Statistics Ouick Reference Card & R Commands

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Get R at: http://www.r-project.org

R commands: bold typewriter text

1 Misc R

To make a vector / store data: $\mathbf{x} = \mathbf{c} (\mathbf{x} \mathbf{1}, \mathbf{x} \mathbf{2}, \ldots)$ Help: general RSiteSearch ("Search Phrase")

Help: function ?functionName

Get column of data from table:

tableName\$columnName

List all variables: 1s()

Delete all variables: rm(list=ls())

$$\sqrt{x} = \text{sqrt}(\mathbf{x})$$
 (1)

$$x^n = \mathbf{x}^{\wedge} \mathbf{n} \tag{2}$$

$$n =$$
length (x) (

$$T = \mathtt{table}(\mathbf{x})$$

2 Descriptive Statistics

2.1 Numerical

Let x=c(x1, x2, x3, ...)

$$total = \sum_{i=1}^{n} x_i = sum(\mathbf{x})$$
 (5)

$$min = min(x)$$

$$\max = \max(\mathbf{x}) \tag{7}$$

six number summary: summary (x)

$$\mu = \frac{\sum x_i}{N} = \text{mean}(\mathbf{x}) \tag{9}$$

$$\bar{x} = \frac{\sum x_i}{n} = \text{mean}(\mathbf{x}) \tag{10}$$

$$\tilde{x} = P_{50} = \text{median}(\mathbf{x}) \tag{11}$$

$$\sigma = \sqrt{\frac{\sum (x_i - \mu)^2}{N}} \tag{12}$$

$$s = \sqrt{\frac{\sum (x_i - \bar{x})^2}{n - 1}} = \operatorname{sd}(\mathbf{x}) \quad (13)$$

$$CV = \frac{\sigma}{\mu} = \frac{s}{\bar{x}} \tag{14}$$

2.2 RELATIVE STANDING

$$z = \frac{x - \mu}{\sigma} = \frac{x - \bar{x}}{\sigma} \tag{15}$$

Percentiles:

$$P_k = x_i$$
, (sorted x)

$$k = \frac{i - 0.5}{n} \cdot 100\% \tag{16}$$

To find x_i given P_k , i is:

- 1. L = (k/100%)n
- 2. if L is an integer: i = L + 0.5; otherwise i=L and round up.

2.3 VISUAL

All plots have optional arguments:

- main="" sets title
- xlab="", ylab="" sets x/y-axis label
- type="p" for point plot
- type="1" for line plot
- type="b" for both points and lines

Ex: plot(x, y, type="b", main="My Plot") Plot Types:

hist(x) histogram

stem(x) stem & leaf

boxplot (x) box plot

plot(T) bar plot, T=table(x)

plot (x, y) scatter plot, x, y are ordered vectors

plot (t, y) time series plot, t, y are ordered vectors curve (expr, xmin, xmax) plot expr involving x

2.4 ASSESSING NORMALITY

Q-Q plot: qqnorm(x); qqline(x)

3 Probability

Number of successes x with n possible outcomes. (Don't double count!)

$$P(A) = \frac{x_A}{n} \tag{17}$$

$$P(\bar{A}) = 1 - P(A) \tag{18}$$

$$P(A \text{ or } B) = P(A) + P(B) - P(A \text{ and } B)$$
 (19)

$$P(A \text{ or } B) = P(A) + P(B) \text{ if } A, B \text{ mut. excl.}$$
 (20)

$$P(A \text{ and } B) = P(A) \cdot P(B|A) \tag{21}$$

$$P(A \text{ and } B) = P(A) \cdot P(B)$$
 if A, B independent (22)

$$n! = n(n-1)\cdots 1 =$$
factorial (n) (23)

$$_{n}P_{k} = \frac{n!}{(n-k)!}$$
 Perm. no elem. alike (24)

$$=\frac{n!}{n_1!n_2!\cdots n_k!}$$
 Perm. n_1 alike, ... (25)

$$_{n}C_{k}=\frac{n!}{(n-k)!k!}=$$
 choose (n,k) (26)

4 Discrete Random Variables

$$P(x_i)$$
: probability distribution (27)

$$E = \mu = \sum x_i \cdot P(x_i) \tag{28}$$

$$\sigma = \sqrt{\sum (x_i - \mu)^2 \cdot P(x_i)}$$
 (29)

4.1 BINOMIAL DISTRIBUTION

$$\mu = n \cdot p \tag{30}$$

$$\sigma = \sqrt{n \cdot p \cdot q} \tag{31}$$

$$P(x) = {}_{n}C_{x}p^{x}q^{(n-x)} = \text{dbinom}(x, n, p)$$
 (32)

4.2 Poisson distribution

$$P(x) = \frac{\mu^x \cdot e^{-\mu}}{x!} = \text{dpois}(\mathbf{x}, \ \mu)$$
 (33)

5 Continuous random variables

CDF F(x) gives area to the left of x, $F^{-1}(p)$ expects p is area to the left.

$$f(x)$$
: probability density (34)

$$E = \mu = \int_{-\infty}^{\infty} x \cdot f(x) \, dx \tag{35}$$

$$\sigma = \sqrt{\int_{-\infty}^{\infty} (x - \mu)^2 \cdot f(x) \, dx} \tag{36}$$

$$F(x)$$
: cumulative prob. density (CDF) (37)

$$F^{-1}(x)$$
: inv. cumulative prob. density (38)

$$F(x) = \int_{-\infty}^{x} f(x') dx'$$
 (39)

$$p = P(x < x') = F(x') \tag{40}$$

$$x' = F^{-1}(p) (41)$$

$$p = P(x > a) = 1 - F(a)$$
(42)

$$p = P(a < x < b) = F(b) - F(a)$$
 (43)

5.1 Uniform distribution

$$p = P(u < u') = F(u')$$

=
$$punif(u', min=0, max=1)$$
 (44)

$$u' = F^{-1}(p) = \text{qunif(p, min=0, max=1)}$$
 (45)

5.2 NORMAL DISTRIBUTION

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \cdot e^{-\frac{1}{2}\frac{(x-\mu)^2}{\sigma^2}} \tag{46}$$

$$p = P(z < z') = F(z') = pnorm(z')$$
 (47)

$$z' = F^{-1}(p) = \text{qnorm}(p)$$
 (48)

$$p = P(x < x') = F(x')$$

= pnorm(
$$\mathbf{x}'$$
, mean= μ , sd= σ) (49)

$$x' = F^{-1}(p)$$

$$=$$
 qnorm(p, mean= μ , sd= σ) (50)

5.3 t-DISTRIBUTION

$$p = P(t < t') = F(t') = pt(t', df)$$
 (51)

$$t' = F^{-1}(p) = qt (p, df)$$
 (52)

5.4 χ^2 -DISTRIBUTION

$$p = P(\chi^2 < \chi^{2'}) = F(\chi^{2'})$$

$$= pchisq(X^2', df)$$
(53)

$$\chi^{2'} = F^{-1}(p) = \text{qchisq(p, df)}$$
 (54)

5.5 F-DISTRIBUTION

$$p = P(F < F') = F(F')$$

$$= pf(F', df1, df2)$$
(55)

$$F' = F^{-1}(p) = qf(p, df1, df2)$$
 (56)

6 Sampling distributions

$$\mu_{\bar{x}} = \mu$$
 $\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$ (57)

$$\mu_{\hat{p}} = p \qquad \qquad \sigma_{\hat{p}} = \sqrt{\frac{pq}{n}} \qquad (58)$$

7 Estimation

7.1 CONFIDENCE INTERVALS

proportion:
$$\hat{p} \pm E$$
, $E = z_{\alpha/2} \cdot \sigma_{\hat{p}}$ (59)

mean (
$$\sigma$$
 known): $\bar{x} \pm E$, $E = z_{\alpha/2} \cdot \sigma_{\bar{x}}$ (60)

mean (
$$\sigma$$
 unknown, use s): $\bar{x} \pm E$, $E = t_{\alpha/2} \cdot \sigma_{\bar{x}}$, (61)

$$df = n - 1$$

variance:
$$\frac{(n-1)s^2}{\chi_R^2} < \sigma^2 < \frac{(n-1)s^2}{\chi_L^2},$$
 (62)

$$df = n - 1$$

2 proportions:
$$\Delta \hat{p} \pm z_{\alpha/2} \cdot \sqrt{\frac{\hat{p_1}\hat{q_1}}{n_1} + \frac{\hat{p_2}\hat{q_2}}{n_2}}$$
 (63)

2 means (indep):
$$\Delta \bar{x} \pm t_{\alpha/2} \cdot \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$
, (64)

$$df \approx \min\left(n_1 - 1, n_2 - 1\right)$$

matched pairs:
$$\bar{d} \pm t_{\alpha/2} \cdot \frac{s_d}{\sqrt{n}}, \quad d_i = x_i - y_i,$$
 (65)

$$df = n-1$$

7.2 CI CRITICAL VALUES (TWO SIDED)

$$z_{\alpha/2} = F_z^{-1}(1 - \alpha/2) = \text{qnorm(1-alpha/2)}$$
 (66)
 $t_{\alpha/2} = F_c^{-1}(1 - \alpha/2) = \text{qt(1-alpha/2, df)}$ (67)

$$y^2 = F^{-1}(\alpha/2) = \cosh(\alpha/2) \cosh(\alpha/2)$$
 (68)

$$\chi_L^2 = F_{\chi^2}^{-1}(\alpha/2) = \text{qchisq(alpha/2, df)}$$
 (68)

$$\chi_R^2 = F_{\chi^2}^{-1}(1 - \alpha/2) = \text{qchisq(1-alpha/2, df)}$$

7.3 REQUIRED SAMPLE SIZE

proportion:
$$n = \hat{p}\hat{q} \left(\frac{z_{\alpha/2}}{E}\right)^2$$
, (70)
 $(\hat{p} = \hat{q} = 0.5 \text{ if unknown})$

mean:
$$n = \left(\frac{z_{\alpha/2} \cdot \hat{\sigma}}{E}\right)^2$$
 (71)

8 Hypothesis Tests

Test statistic and R function (when available) are listed for each.

Optional arguments for **hypothesis tests**:

alternative="two.sided" can be:

"two.sided", "less", "greater"

conf.level=0.95 constructs a 95% confidence interval. Standard CI
 only when alternative="two.sided".

Optional arguments for power calculations & Type II error:

alternative="two.sided" can be:

"two.sided" or "one.sided"

sig.level=0.05 sets the significance level α .

8.1 1-SAMPLE PROPORTION

 $H_0: p = p_0$

prop.test(x, n, $p=p_0$, alternative="two.sided")

$$z = \frac{\hat{p} - p_0}{\sqrt{p_0 q_0/n}} \tag{72}$$

8.2 1-SAMPLE MEAN (σ KNOWN)

 $H_0: \mu = \mu_0$

$$z = \frac{\bar{x} - \mu_0}{\sigma / \sqrt{n}} \tag{73}$$

8.3 1-SAMPLE MEAN (σ UNKNOWN)

 $H_0: \mu = \mu_0$

t.test(x, $mu=\mu_0$, alternative="two.sided")

Where **x** is a vector of sample data.

$$t = \frac{\bar{x} - \mu_0}{s / \sqrt{n}}, \quad df = n - 1 \tag{74}$$

Required Sample size:

power.t.test(delta=h, sd = σ , sig.level= α , power=1 - β , type ="one.sample", alternative="two.sided")

8.4 2-SAMPLE PROPORTION TEST

 $H_0: p_1 = p_2$ or equivalently $H_0: \Delta p = 0$

prop.test(x, n, alternative="two.sided")

where: $\mathbf{x}=\mathbf{c}(x_1, x_2)$ and $\mathbf{n}=\mathbf{c}(n_1, n_2)$

$$z = \frac{\Delta \hat{p} - \Delta p_0}{\sqrt{\frac{\hat{p}q}{n_1} + \frac{\hat{p}q}{n_2}}}, \quad \Delta \hat{p} = \hat{p}_1 - \hat{p}_2$$
 (75)

$$\bar{p} = \frac{x_1 + x_2}{n_1 + n_2}, \quad \bar{q} = 1 - \bar{p}$$
 (76)

Required Sample size:

power.prop.test(p1= p_1 , p2= p_2 , power= $1-\beta$, sig.level= α , alternative="two.sided")

8.5 2-SAMPLE MEAN TEST

 $H_0: \mu_1 = \mu_2$ or equivalently $H_0: \Delta \mu = 0$

t.test(x1, x2, alternative="two.sided")

where: x1 and x2 are vectors of sample 1 and sample 2 data.

$$t = \frac{\Delta \bar{x} - \Delta \mu_0}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}} \quad df \approx \min(n_1 - 1, n_2 - 1), \quad \Delta \bar{x} = \bar{x}_1 - \bar{x}_2$$
 (77)

Required Sample size:

power.t.test(delta=h, sd = σ , sig.level= α , power=1 - β , type ="two.sample", alternative="two.sided")

8.6 2-SAMPLE MATCHED PAIRS TEST

 $H_0: \mu_d = 0$

t.test(x, y, paired=TRUE, alternative="two.sided") where: x and y are ordered vectors of sample 1 and sample 2 data.

$$t = \frac{\bar{d} - \mu_{d0}}{s_d / \sqrt{n}}, \quad d_i = x_i - y_i, \quad df = n - 1$$
 (78)

Required Sample size:

power.t.test(delta=h, sd = σ , sig.level= α , power=1 - β , type ="paired", alternative="two.sided")

8.7 Test of homogeneity, test of independence

 $H_0: p_1 = p_2 = \cdots = p_n$ (homogeneity)

 $H_0: X$ and Y are independent (independence)

chisq.test(D)

Enter table: D=data.frame(c1, c2, ...), where c1, c2, ... are column data vectors.

Or generate table: D=table(x1, x2), where x1, x2 are ordered vectors of raw categorical data.

$$\chi^2 = \sum \frac{(O_i - E_i)^2}{E_i}, \quad df = (\text{num rows - 1})(\text{num cols - 1})$$
 (79)

$$E_i = \frac{\text{(row total)(column total)}}{\text{(grand total)}} = np_i \tag{80}$$

For 2×2 contingency tables, you can use the Fisher Exact Test:

fisher.test(D, alternative="greater")

(must specify alternative as greater)

9 Linear Regression

9.1 LINEAR CORRELATION

 $H_0: \rho = 0$

cor.test(x, y)

where: \mathbf{x} and \mathbf{y} are ordered vectors.

$$r = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{(n-1)s_x s_y}, \quad t = \frac{r-0}{\sqrt{\frac{1-r^2}{n-2}}}, \quad df = n-2$$
 (81)

9.2 MODELS IN R

EQUATION	R MODEL
$y = b_0 + b_1 x_1$	y∼x1
$y = 0 + b_1 x_1$	y~0+x1
$y = b_0 + b_1 x_1 + b_2 x_2$	y~x1+x2
$y = b_0 + b_1 x_1 + b_2 x_2 + b_{12} x_1 x_2$	y~x1+x2+x1*x2
$y = b_0 + b_1 x_1 + b_2 x_2^2$	y~x1+I (x2 [^] 2)
	$y = b_0 + b_1 x_1$ $y = 0 + b_1 x_1$ $y = b_0 + b_1 x_1 + b_2 x_2$ $y = b_0 + b_1 x_1 + b_2 x_2 + b_{12} x_1 x_2$

9.3 REGRESSION

Simple linear regression steps:

- 1. Make sure there is a significant linear correlation.
- 2. results= $lm(y \sim x)$ Linear regression of y on x vectors
- 3. results View the results
- 4. plot(x, y); abline(results) Plot regression line on data
- 5. plot(x, results\$residuals) Plot residuals

$$y = b_0 + b_1 x_1 \tag{82}$$

$$b_1 = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2}$$
 (83)

$$b_0 = \bar{\mathbf{y}} - b_1 \bar{\mathbf{x}} \tag{84}$$

9.4 PREDICTION INTERVALS

To predict y when x = 5 and show the 95% prediction interval with regression model in results:

predict(results, newdata=data.frame(x=5),
int="pred")

10 ANOVA

10.1 ONE WAY ANOVA

- results=aov(depVarColName~indepVarColName, data=tableName) Run ANOVA with data in TableName, factor data in indepVarColName column, and response data in depVarColName column.
- 2. summary (results) Summarize results
- boxplot (depVarColName~indepVarColName, data=tableName) Boxplot of levels for factor

$$F = \frac{MS(\text{treatment})}{MS(\text{error})}, \quad df_1 = k - 1, \, df_2 = N - k$$
 (85)

To find required sample size and power see power.anova.test(...)

11 Loading and using external data and tables

11.1 LOADING EXCEL DATA

- 1. Export your table as a CSV file (comma seperated file) from Excel.
- Import your table into MyTable in R using:
 MyTable=read.csv(file.choose())

11.2 LOADING AN .RDATA FILE

You can either double click on the .RData file or use the menu:

- Windows: File→Load Workspace...
- Mac: Workspace→Load Workspace File...

11.3 Using tables of data

- 1. To see all the available variables type: 1s ()
- 2. To see what's inside a variable, type its name.
- If the variable tableName is a table, you can also type names (tableName) to see the column names or type head (tableName) to see the first few rows of data.
- 4. To access a column of data type tableName\$columnName

An example demonstrating how to get the women's height data and find the mean:

- > ls() # See what variables are defined
- [1] "women" "x"
- > head(women) #Look at the first few entries
 height weight
- 58 115
- 2 59 117
- 3 60 120
- > names(women) # Just get the column names
- [1] "height" "weight"
- > women\$height # Display the height data
- [1] 58 59 60 61 62 63 64 65 66 67 68 69 70 71 72
- > mean(women\$height) # Find the mean of the heights [1] 65