

# OCR Further Maths AS-level

## Statistics

### Formula Sheet

Provided in formula book

Not provided in formula book

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## Probability

### Permutations and Combinations

Number of permutations of $n$ distinct objects	$n! = n \times (n - 1) \times (n - 2) \dots \times 2 \times 1$
Number of combinations when choosing $r$ objects from $n$ objects	$\binom{n}{r} = {}^n C_r = \frac{n!}{r!(n-r)!}$
Number of permutations of a subset of size $r$ from a set of $n$ distinct objects	${}^n P_r = {}^n C_r \times r! = \frac{n!}{(n-r)!}$
Number of permutations from $n$ objects with $r_A$ of type $A$ , $r_B$ of type $B$ etc.	$\frac{n!}{r_A! r_B! \dots}$

### Probability Problems

$n(A)$	The number of ways of making a choice about $A$
Product principle	$n(A \text{ and } B) = n(A) \times n(B)$
Addition principle (given that $A$ and $B$ are mutually exclusive)	$n(A \text{ or } B) = n(A) + n(B)$
Counting principles in probability (given that all outcomes are equally likely)	$P(A) = \frac{\text{number of outcomes in which } A \text{ occurs}}{\text{total number of possible outcomes}}$



## Discrete Random Variables

### Average and Spread of Discrete Random Variables

For the random variable taking the values $x_i$ with $P(X = x_i) = p_i$	
Expectation	$\mu = E(X) = \sum x_i p_i$
Variance	$\begin{aligned} \sigma^2 = \text{Var}(X) &= \sum (x_i - \mu)^2 p_i \\ &= \sum x_i^2 p_i - \mu^2 \\ &= E(X^2) - (E(X))^2 \end{aligned}$

### Linear Coding with the Mean and Variance

$$\begin{aligned} \text{For } Y &= aX + b: \\ E(Y) &= aE(X) + b \\ \text{Var}(Y) &= a^2 \text{Var}(X) \end{aligned}$$

### Binomial Distribution

For $X \sim B(n, p)$
$P(X = x) = \binom{n}{x} p^x (1-p)^{n-x}$
$E(X) = np$
$\text{Var}(X) = np(1-p)$

### Discrete Uniform Distribution

For $X \sim U(n)$
$P(X = x) = \frac{1}{n} \text{ for } x = 1, 2, \dots, n$
$E(X) = \frac{n+1}{2}$
$\text{Var}(X) = \frac{n^2 - 1}{12}$



### Geometric Distribution

For  $X \sim \text{Geo}(p)$

$$P(X = x) = p(1 - p)^{x-1}$$

for  $x = 1, 2, 3 \dots$

$$P(X > x) = (1 - p)^x$$

$$E(X) = \frac{1}{p}$$

$$\text{Var}(X) = \frac{1 - p}{p^2}$$

### Poisson Distribution

For  $X \sim \text{Po}(\lambda)$

$$P(X = x) = \frac{e^{-\lambda} \lambda^x}{x!} \text{ for } x = 0, 1, 2 \dots$$

$$E(X) = \lambda$$

$$\text{Var}(X) = \lambda$$

$$\begin{aligned} X &\sim \text{Po}(\lambda) \\ Y &\sim \text{Po}(\mu) \end{aligned}$$

$$\begin{aligned} \text{When } Z &= X + Y, \\ Z &\sim \text{Po}(\lambda + \mu) \end{aligned}$$



## Chi-squared Tests

### Contingency Tables

Expected value in cell $i$	$E_i = \frac{\text{row total} \times \text{column total}}{\text{overall total}}$
Chi-squared value	$\chi_{\text{calc}}^2 = \frac{\sum(O_i - E_i)^2}{E_i}$
Degrees of freedom in an $n \times m$ contingency table	$v = (n - 1)(m - 1)$

### Hypothesis Testing

If variables are independent and

$E_i > 5$  for all  $i$

$$\chi_{\text{calc}}^2 = \frac{\sum(O_i - E_i)^2}{E_i} \approx \chi_v^2$$

### Yates' Correction

When  $v = 1$

$$\chi_{\text{Yates}}^2 = \frac{\sum |(O_i - E_i) - 0.5|^2}{E_i}$$

### Goodness of Fit Test

$v = \text{number of bins} - \text{number of constraints}$



## Correlation

### Pearson's Product Moment Correlation Coefficient

For a set of bivariate data with variables  $X$  and  $Y$

$$r = \frac{\sum x_i y_i - \frac{\sum x_i \sum y_i}{n}}{\sqrt{\left(\sum x_i^2 - \frac{(\sum x_i)^2}{n}\right) \left(\sum y_i^2 - \frac{(\sum y_i)^2}{n}\right)}}$$

$$= \frac{S_{xy}}{\sqrt{S_{xx} S_{yy}}}$$

$$S_{xx} = \sum (x_i - \bar{x})^2 = \sum x_i^2 - \frac{(\sum x_i)^2}{n}$$

$$S_{yy} = \sum (y_i - \bar{y})^2 = \sum y_i^2 - \frac{(\sum y_i)^2}{n}$$

$$S_{xy} = \sum (x_i - \bar{x})(y_i - \bar{y}) = \sum x_i y_i - \frac{\sum x_i \sum y_i}{n}$$

### Spearman's Rank Correlation Coefficient

$$r_s = 1 - \frac{6 \sum d^2}{n(n^2 - 1)}$$

where  $d$  = difference in ranks and  $n$  = number of data pairs

## Linear Regression

### Least Squares Regression Line

$$y = ax + b$$

$$a = \bar{y} - b\bar{x}$$

$$b = \frac{S_{xy}}{S_{xx}} = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2}$$

