## Many useful Excel and R functions

Notes：NA＝not applicable；1：，2：．．＝alternative methods to do the same thing；\＃．．．＝comments if necessary

| Description | Excel（Calc usually too） | R |
| :---: | :---: | :---: |
| Getting started |  |  |
| Install program | you probably already have it | http：／／cran．r－project．org／ |
| Update program | spend money | ```1: http://cran.r-project.org/ 2: Within R (Windows): \# install, move, update, quit: if(!require(installr)) \{ install.packages("installr");requir e(installr) \(\}\) updateR(F, T, T, F, T, F, T)``` |
| Run command C | ＝C \＃＂＝＂everywhere below too | C \＃no need for＂＝＂before it |
| Get help on function x | 1：click $f x$ symbol，find function， double－click it，click 函數說明 <br> 2：search the web | 1：？x \＃Needs exact match <br> 2：help（＂x＂）\＃same as ？x <br> 3：？？x \＃Fuzzy match <br> 4：search the web <br> \＃Usage shows syntax and <br> \＃defaults；Arguments shows <br> \＃input；Value shows output |
| Get help on general function F that works differently for object types A vs．B | NA | $\begin{array}{\|l\|} \hline \text { ?F.A vs. ?F.B } \\ \text { \# For example: } \\ \text { ?summary.lm } \\ \text { ?summary.glm } \\ \text { ?summary.aov } \\ \text { ?summary.table } \\ \text { ?plot.table } \\ \hline \end{array}$ |
| Put value V into variable x | type V into cell x | $\begin{aligned} & 1: x=V \\ & 2: x<-V \\ & 3: V->x \end{aligned}$ |
| Put value V into both x and y | type V into cell $x$ ，drag to y | $\mathrm{x}=\mathrm{y}=\mathrm{V}$ \＃Cool！ |
| Load tab－delimited file＂F＂into data frame D ，first row as variable names | 1：copy／paste from text file <br> 2：open file within Excel | $\begin{aligned} & 1: \mathrm{D}=\text { read.delim("F") } \\ & \text { 2: } \mathrm{D}=\text { read.table("F",T) } \end{aligned}$ |
| Load tab－delimited file on the web at http：／／www／F into data frame D | use File／Open，then write／paste http：／／www／F | D＝read．delim（＂http：／／www／F＂） |
| Load space－delimited file＂F＂into data frame D，first row as variable names | same as above，but then split by <br> space（空格） | $\begin{aligned} & \text { 1: D = read.table("F",T) } \\ & \text { 2: D = read.delim("F",sep=" ") } \end{aligned}$ |
| Load comma－delimited file＂F＂into data frame $D$ ，first raw as variable names | 1：open within Excel， splitting columns by＂，＂ 2：copy／paste from text file，then split columns by ＂，＂ | $\begin{array}{\|l} \hline \text { 1: } \mathrm{D}=\text { read.csv("F") } \\ \text { 2: } \mathrm{D}=\text { read.table("F", } \\ \text { sep=",", header=T) } \end{array}$ |
| Object x inside object O（e．g．data frame） | click appropriate row or column | $\begin{array}{\|l\|} \hline \text { 1: O\$x } \\ \text { 2: } \operatorname{attach}(\mathrm{O}) ; \mathrm{x} \text {; detach(O) } \\ \text { \# "\$" also applies to function } \\ \text { \# outputs, e.g.: } \\ \text { summary( } \operatorname{lm}(\mathrm{y} \sim \mathrm{x})) \text { ) } \text { residuals } \\ \hline \end{array}$ |
| Show the local file directory | NA | $\operatorname{dir}()$ |
| Timing script S | NA | $\begin{aligned} & \text { now }=\text { proc.time() } \\ & \mathrm{S} \\ & \text { proc.time() }- \text { now } \end{aligned}$ |


| Vectors，matrices，lists and data frames |  |  |
| :---: | :---: | :---: |
| Create the vector of numbers $\mathrm{x}, \mathrm{y}, \mathrm{z}$ | $\mathrm{x}, \mathrm{y}, \mathrm{z}$ in adjacent cells （vertical or horizontal） | $\mathrm{c}(\mathrm{x}, \mathrm{y}, \mathrm{z})$ |
| Omit NA（not available）data in object O | Math functions automatically ignore strings like＂NA＂ | na．omit（O） |
| Create number series $1,2, \ldots \mathrm{n}$ | type $1 \& 2$ ，select，then drag lower right corner | 1：n |
| Create number sequence $1,3,5, \ldots, \mathrm{n}$ | type $1 \& 3$ ，then drag corner | seq（1，n，by＝2） |
| Repeat number x for n times | drag cell with x from corner | rep（x，n） |
| Add 1 to the numbers 2，5，7 to get 3，6，8 | NA | 1＋c（ $2,5,7)$ \＃grammatical！ |
| Number of values in vector x | COUNT（x）\＃only numbers | length（x）\＃numbers or strings（all same type） |
| Number of values in vector x that are greater than y | COUNTIF（x，＂＞y＂） | length（ $\mathrm{x}[\mathrm{x}>\mathrm{y}$ ］） |
| Convert number x into string＂x＂ | $\begin{aligned} & \text { TEXT(x,"\#.\#\#") \# } 2 \\ & \text { decimals } \end{aligned}$ | as．character（x） |
| Look up x in table T，find what＇s in column C（in x＇s row） | VLOOKUP（x，T，C，FALSE <br> ） | $\mathrm{C}[\mathrm{T}=\mathrm{=} \mathrm{x}]$ |
| Create data frame D with columns x \＆y | NA | $\mathrm{D}=$ data．frame（ $\mathrm{x}, \mathrm{y}$ ） |
| Create data frame D1 that＇s a subset of data frame D，such that $\mathrm{x}>1$ | NA | D1 $=\operatorname{subset}(\mathrm{D}, \mathrm{D} \$ \mathrm{x}>1)$ |
| Count number of rows in data frame D | COUNT（D）\＃select a column | nrow（D） |
| Count number of columns in data frame D | COUNT（D）\＃select a row | $\mathrm{ncol}(\mathrm{D})$ |
| Put columns x and y side by side | copy／paste them as you like | cbind（x，y） |
| Put rows x and y one on top of the other | copy／paste them as you like | rbind（x，y） |
| Create a vector V with N zeros | type 0 ，drag corner | $\mathrm{V}=$ numeric（N） |
| Create an empty matrix M with C columns and R rows（all＂NA＂＝not available） | type＂NA＂，drag corner down to make column， then drag again rightward （or vice versa） | $\mathrm{M}=$ matrix（ （ $\mathrm{col}=\mathrm{C}$ ，nrow $=\mathrm{R}$ ） |
| Create the matrix a c <br> b d | type a，b，c，d into the appropriate cells | $\begin{aligned} & \text { 1: matrix(c(a,b,c,d),nrow=2) } \\ & \text { 2: matrix(c(a,b,c,d),ncol=2) } \\ & \hline \end{aligned}$ |
| Flip（transpose） $\mathrm{n} \times \mathrm{m}$ matrix M （ n rows， m columns）into $\mathrm{m} \times \mathrm{n}$ matrix $\mathrm{M}^{\prime}$ | copy matrix，paste in new place using Paste Special （選擇性貼上）and Transpose（轉置） | t（M） |
| Add column names＂A＂\＆＂B＂to two－ column matrix M | NA | colnames（M）＝c（＂A＂，＂B＂） |
| Add row names＂A＂\＆＂B＂to two－row matrix M | NA | rownames（M）＝c（＂C＂，＂D＂） |
| Show column and row names of matrix M | NA | colnames（M）；rownames（M） |
| Show column names in data frame D | NA | $\begin{aligned} & \hline \text { 1: names(D) } \\ & \text { 2: colnames(D) } \\ & \hline \end{aligned}$ |
| For vector x ，find the ith position | click on the appropriate cell | x［i］ |
| For the data frame（or matrix） x ，find the ith row and jth column | click on the appropriate cell | $\mathrm{x}[\mathrm{i}, \mathrm{j}]$ |
| All values in data frame D on row x | click row number x | D［x，］ |
| All values in data frame D in ith column named＂x＂ | click column letter＂x＂ | 1：D［，i］\＃Using number <br> 2：D［，＂x＂］\＃Using name |
| Show first six rows of data frame D | scroll to the top of the sheet | head（D） |


| Show last six rows of data frame D | scroll to the bottom of the sheet | tail(D) |
| :---: | :---: | :---: |
| Sort column x into alphanumerical order | use $\mathrm{A} \rightarrow \mathrm{Z}$ dialog box | sort(x) |
| Sort columns x and y in data frame D into the order defined by x | use $\mathrm{A} \rightarrow \mathrm{Z}$ dialog box | 1: $\mathrm{D}[\operatorname{order}(\mathrm{D} \$ \mathrm{x})$, <br> 2: library (dplyr) arrange(D, x) |
| Remove repeats in vector A | sort column A , then in B 2 : IF(A2=A1,"",A2), drag down, copy/paste value, sort column B | unique(A) |
| Combine tables D1 and D2 by matching same-named column x | use VLOOKUP cleverly | merge(D1,D2) |
| Combine smaller item, subject, and response files I, S, R into one large file | use VLOOKUP cleverly | merge(R,I) \# Same item IDs <br> merge(R,S) \# Same subj IDs |
| Create a list with number 1 and string "a" | just type/paste into cells | list(1,"a") \# c() can't do this |
| Create a list L of vectors (1,2) and ( $3,4,5$ ) | NA | $\mathrm{L}=\operatorname{list}(\mathrm{c}(1,2), \mathrm{c}(3,4,5))$ |
| Second element in first element in list L | NA | L[[1]][2] \# e.g. = 2 for above |
| Split vector or data frame X by factor F | NA | split(X,F) \# Outputs a list |
| Check if vector x elements are in vector y | NA | is.element(x,y) |
| Remove elements that are in vector x from vector y | NA | $\begin{aligned} & \text { 1: } \mathrm{y}[\mathrm{y}!=\mathrm{x}] \\ & \text { 2: } \operatorname{setdiff(y,x)} \end{aligned}$ |
| Cut continuous values in vector x into n equal-sized bins, creating new factor $B$ | NA | $\mathrm{B}=\operatorname{cut}(\mathrm{x}, \mathrm{n})$ |
| Logic |  |  |
| True | TRUE | $\begin{aligned} & \text { 1: TRUE } \\ & \text { 2: T \# never "true" or "t" } \end{aligned}$ |
| False | FALSE | $\begin{aligned} & \text { 1: FALSE } \\ & \text { 2: F \# never "false" or "f" } \end{aligned}$ |
| If $x$ is true then value $y$, otherwise value $z$ | IF(x,y,z) | if (x) $\{\mathrm{y}\}$ else $\{\mathrm{z}\}$ |
| If $x$ is true then command $y$, otherwise command z | NA | if (x) $\{\mathrm{y}\}$ else $\{\mathrm{z}\}$ |
| x equals y (true or false) | $\mathrm{x}=\mathrm{y}$ | $\mathrm{x}==\mathrm{y}$ |
| $x$ doesn't equal y (true or false) | $\mathrm{x}<>\mathrm{y}$ | x! $=\mathrm{y}$ |
| $x$ and $y$ (true only if both $x$ and $y$ are true) | AND(x,y) | $x$ \& y |
| x or y (true if either x and/or y is true) | OR(x,y) | $\mathrm{x} \mid \mathrm{y}$ |
| Convert logical x into 0 ( F ) or 1 (T) | if(x, 1,0) | $\begin{aligned} & \text { 1:1*x } \\ & \text { 2: as.numeric }(\mathrm{x}) \\ & \hline \end{aligned}$ |
| Functions and packages |  |  |
| Add comment y after R code line x | NA | x \# y |
| Run command x , then command y | NA | $\begin{gathered} 1: x \\ y \\ 2: x ; y \end{gathered}$ |
| Repeat command x for n times (for-loop) | NA | for (i in 1:n) $\{\mathrm{x}\}$ \# 1:n is vector! |
| Print out "JM" one letter at a time | NA | for (i in c("J","M")) \{print(i)\} |
| Create a new function Fun that takes argument x and outputs value y | need to use VBA to create a macro (search web for help) | $\begin{aligned} & \text { Fun }=\text { function }(x)\{ \\ & \text { return }(\mathrm{y}) \\ & \} \end{aligned}$ |
| Compute means of rows in matrix M | AVERAGE(row), drag down | apply(M,1,mean) |
| Find sum of columns in matrix M | AVERAGE(col), drag right | apply(M,2,sum) |
| Compute by-subj means for variable x in data set D <br> \# or any one-argument function (e.g. sum) | AVERAGE(x) \# assumes subj defines rows (columns) and x values are in a matrix | ```1: apply(D$x,D$subj,mean) 2: library(dplyr) summarize(group_by(D, subj), mean(x))``` |
| Compute means for variable $y$ when another variable $\mathrm{x}>23$ | AVERAGEIFS(y,x,">23") \# assumes x and y are columns like in R | mean(y[x>23]) |


| Compute means for variable y for factors <br> A (A1 vs. A2) \& B (B1 vs. B2), and put <br> them in a table | DAVERAGE(database,field, <br> criteria) <br> \# "database" = R-style data <br> \# "field" = factor name <br> \# "criteria" = minitable like: <br> A B (factor names) <br> A1 B2 (one level each) | tapply(y,list(A,B),mean) <br> \# more factors and more levels <br> also work |
| :--- | :--- | :--- |


| Put six plots into a 2－row by 3－column arrangement | lots of plotting | 1：par（mfrow＝c（2，3）） <br> for（i in 1：6）$\{\operatorname{plot}(\operatorname{runif}(10))\}$ <br> 2：layout（matrix（1：6，nrow＝2）） <br> for（i in 1：6）$\{\operatorname{plot}($ runif（10））$\}$ |
| :---: | :---: | :---: |
| Make a scatterplot of vectors x and y with x －axis label＂Age＂， y －axis label <br> ＂Accuracy＂，with $x$ values from 0.5 to 1 and y values from 0 to 0.5 <br> \＃Same methods work for most plots， including histograms | poke around chart menu | ```plot(x,y, xlab = "Age", ylab = "Accuracy", xlim=c(0.5,1), ylim=c(0,0.5) )``` |
| Make x \＆y scatterplot with no numbers | poke around chart menu | plot（x，y，xaxt＝＂n＂，yaxt＝＂n＂） |
| Add horizontal line to existing plot at $\mathrm{y}=3$ | NA | abline（h＝3） |
| Add vertical line to existing plot at $\mathrm{x}=7$ | NA | abline（v＝7） |
| Make a line plot of $x$（on $x$－axis）and $y$（on $y$－axis） | poke around chart menu | ```plot(x,y,type="l") \# Make sure x is sorted first! \# type="1": line; \# type="p" (points) default; \# ?plot for other types \# ?points pch for other dot \# shapes``` |
| Make a bar graph of crossed values of $\mathrm{Y}=$ $\mathrm{a}, \mathrm{b}, \mathrm{c}, \mathrm{d}$ as a function of factors $\mathrm{F}=\mathrm{F} 1, \mathrm{~F} 2$ and $\mathrm{G}=\mathrm{G} 1, \mathrm{G} 2$ in matrix M ，with y －axis starting at zero，and M and barplot like so： | adjust Excel＇s automatic $y$－axis to start it at zero （recommended by many statisticians to make scale clear）；R acts like Excel here，in treating columns $(\mathrm{F})$ as the main label（in names．arg）and rows（G） as legend label（in legend．test） | ```1:M = matrix(c(a,b,c,d), nrow=2) \(\operatorname{barplot}(\mathrm{M}\), beside \(=\mathrm{T}\), names.arg=c("F1","F2"), legend.text=c("G1","G2"), ylab = "Y") 2: search web or books for help on using ggplot2``` |
| Make same bar graph as above，but use $y$－ axis range $a$ to $b$ ，where $a$ is not zero | poke around chart menu | $\begin{aligned} & \text { barplot(M, beside=T, } \\ & \text { legend.text=c("F1","F2"), } \\ & \text { ylab = "Y", ylim=c(a,b), } \\ & \text { xpd=F, \# Keep bars inside } \\ & \text { xaxt="n") \# No x label yet } \\ & \text { axis(side=1, at=c(2,5), } \\ & \text { labels=c("G1", "G2")) } \\ & \text { box(bty="1") \# lower-case L } \\ & \hline \end{aligned}$ |
| Plot standard histogram of sample $\mathrm{S}(\mathrm{S}=$ vector of numbers） | Analysis toolbox：直方圖 \＃use about 10 equal－sized bins | hist（S） |
| Change number／size of bars in histogram for S （remember for histograms，bar area is what matters，not bar height） | Analysis toolbox：直方圖 \＃enter different bins | 1：hist（S，breaks＝3）\＃ 3 bars <br> 2：hist（S，breaks＝c $(0,10))$ <br> \＃Breaks at points 0 and 10 |
| Make box（and whiskers）plot | NA | boxplot（．．．）\＃cf．？boxplot |
| Plot density of sample S | NA | plot（density（S）） |
| Make line plot with solid line for variable x 1 and dashed line for variable x 2 with dependent variable y | poke around chart menu | $\begin{aligned} & \text { plot(x1,y,type="1") } \\ & \text { lines(x2,y,lty=2) } \\ & \text { \# lty is line type } \\ & \text { \# lty=1 (solid) is default } \\ & \text { \# lty=2 is dashed } \\ & \text { \# lty }=3 \text { is dotted } \\ & \text { \# lwd=2 is wider } \end{aligned}$ |
| Add a legend at the top right for a line plot showing that the solid line represents Cats and the dashed line represents Dogs | poke around chart menu | $\begin{aligned} & \hline \text { legend("topright", } \\ & \text { lty=c(1,2), } \\ & \text { legend=c("Cats","Dog")) } \end{aligned}$ |


| Add upper＋lower error bars to bar plot B with n means M （vector），where each half of the error bar has length E （e．g．， $\mathrm{E}=1$ sd，or $E=S E$ ，or $E=$ one half of the $95 \%$ confidence interval） | make bar plot，search menu for error bars，enter values you want | ```1:source("http://www.ccunix. ccu.edu.tw/~Ingproc/ errorbar_Rcode.txt") E.bars \(=\operatorname{rep}(E, n)\) error.bar(B,M,E.bars) 2: library(ggplot2) B + geom_errorbar( aes(ymin=M-E, \(y \max =\mathrm{M}+\mathrm{E})\) )``` |
| :---: | :---: | :---: |
| Add linear regression line to scatter plot（ x on x －axis， y on y －axis） | right－click dots，choose 加上趨勢線，then keep 線性 default | $\begin{aligned} & \text { plot }(x, y) \\ & \text { abline }(\operatorname{lm}(y \sim x)) \end{aligned}$ |
| Add local regression line to scatter plot（x on x －axis， y on y －axis） | right－click dots，choose 加上趨勢線，then choose 移動平均 | $\begin{aligned} & \text { plot(x,y) } \\ & \text { lines(predict }(\operatorname{loess}(\mathrm{y} \sim \mathrm{x})) \text { ) } \\ & \text { \# sort } \mathrm{x} \text { first } \end{aligned}$ |
| Make trellis plot for scatterplot $\mathrm{y} \sim \mathrm{x} 1$＊ x 2 （ $\mathrm{y}, \mathrm{x} 1$ ，x2 all numerical，and you want to visualize the $\mathrm{x} 1 \times \mathrm{x} 2$ interaction）in data frame D，with linear best－fit lines for each | sort data by x 1 ，divide x 1 into a few（3－6）subsets， plot $\mathrm{y} \sim \mathrm{x} 2$ for each subset （like method 3 for R ） | ```1: library(lattice) \(\mathrm{x} 1 . \mathrm{eq}=\) equal.count \((\mathrm{D} \$ \mathrm{x} 1)\) xyplot( \(\mathrm{D} \$ \mathrm{y} \sim \mathrm{D} \$ \mathrm{x} 2 \mid \mathrm{x} 1 . \mathrm{eq}\), panel \(=\) function \((x, y)\{\) panel.xyplot( \(\mathrm{x}, \mathrm{y}\) ) panel.abline \((\operatorname{lm}(y \sim x))\) \} ) 2: library(ggplot2) D\$x1cuts \(=\operatorname{cut}(D \$ x 1,7)\) qplot( \(\mathrm{y}, \mathrm{x} 2\), data \(=\mathrm{D}\), facets \(=\sim x 1\) cuts \()+\) stat_smooth(method =" 1 m ") 3: \(\operatorname{par}(\) mfrow \(=c(2,3)) \# 6\) plots \(\mathrm{D}=\mathrm{D}[\operatorname{order}(\mathrm{D} \$ \mathrm{x} 1)\),] \(\mathrm{N}=\operatorname{nrow}(\mathrm{D})\) \(\mathrm{n}=\) ceiling(N/6) rangey \(=\operatorname{range}(\mathrm{D} \$ \mathrm{y})\) rangex2 \(=\operatorname{range}(\mathrm{D} \$ \mathrm{x} 2)\) for (i in 1:6) \{ \(\operatorname{minx} 1=\mathrm{D} \$ \mathrm{x} 1[\mathrm{n} *(\mathrm{i}-1)+1]\) \(\operatorname{maxx} 1=\mathrm{D} \$ \mathrm{x} 1[\min (\mathrm{n} * \mathrm{i}, \mathrm{N})]\) D. \(\mathrm{i}=\operatorname{subset}(\mathrm{D}\), (D\$x1>= minx1 \& D \(\$ \mathrm{x} 1<=\operatorname{maxx} 1\) )) \(\operatorname{plot}(\mathrm{D} . \mathrm{i} \$ \mathrm{x} 2, \mathrm{D} . \mathrm{i} \$ \mathrm{y}\), xlab="x2", ylab="y", main = paste("x1: from",minx1, "to", \(\max 1\) 1)) abline \((\operatorname{lm}(\mathrm{y} \sim \mathrm{x} 2\), data=D.i)) \}``` |


| Make trellis plot for scatterplot of $\mathrm{y} \sim \mathrm{x}$ with linear best－fit lines，with grouping unit g （ y \＆ x numerical）in data frame D \＃Useful for LME and GLMM too | sort data by g ，plot $\mathrm{y} \sim \mathrm{x}$ for each $g$（like method 3 for R ） | ```1: library(lattice) xyplot( \(\mathrm{y} \sim \mathrm{x} \mid\) factor \((\mathrm{g})\), data = D) 2: library (ggplot2) qplot( \(\mathrm{x}, \mathrm{y}\), data=D, facets \(=\sim \mathrm{g})+\) stat_smooth(method =" 1 m ") 3: \(\operatorname{par}(m f r o w=c(n, m))\) \(\# \mathrm{n} \& \mathrm{~m}\) divide up g neatly rangex \(=\) range \((\mathrm{D} \$ \mathrm{x})\) rangey \(=\operatorname{range}(\mathrm{D} \$ \mathrm{y})\) for (i in 1:length \((\mathrm{g})\) ) \{ D. \(\mathrm{i}=\operatorname{subset}(\mathrm{D}, \mathrm{D}[\mathrm{D} \$ \mathrm{~g}==\mathrm{i}])\) \(\operatorname{plot}(\mathrm{D} \$ \mathrm{x}, \mathrm{D} \$ \mathrm{y}\), main \(=\mathrm{i}\), xlim = rangex, ylim=rangey) \}``` |
| :---: | :---: | :---: |
| Make 3D scatterplot（x on x－axis， y on y － axis， z on z －axis） | NA | $\begin{aligned} & \text { library }(\mathrm{rgl}) \\ & \operatorname{plot} 3 \mathrm{~d}(\mathrm{x}, \mathrm{y}, \mathrm{z}) \end{aligned}$ |
| Make 3D scatterplot，split into a series of 2D scatterplots（ $x \& y=$ independent variables， $\mathrm{z}=$ dependent variable） | NA | library（ggplot2） $y . \operatorname{cut}=\operatorname{cut}(\mathrm{y}, 7)$ $\mathrm{qplot}(\mathrm{z}, \mathrm{x}$, facets $=\sim \mathrm{y} . \mathrm{cut})$ |
| Make mosaic plot of contingency table T | NA | mosaicplot（T） |
| Plot logistic regression model L for $\mathrm{y} \sim \mathrm{x}$ | sort data by x ，divide y into bins，within each bin convert y to logits： LN（AVERAGE（y）／ （1－AVERAGE（y）） make scatterplot of $\log \operatorname{it}(\mathrm{y}) \sim \mathrm{x}$ ， right－click dots，choose 加上趨勢線，then keep 線性 default <br> \＃Like method 2 for R | ```1: plot(x,y) curve(predict(L, data.frame(x=x), type="response"), add=T) 2: bins = cut(x,10) # Or more logit.bin = function(x){ prob1 = mean(c(x,0,1)) prob0 = 1-prob1 return(log(prob1/prob0)) } meanx = tapply(x, bins, mean) logity = tapply(y, bins, logit.bin) plot(meanx, logity) abline(lm(logity~meanx))``` |
| Descriptive statistics |  |  |
| Make a frequency table for sample S | 1：Analysis toolbox：直方 <br> 2：see handout for word frequency example | $\begin{aligned} & \hline \text { 1: xtabs }(\sim \mathrm{S}) \\ & \text { 2: table }(\mathrm{S}) \end{aligned}$ |
| Make a frequency table cross－classified by factors x and y （ $\mathrm{x}=$ row， $\mathrm{y}=$ columns） | basically do it by hand | $\begin{aligned} & \text { 1: } \operatorname{xtabs}(\sim \mathrm{x}+\mathrm{y}) \\ & \text { 2: table }(\mathrm{x}, \mathrm{y}) \\ & \hline \end{aligned}$ |
| Mean of sample S | $\begin{aligned} & \text { 1: AVERAGE(S) } \\ & \text { 2: } \operatorname{SUM}(\mathrm{S}) / \operatorname{COUNT}(\mathrm{S}) \\ & \hline \end{aligned}$ | $\begin{aligned} & \text { 1: mean(S) } \\ & \text { 2: } \operatorname{sum}(\mathrm{S}) / \text { length }(\mathrm{S}) \\ & \hline \end{aligned}$ |
| Median of sample S | MEDIAN（S） | median（S） |
| Mode of sample S | MODE（S） | $\begin{aligned} & \text { as.numeric(names(sort( } \\ & \text {-table(S)))[1]) \# Handout code } \\ & \text { doesn't work (sorry) } \end{aligned}$ |
| Sample standard deviation of sample S | STDEV（S） | $\begin{aligned} & \text { 1: } \operatorname{sd}(\mathrm{S}) \\ & \text { 2: } \operatorname{sqrt(\operatorname {sum}((\mathrm {S}-\operatorname {mean}(\mathrm {S}))^{\wedge }2)/} \\ & (\operatorname{length}(\mathrm{S})-1)) \end{aligned}$ |
| Sample variance of sample S | $\begin{aligned} & \text { 1: } \operatorname{VAR}(\mathrm{S}) \\ & \text { 2: } \operatorname{STDEV}(\mathrm{S})^{\wedge} 2 \end{aligned}$ | $\begin{aligned} & \text { 1: } \operatorname{var}(\mathrm{S}) \\ & \text { 2: } \operatorname{sd}(\mathrm{S})^{\wedge} 2 \\ & \hline \end{aligned}$ |


| Randomness and permutations |  |  |
| :--- | :--- | :--- |
| Reset randomizer | Microsoft won＇t say | set．seed（1）\＃or any number |
| Given $x$ things，calculate how many ways <br> to choose y things | COMBIN（x，y） | choose（x，y） |
| Randomly select x values between 0 and 1 <br> with equal probability | copy and paste RAND（）x <br> times | runif（x） |
| Randomly select x values from a normal <br> （Gaussian）distribution with mean M and <br> standard deviation s | Analysis toolbox： <br> 亂數產生器 | rnorm（x，M，s） |
| Distributions |  |  |


| Convert factor F into an ordinal factor | NA | $\begin{aligned} & \text { 1: } \mathrm{F}=\operatorname{ordered}(\mathrm{F}) \\ & 2: \mathrm{F}=\text { factor }(\mathrm{F}, \text { ordered }=\mathrm{T}) \\ & \text { \# Creates polynomial coding: } \\ & \text { \# F.L }=\text { linear component; } \\ & \text { \# F.Q }=\mathrm{F}^{\wedge} 2 \text { (quadratic) } \\ & \text { \# F.C }=\mathrm{F}^{\wedge} 3 \text { (cubic) } \end{aligned}$ |
| :---: | :---: | :---: |
| Relevel factor F （levels A and B）so that B is the reference level（ 0 in dummy coding） | NA | $\mathrm{F}=\operatorname{relevel}(\mathrm{F}, \mathrm{B}$＂） |
| Convert factor F（levels A，B，C）into effect（sum）coding，splitting F into FB （ $\mathrm{A}=0, \mathrm{~B}=1, \mathrm{C}=-1$ ）and $\mathrm{FC}(\mathrm{A}=0, \mathrm{~B}=-1$ ， $\mathrm{C}=1$ ）\＃Effect coding is better if you want to test interactions with F | NA | $\begin{aligned} & \hline \text { contrasts }(\mathrm{F})= \\ & \text { contr.sum }(\text { levels }(\mathrm{F})) \end{aligned}$ |
| Convert factor F（levels A，B）into effect （sum）coding，changing F into FA（ $\mathrm{A}=1$ ， $\mathrm{B}=-1$ ） | NA | $\begin{aligned} & \text { contrasts }(\mathrm{F})= \\ & \text { contr.sum }(\text { levels }(\mathrm{F})) \end{aligned}$ |
| Convert factor F （levels $\mathrm{A}, \mathrm{B}$ ）into effect （sum）coding，changing $F$ into $F B$（ $A=-1$ ， $\mathrm{B}=1$ ）\＃Safer than above，in my experience | NA | $\mathrm{FB}=2 *\left(\mathrm{~F}=={ }^{\text {＂}} \mathrm{B}\right.$＂$)-1$ |
| $z, t$, and $F$ tests |  |  |
| Two－tailed p value for one－sample $z$ test with population $\mu$ and $\sigma$ and sample S | $\begin{aligned} & \hline \hline \mathrm{z}=(\text { AVERAGE(S) }-\mu) / \\ & (\sigma / \text { SQRT(COUNT(S) })) \\ & \mathrm{p}=2 * \operatorname{NORMSDIST}(- \\ & \text { ABS }(\mathrm{z})) \end{aligned}$ | $\begin{gathered} \hline \hline \mathrm{z}=(\operatorname{mean}(\mathrm{S})-\mu) / \\ (\sigma / \operatorname{sqrt}(\operatorname{length}(\mathrm{S}))) \\ \mathrm{p}=2 * \operatorname{pnorm}(-\operatorname{abs}(\mathrm{z})) \end{gathered}$ |
| Two－tailed p value for one－sample $t$ test with population $\mu$ and sample S | $\begin{aligned} \mathrm{t}= & (\text { AVERAGE(S)- } \mu) / \\ & (\text { STDEV }(\mathrm{S}) / \\ & \text { SQRT(COUNT(S))) } \\ \mathrm{p}= & \text { TDIST(ABS(t), } \\ & \text { COUNT(S)-1,2) } \end{aligned}$ | $\begin{gathered} 1: \mathrm{t}=(\text { mean }(\mathrm{S})-\mu) / \\ (\operatorname{sd}(\mathrm{S}) / \mathrm{sqrt}(\mathrm{length}(\mathrm{~S}))) \\ \mathrm{p}=2 * \mathrm{pt}(-\mathrm{abs}(\mathrm{t}), \\ \text { df=}=\operatorname{length}(\mathrm{S})-1) \\ \text { 2: t.test }(\mathrm{S}, \mathrm{mu}=\mu) \end{gathered}$ |
| Unpaired $t$ test assuming equal variance （homoscedastic）for a vs． b （levels of factor X ，with dependent variable Y ） | Analysis toolbox： <br> t 檢定：兩個母體平均數差的檢定，假設變異數相等 | $\begin{aligned} & \text { 1: } \operatorname{t.test}(\mathrm{a}, \mathrm{~b}, \text { var.equal }=\mathrm{T}) \\ & \text { 2: } \operatorname{t.test}(\mathrm{Y} \sim \mathrm{X}, \text { var.equal }=\mathrm{T}) \end{aligned}$ |
| Unpaired $t$ test not assuming equal variance（heteroscedastic）for a vs． b （levels of X，with dependent variable Y） | Analysis toolbox： <br> t 檢定：兩個母體平均數差的檢定，假設變異數不相等 | $\begin{aligned} & \text { 1: } \operatorname{t.test}(\mathrm{a}, \mathrm{~b}) \\ & \text { 2: } \mathrm{t} . \operatorname{test}(\mathrm{Y} \sim \mathrm{X}) \end{aligned}$ |
| Paired $t$ test for a vs． b （levels of factor X ，with dependent variable Y ） | Analysis toolbox： <br> t 檢定：成對母體平均數差 <br> 異檢定 | $\begin{aligned} & \text { 1: } \operatorname{t.test}(\mathrm{a}, \mathrm{~b}, \text { paired }=\mathrm{T}) \\ & \text { 2: } \operatorname{t.test}(\mathrm{Y} \sim \mathrm{X}, \text { paired }=\mathrm{T}) \end{aligned}$ |
| One－tailed $p$ value for a certain $F$ value and $d f_{\text {numerator }}$ and $d f_{\text {denominator }}$ | FDIST（F， $\mathrm{df}_{\mathrm{n}}, \mathrm{df}_{\mathrm{d}}$ ） | $\begin{aligned} & \text { 1: } 1-\mathrm{pf}\left(\mathrm{~F}, \mathrm{df}_{\mathrm{n}}, \mathrm{df}_{\mathrm{d}}\right) \\ & \text { 2: } \mathrm{pf}\left(\mathrm{~F}, \mathrm{df} \mathrm{f}_{\mathrm{n}}, \mathrm{df} \mathrm{~d} \text {, lower.tail }=\mathrm{F}\right) \end{aligned}$ |
| One－tailed $F$ test to test if samples a and b come from populations with equal variances，where $\mathrm{S}_{\mathrm{a}}>\mathrm{s}_{\mathrm{b}}$ | Analysis toolbox： <br> F 檢定：兩個常態母體變異數的檢定（a must be to the left b） | $\begin{aligned} & \text { 1: } 1-\operatorname{pf}\left(\mathrm{F}, \mathrm{df}_{\mathrm{a}}, \mathrm{df}_{\mathrm{b}}\right) \\ & \text { 2: } \mathrm{pf}\left(\mathrm{~F}, \mathrm{df}_{\mathrm{a}}, \mathrm{df}_{\mathrm{b}} \text {, lower.tail }=\mathrm{F}\right) \end{aligned}$ |
| Two－tailed $F$ test to test if samples a and b come from populations with equal variances | FTEST（a，b） | var．test（x，y） |
| 95\％confidence interval for $t$ tests | Run analysis toolbox，get critical value and variance to compute using handout formulas | t．test（．．．）gives upper and lower value of confidence interval automatically；to use in graph， must find half its range：（max－ $\min ) / 2$ |


| $\mathrm{x} \%$ confidence interval for $t$ tests | Run analysis toolbox using alpha $=1-x / 100$ ，get critical value and variance to compute using handout formulas | t．test（.. ，conf．level $=x / 100$ ）gives $\mathrm{x} \%$ confidence interval automatically |
| :---: | :---: | :---: |
| Correlation and linear regression analysis |  |  |
| Pearson＇s correlation coefficient $r$（for variables x and y ） | CORREL（x，y） | cor（x，y） |
| Test significance of Pearson＇s correlation coefficient（between x and y ） | use correl－sig．xls or search the Web for tools | $\begin{aligned} & \text { 1: cor.test( } x, y) \\ & \text { 2: summary }(\operatorname{lm}(y \sim x)) \end{aligned}$ |
| Multiple linear regression（ $\mathrm{y}=\mathrm{dep}$ ； x 1 ， x 2 ＝indeps），with data in D | Analysis toolbox：迴歸 | $\begin{aligned} & \text { summary }(\operatorname{lm}(\mathrm{y} \sim \mathrm{x} 1+\mathrm{x} 2 \text {, data }=\mathrm{D})) \\ & \text { \# data argument also used below } \end{aligned}$ |
| Likelihood ratio test for fit of simpler model L0 vs．fit of more complex L1 | NA | $\begin{aligned} & \text { anova(L0,L1) } \\ & \text { \# L0 and L1 created by } \operatorname{lm}(\ldots) \\ & \hline \end{aligned}$ |
| Test significance of indep x 1 in linear model $\mathrm{y} \sim \mathrm{x} 1+\mathrm{x} 2$ | Analysis toolbox：迴歸 | $\begin{aligned} & \text { 1: summary }(\operatorname{lm}(\mathrm{y} \sim \mathrm{x} 1+\mathrm{x} 2)) \\ & \text { 2: } \operatorname{anova(M.no\_ x1,M.has\_ x1)~} \end{aligned}$ |
| Stepwise regression for $\mathrm{y} \sim \mathrm{x} 1+\mathrm{x} 2$ in dataframe D | NA | attach（D） <br> base． $\operatorname{lm}=\operatorname{lm}(\mathrm{y} \sim 1)$ <br> summary（step（base．lm， $\mathrm{y} \sim \mathrm{x} 1+\mathrm{x} 2)$ ） |
| Test independent variables $\mathrm{x} 1, \mathrm{x} 2, \mathrm{x} 3$ for collinearity in dataframe D （dependent variable $=y$ ） | Analysis toolbox：迴歸 Then compute $R^{2}$ for $\mathrm{x} 1 \sim \mathrm{x} 2+\mathrm{x} 3$ ，then use VIF formula in handout | ```1: library(car) \(\operatorname{vif}(\operatorname{lm}(y \sim x 1+x 2+x 3))\) \(\#<5\) is good 2: \(\operatorname{kappa(D[c("x1","x2","x3")])~}\) \# < 30 is good 3: library(languageR) collin.fnc(D[c("x1", "x2", "x3"))\$cnumber``` |
| Get predictions（y－hat）of simple linear model predicting y from x for new data $\mathrm{x}^{\prime}$ | FORECAST（ $\mathrm{x}^{\prime}, \mathrm{y}, \mathrm{x}$ ） <br> \＃ $\mathrm{x}^{\prime}$ is just one value | predict $(\operatorname{lm}(\mathrm{y} \sim \mathrm{x})$ ，newdata $=$ data．frame（ $\left.\mathrm{x}^{\prime}\right)$ ） \＃ $\mathrm{x}^{\prime}$ is a vector；also works for multiple regression |
| Get residuals of a linear model L for dependent variable Y | Analysis toolbox：迴歸， then use coefficients to write equation to predict $y$－ hat，then subtract $y$－hat from real values Y | $\begin{aligned} & \text { 1: } \operatorname{resid}(\mathrm{L}) \\ & \text { 2: Y-predict(L) } \end{aligned}$ |
| Standardize regression coefficients for regression model $\mathrm{y} \sim \mathrm{x} 1+\mathrm{x} 2$（also works for generalized linear models and mixed－ effects models） | 1：Use STANDARDIZE on $\mathrm{x} 1 \& \mathrm{x} 2$ then Analysis toolbox：迴歸 on these $z$ scores <br> 2：Analysis toolbox：迴歸， then for x 1 coefficient B1：B1＊STDEV（x1）／ STDEV（y）\＃Same for x2 | $\begin{aligned} & \text { 1: } \operatorname{summary}(\operatorname{lm}(\mathrm{y} \sim \operatorname{scale}(\mathrm{x} 1)+ \\ & \text { scale }(\mathrm{x} 2) \\ & \text { 2: For } \mathrm{x} 1 \text { coefficient B1: } \\ & \text { B1*sd }(\mathrm{x} 1) / \text { sd }(\mathrm{y}) \text { \# Same for } \mathrm{x} 2 \end{aligned}$ |
| Repeated－measures regression $\mathrm{y} \sim \mathrm{x}$ with grouping variable g in data frame D （also applies to logistic \＆Poisson regression）， where $\mathrm{B}=$ by－unit coefficients（e．g．， $\mathrm{B}=$ B 0 for intercept，or $\mathrm{B}=\mathrm{B} 1$ for x slope） | Analysis toolbox：迴歸 AVERAGE（B）\＃Coef． STDEV（B）\＃SE \＃t，df，p from one－sample \＃t test（see above） | ```B.coef = numeric(length(g)) for (i in 1:length(g)) \{ D. \(\mathrm{i}=\operatorname{subset}(\mathrm{D}, \mathrm{D}[\mathrm{D} \$ \mathrm{~g}==\mathrm{i}])\) \(\operatorname{lm} . \mathrm{i}=\operatorname{lm}(\mathrm{y} \sim \mathrm{x}\), data \(=\mathrm{D} . \mathrm{i})\) B.coef[i] = summary(lm.i) \$coefficients["B","Estimate"] \} t.test(B.coef) \# gives all but SE \# I'll add SE info after HW3...``` |
| ANOVA |  |  |
| One－way independent－measures ANOVA （ $\mathrm{y}=$ dependent； $\mathrm{x}=$ independent） | Analysis toolbox：單因子變異數分析 | $\begin{aligned} & \hline \text { 1: summary }(\operatorname{aov}(y \sim x)) \\ & \text { 2: } \operatorname{anova}(\operatorname{lm}(y \sim x)) \end{aligned}$ |


| Two－way independent－measures ANOVA （ $\mathrm{y}=$ dependent； $\mathrm{x} 1, \mathrm{x} 2=$ independents） | Analysis toolbox： <br> 雙因子變異數分析：重複試驗 | $\begin{aligned} & \hline \text { 1: summary }(\operatorname{aov}(y \sim x 1 * x 2)) \\ & \text { 2: } \operatorname{anova}(\operatorname{lm}(y \sim x 1 * x 2)) \end{aligned}$ |
| :---: | :---: | :---: |
| One－way repeated－measures ANOVA（ $\mathrm{y}=$ dependent； $\mathrm{x}=$ indep； $\mathrm{S}=$ grouping unit） | Analysis toolbox： <br> 雙因子變異數分析：無重複試驗 | $\begin{aligned} & \text { S = as.factor(S) \# Make sure! } \\ & \text { summary }(\operatorname{aov}(\mathrm{y} \sim \mathrm{x}+\operatorname{Error}(\mathrm{S} / \mathrm{x}))) \end{aligned}$ |
| Two－way repeated－measures ANOVA（ $\mathrm{y}=$ dep；x1，x2＝indeps；grouped by S） | use repeated－measures regression by hand | $\begin{aligned} & \text { summary(aov(y } \sim \mathrm{x} 1 * \mathrm{x} 2 \\ & +\operatorname{Error}(\mathrm{S} /(\mathrm{x} 1 * \mathrm{x} 2)))) \\ & \hline \end{aligned}$ |
| One type of mixed ANOVA（ $\mathrm{y}=\mathrm{dep}$ ； $\mathrm{x} 1=$ between－group indep；$x 2$＝within－group indep，grouped by S） | probably NA | ```1: summary(aov(y ~ x1*x2 +Error(S/x2))) 2: library(ez) # Likewise above ezANOVA(dv = y, wid = S, within = x2, between = x1)``` |
| Tukey HSD test［formula＝any ANOVA formula，e．g． $\mathrm{y} \sim \mathrm{x}$ ，or $\mathrm{y} \sim \mathrm{x}+\operatorname{Error}(\mathrm{S} / \mathrm{x})$ ］ | use equation in handout and find table of Studentized range statistic $q$ on the Web | ```1: TukeyHSD(aov(formula)) 2: library(emmeans) emmeans(aov(formula), list(pairwise \(\sim \mathrm{x}\) ), adjust="tukey")``` |
| Correct for sphericity violations in repeated－measures ANOVA in factors with more than two levels（ $\mathrm{y}=\mathrm{dep}$ ； $\mathrm{x}=$ within－ group indep with 3 or more levels； $\mathrm{S}=$ grouping unit； $\mathrm{df}=\mathrm{df}_{\text {denominator }}$ ） | NA | ```library(ez) ezANOVA(dv = y, wid \(=S\), within = x ) \# HFe = Huynh-Feldt epsilon \(\# \mathrm{p}[\mathrm{HF}]=\) its p value \# correct \(\mathrm{df}=\) original df * HFe``` |
| Compute minF＇for independent variable x ，using the following ANOVA results： <br> By－participant ANOVA： <br> x．F1： F value for x <br> x．dfn1：df for x levels（numerator） <br> x．dfd1：df for random（denominator） <br> By－item ANOVA： <br> x．F2： F value for x <br> x．dfn2：df for x levels（numerator） <br> x．dfd2：df for random（denominator） <br> Then you get the following： <br> $\operatorname{minF} . F:$ minF $^{\prime}$ <br> minF．dfn：df for $x$ levels <br> minF．dfd：df for random <br> minF．p：$p$ value | ```\(\operatorname{minF} . \mathrm{F}=\) (x.F1*x.F2/(x.F1+x.F2) minF.dfn = x.dfn1 \# (= xdfn2) \(\operatorname{minF} . d f d=\) (x.F1+x.F2) \()^{\wedge}\) / (x.F1^2/x.dfd2 + x.F2^2/x.dfd1) \(\operatorname{minF} . p=\) FDIST(minF.F, minF.dfn, minF.dfd)``` | ```\(\operatorname{minF} . \mathrm{F}=\) (x.F1*x.F2/(x.F1+x.F2) minF.dfn = x.dfn1 \# (= xdfn2) \(\operatorname{minF} . d f d=\) (x.F1+x.F2) \({ }^{\wedge} 2\) / (x.F1^2/x.dfd2 + x.F2^2/x.dfd1) \(\operatorname{minF} . p=p f(\operatorname{minF} . F, \operatorname{minF} . d f n\), minF.dfd, lower.tail=F)``` |
| Contingency tables（and other simple categorical tests） |  |  |
| one－tailed $p$ value for binomial test on getting at most x in n binary events | BINOMDIST（x，n，0．5，TRUE） | ```1: pbinom(x, n, 0.5) 2: binom.test(x,n,alternative="left")``` |
| One－way chi－squared test on vector V， where $\mathrm{H}_{0}$ ：all counts the same | CHITEST（observed，expected） <br> \＃Must compute expected first | chisq．test（V） |
| One－way chi－squared test on vector V， where $\mathrm{H}_{0}$ ：counts $=$ vector W | CHITEST（observed，expected） \＃Also，only gives p value | $\operatorname{chisq} . \operatorname{test}(\mathrm{V}, \mathrm{p}=\mathrm{W})$ |
| Two－way chi－squared test for column $\times$ row interaction in $2 \times 2$ matrix M | CHITEST（observed，expected） <br> \＃Doesn＇t use Yate＇s correction | 1：chisq．test（M）\＃With Yate＇s <br> 2：summary（as．table（M）） <br> \＃Without Yate＇s correction |
| Two－way chi－squared test for column $\times$ row interaction in larger matrix M | CHITEST（observed，expected） <br> \＃Basically，forget this method | $\begin{aligned} & \text { 1: chisq.test(M) } \\ & \text { 2: summary(as.table(M)) } \\ & \text { \# Same: Yate's irrelevant } \end{aligned}$ |
| Two－tailed $p$ value testing for column $\times$ row interaction in contingency table M | NA | fisher．test（M） |
| Exact McNemar test for paired binary data，with a $(1,0)$ pairs and $b(0,1)$ pairs | $\begin{aligned} & \text { BINOMDIST(MIN }(\mathrm{a}, \mathrm{~b}), \mathrm{a}+\mathrm{b}, \\ & 0.5, \text { TRUE) } \end{aligned}$ | $\operatorname{pbinom}(\min (\mathrm{a}, \mathrm{b}), \mathrm{a}+\mathrm{b}, 0.5)$ |


| Logistic regression (and other generalized linear models) |  |  |  |
| :---: | :---: | :---: | :---: |
| Convert probability P into log odds (logit) | LN(P/(1-P)) |  | $\begin{aligned} & \text { 1: } \ln (\mathrm{P} /(1-\mathrm{P})) \\ & \text { 2: } \operatorname{library}(\text { gtools }) \\ & \operatorname{logit}(\mathrm{P}) \end{aligned}$ |
| Convert log odds L into probability | EXP(L)/(1+EXP(L)) |  | $\begin{aligned} & \text { 1: } \exp (\mathrm{L}) /(1+\exp (\mathrm{L})) \\ & \text { 2: library }(\mathrm{gtools}) \\ & \quad \text { inv.logit( } \mathrm{L}) \end{aligned}$ |
| Logistic regression model $\mathrm{y} \sim \mathrm{x} 1+\mathrm{x} 2$ ( y is binary variable, all data are independent), with data in data frame D | NA |  | $\begin{aligned} & \operatorname{glm}(y \sim x 1+x 2, \\ & \text { family=binomial, data }=D) \\ & \text { \# data argument also below } \\ & \hline \end{aligned}$ |
| Show coefficients table for logistic regression model L | NA |  | ```summary(L) \# p-values based on Wald test``` |
| Predict log odds from logistic regression model L | NA |  | predict(L) |
| Predict binary observations ( 0 vs. 1) from logistic regression model L | NA |  | predict(L, type="response") |
| Likelihood ratio test for simpler generalized linear regression model L0 vs. more complex L1 (applies to both logistic regression and Poisson regression) | NA |  | anova(L0, L1 test="Chisq") |
| Test parameter x 1 of logistic regression model $\mathrm{y} \sim \mathrm{x} 1+\mathrm{x} 2$ using likelihood ratio test | NA |  | $\begin{aligned} & \hline \mathrm{L} 1=\mathrm{g} \operatorname{lm}(\mathrm{y} \sim \mathrm{x} 1+\mathrm{x} 2, \\ & \text { family }=\text { binomial }) \\ & \mathrm{L} 0=\operatorname{glm}(\mathrm{y} \sim \mathrm{x} 2, \\ & \text { family=binomial }) \\ & \text { anova(L0, L1 test="Chisq" }) \\ & \hline \end{aligned}$ |
| Ordinal logistic regression $y \sim x$ ( $y$ is ordinal variable, all data are independent) | NA |  | ```library(MASS) summary \((\operatorname{polr}(\mathrm{y} \sim \mathrm{x})\) \# Table compares each level \# with next level``` |
| Multinomial logistic regression $\mathrm{y} \sim \mathrm{x}$ ( y has three nominal values " A ", " B ", " C ", all data independent) | \# Wald test only: $\mathrm{z}=\mathrm{B} / \mathrm{SE}$ <br> $\mathrm{p}=2 *$ NORMSDIST( <br> -ABS(z)) <br> \# Likewise below |  | library(nnet) <br> summary(multinom( $\mathrm{y} \sim \mathrm{x})$ ) <br> \# Table treats A as baseline <br> \# Wald test for each row: <br> $z=B / S E \quad B=$ coefficient <br> $\mathrm{p}=2$ *pnorm $(-\mathrm{abs}(\mathrm{z}))$ |
| Poisson regression $\mathrm{y} \sim \mathrm{x}$ ( y is count data) | NA |  | summary(glm(y $\sim \mathrm{x}$, family=poisson)) |
| Mixed-effects modeling (linear and generalized linear) |  |  |  |
| Maximal one-random-factor LME: $\mathrm{y}=$ dependent (continuous, normal) $\mathrm{x} 1, \mathrm{x} 2=$ independent $\mathrm{g}=$ grouping unit ( x 1 grouped by g ) | NA | ```1: library(nlme) lme \((\mathrm{y} \sim \mathrm{x} 1+\mathrm{x} 2\), random \(=\sim \mathrm{x} 1 \mid \mathrm{g})\) 2: library(lme4) \# Assumed elsewhere below \(\operatorname{lmer}(\mathrm{y} \sim \mathrm{x} 1+\mathrm{x} 2+(\mathrm{x} 1 \mid \mathrm{g}))\)``` |  |
| Show results of LME model L | NA | ```summary(L) \# lme shows p , Imer doesn't \# Always build/name model first, before using summary \# because the model may take a long time to build``` |  |
| Get p values for LME model derived from formula structure $\mathrm{y} \sim \mathrm{x}+(\mathrm{x} \mid \mathrm{g})$ | NA | 1: Trust lme o <br> 2: 2 *pnorm(-a <br> 3: library(afex) $\mathrm{L}=\operatorname{mixed}(\mathrm{y}$ summary(L) <br> 4: library(afex) \# Likelihoo $\mathrm{L}=\operatorname{mixed}($ summary(L) <br> \# method="PB <br> \# Forget abou <br> \# changes sum | ut (controversial) <br> (t)) \# Claims $\mathrm{t}=\mathrm{z}$ (needs large N ) <br> Loads lme4 for you <br> $\mathrm{x}+(\mathrm{x} \mid \mathrm{g}))$ \# Kenward-Roger p <br> Loads lme4 for you <br> atio tests (needs large N ) <br> $\mathrm{x}+(\mathrm{x} \mid \mathrm{g})$, method="LRT") <br> not working for afex's mixed function; nerTest (worse than Kenward-Roger, ary.lmer behavior) |


| Maximal two-random-factor additive LME (recommended by Barr et al., 2013): <br> $\mathrm{y}=$ dependent (continuous, normal) <br> $\mathrm{x} 1, \mathrm{x} 2$ = independent <br> $\mathrm{g} 1=$ grouping unit for x 1 (random effect) <br> $\mathrm{g} 2=$ grouping unit for x 2 (random effect) | NA | 1: $\operatorname{lmer}(\mathrm{y} \sim \mathrm{x} 1+\mathrm{x} 2+(\mathrm{x} 1 \mid \mathrm{g} 1)+(\mathrm{x} 2 \mid \mathrm{g} 2))$ <br> 2: $\operatorname{lmer}(\mathrm{y} \sim \mathrm{x} 1+\mathrm{x} 2+(1+\mathrm{x} 1 \mid \mathrm{g} 1)+(1+\mathrm{x} 2 \mid \mathrm{g} 2))$ <br> \# R assumes the intercepts automatically |
| :---: | :---: | :---: |
| Maximal one-random-factor LME with interaction: <br> $y=$ dependent (continuous, normal) <br> $\mathrm{x} 1, \mathrm{x} 2=$ independent <br> $\mathrm{g}=$ grouping unit for $\mathrm{x} 1 \& \mathrm{x} 2$ | NA | $\operatorname{lmer}(\mathrm{y} \sim \mathrm{x} 1 * x 2+((\mathrm{x} 1 * \mathrm{x} 2) \mid \mathrm{g}))$ |
| Likelihood ratio test to compare fit of simper LME model L0 vs. complex L1 | NA | anova(L0,L1) |
| Likelihood ratio test for above to see if random g 2 variable is really necessary (not recommended by Barr et al., 2013, but cf. Raaijmakers et al., 1999) | NA | L.1.2 $=\operatorname{lmer}(\mathrm{y} \sim \mathrm{x} 1+\mathrm{x} 2+(\mathrm{x} 1 \mid \mathrm{g} 1)+(\mathrm{x} 2 \mid \mathrm{g} 2))$ $\mathrm{L} .1=\operatorname{lmer}(\mathrm{y} \sim \mathrm{x} 1+\mathrm{x} 2+(\mathrm{x} 1 \mid \mathrm{g} 1))$ anova(L.1, L.1.2) |
| LME without random intercepts (if maximal model fails to converge) <br> \# This and below also work for GLMM | NA | $\operatorname{lmer}(\mathrm{y} \sim \mathrm{x}+(\mathrm{x} \mid \mathrm{g}))$ \# Maximal model <br> $\operatorname{lmer}(\mathrm{y} \sim \mathrm{x}+(0+\mathrm{x} \mid \mathrm{g}))$ \# Next-best model |
| LME without random intercept $\times$ slope interaction (if above also fails) | NA | $\operatorname{lmer}(\mathrm{y} \sim \mathrm{x}+(0+\mathrm{x} \mid \mathrm{g})+(1 \mid \mathrm{g}))$ \# Slope \& intercept separate |
| LME without random slopes (if all fails) | NA | $\operatorname{lmer}(\mathrm{y} \sim \mathrm{x}+(1 \mid \mathrm{g})$ ) \# Worst LME model (Barr et al., 2013) |
| Maximal one-random-factor mixed-effects logistic regression (a kind of GLMM): <br> $\mathrm{y}=$ dependent (binary) <br> $\mathrm{x} 1, \mathrm{x} 2$ = independent <br> g = grouping unit ( x 1 grouped by g ) | NA | ```1: library(MASS) glmmPQL(y~x1+x2, random=~x1 \|g, family=binomial) 2: library(lme4) # Assumed elsewhere below glmer(y~x1+x2+(x1|g), family=binomial)``` |
| Maximal two-random-factor mixed-effects logistic regression: <br> $\mathrm{y}=$ dependent (binary) <br> $\mathrm{x} 1, \mathrm{x} 2$ = independent <br> $\mathrm{g} 1=$ grouping unit for x 1 (random effect) <br> $\mathrm{g} 2=$ grouping unit for x 2 (random effect) | NA | $\operatorname{glmer}(\mathrm{y} \sim \mathrm{x} 1+\mathrm{x} 2+(\mathrm{x} 1 \mid \mathrm{g} 1)+(\mathrm{x} 2 \mid \mathrm{g} 2)$, family=binomial $)$ |
| Likelihood ratio test to compare fit of simper GLMM model L0 vs. more complex L1 | NA | anova(L0, L1, test="Chisq") |

