

# 3Y03 - PROBABILITY AND STATISTICS FOR ENGINEERING

WS19

Lecture 19

Last Time

## Box (& Whisker) Plots

OUTLIERS

↓  
X X



$q_1$       m       $q_3$

Interquartile Range

$1.5 \times IQR$

$IQR = q_3 - q_1$

$1.5 \times IQR$

$3 \times IQR$

$3 \times IQR$

OUTLIERS

↓      ↓  
X      X X

EXTREME  
OUTLIER

$m$  = median = middle value ( $n$  odd) or average of 2 middle values ( $n$  even)

$q_1$  = 1st quartile = median of the values below  $m$

$q_3$  = 3rd quartile = median of the values above  $m$ .

$q_3 = 50.5$

Example Data Set 1

$n = 12$

10    11    18    19    23    31    33    39    50    51  
 $q_1 = 18.5$                    $m = 32$                    $72 \quad 105$

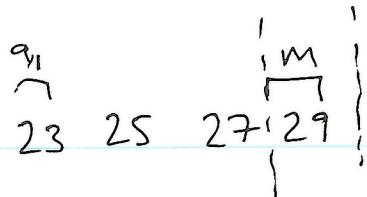
$$IQR = q_3 - q_1 = 50.5 - 18.5 = 32$$

$$1.5 \times IQR = 1.5 \times 32 = 48$$

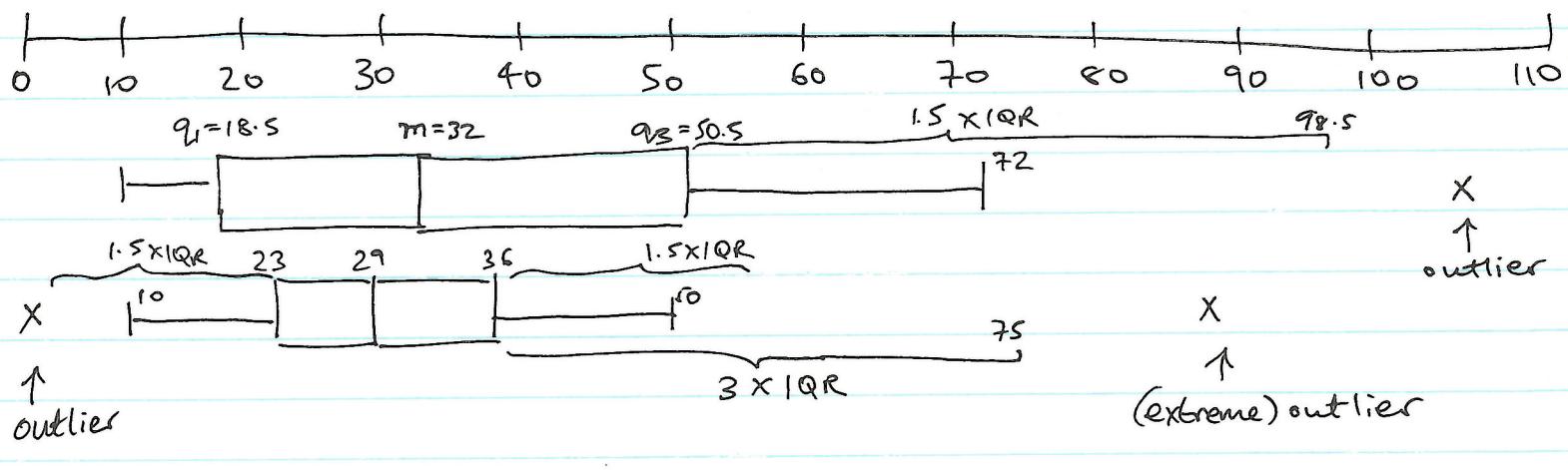
$$3 \times IQR = 3 \times 32 = 96$$

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Data Set #2  $n=11$  1 10 23 25 27 29 30 35



$IQR = Q_3 - Q_1 = 30 - 10 = 20$   
 $1.5 \times IQR = 1.5 \times 20 = 30$



## 6.7 Probability Plot

- histograms indicate underlying distribution
- in some situations you want to assume a certain distr.

We hypothesize a prob. distr. (with pdf  $f(x)$  & cdf  $F(x)$ ) & check hypothesis with a prob. plot.

Sample  $\{x_1, \dots, x_n\}$

Rank / Order the sample & rename:  ~~$x_{(1)} \leq x_{(2)} \leq \dots \leq x_{(n)}$~~

Idea  $x_{(j)}$  (jth observation in the list) should approximate the  $100 \left(\frac{j}{n}\right)$  th percentile

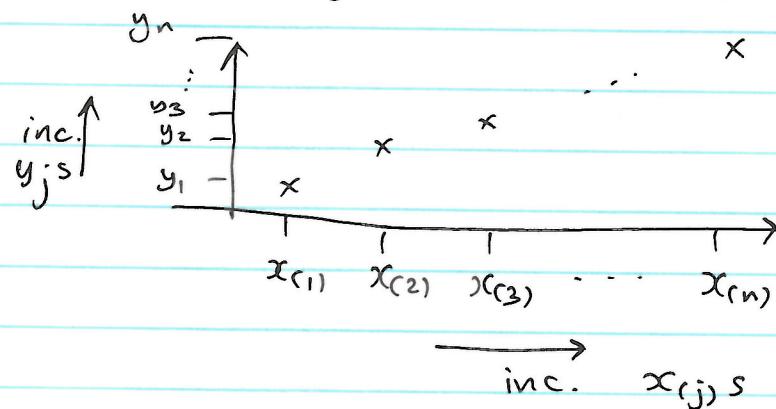
Want  $y_j$  s.t.

$$P(X \leq y_j) = F(y_j) = \frac{j-0.5}{n}$$

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Plot  $y_j$ 's against  $x_{(j)}$ 's

— should be a straight line  
if our hypothesis is correct

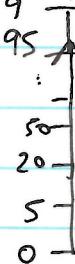


Subjective judgment as to  
whether the line is  
straight or not!

Usually for us the hypothesis will be Normal distr.

i.e.  $F(x) = \Phi(x)$  & values  $y_j$  are  $z_j$ -values from  
(reverse) normal table.

There is "normal prob. paper" with



— get to plot  $\frac{j-0.5}{n}$   
against  $x_{(j)}$ 's.

Still the case that straight  
line = good hypothesis.

Even if line is not straight, plot can still indicate  
features of prob. distr. e.g. Symmetry / skew — see textbook  
diagrams

## 7. Point Estimation of Parameters

Recall: Sample  $\{x_1, \dots, x_n\}$  (data)

thought of as realizations of independent r.v.s.  
with common underlying distribution  $\{X_1, \dots, X_n\}$   
(also called random sample)

$\{x_1, \dots, x_n\}$  → there is one set of possible values of  
 $\{X_1, \dots, X_n\}$  with some samples more likely ~~less~~ depending  
on prob. distr.

2 big concepts of statistical analysis/inference:

- { ① Point Estimation = parameter estimation
- ② Hypothesis Testing

Parameter : some feature/function of underlying distr.

↑  
usually e.g. pop. mean  $\mu$ , pop. variance  $\sigma^2$

little Greek letter: default is  $\theta$

A statistic is a function of  $\{X_1, \dots, X_n\}$

(Remember - any function of r.v.s. is also a r.v.)

e.g.  $\bar{X}$  = sample mean  $S^2$  = sample variance

The prob. distr. of a statistic is called a sampling distribution

Estimation Given a parameter  $\theta$

- an estimator is a statistic used to estimate  $\theta$

↳ notation often  $\hat{H} \left[ = h(X_1, \dots, X_n) \right]_{\text{some } h}$

e.g.  $\bar{X} = \frac{1}{n}(X_1 + \dots + X_n)$  an estimator for  $\mu$

$S^2 = \frac{1}{n-1}(X_1^2 + \dots + X_n^2 - n\bar{X}^2)$  an estimator for  $\sigma^2$

After we take a sample  $\{x_1, \dots, x_n\}$  & calculate a value for  $\hat{H}$ , ~~this~~ the value it takes is called the point estimate for  $\theta$ , denoted  $\hat{\theta}$

e.g.  $\bar{x}$  is a point estimate for  $\mu$

$$S^2$$

$$\sigma^2$$