

# Formulae: Statistical Computing

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## 1 Data Types

- Descriptive Statistics
  - Qualitative variables
    - Nominal
    - Ordinal
  - Numeric or Quantitative Variables
    - Discrete
    - Continuous
- Object types in R
  - Factor (Qualitative)
  - Numeric (Quantitative)
  - Logical
  - Character
  - List

## 2 Frequency

- Absolute Frequency  $h_i$  (`table()`)
- Relative Frequency  $f_i = \frac{h_i}{n}$   
(`prop.table(table())`)
- Absolute cumulative frequency  $H_i = \sum_{j=1}^i h_j$   
(`cumsum(table())`)
- Relative cumulative frequency  $F_i = \sum_{j=1}^i f_j = \frac{H_i}{n}$   
(`cumsum(prop.table(table()))`)

## 3 Descriptive Statistics

- Mean (arithmetic mean)  $\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$ 
  - If  $y_i = ax_i + b$  ( $a$  &  $b$  constant), then  $\bar{y} = a\bar{x} + b$ .
  - If  $z_i = x_i + y_i$ , then  $\bar{z} = \bar{x} + \bar{y}$ .
- Median  $x_{0.5}$   
The ordered data values are  $x_{(1)}, \dots, x_{(n)}$ 
  - odd  $n$ :  $x_{0.5} = x_{(\frac{n+1}{2})}$
  - even  $n$ :  $x_{0.5} = \frac{1}{2} \left( x_{(\frac{n}{2})} + x_{(\frac{n}{2}+1)} \right)$

- Mode  $x_D$  is the most frequent value.
- Variance  $s_x^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$
- Standard deviation (SD)  $s_x = \sqrt{s_x^2}$   
If  $y_i = ax_i + b$  ( $a$  &  $b$  constant), then
  - $\text{Var } s_y^2 = a^2 s_x^2$
  - $\text{SD } s_y = a s_x$
- Range  $R = x_{\max} - x_{\min}$
- Interquartile range  $IQR = Q_3 - Q_1$
- Coefficient of variation  $CV = \frac{s}{\bar{x}}$
- First quartile ( $Q_1$ )
  - If  $n$  is divisible by 4  
 $Q_1 = x_{0.25} = \frac{1}{2} \left( x_{(\frac{n}{4})} + x_{(\frac{n}{4}+1)} \right)$
  - If  $n$  is not divisible by 4  $Q_1 = x_{0.25} = x_{(\lceil \frac{n}{4} \rceil)}$   
 $\lceil \cdot \rceil$  means round up.
  - R: `quantile(x, 0.25)`
- Third quartile ( $Q_3$ )
  - If  $n$  is divisible by 4  
 $Q_3 = x_{0.75} = \frac{1}{2} \left( x_{(\frac{3n}{4})} + x_{(\frac{3n}{4}+1)} \right)$
  - If  $n$  is not divisible by 4  $Q_3 = x_{0.75} = x_{(\lceil \frac{3n}{4} \rceil)}$
  - R: `quantile(x, 0.75)`
- $p$ -quantile
  - If  $pn$  is an integer  $x_p = \frac{1}{2} (x_{(pn)} + x_{(pn+1)})$
  - If  $pn$  is not an integer  $x_p = x_{(\lceil pn \rceil)}$
  - R: `quantile(x, p)`
- Skewness (Symmetry):  $g_1$ 
  - $g_1 \gg 0$  right-skewed, right-tailed, leaning to the left
  - $g_1 \ll 0$  left-skewed, left-tailed, leaning to the right
  - $g_1 \approx 0$  symmetric.

- Covariance  $s_{xy} = \frac{1}{n-1} \sum (x_i - \bar{x})(y_i - \bar{y})$

- Correlation coefficient

$$r_{x,y} = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum (x_i - \bar{x})^2 \sum (y_i - \bar{y})^2}} = \frac{s_{xy}}{s_x \cdot s_y}$$

- Empirical distribution function

$$F_n(b) = P(X \leq b) = \frac{\#x_i \leq b}{n}$$

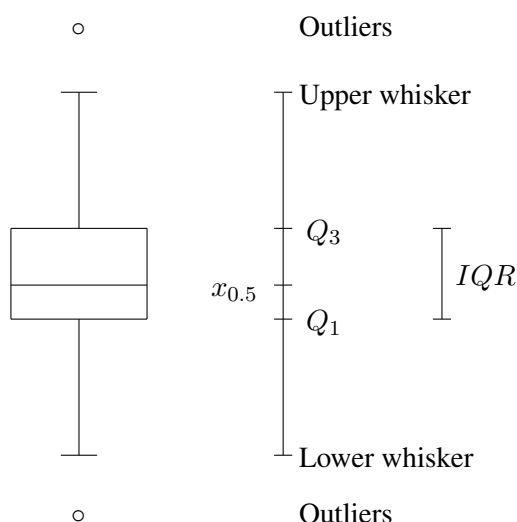
- R: `ecdf (x)`

## 4 Graphics

### Histogram

Height of  $i$ -th Column is the “density”  $y_i = \frac{h_i}{b_i \cdot n}$ , where  $h_i$  is the absolute frequency in the  $i$ -th interval and  $b_i$  is the interval width. R: `hist (x)`

### Box plot



Upper whisker is the largest data value  $\leq Q_3 + 1.5IQR$

Lower whisker is the smallest data value  $\geq Q_1 - 1.5IQR$

R: `boxplot (y)` or `boxplot (y~x)`

## 5 Normal Distribution

- Let  $Z \sim N(0, 1)$  be a random variable with the standard normal distribution,  $P(Z \leq z) = \Phi(z)$

R: `pnorm (z)`

- Let  $X$  have a general normal  $N(\mu, \sigma^2)$  distribution.  $Z = \frac{X - \mu}{\sigma}$  has a standard normal distribution.

- Central limit theorem: Let  $X_1, X_2, \dots, X_n$  be an iid. random sample, from an arbitrary distribution with expectation  $\mu$  and variance  $\sigma^2$

For large  $n$ , the distribution of the random variable  $Z = \frac{\bar{X} - \mu}{\sqrt{\sigma^2/n}}$  is well approximated by the standard normal  $N(0, 1)$  distribution.

$$\Rightarrow \frac{\bar{X} - \mu}{\sqrt{\sigma^2/n}} \overset{a}{\sim} N(0, 1) \quad \text{or equivalently} \quad \bar{X} \overset{a}{\sim} N(\mu, \sigma^2/n)$$

## 6 Regression

Regression line for paired data  $(x_i, y_i)$ :

$$y_i = \hat{a} + \hat{b}x_i + \hat{\epsilon}_i,$$

where  $\hat{a}$  is the least squares estimator for the intercept and  $\hat{b}$  is the least squares estimate for the gradient.  
 $\hat{\epsilon}_i$  is the  $i$ -th residual or  $i$ -th error term.

The regression coefficients are calculated using:

$$\begin{aligned} \blacktriangleright \hat{b} &= \frac{\sum (y_i - \bar{y})(x_i - \bar{x})}{\sum (x_i - \bar{x})^2} = \frac{s_{xy}}{s_x^2} \\ \blacktriangleright \hat{a} &= \bar{y} - \hat{b}\bar{x} \end{aligned}$$

The fitted values are  $\hat{y}_i = \hat{a} + \hat{b}x_i$ .

The residuals are  $\hat{\epsilon}_i = y_i - \hat{y}_i$ .

## 7 Confidence intervals

- A confidence interval for  $\mu$  with 95% confidence level, based on the normal distribution

$$\left[ \bar{x} \pm 1.96 \frac{\sigma}{\sqrt{n}} \right],$$

estimate  $\sigma$  using  $s_x$  if  $\sigma$  is unknown. For other confidence levels  $(1 - \alpha)$  use `qnorm(1-alpha/2)`.

- A confidence interval for  $\mu$  with 95% confidence level, based on the  $t$  distribution

$$\left[ \bar{x} \pm t \frac{s_x}{\sqrt{n}} \right]$$

$t$  depends on the confidence level and the sample size `qt(1-alpha/2, n-1)`.

- **Confidence interval for a proportion  $p$**

$\hat{p} = \bar{x} = \frac{\sum_{i=1}^n x_i}{n}$  is an estimate for  $p$ .

$$\left[ \bar{x} \pm 1.96 \frac{\sqrt{\bar{x}(1 - \bar{x})}}{\sqrt{n}} \right]$$

is an approximate 95% confidence interval for  $p$ , provided  $n > 30$ .

## 8 Hypothesis tests

**One sample  $t$ -tests** `t.test(x)`

- Two sided test for an expectation  $\mu$  with significance level  $\alpha$

$$H_0 : \mu = \mu_0 \quad \text{vs} \quad H_1 : \mu \neq \mu_0$$

Critical region:  $H_0$  is rejected iff  $t_{\text{stat}} = \frac{\bar{x} - \mu_0}{s/\sqrt{n}} > t_{cr}$  or  $t_{\text{stat}} < -t_{cr}$ ,

where  $t_{cr}$  is a quantile from the  $t$ -distribution `qt(1-alpha/2, n-1)`.

- One sided  $t$ -test for an expectation  $\mu$  with significance level  $\alpha$

a)  $H_0 : \mu \geq \mu_0$  vs  $H_1 : \mu < \mu_0$  Critical region:  $H_0$  is rejected iff  $t_{\text{stat}} = \frac{\bar{x} - \mu_0}{s/\sqrt{n}} < -t_{cr}$ ,  
 $t_{cr}$  is qt (1-alpha, n-1)

b)  $H_0 : \mu \leq \mu_0$  vs  $H_1 : \mu > \mu_0$   $H_0$  is rejected iff  $t_{\text{stat}} = \frac{\bar{x} - \mu_0}{s/\sqrt{n}} > t_{cr}$   $t_{cr}$  is qt (1-alpha, n-1)

- $p$ -Value: reject the null hypothesis iff  $p < \alpha$  the significance level.

**Two sample tests** `t.test(x, y)` `t.test(x~y)`

- For two unpaired samples:

Test statistic:  $t_{\text{stat}} = \frac{\bar{x} - \bar{y}}{s_p \sqrt{\frac{1}{n_x} + \frac{1}{n_y}}}$ , with pooled variance  $s_p^2 = \frac{(n_x - 1)s_x^2 + (n_y - 1)s_y^2}{n_x + n_y - 2}$ .

Degrees of freedom:  $m = n_x + n_y - 2$ .

- For two unpaired samples: Calculate  $d_i = x_i - y_i$  and carry out a one sample  $t$ -test on  $d_i$ .
- Critical region for two sided tests  $H_0 : \mu_x = \mu_y$  vs  $H_1 : \mu_x \neq \mu_y$   
 $H_0$  is rejected iff  $t_{\text{stat}} > t_{cr}$  or  $t_{\text{stat}} < -t_{cr}$ .
- Critical region for one sided tests
  - (a)  $H_0 : \mu_x \geq \mu_y$  vs  $H_1 : \mu_x < \mu_y$ .  $\Rightarrow H_0$  is rejected iff  $t_{\text{stat}} < -t_{cr}$ .
  - (b)  $H_0 : \mu_x \leq \mu_y$  vs  $H_1 : \mu_x > \mu_y$ .  $\Rightarrow H_0$  is rejected iff  $t_{\text{stat}} > t_{cr}$ .

## $\chi^2$ Test of independence

For variables  $X$  and  $Y$  with values  $X_1, \dots, X_m$  and  $Y_1, \dots, Y_n$ . Joint frequency table:

	$Y_1$	$\dots$	$Y_n$	Total
$X_1$	$h_{11}$	$\dots$	$h_{1n}$	$h_{1\cdot}$
$\vdots$	$\vdots$		$\vdots$	$\vdots$
$X_m$	$h_{m1}$	$\dots$	$h_{mn}$	$h_{m\cdot}$
Total	$h_{\cdot 1}$	$\dots$	$h_{\cdot n}$	$n$

The expected frequencies are:  $e_{ij} = \frac{h_{i\cdot} h_{\cdot j}}{n}$

The test statistic is:  $\chi_{\text{stat}}^2 = \sum_{i,j} \frac{(h_{ij} - e_{ij})^2}{e_{ij}}$

Degrees of freedom:  $k = (m_X - 1)(m_Y - 1)$

Critical value is the  $1 - \alpha$ -quantile from the  $\chi_k^2$  distribution `qchisq(1-alpha, k)`.

$H_0$  is rejected iff  $\chi_{\text{stat}}^2 > \text{critical value}$ .

## Test of equality of two variances

The test statistic is  $f_{\text{stat}} = \frac{s_x^2}{s_y^2}$

R: `var.test(x, y)` or `var.test(x ~ y)`