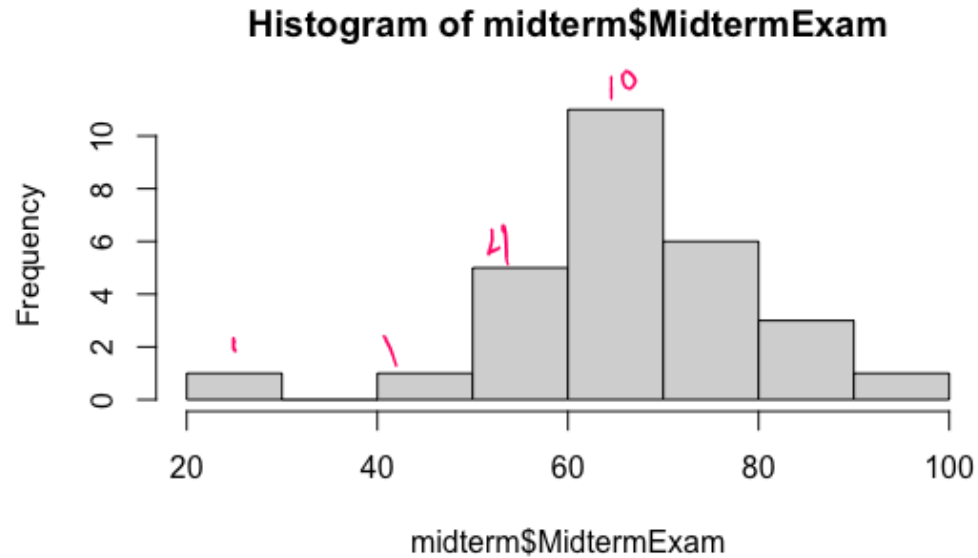
- A histogram is a barchart displaying the frequency distribution information
- There are three types of histograms: frequency, relative frequency and cumulative relative frequency
- R code

```
hist(midterm$MidtermExam)
```

→ needs lots of data to get shape

Histogram E

- R output



Histogram of Midterm Exam Scores

estimators to be robust \rightarrow not overly influenced by extreme values

Boxplot

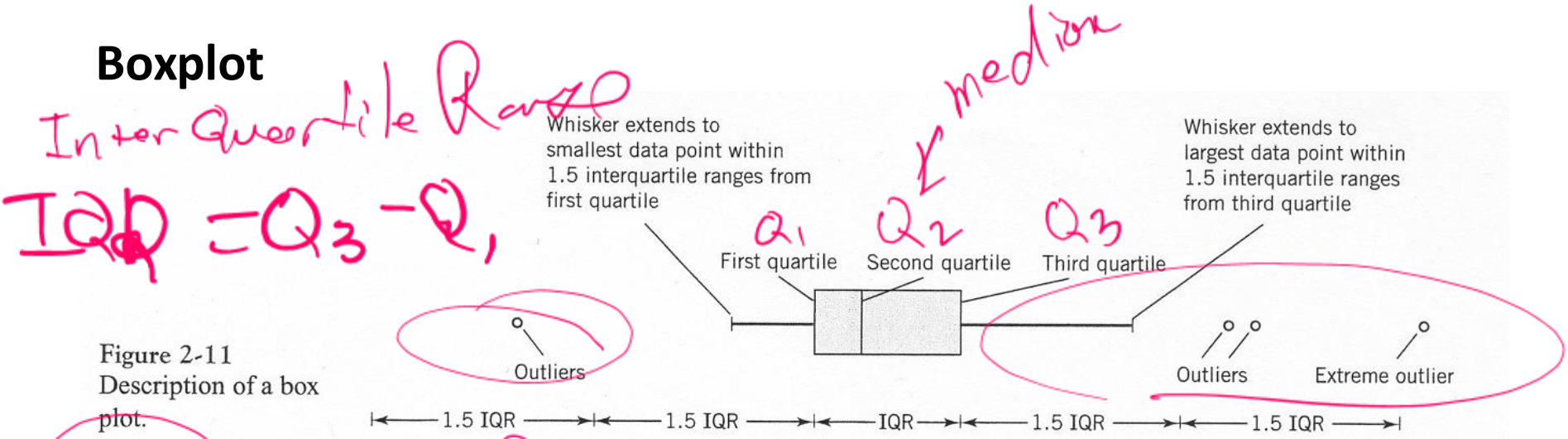


Figure 2-11
Description of a box plot.

Box Plot 1

Box plot with explanation

Box & whiskers

Samuel Stastich

Statistician

Q₁: which plant has highest Quality index? Plant 1 has highest Q₁ ✓

Box Plot 2

Q₂: what is difference in Q₁ between plants 2 & 3? by location same median, so no difference

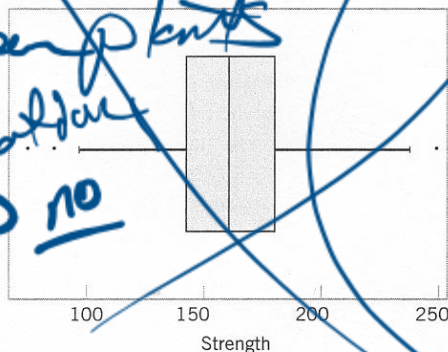


Figure 2-12 Box plot for compressive strength data in Table 2-2.

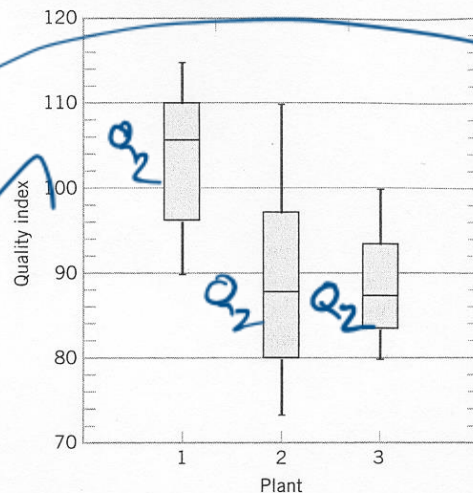


Figure 2-13 Comparative box plots of a quality index at three plants.

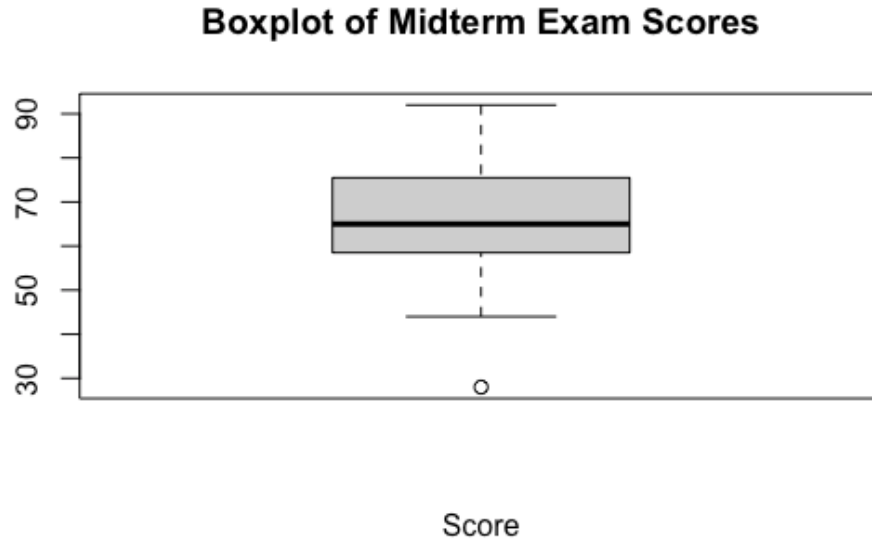
examples of boxplots

Q₃: which plant has most variability?
Q₂ → rectangle is largest & whiskers are longest

Box Plot 3

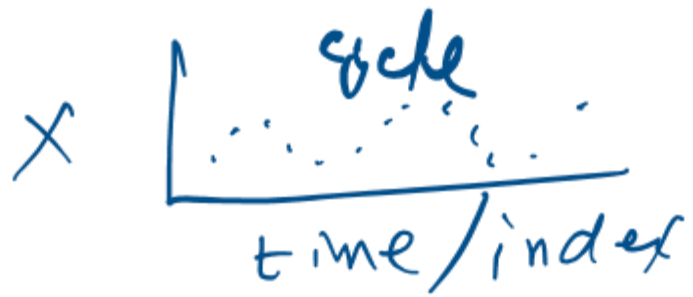
- R code for l
- R Box Plot c

```
boxplot  
ain='Box
```



```
:'Score',m  
es')
```

Boxplot of Midterm Exam Scores

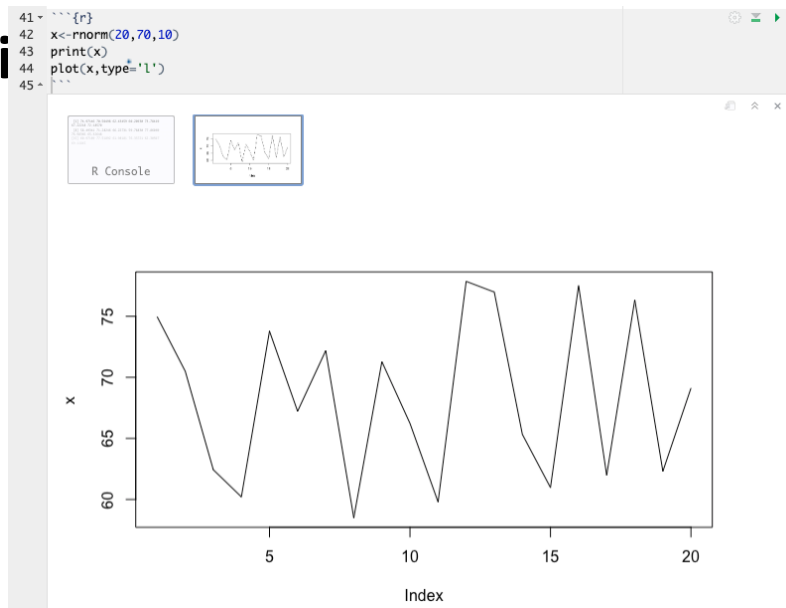


Time Series Plot

- A **time series plot** is a graph in which the vertical axis denotes the observed value of the variable (say x) and the horizontal axis denotes time
- Excellent tool for detecting:
 - trends, ✓
 - cycles, ✓
 - other non-random patterns



Time Series Plot i



Time Series Plot

Probability Plotting

- **Probability plotting** is a graphical method of determining whether sample data conform to a hypothesized distribution
- Used for validating assumptions
- Alternative to hypothesis testing

Construction

1. Sort the data from smallest to largest, .
2. $x_{(1)}, x_{(2)}, \dots, x_{(n)}$
3. Calculate the observed cumulative frequency $(j - 0.5)/n$

For the normal distribution find z_j that satisfies

$$\frac{j - 0.5}{n} = P(Z \leq z_j) = \Phi(z_j)$$

3. Plot z_j versus $x_{(j)}$ on special graph paper

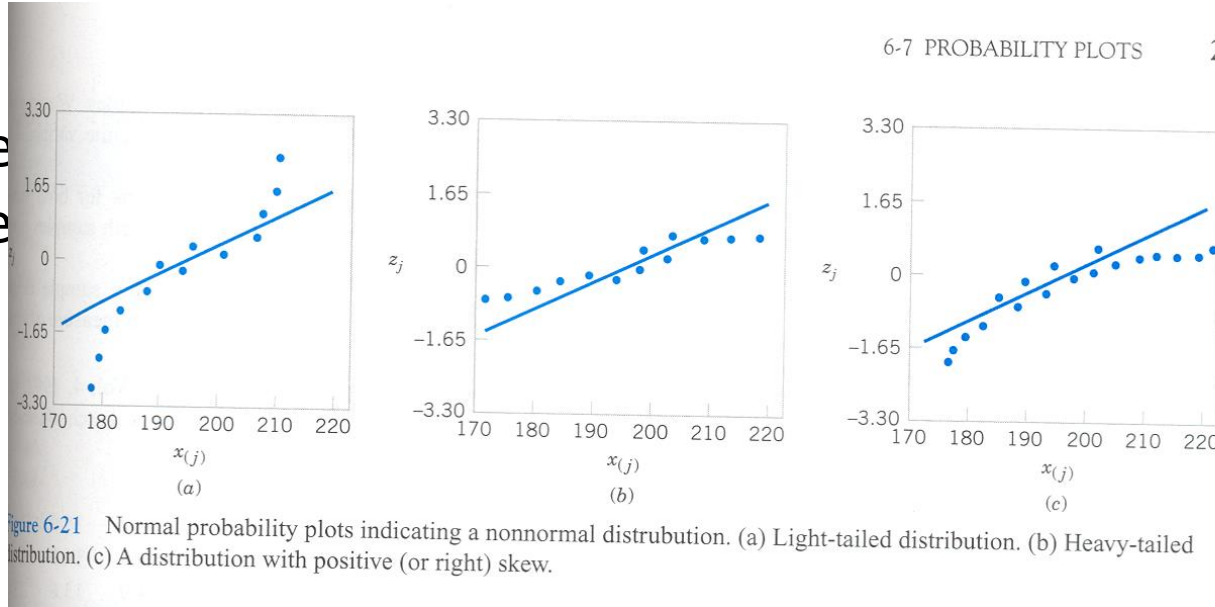
ignore this ancient history

replaced by computer graphics

weakness: Subjectivity

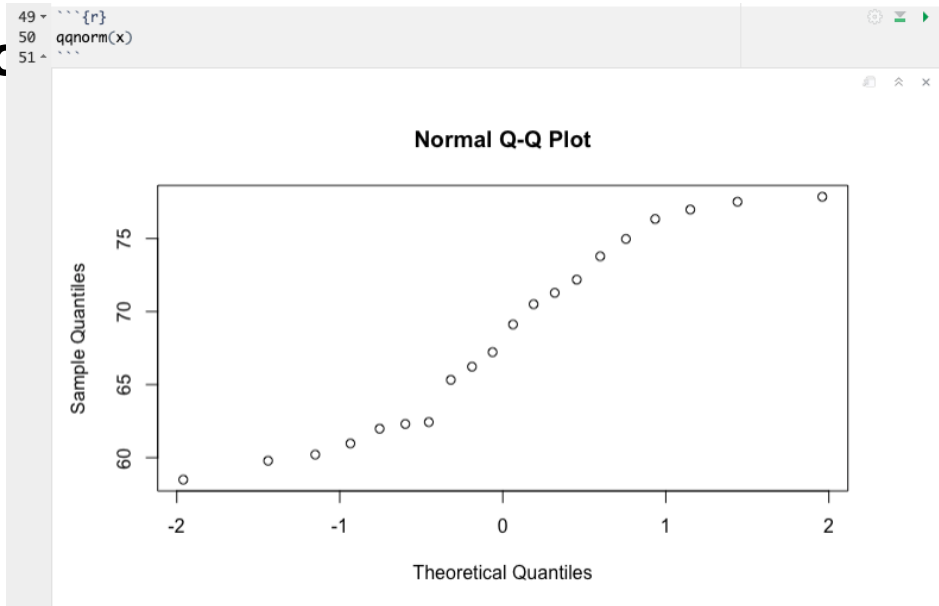
Usage

- If the correlation is



normal probability plots from textbook, figure 6.21 on page 215

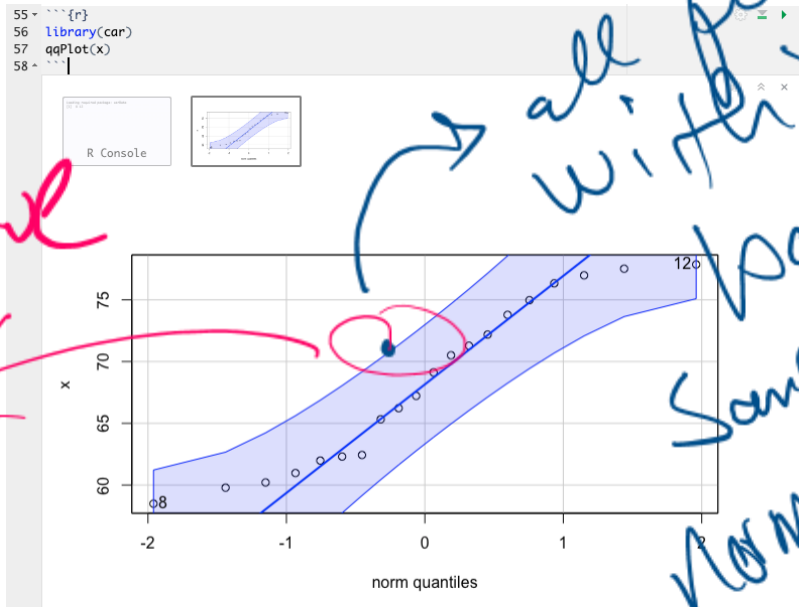
Probability Plots



Normal Probability Plot

Probability Plot Example 2

- Difficulty from example one is how close to straight is “good enough”
- Add confidence bands to normal probability plot
 - Requires package car to be added to R
 - If all points are within the band, we are 95% confident that the sample is from a normal distribution. However if one or more points are not within band, the data is not from a normal distribution



this would
be subjective

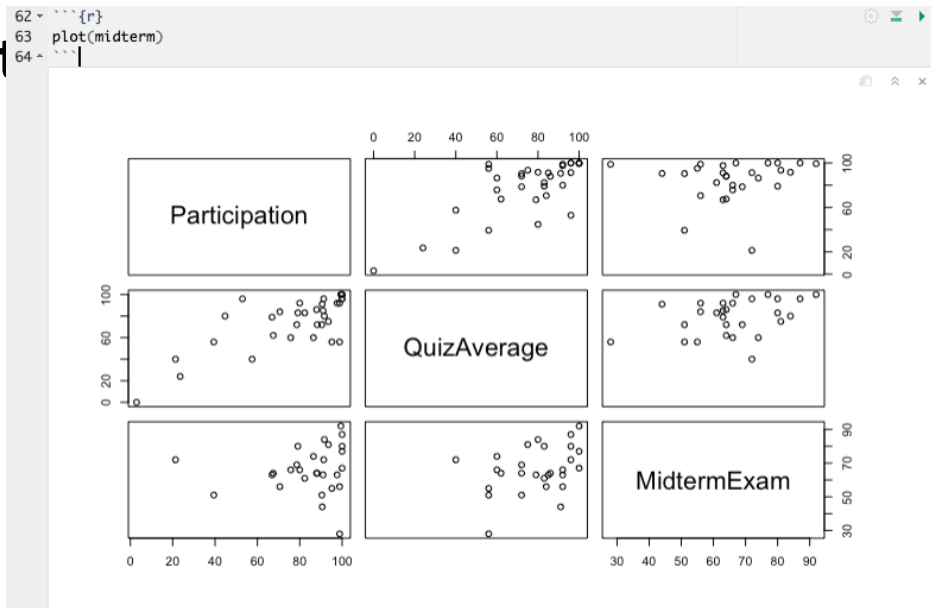
all points
within the
band so
sample is
normal

QQ Plot with band

not cover
need Chapter 5

Multivariate Data

Matrix of Scatter



Scatter Plots

Covariance in R

```
67 ~ ``{r}  
68 midterm_NA <- na.omit(midterm)  
69 print(cov(midterm_NA))  
70 ~ ``
```

	Participation	QuizAverage	MidtermExam
Participation	340.16778	193.7847	28.75699
QuizAverage	193.78474	269.0899	81.17460
MidtermExam	28.75699	81.1746	188.43915

Covariance Matrix

Correlation

```
74 ~~~{r}  
75 print(cor(midterm_NA))  
76 ~~~
```

	Participation	QuizAverage	MidtermExam
Participation	1.0000000	0.6405076	0.1135825
QuizAverage	0.6405076	1.0000000	0.3604839
MidtermExam	0.1135825	0.3604839	1.0000000

Correlation Matrix