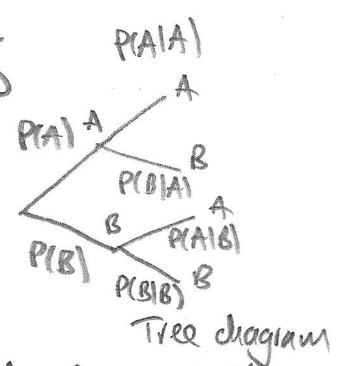


If  $A_1, A_2 \dots A_n$  are  $n$  independent events  
 $P(A_1 \& A_2 \& A_3 \dots A_n) = \prod_{i=1}^n P(A_i)$

MAX  $P(A_i) = 1$   
 MIN  $P(A_i) = 0$



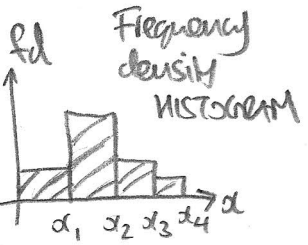
$P(A \& B) = P(A)P(B|A) = P(B)P(A|B)$  conditional probability

**BAYES THEOREM**  
 If  $A_1 \dots A_n$  are mutually exclusive events  
 $P(A_1 \text{ OR } A_2 \text{ OR } \dots A_n) = \sum_{i=1}^n P(A_i)$

# Statistics 1

## Probability

See over \* (Probability distributions)



$fd_j = \frac{\text{frequency}_j}{\alpha_j - \alpha_{j-1}}$  ←  $\alpha_{j-1} \leq \alpha < \alpha_j$

## Data representation

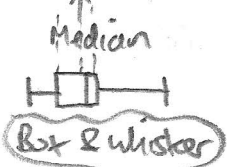
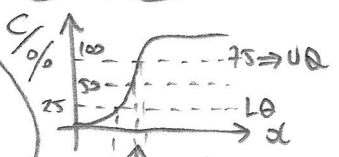
### Stem & leaf

"Back to Back"

1	3	10	6	7
	2	11	1	1
6	4	4	12	2
9	8	8	13	0
			14	3

7/13 means 13.7 or 137  
 13/3 " 13.3 or 133  
 i.e. leaf is last digit.

## Cumulative frequency



Inter-quartile range = UQ - LQ

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

For a frequency table, estimate  $\bar{x}$  and  $\sigma$  using mid range  
 i.e.  $\frac{x_i + x_{i-1}}{2}$

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

↑  $f_i$  instances of each  $x_i$  (frequency)

## Permutations and combinations

# combinations (order not important) of  $r$  objects from  $n$  objects  
 $= {}^n C_r$

# different permutations of  $r$  objects which can be made from  $r$  distinct objects is

${}^n P_r = \frac{n!}{(n-r)!}$

# permutations of  $n$  distinct objects is  $n!$

# distinct permutations of  $n$  objects  
 -  $p$  are identical  
 -  $q$  or remainder are "  
 -  $r$  " " are "

$= \frac{n!}{p! q! r! \dots}$



## Correlation and Regression

Covariance of  $x_i, y_i$   
 $C = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{n}$

Product moment correlation coefficient

+1 or +ve  
 -1 or -ve  
 0 No correlation

not changed by linear transform of  $x, y$

**Standard deviation** =  $\sqrt{\text{Variance}}$

$\sigma_x = \sqrt{\frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n}}$  ↓  $\sigma^2$

$\sigma_x = \sqrt{\frac{\sum_{i=1}^n (f_i x_i - \bar{x})^2}{\sum_{i=1}^n f_i}}$

Spearman's rank correlation coefficient

$$r_s = 1 - \frac{6 \sum_{i=1}^n d_i^2}{n(n^2-1)}$$

Similar range -1 to 1 (product moment).

$d_i$  = difference between the RANKS for the  $i$ th item  $x_i - y_i$

$x_i$	$y_i$
A 5	3
B 4	2
C 2	5
D 1	4
E 3	1

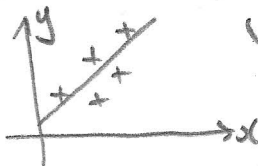
Rank correlation

Linear Regression

HARDEST QUESTIONS ARE PROBABLY COMBINATORICS - PRACTICE!

"least squares regression"

"line of best fit"



Probability of Event E  
=  $\frac{\text{\# of outcomes which result in E}}{\text{total \# of possible outcomes}}$

More Statistics 1

i.e. minimize  $\sum_{i=1}^n (y_i - \bar{y})^2$

i.e. minimize  $\sum_{i=1}^n (x_i - \bar{x})^2$

"y on x" (vertical)  
 $y = a + bx$   
 $a = \bar{y} - b\bar{x}$   
 $b = \frac{C}{\sigma_x^2} = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})/n}{\sum_{i=1}^n (x_i - \bar{x})^2/n}$

"x on y" (horizontal)  
 $y = a + bx$   
 $a = \bar{x} - b\bar{y}$   
 $b = \frac{C}{\sigma_y^2}$

[Note  $S_{xx} = \sigma_x^2 n$ ,  $S_{yy} = \sigma_y^2 n$   
 $S_{xy} = cn$ ]

Discrete

Probability Distributions

$X \sim D(a, b, \dots)$

"X is a random variable distributed by D, parameterised by a, b, ..."

Expectation

$$E[X] = \mu = \sum_{i=1}^n x_i p_i$$

$$p_i = P(X = x_i)$$

$$\text{Var}[X] = \sigma^2 = \sum_{i=1}^n (x_i - \mu)^2 p_i = \sum_{i=1}^n x_i^2 p_i - \mu^2$$

BINOMIAL

$$\binom{n}{x} p^x (1-p)^{n-x}$$

$np$

$np(1-p)$

$X \sim B(n, p)$

POISSON<sup>+</sup>

$$e^{-\lambda} \frac{\lambda^x}{x!}$$

$\lambda$

$\lambda$

$X \sim P(\lambda)$

GEOMETRIC

$$p(1-p)^{x-1}$$

$p^{-1}$

$\frac{1-p}{p^2}$

$X \sim G(p)$

$p$  is prob. of success of a trial.  
 $X = \#$  of successes in  $n$  trials

$X = \#$  of events in given time interval

$p$  is prob. of success of a trial  
 $X = \#$  of trials till first success.

Probability  $X=x$  ↑

MEAN

STANDARD

+ Stats 2, 3