

# R Shortcuts for Statistics

## Useful commands

<code>which()</code>	filter data frame or factor by creating a filter list
<code>table(df)</code>	number of entries for each value in dataframe
<code>prop.table</code>	table in terms of probabilities
<code>str(df)</code>	Type of variables present at df
<code>mean(factor/df)</code>	means of dataframe columns or factor
<code>sd(factor/df)</code>	standard deviation of dataframe columns or factor
<code>paste0</code>	concatenate strings
<code>Levels</code>	return levels of attributes

## Tests

Name	Use Case	Comments
<code>acf()</code>		
<code>dwtest()</code>		
<code>shapiro.test()</code>	Test if data is normally distributed	
<code>t.test()</code>		
<code>chisq()</code>		
<code>varTest()</code>		
<code>cor(df[,c(4,1:3)],method="pearson")</code>		Numeric Insights Parametric version for normal-like
<code>cor(df[,c(4,1:3)],method="spearman")</code>		Numeric Insights Non Parametric
<code>cor.test(df\$prestige,df\$income,method="pearson")</code>	Test Hypothesis to test whether my population rho equals 0 or not	Inferential
<code>cor.test(df\$prestige,df\$income,method="spearman",data=df)</code>	Test Hypothesis to test whether my population rho equals 0 or not	

Question	Solution
Is serial correlation present?	<code>acf(df\$cal)</code> <code>library(lmtest)</code> <code>dwtest(df\$cal~1)</code>
Determine univariant severe outliers.	See Excel
Multivariate Outliers:	See Excel, then:

	<pre>abline( h=res.mout\$cutoff, lwd=2, col="red") abline( v=res.mout\$cutoff, lwd=2, col="red")  llmout &lt;- which( ( res.mout\$md &gt; res.mout\$cutoff ) &amp; (res.mout\$rd &gt; res.mout\$cutoff ));llmout df[llmout,] res.mout\$md[llmout] df\$mout &lt;- 0 df\$mout[ llmout ] &lt;- 1 df\$mout &lt;- factor( df\$mout, labels = c("MvOut.No","MvOut.Yes"))</pre>		
Using EDA which are the most associated variables with the numeric response variable? Use also FactoMineR profiling tools at 99% significance level	<p>See Excel for FactoMineR</p> <p>EDA:</p> <pre>plot(df[,c(TARGET_VAR,EXPLAIN_VARS)]) cor(df[,c(TARGET_VAR, EXPLAIN_VARS)], method="spearman") corrplot(cor(df[,c(9,3:8)], method="spearman"), is.cor=T)</pre>		
determine the most relevant global associations at 99% CI for categorial Target var.	See Excel		
Say Something about distribution that was assumed. Graphical and inferential	graphical hist(df\$cal,30) hist(log(df\$cal),30)	Inferential shapiro.test( log(df\$cal) )	
Num_var variate dispersion behavior according to the categ_var. numeric, graphics and inferential	Graphical Boxplot( num_var~cat_var, data = df )	Inferential See Excel for correct test (29-33)	Numerical tapply( df\$cal, df\$brand, sd )
Num_var variate mean behavior according to the cat_var. numeric, graphics and inferential	Boxplot( cal~brand, data = df )	See Excel for correct test (23-26)	tapply( df\$cal, df\$brand, mean )
which brands show a remarkable difference in mean behavior among them. Use one-sided tests	<pre>pairwise.wilcox.test( df\$cal, df\$brand, alternative="less" ) pairwise.wilcox.test( df\$cal, df\$brand, alternative="greater" )</pre>		
test at the 1% level the null hypothesis that the population standard deviation is not larger than 0.15cal against the alternative that it is	<pre>tapply(df\$cal,df\$brand,sd) table(df\$brand) ss &lt;- 0.16362880 # H0: sigma^2= 0.15^2 H1: sigma &gt; 0.15^2 Normal population (n-1)ss^2/sigma^2 ~ X2(n-1) # (n-1)ss^2/sigma^2  chi&lt;-(29-1)*(ss^2)/(0.15^2);chi 1-pchisq(chi,28) # pvalue &gt; 0.01 H0 can not be rejected  Or</pre>		

	<pre>varTest(df[which(df\$brand == "A"),]\$cal, alternative="greater", conf.level =0.99, sigma.squared = 0.15^2)</pre>
<p>99% upper threshold for the number of calories for brand A population variance. Normal distribution for calories is assumed to hold.</p>	<pre>varTest(x, sigma.squared=0.15^2, alternative="less",conf.level=0.99) sqrt(variance) to obtain standard deviation</pre>
<p>Build a 99% confidence interval for the difference in the mean of 100 g calories between brands A and C. Assume that equal variances in the population calories per brand does not hold</p>	<pre>ll &lt;- which( df\$brand %in% c("A","C")) dff &lt;- df[ll,]  t.test(dff\$cal~dff\$brand, conf.level=0.99) – used to create confidence interval fligner.test(dff\$cal,dff\$brand, conf.level=0.99) – used to check if variances are the same t.test(dff\$cal~dff\$brand, conf.level=0.99, var.equal = T)</pre>
<p>Out of 100 people, 60 prefer A to C. Determine a 99% confidence interval for the population proportion that favors A in front of C. Test the null hypothesis that selecting A and C has equal probability.</p>	<pre>prop.test(60, n=100, p=0.5, conf.level=0.99, correct=F)</pre>
<p>Determine a 99% confidence interval for the difference in the population proportion that favors A in front of C accounting the two surveys. Test the null hypothesis that selecting A brand has a lower probability in the second of the surveys</p>	<pre>prop.test(c(60,110), n=c(100,200), conf.level=0.99, correct=F, alternative="greater")</pre>