

# Distribution - STAT110 Otago

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## Terms in this set (21)

|  |   |
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| mean of the binary (Bernoulli) distribution                      | $\mu = p$   |
| variance(方差) of the binary (Bernoulli) distribution              | $\sigma^2 = p(1 - p)$   |
| difference between binary distribution and binomial distribution | $n = 1 \Rightarrow$ <b>binary</b> distribution<br>$n > 1 \Rightarrow$ <b>binomial</b> distribution  |
| mean of the binomial distribution                                | $\mu = np$  |
| variance of the binomial distribution                            | $\sigma^2 = np(1 - p)$  |
| Conditions for binomial distribution                             | Outcome is binary.<br>We have $n$ independent trials.<br>The number of trials is fixed.<br>Probability of success $\pi$ must stay constant. |
| Probability of $x$ successes in $n$ trials                       | $= \binom{n}{x} \pi^x$  |

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| binomial coefficient (n choose k)  | $n! / (k!(n-k)!)$ <p><i>no need to memorise</i></p>  |
| standard normal distribution(Z):   | $Z \sim N(\mu = 0, \sigma^2 = 1)$  |
| $\mu$ (normal distribution)  | <i>moves the curve but does not change its shape</i>   |
| $\sigma$ (moves the curve but does not change its shape)   | <i>spreads the curve more widely about <math>X = \mu</math> but does not alter the centre</i>  |
| Compare a relative frequency histogram with a probability distribution   | <p><b>Relative frequency histogram</b> represents a <b>sample</b> (smaller number of individuals).</p> <p><b>Probability density function</b> represents a <b>population</b> (large number of individuals).</p>  |
| how to estimate the value of the parameters if estimate a probability distribution curve from a relative frequency histogram | <p><math>\mu</math> is estimated by the sample mean</p> <p><math>\sigma</math> is estimated by the sample standard deviation, <math>s</math>.</p>  |
| what does the areas under the normal distribution curve represent?   | probabilities  |
| what is Z-score (Z-value)  | <p>number of standard deviations away from the mean</p> <p>Any normal distribution value, <math>X \sim N(\mu_X, \sigma^2_X)</math>, can be put on the standard normal scale, <math>Z \sim N(0, 1)</math>.</p> <p><b>Z-score follows a standard normal distribution</b></p> |
| formula for Z-Value  | $Z = (X - \mu_X) / \sigma_X$   |
| when will the sampling distribution of the mean will follow a normal distribution?   | if $n$ (the samples, not $X$ ) is large enough   |

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| Central Limit Theorem (CLT)                    | the <b>sampling</b> distribution derived from a simple random sample will <b>be approximately normally distributed</b>   |
| What is the mean of the sampling distribution? | population mean<br>$\mu_{\bar{X}} = \mu_X$   |
| Variance of the sampling distribution          | the variability of sample means<br>$\sigma_{\bar{X}} = \frac{\sigma_X}{\sqrt{n}}$  |
| Notes on the sampling distribution             | <p>If sample size <math>n</math> is greater, then the standard error of the mean is smaller (<b>more compact distribution, greater precision</b>).</p> <p>If <math>X</math> is normal, then <math>\bar{X}</math> is normal (for any <math>n</math>)</p> <p>If <math>X</math> is not normal, then <math>\bar{X}</math> is approximately normal for large <math>n</math> (central limit theorem)</p> |