# Mathematics HL

# 1 Algebra

## 1.1 Sequences and Series

Arithmetic progressions

- $T_n = U_n = a + (n-1)d$ .
- A sequence is an A.P if  $T_n T_{n-1} = d = \text{constant}$ .
- $S_n = \frac{n}{2}(a+l) = \frac{n}{2}(2a+(n-1)d).$
- $\bullet \ T_n = S_n S_{n-1}.$

Geometric progressions

- $\bullet \ T_n = ar^{n-1}.$
- A sequence is a G.P if  $\frac{T_n}{T_{n-1}} = r = \text{constant.}$
- $\bullet S_n = \frac{a(1-r^n)}{1-r}.$
- $|r| < 1 \implies S_{\infty} = \frac{a}{1-r}$ .
- $|r| > 1 \implies$  divergent.

#### 1.2 Summation

For  $\sum_{r=m}^{n} u_r$ , the number of terms is (n-m+1).

$$\sum_{r=1}^{n} (x_r \pm y_r) = \sum_{r=1}^{n} x_r \pm \sum_{r=1}^{n} y_r$$

$$\sum_{r=1}^{n} k u_r = k \sum_{r=1}^{n} u_r$$

$$\sum_{r=m}^{n} u_r = \sum_{r=1}^{n} u_r - \sum_{r=1}^{m-1} u_r$$

Useful sums:

$$\sum_{r=1}^{n} r = \frac{1}{2}n(n+1)$$

$$\sum_{r=1}^{n} r^2 = \frac{1}{6}n(n+1)(2n+1)$$

$$\sum_{r=1}^{n} r^{3} = (\sum_{r=1}^{n} r)^{2} = \frac{1}{4} n^{2} (n+1)^{2}$$

## 1.3 Permutations and combinations

$${}^{n}C_{r} = \frac{n!}{r!(n-r)!} \qquad {}^{n}P_{r} = {}^{n}C_{r} \cdot r!$$

If m objects are identical and the remaining are distinct (a total of n objects), permutations =  $\frac{n!}{m!}$ 

#### 1.4 The Binomial Thoerem

$$(1+x)^n = 1 + nx + \frac{n(n-1)}{2!}x^2 + \frac{n(n-1)(n-2)}{3!}x^3 + \cdots$$

#### 1.5 Mathematical induction

- 1. Let  $P_n$  be the statement: *ello* for all  $n \in \mathbb{Z}^+$ .
- 2. For n = 1: LHS = something. RHS = something  $\implies P_1$  is true.
- 3. Assume  $P_k$  is true for some  $k \in \mathbb{Z}^+$ .
- 4. Showing that  $P_{k+1}$  is true: it is true!
- 5. Since  $P_1$  is true, and  $P_k$  is true  $\implies P_{k+1}$  is true, by Mathematical Induction,  $P_n$  is true for all  $n \in \mathbb{Z}^+$ .

To do the inductive step:

$$\bullet \, \frac{d^{k+1}y}{dx^{k+1}} = \frac{d}{dx} (\frac{d^ky}{dx^k})$$

 $\bullet$  For divisibility, let the expression = a multiple of m. You can always rearrange the inductive hypothesis.

# 2 Functions and equations

- A function is a to-one relationship.
- If the vertical line x = a cuts the graph at one point only, then f is a function. If it cuts more than once, give an example.
- If a function passes the horizontal line test, it will have an inverse.
- The inverse is just a reflection of the graph in the line y = x.
- For inverse functions,  $R_f = D_{f^{-1}}$  and  $D_f = R_{f^{-1}}$ .
- For gf to exist,  $R_f \subseteq D_g$ .
- $D_{gf} = D_f$ .
- $R_{qf} = R_q | (D_q = R_f).$
- $(g \circ f)^{-1}(x) = (f^{-1} \circ g^{-1})(x)$ .
- $ff^{-1}(x)$  may not necessarily intersect with  $f^{-1}f(x)$ , it depends on the domain.
- For a periodic function, f(x) = f(x+c).

## 2.1 Graphs

- To transform, TSST. (translate and stretch)x then (translate and stretch)y.
- For y = |f(x)|, retain  $y \ge 0$ , then reflect y < 0.
- For y = f(|x|), retain  $x \ge 0$ , then reflect  $x \ge 0$  to the left of the x-axis.
- $\bullet$  For each transformation, you're allowed to replace x by something else.

## 2.2 Polynomials

- For a polynomial of degree n:
  - The sum of individual roots =  $-\frac{a_{n-1}}{a_n}$
  - The sum of (choose 2) roots =  $\frac{a_{n-2}}{a_n}$
  - The sum of (choose 3) roots =  $-\frac{a_{n-3}}{a_n}$
  - The product of roots, i.e the sum of (choose n) roots =  $(-1)^n \frac{a_0}{a_n}$
- For the special case of a quadratic:  $\alpha + \beta = -\frac{b}{a}$ ,  $\alpha\beta = \frac{c}{a}$
- $\bullet$  A polynomial of degree n has a maximum of n roots, but some of these may be complex.

## 2.3 Circular functions and Trigonometry

- The ambiguous case of the sine rule occurs when the angle you are trying to find is opposite the longest side.
- $\sin(-\theta) = -\sin\theta$   $\tan(-\theta) = -\tan\theta$  (odd functions).
- $\cos(-\theta) = \cos\theta$  (even function).
- For  $\pi \pm \theta$  or  $2\pi \pm \theta$ : sin-sin, cos-cos, tan-tan.
- For  $\frac{\pi}{2} \pm \theta$  or  $\frac{3\pi}{2} \pm \theta$ : sin-cos, cos-sin, tan-cot.
- $\tan x = \cot(\frac{\pi}{2} x)$ .
- $\sec x = \csc(\frac{\pi}{2} x)$ .
- The domain of  $\arcsin x$  and  $\arccos x$  are [-1,1].
- $\cos(\arcsin x) = \sin(\arccos x) = \sqrt{1 x^2}$ .
- A circle with centre (h, k) and radius r is described by:

$$(x-h)^2 + (y-k)^2 = r^2$$

• To simplify an expression with trig, it may help to use the half angle formula.

$$\frac{\sin\theta}{1+\cos\theta} = \frac{2\sin\frac{\theta}{2}\cos\frac{\theta}{2}}{1+2\cos^2\frac{\theta}{2}-1} = \tan\frac{\theta}{2}$$

## 2.4 Systems of equations

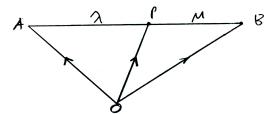
• A system of equations can be written as an augmented matrix:

$$2x + 3y + 4z = 2 
3x - 2y + z = -3 \rightarrow \begin{pmatrix} 2 & 3 & 4 & 2 \\ 3 & -2 & 1 & -3 \\ 1 & 4 & -1 & 5 \end{pmatrix}$$

- A system is **consistent** if it has solutions.
- A system is **inconsistent** if one of the rows reduces to 0 = a.
- If the last row reduces to 0 = 0, there are infinitely many solutions and the general solution can be found by setting  $z = \lambda$  where  $\lambda$  is a parameter.
- If the determinant of the  $3 \times 3$  matrix is zero, then there is no unique solution (i.e either no solutions or infinite solutions).
- This links to planes, since the Cartesian equation of a plane is ax + by + cz = d.

#### 3 Vectors

- A vector  $\overrightarrow{AB}$  can be represented by a straight line, with an arrow, joining A and B.
- A vector can also be denoted with a lower case letter, e.g a, which is written with a tilde below it.
- A position vector defines the position of a point relative to the origin.  $\mathbf{a} = \overrightarrow{OA}$ .
- The Cartesian form of a vector:  $\mathbf{r} = x\mathbf{i} + y\mathbf{j} + z\mathbf{k}$ , or  $\mathbf{r} = \begin{pmatrix} x \\ y \\ z \end{pmatrix}$
- $\bullet \ |\mathbf{r}| = \sqrt{x^2 + y^2 + z^2}$
- A unit vector:  $\hat{\mathbf{r}} = \frac{\mathbf{r}}{|\mathbf{r}|}$
- The Ratio Thoerem:  $\overrightarrow{OP} = \frac{\mu \overrightarrow{OA} + \lambda \overrightarrow{OB}}{\mu + \lambda}$



#### 3.1 Scalar products

- The scalar product of two vectors is defined as  $\mathbf{a} \cdot \mathbf{b} = |\mathbf{a}| |\mathbf{b}| |\cos \theta$ .
- The vectors must both converge or diverge from one point.
- $\bullet \ \mathbf{a} \cdot \mathbf{a} = |\mathbf{a}|^2.$
- Most algebra works, except for cancellation and division.

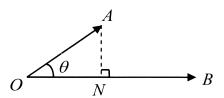
•  $\mathbf{a} \perp \mathbf{b} \iff \mathbf{a} \cdot \mathbf{b} = 0.$ 

$$\bullet \ \begin{pmatrix} a_1 \\ a_2 \\ b_2 \end{pmatrix} \cdot \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix} = a_1b_1 + a_2b_2 + a_3b_3.$$

## 3.2 Vector products

- The vector product of two vectors is defined as  $\mathbf{a} \times \mathbf{b} = |\mathbf{a}| |\mathbf{b}| \sin \theta \hat{\mathbf{n}}$ .
- $\hat{\mathbf{n}}$  is a unit vector perpendicular to  $\mathbf{a}$  and  $\mathbf{b}$ .
- $\bullet \ (\mathbf{a} \times \mathbf{b}) = -(\mathbf{b} \times \mathbf{a}).$
- $(\lambda \mathbf{a}) \times (\mu \mathbf{b}) = (\lambda \mu)(\mathbf{a} \times \mathbf{b}).$
- $|\mathbf{a} \times \mathbf{b}| = |\mathbf{a}||\mathbf{b}|\sin \theta$ .
- $\mathbf{a} \parallel \mathbf{b} \iff \mathbf{a} \times \mathbf{b} = 0$ , hence  $\mathbf{a} \times \mathbf{a} = 0$ .
- $\mathbf{a} \perp \mathbf{b} \iff \mathbf{a} \times \mathbf{b} = |\mathbf{a}||\mathbf{b}|.$
- $\mathbf{a} \cdot (\mathbf{a} \times \mathbf{b}) = 0$ .
- $\begin{pmatrix} a_1 \\ a_2 \\ b_2 \end{pmatrix} \times \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix} = \begin{pmatrix} a_2b_3 b_2a_3 \\ -(a_1b_3 b_1a_3) \\ a_1b_2 b_1a_2 \end{pmatrix}$ . Cover top find det, cover mid find negative det, cover bot find det.
- Area  $\triangle ABC = \frac{1}{2} |\overrightarrow{AB}| |\overrightarrow{AC}| \sin \theta = \frac{1}{2} |\overrightarrow{AB} \times \overrightarrow{AC}|.$

## 3.3 Projections and resolving vectors



- The length of the horizontal projection of **a** onto  $\mathbf{b} = \overrightarrow{ON} = |\mathbf{a}| |\hat{\mathbf{b}}| \cos \theta = \mathbf{a} \cdot \hat{\mathbf{b}}$
- The length of the vertical projection is given by  $|AN| = |\mathbf{a} \times \hat{\mathbf{b}}|$
- The horizontal projection vector is then  $\mathbf{u} = (\mathbf{a} \cdot \hat{\mathbf{b}})\hat{\mathbf{b}}$ , which is the same as the resolved component of  $\mathbf{a}$  parallel to  $\mathbf{b}$ .
- The perpendicular component of  $\mathbf{a}$  is  $\mathbf{v} = \mathbf{a} \mathbf{u}$ .

#### 3.4 Straight lines

$$l: \mathbf{r} = \mathbf{a} + \lambda \mathbf{d}, \quad \lambda \in \mathbb{R}.$$

- The vector equation of a line uses a position vector  $\mathbf{a}$  of a fixed point on l, and a direction vector  $\mathbf{d}$  parallel to l, to find the position vector of any point on the line  $(\mathbf{r})$ .
- $\lambda$  is a real parameter, which means that the vector equation of a line is not unique.

• To get the **parametric form**, we write the equation as column vectors then equate components:

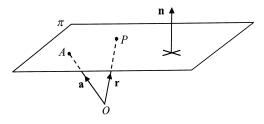
$$\begin{cases} x = a_1 + \lambda d_1, \\ y = a_2 + \lambda d_2, \quad \lambda \in \mathbb{R} \\ z = a_3 + \lambda d_3, \end{cases}$$

• To get the Cartesian form, make  $\lambda$  the subject then eliminate it.

$$\begin{cases} x = \mathbf{a_1} + \lambda \mathbf{d_1}, \\ y = \mathbf{a_2} + \lambda \mathbf{d_2}, \\ z = \mathbf{a_3} + \lambda \mathbf{d_3} \end{cases} \implies \begin{cases} \frac{x - \mathbf{a_1}}{\mathbf{d_1}} = \lambda, \\ \frac{y - \mathbf{a_2}}{\mathbf{d_2}} = \lambda, \\ \frac{z - \mathbf{a_3}}{\mathbf{d_3}} = \lambda \end{cases} \implies \frac{x - \mathbf{a_1}}{\mathbf{d_1}} = \frac{y - \mathbf{a_2}}{\mathbf{d_2}} = \frac{z - \mathbf{a_3}}{\mathbf{d_3}} \quad (= \lambda)$$

- $l_1$  and  $l_2$  are parallel  $\iff$   $\mathbf{d_1}$  and  $\mathbf{d_2}$  are parallel  $\iff$   $\mathbf{d_1} = k\mathbf{d_2}$ , for some  $k \in \mathbb{R}$ .
- $l_1$  and  $l_2$  intersect  $\iff$ 
  - $\mathbf{d_1}$  is not parallel to  $\mathbf{d_2}$  AND
  - there exist unique values of  $\lambda$  and  $\mu$  such that  $\mathbf{a_1} + \lambda \mathbf{d_1} = \mathbf{a_2} + \mu \mathbf{d_2}$ .
- The lines are skew  $\iff$  the direction vectors aren't parallel and there aren't unique values of  $\lambda$  and  $\mu$ .
- The acute angle between two lines is giveb by  $\cos^{-1} \left| \frac{\mathbf{d_1} \cdot \mathbf{d_2}}{|\mathbf{d_1}| |\mathbf{d_2}|} \right|$

#### 3.5 Planes



$$\overrightarrow{AP} \perp \mathbf{n} \implies \overrightarrow{AP} \cdot \mathbf{n} = 0 \implies (\mathbf{r} - \mathbf{a}) \cdot \mathbf{n} = 0.$$

- The scalar product form of the vector equation of the plane is  $\mathbf{r} \cdot \mathbf{n} = \mathbf{a} \cdot \mathbf{n}$ , where  $\mathbf{a}$  is a fixed point on the plane.
- n can be found by taking the cross product of two known vectors parallel to the plane.
- The shortest distance between the origin and the plane:  $|d| = |\mathbf{a} \cdot \hat{\mathbf{n}}|$
- The **parametric form** of the vector equation of the plane:

$$\pi: \mathbf{r} = \mathbf{a} + \lambda \mathbf{d_1} + \mu \mathbf{d_2}, \quad \lambda, \mu \in \mathbb{R}$$

• By expanding the scalar product form, we can arrive at the Cartesian form:

$$\mathbf{r} \cdot \mathbf{n} = \begin{pmatrix} x \\ y \\ z \end{pmatrix} \cdot \begin{pmatrix} a \\ b \\ c \end{pmatrix} = ax + by + cz = D$$

• A line will be parallel to a plane if it is perpendicular to  $\mathbf{n}$ , i.e  $\mathbf{n} \cdot \mathbf{d} = 0$  and there is no common point.

- If not parallel, it will intersect at a point, which can be found by substituting the line equation into the plane equation.
- The acute angle between l and  $\pi$ :  $\sin \theta = \left| \frac{\mathbf{d} \cdot \mathbf{n}}{|\mathbf{d}||\mathbf{n}|} \right|$
- When planes intersect, their Cartesian forms can be combined to form a system of simultaneous equations
  - If there is a unique solution, the planes intersect at a point.
  - If there are infinitely many solutions, the planes intersect in a line.
  - If there are no solutions, the three planes do not intersect.

## 4 Calculus

## 4.1 Differentiation

• If the limit of the denominator of a rational function is zero, you cannot substitute to find the limit: either 'juggle' or use l'Hopital's rule, e.g:

$$\lim_{x \to 0} \left( \frac{\sin x}{x} \right) = \lim_{x \to 0} \left( \frac{\cos x}{1} \right) = 1$$

• The definition of the derivative:

$$f'(x) = \lim_{\delta x \to 0} \left( \frac{f(x + \delta x) - f(x)}{\delta x} \right)$$

• Special derivatives:

$$\frac{d}{dx}(\sec x) = \sec x \tan x$$

$$\frac{d}{dx}(\csc x) = -\csc x \cot x$$

$$\frac{d}{dx}(\cot x) = -\csc^2 x$$

$$\frac{d}{dx}(\arcsin x) = \frac{1}{\sqrt{1-x^2}}$$

$$\frac{d}{dx}(\arccos x) = -\frac{1}{\sqrt{1-x^2}}$$

$$\frac{d}{dx}(\arctan x) = \frac{1}{1+x^2}$$

$$\frac{d}{dx}(a^x) = a^x \ln a$$

$$\frac{d}{dx}(\log_a x) = \frac{1}{x \ln a}$$

- f(x) is an increasing function on (a,b) if  $\frac{dy}{dx} > 0$  on that interval, or strictly increasing function if  $\frac{dy}{dx} \ge 0$ .
- f(x) is **concave upwards** on (a,b) if  $\frac{d^2y}{dx^2} > 0$ .
- If the derivative at a point is zero, the function is stationary.
- If the derivative at a point is  $\infty$ , there is a vertical line.
- For a point of inflexion,  $\frac{d^2y}{dx^2} = 0$  AND the sign of  $\frac{d^2y}{dx^2}$  changes, i.e concativity changes.
- Sketching the graph of f'(x) given f(x):

- Stationary point  $\rightarrow x$ -intercept.
- -f(x) increasing  $\to f'(x)$  above x-axis.
- Point of inflexion  $\rightarrow$  turning point.
- The gradient at any point on the curve:  $m = \frac{dy}{dx}|_{x=x_0}$ .
- The equation of a tangent to the curve at  $(x_0, y_0)$ :  $y y_0 = m(x x_0)$ .
- If two variabels are related, their rates of change are also related:

$$\frac{dy}{dt} = \frac{dy}{dx} \cdot \frac{dx}{dt}$$

• In kinematics especially:

$$a = \frac{dv}{dt} = \frac{dv}{dx} \cdot \frac{ds}{dt} = v\frac{dv}{ds}$$

## 4.2 Integration

$$\int (px+q)^n dx = \frac{(px+q)^{n+1}}{p(n+1)} + C$$

$$\int f'(x)(f(x))^n dx = \frac{(f(x))^{n+1}}{n+1} + C$$

$$\int \frac{f'(x)}{f(x)} dx = \ln|f(x)| + C$$

$$\int \frac{1}{x \ln x} dx = \int \frac{x^{-1}}{\ln x} + C = \ln|\ln|x|| + C$$

$$\int \tan x \, dx = \ln|\sec x| + C$$

$$\int \sec x \, dx = \ln|\sec x + \tan x| + C$$

$$\int \csc x \, dx = -\ln|\csc x + \cot x| + C$$

$$\int \frac{1}{(x+k)^2 + a^2} \, dx = \frac{1}{a} \arctan(\frac{x+k}{a}) + C$$

$$\int \frac{1}{\sqrt{a^2 - (x+k)^2}} \, dx = \arcsin(\frac{x+k}{a}) + C$$

- To integrate  $\sin^2 x$  or  $\cos^2 x$ , we expand  $\cos(2x)$  and rearrange.
- To integrate  $\sin^3 x$ , split into  $\int \sin x (\sin^2 x) dx$ , then use  $\sin^2 x + \cos^2 x = 1$ .
- If the integral is of the form:

$$\int \frac{px+q}{\sqrt{Ax^2+Bx+C}} dx \quad \text{or} \quad \int \frac{px+q}{Ax^2+Bx+C} dx$$

use sorcery to change it into  $\int \frac{f'(x)}{f(x)} dx$  or  $\int f'(x) (f(x))^n dx$ .

- Integration by substitution:
  - 1. Replace dx by  $\frac{dx}{dt} \cdot dt$ .
  - 2. Substitute by replacing all x with g(t).

Then: 
$$\int f(x) dx = \int f(g(t)) \frac{dx}{dt} \cdot dt$$

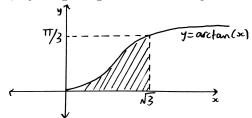
• Integration by parts:

$$\int u dv = uv - \int v du$$

• To choose which one to differentiate, use LIATE: Logs, Inverse trig, Algebraic, Trig, Exponentials.

## 4.3 Definite integrals

- $\int_a^b f(x)dx = -\int_b^a f(x)dx$
- The definite integral  $\int_a^b f(x)dx$  can only be found if f(x) is defined for all  $x \in (a,b)$ .
- The area between a curve and the y-axis:  $\int_a^b f(y)dy$
- $\bullet$  If a function is difficult to integrate, try integrating its inverse w.r.t y then subtract from a rectangle. e.g.



$$\int_0^{\sqrt{3}} \arctan x \ dx = \frac{\pi\sqrt{3}}{3} - \int_0^{\frac{\pi}{3}} \tan y \ dy$$

- The area between the curve and the axis is always  $\int_a^b |f(x)| dx$ .
- The area between two curves is always  $\int_a^b y_1 y_2 dx$ .
- The volume of revolution:

$$V = \pi \int_{a}^{b} y^{2} dx$$

• The volume of revolution of the area enclosed by two curves:

$$V = \pi \int_{a}^{b} (y_1)^2 dx - \pi \int_{a}^{b} (y_2)^2 dx$$

# 5 Probability and Statistics

## 5.1 Probability

- Two events A and B are mutually exclusive if  $P(A \cap B) = 0$ .
- $\bullet \ P(A \cup B) = P(A) + P(B) P(A \cap B).$
- $P(A|B) = \frac{P(A \cap B)}{P(B)}$ .
- A and B are **independent** if P(A|B) = P(A), so if they are independent  $P(A \cap B) = P(A)P(B)$ .

#### 5.2 Discrete random variables

- P(X = x) is the probability that the r.v X will assume a value of x.
- A discrete r.v can assume a countable number of values.
- For a d.r.v taking values  $x_1, x_2, x_3, ..., x_n$ , the **probability distribution** is defined as  $P(X = x_i)$ , such that:

$$0 \le P(X = x_i) \le 1$$
 and  $\sum_{\text{all } i} P(X = x_i) = 1$ 

• The expectation of a d.r.v:

$$E(X) = \mu = \sum xP(X = x)$$

$$E(g(X)) = \sum g(x)P(X = x)$$

$$E(a) = a$$

$$E(aX \pm B) = aE(X) \pm b$$

$$E(X \pm Y) = E(X) \pm E(Y)$$

• The variance of a d.r.v:

Var
$$(X) = \sigma^2 = E((x - \mu)^2) = E(X^2) - [E(X)]^2$$
  
Var $(a) = 0$   
Var $(ax + b) = a^2 \text{Var}((X))$   
Var $(X \pm Y) = \text{Var}(X) + \text{Var}(Y)$  (only if X and Y are independent)

• Note: never subtract variance.

#### 5.3 Discrete distributions

The Binomial distribution

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

- There are n independent trials, two possible outcomes (either 'success' or 'failure'), with constant probability of success p, X is the number of 'successes'.
- $\bullet$  The Binomial distribution is a combination of n Bernoulli trials.
- For  $P(X \le x)$ , we find P(X = 0) + P(X = 1) + P(X = 2) + ... + P(X = x).

## The Poisson distribution

$$X \sim Po(\lambda)$$
  $P(X = x) = \frac{e^{-\lambda}\lambda^x}{x!}$   $E(X) = Var(X) = \lambda$ 

- For a random variable in time or space, if there is no chance of simultaneous events, the events are independent, and the events have a constant probability of occurring, it is a Poisson process.
- $\lambda$  is the parameter, and defines the number of events in a given time/space.
- If  $X \sim Po(\lambda)$  and  $Y \sim Po(\mu)$ , then  $X + Y \sim Po(\lambda + \mu)$ .

#### The Geometric distribution

$$X \sim Geo(p)$$
  $P(X = x) = pq^{x-1}, \ x \ge 1$   $E(X) = \frac{1}{p}$   $Var(X) = \frac{q}{p^2}$ 

If we perform a series of independent trials with a probability p of success, X is the number of trials up to and including the first success.

$$P(X > x) = P(X = x + 1) + P(X = x + 2) + \dots$$

$$= pq^{x} + pq^{x+1} + pq^{x+2} + \dots$$

$$= pq^{x}(1 + q + q^{2} + \dots) = pq^{x}(\frac{1}{1 - q}) = q^{x}$$

$$P(X > a + b|X > a) = P(X > b) = q^{b}$$

#### The Negative Binomial distribution

$$X \sim NB(r,p)$$
  $P(X = x) = {x-1 \choose r-1} p^r q^{x-r}, \ r \ge 1, \ x \ge 1$   $E(X) = \frac{r}{p}$   $Var(X) = \frac{rq}{p^2}$ 

- $\bullet$  X is the number of trials needed to achieve r successes.
- The Negative Binomial distribution is just a combination of r geometric trials.

#### 5.4 Continuous random variables and CDFs

- Instead of probability distributions, we have probability density functions (PDFs), denoted by f(x).
  - $-f(x) \ge 0$  for all  $x \in \mathbb{R}$  $-\int_{-\infty}^{\infty} f(x) dx = 1$
- Continuous  $\implies$  uncountable, so P(X = x) = 0. Therefore,  $\geq$  or > is irrelevant.

$$P(a < X < b) = \int_{a}^{b} f(x) dx$$

$$E(X) = \mu = \int_{-\infty}^{\infty} x f(x) dx$$

$$E(g(X)) = \int_{-\infty}^{\infty} g(x) f(x) dx$$

$$P(|X - a| < b) = P(-b < X - a < b)$$

- The mode of a c.r.v is the value of x which gives the maximum probability, i.e the x coordinate of the highest point in the domain.
- The cumulative distribution function (CDF):

$$F(x) = P(X \le x) = \int_{-\infty}^{x} f(t) dt$$

$$\lim_{x \to -\infty} F(x) = 0 \qquad \lim_{x \to \infty} F(x) = 1$$

$$P(a < X < b) = F(b) - F(a)$$

$$\frac{d}{dx}F(x) = f(x)$$

- F(x) is continuous and increasing (since f(x) > 0).
- To find the median m, set  $F(m) = \frac{1}{2}$  and solve for m, i.e.  $\int_{-\infty}^{m} f(t)dt = 0.5$

#### 5.5 The Normal distribution

$$X \sim N(\mu, \sigma^2)$$

- The Normal distribution is a bell curve symmetrical about  $x = \mu$ .
- The mean = median = mode =  $\mu$ .
- $\mu$  affects the location of the curve, whereas  $\sigma^2$  affects the spread.
- The standard normal distribution is denoted by  $Z \sim N(0,1)$ .
- Any normal distribution can be standardised:  $Z = \frac{X \mu}{\sigma}$
- The Z score represents the number of standard deviations away from the mean.
- To find c given P(X < c) = p, use invNorm.
- If  $X \sim N(\mu_1, \sigma_1^2)$  and  $Y \sim N(\mu_2, \sigma_2^2)$ , then aX + bY also has a normal distribution.

$$E(aX + bY) = aE(X) + bE(Y)$$

$$= a\mu_1 + b\mu_2$$

$$Var(aX + bY) = a^2\sigma_1^2 + b^2\sigma_2^2$$

$$aX + bY \sim N(a\mu_1 + b\mu_2, \ a^2\sigma_1^2 + b^2\sigma_2^2)$$

#### 5.6 Sampling

- If X is a random variable,  $X_1, X_2, X_3, ..., X_n$  are a sample of n independent observations.
- The sample mean:

$$\bar{X} = \frac{X_1 + X_2 + \dots + X_n}{n}$$
 
$$E(\bar{X}) = E\left(\frac{X_1 + X_2 + \dots + X_n}{n}\right) = \frac{nE(X)}{n} = E(X) = \mu$$
 
$$\operatorname{Var}(\bar{X}) = \operatorname{Var}\left(\frac{X_1 + X_2 + \dots + X_n}{n}\right) = \frac{1}{n^2}\operatorname{Var}(X_1 + X_2 + \dots + X_n) = \frac{n\operatorname{Var}(X)}{n^2} = \frac{\sigma^2}{n}$$

- For the sample sum:  $E(S) = n\mu$ ,  $Var(S) = n\sigma^2$
- Therefore, in a normal population:

$$\bar{X} \sim N(\mu, \frac{\sigma^2}{n})$$
  $\sum_{r=1}^{n} X_r \sim N(n\mu, n\sigma^2)$ 

• The Central Limit Theorem states that, for a large sample size  $(n \ge 50)$ , the sample mean/sum of a sample from any distribution (e.g not normal), will approximately follow the normal distribution.

#### 5.7 Estimators

- An estimator is a test statistic T based on observed data that estimates an unknown parameter  $\theta$ .
- The estimator is **unbiased** if  $E(T) = \theta$ .
- The sample mean is an unbiased estimator of  $\mu$  since  $E(\bar{X}) = \mu$ .
- However, the sample variance is not an unbiased estimator for  $\sigma^2$  since  $E(S_n^2) = \frac{n-1}{n}\sigma^2$ .
- An unbiased estimator for  $\sigma^2$ :

$$\begin{split} s_{n-1}^2 &= \frac{n}{n-1} \times S_n^2 = \frac{n}{n-1} \left( \frac{1}{n} \sum x^2 - (\bar{x})^2 \right) \\ &= \frac{1}{n-1} \left( \sum x^2 - \frac{(\sum x)^2}{n} \right) \end{split}$$

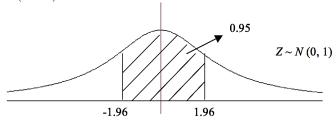
• An unbiased estimator is more **efficient** than another if it has a lower variance.

#### 5.8 Confidence intervals

- A 95% confidence interval (CI) means that there is a 95% chance that the interval includes  $\mu$ .
- For  $X \sim N(\mu, \sigma^2)$ , if we take a sample:  $\bar{X} \sim N(\mu, \sigma^2)$ .

Confidence limits = 
$$\bar{X} \pm Z_k \frac{\sigma}{\sqrt{n}}$$
  
CI =  $\left[\bar{X} - Z_k \frac{\sigma}{\sqrt{n}}, \ \bar{X} + Z_k \frac{\sigma}{\sqrt{n}}\right]$ 

- $Z_k$  is the **critical value**, and is found using invNorm.
- For a 95% CI: invNorm(0.025) = -1.96



- The width of a CI is  $2Z_k \frac{\sigma}{\sqrt{n}}$
- If we have a large sample from any population ( $\mu$  and  $\sigma^2$  unknown), we can use the CLT.

$$CI = \left[\bar{x} - Z_k \frac{s_{n-1}}{\sqrt{n}}, \ \bar{x} + Z_k \frac{s_{n-1}}{\sqrt{n}}\right]$$

- $\bullet$  If the population is normal but we do not know the variance, we use the t-distribution.
  - $T = \frac{\bar{X} \mu}{s_{n-1}/\sqrt{n}}$  follows a t-distribution with n-1 degrees of freedom.

$$CI = \left[ \bar{x} - t_k \frac{s_{n-1}}{\sqrt{n}}, \ \bar{x} + t_k \frac{s_{n-1}}{\sqrt{n}} \right]$$

$\sigma^2$	n	Assumptions	Test Statistic
known	large	CLT	$Z = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \sim N(0, 1)$
	small	normal	$Z = \frac{1}{\sigma/\sqrt{n}} \approx W(0,1)$
unknown	large	$\operatorname{CLT}$	$Z = \frac{\bar{X} - \mu}{\frac{s_{n-1}/\sqrt{n}}{\sqrt{n}}} \sim N(0, 1)$
	small	normal	$T = \frac{X - \mu}{\frac{s_{n-1}/\sqrt{n}}{\sqrt{n}}} \sim t_{n-1}$

# 5.9 Hypothesis testing

- 1. State  $H_0$  and  $H_1$ .
- 2. Test statistic.
- 3. Level of significance and rejection criteria.
- 4. Compute *p*-value (or *z*-value or *t*-value).
- 5. Conclusion in context.

e.g

 $H_0: \mu = 3$ 

 $H_1: \mu > 3$ 

Test statistic:  $Z = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}}$ 

Sig level = 5%, one tailed.

Reject  $H_0$  if p < 0.05

Since p-value = 0.03 < 0.05, we reject  $H_0$  and conclude that there is significant evidence at the 5% level that...

- $P(\text{Type I Error}) = P(H_0 \text{ rejected}|H_0 \text{ true}) = \alpha\%$ . i.e P(Type I Error) = significance level.
- $P(\text{Type II Error}) = P(H_0 \text{ accepted}|H_1 \text{ true}).$
- For example, for  $H_0: \mu = \mu_0$   $H_1: \mu = \mu_1$ ,

 $P(\text{Type II Error}) = P(H_0 \text{ accepted}|H_1 \text{ true}) = P(\bar{X} < \text{critical value } |\bar{X} \sim N(\mu_1, \sigma^2))$ 

#### 5.10 PGFs

$$G(t) = E(t^{X}) = \sum t^{x} P(X = x)$$

$$G(1) = 1$$

$$G'(t) = \sum x t^{x-1} P(X = x) :: E(X) = G'(1)$$

$$G''(t) = \sum x (x - 1) t^{x-2} P(X = x)$$

$$G''(1) = \sum x^{2} P(X = x) - \sum x P(X = x) = E(X^{2}) - E(X)$$

$$:: E(X^{2}) = G''(1) + G'(1)$$

$$:: Var(X) = G''(1) + G'(1) - [G'(1)]^{2}$$

If 
$$Z = X + Y$$
,  $G_Z(t) = E(t^Z) = E(T^{X+Y}) = E(t^X)E(t^Y) = G_X(t)G_Y(t)$ 

- To find P(X=n), we use the Maclaurin series:  $P(X=n) = \frac{G^n(0)}{n!}$ .
- To prove most things about PGFs, differentiation will be involved (sometimes using the product rule and chain rule).

### **Binomial**

If  $Y \sim B(n, p)$ , we can say that  $Y = X_1 + X_2 + X_3 + ... + X_n$  where X is a Bernoulli trial.

$$\begin{array}{|c|c|c|c|c|} \hline x & 0 & 1 \\ \hline P(X=x) & q & p \\ \hline \end{array}$$

$$G_X(t) = \sum t^x P(X = x) = q + pt$$

$$G_Y(t) = E(t^Y) = E(t^{X_1 + \dots + X_n}) = [E(t^X)]^n = [G_X(t)]^n = (q + pt)^n$$

#### Poisson

If 
$$X \sim Po(\lambda)$$
,  $P(X = x) = \frac{e^{-\lambda} \lambda^x}{x!}$ .

$$\begin{split} G(t) &= E(t^X) = \sum t^x P(X=x) \\ &= \sum t^x \frac{e^{-\lambda} \lambda^x}{x!} \\ &= e^{-\lambda} \sum \frac{(\lambda t)^x}{x!} = e^{-\lambda} e^{\lambda t} = e^{\lambda(t-1)}. \end{split}$$

#### Geometric

If 
$$X \sim Geo(p)$$
,  $P(X = x) = pq^{x-1}$ .

$$G(t) = E(t^{X}) = \sum t^{x} P(X = x)$$

$$= \sum t^{x} p q^{x-1}$$

$$= pt + pt^{2} q + pt^{3} q^{2} + pt^{4} q^{3} + \dots + pt^{n} q^{n-1} + \dots$$

$$S_{\infty} = \frac{a}{1-r} = \frac{pt}{1-at}$$

#### Negative Binomial

If  $Y \sim NB(r, p)$ , we can say that  $Y = X_1 + X_2 + X_3 + ... + X_r$ , where  $X \sim Geo(p)$ .

$$G_Y(t) = E(t^Y) = E(t^{X_1 + \dots + X_r}) = [E(t^X)]^r = [G_X(t)]^r = \left(\frac{pt}{1 - qt}\right)^r$$

#### 5.11 Bivariate data and correlations

- If X and Y are random variables, the joint probability distribution is  $P(X = x \cap Y = y)$ .
- $\sum \sum p(x,y) = 1$
- $E(XY) = \sum \sum xy \ p(x,y)$
- Cov(X,Y) = E(XY) E(X)E(Y). X and Y independent  $\implies Cov(X,Y) = 0$ .
- Var(X + Y) = Var(X) + Var(Y) 2Cov(X, Y).
- $\bullet$  The correlation coefficient measures the linear relationship between X and Y

$$\rho = \operatorname{Corr}(X, Y) = \frac{\operatorname{Cov}(X, Y)}{\sigma_X \sigma_Y} = \frac{\sigma_{XY}}{\sigma_X \sigma_Y}$$

• A bivariate sample consists of pairs of data  $(x_1, y_1)$ . For a bivariate sample, the above points do not apply.

$$r = \frac{\sum xy - n\bar{x}\bar{y}}{\sqrt{(\sum x^2 - n\bar{x}^2)(\sum y^2 - n\bar{y}^2)}} = \frac{S_{xy}}{\sqrt{S_{xx}S_{yy}}}, \text{ where } S_{xx} = \sum x^2 - \frac{(\sum x)^2}{n}$$

- If r=0, there is no linear relationship, but it does not imply that X and Y are independent.
- r is independent of the units, and does not show any causality.
- In maths, controlled variable = independent variable.
- The y-on-x regression line y = a + bx will always pass through  $(\bar{x}, \bar{y})$ .

$$y - \bar{y} = b(x - \bar{x})$$
, where  $b = \frac{S_{xy}}{S_{xx}}$ 

• The x-on-y regression line is denoted by x = c + dy.

$$bd = r^2$$
  $r = \pm \sqrt{bd}$ , the sign depends on whether the gradient is positive or negative.

• We can statistically test evidence of a correlation by assuming both variables follow a bivariate normal distribution with correlation coefficient  $\rho$ .

e.g

 $H_0: \rho = 0$ 

 $H_1: \rho \neq 0$ 

Test statistic: 
$$T = r\sqrt{\frac{n-2}{1-r^2}} \sim t_{n-2}$$

Sig level = 5%, two tailed

Reject  $H_0$  if |T| > invt(0.975, n-2)

Note: 
$$T = r\sqrt{\frac{n-2}{1-r^2}}$$
 (sub in values)

Since |T| = 0.08 > 0.05, we reject  $H_0$  and conclude that there is a significant evidence at the 5% level that there is a correlation between...

# 6 Miscellaneous

- $(a \pm b)^3 = a^3 \pm 3a^2b + 3ab^2 \pm b^3$
- $a^3 + b^3 = (a+b)(a^2 ab + b^2)$
- $a^3 b^3 = (a b)(a^2 + ab + b^2)$
- |x+3||x+2| = |(x+3)(x+2)|