Introduction to R Lecture 2 September 21st, 2020

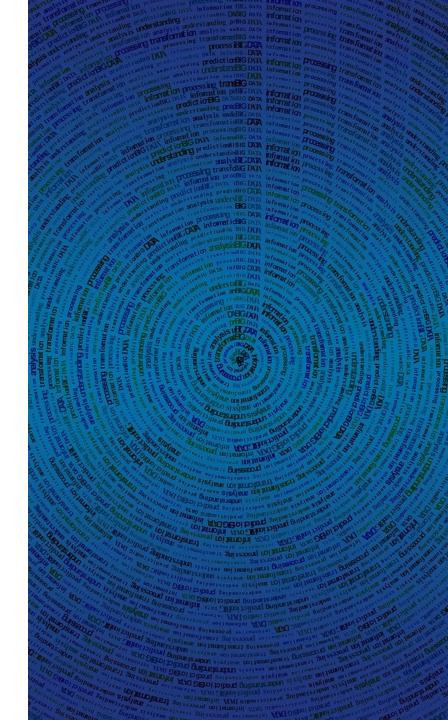
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https://github.com/kristineccles/Introduction_to_R





Overview

- Crash course in statistics
 - Probability distribution and the p-value
 - Descriptive statistics
 - Parametric Tests
 - -Difference of means: T-Test and ANOVA
 - -Relational Statistics: correlations, linear regression
 - -PCA



Steps of Statistical Analysis

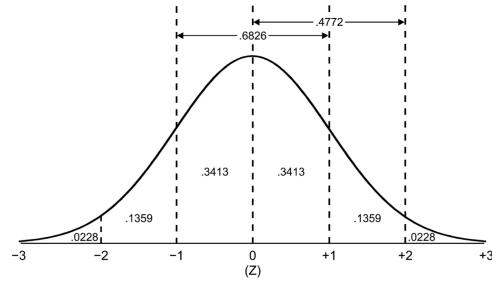
Step 1	State the null and alternative hypothesis
Step 2	Select the appropriate statistical test
Step 3	Select level of significance
Step 4	Delineate regions of rejection and non-rejection of the null (α)
Step 5	Calculate test statistic
Step 6	Make decision regarding null and alternative hypothesis



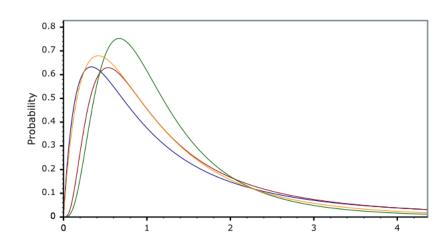
Statistics

- Descriptive statistics uses the data to provide descriptions of the population, either through numerical calculations, graphs, or tables.
- Inferential statistics makes inferences and predictions about a population based on a sample of data taken from the population in question.
 - Based on probability distributions
 - Probability theory deals with random events
 - Random/probabilistic independent sample(s)

T/z-distribution



F-distribution





Probability Distributions

- The area under the curve gives the probability of finding a given value in that range
- Total area under the curve is = 1 (probability is 1)
- The normal distribution
 - What the probability is of getting an extreme value (the tail)
 - 68% of the values are within 1 s.d. of the mean
 - Or, there is a probability of 0.68 that a given value is within 1 s.d. of the mean
 - 95% of the area under the curve is within 1.96 s.d. of the mean
 - 99.7% of the area is within 3 standard deviations of the mean
- Empirical probability relies on the law of large numbers: the relative frequency will eventually converge on the true probability



P-value

- Evaluates the validity of our prediction using a statistical test
 - This will say if the result is statistically significant or not
- Two hypotheses:
 - H_0 the null hypothesis
 - Status quo, no change, nothing special going on, conservative
 - H_A (aka H_1) the alternate hypothesis
 - Your hypothesis there is a difference/change.
- They are mutually exclusive and exhaustive

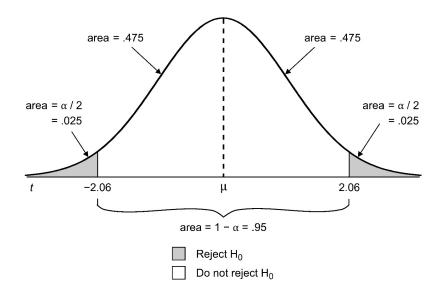
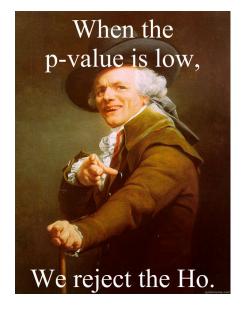


FIGURE 9.2

Normal Distribution Values Associated with a Significance Level (α) = .05: Two-Tailed Case





Descriptive Statistics

- Characteristics of a normal (Gaussian) distribution:
 - Mean, median, mode are all the same & in the middle of the distribution
- Symmetrical, smooth, continuous distribution
- Is my data normal?
 - Make a histogram
 - Look at mean, median, and modal values
 - Check the shape of the distribution:
 - Skewness and Kurtosis

summary(df)

- no st.dev in summary
- -sd(df)

library(psych)

describe(df)



Unpaired T-Test

- Objective: Test if there is a difference between the mean of two groups
 - Boxplots show the median
- Data type: one continuous and one categorical variable
- Assumptions: variable normality, equal variance between the groups
 - For unequal variance use Welch's T-Test



ANOVA

- Test: If there is a difference between the mean of 3 or more groups
 - This only tells you if there is a different NOT where the difference is
- Data: Three or more groups and a continuous variable
- Assumptions (F-test):variable normality, equal variance between the groups (ish)



ANOVA Post-hoc test (Tukey)

- Test: Multiple comparison to determine the differences in means after an ANOVA
- Data: Three or more groups and a continuous variable
- Assumptions (t-test):variable normality, equal variance between the groups
- Note the p adj: This p-value is adjusted for multiple comparison
 - By default R uses Bonferroni (α /n), where n= number of comparisons



Correlation (Pearson Product-Moment)

- Test: Association between variables
 - correlation coefficient ρ (Greek letter rho)
- Data: Two continuous variables
 - Pearson product-moment correlation
 - Spearman rank and Kendall Tau are the non-parametric version
 - Contingency analysis (categorical data)
- Assumptions (t-test): variable normality, linear association between variables



Linear Regression (Univariate)

- Test: the slope coefficient is significant
- Data: two continuous variables
 - Generalized linear model is used for non-parametric (e.g. logit)
- Assumptions (t-test): , Test statistic: t, Assumptions: Variable normality , Random/probabilistic sample of paired variables, Variables have a linear association
- Test: Goodness of fit (R²)
 - test that the model predicts a significant amount of the variance in y
 - Coefficient of determination (not the same as ρ)
- Assumptions (F-test): Variable normality ,Random/probabilistic sample of paired variables, Variables have a linear association



Assumptions of the residuals

- e = y Y-hat
- In linear regression we have assumptions on the residuals too
 - Normality (Shapiro-Wilk)
 - Linearity (RESET test)
 - Lack of serial autocorrelation (Durbin-Watson test)
 - Homogeneity (Breusch-Pagan test)
- Can use individual test or plot(lm1)



Readings a Regression Output

```
Call:
Im(formula = y \sim x, data = dataset)
Residuals:
      1Q Median 3Q Max
-212.04 -62.52 30.87 86.77 121.88
```

Coefficients:

```
Estimate Std. Error t value Pr(>|t|)
(Intercept) 28.149 180.756 0.156 0.88010
           39.057 9.172 4.258 0.00277 **
Χ
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 121.8 on 8 deg. of freedom
Multiple R-squared: 0.69, Adjusted R-squared: 0.65
F-statistic: 18.13 on 1 and 8 DF, p-value: 0.002767
```

- What you asked for:
 - the formula here says predict 'y' from 'x' with data from 'dataset'
- Residuals:
 - 5 number summary of the residuals



Reading a Regression Output

```
Call:
Im(formula = y ~ x, data = dataset)
Residuals:
Min 1Q Median 3Q Max
-212.04 -62.52 30.87 86.77 121.88

Coefficients:
Estimate Std. Error t value Pr(>|t|)
(Intercept) 28.149 180.756 0.156 0.88010
x 39.057 9.172 4.258 0.00277 **
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Model coefficients (aka parameters)

- cols are the coefficient estimates, the standard error of the coefficient, the t value for the coefficient & the p-value for this t value
- y-intercept (intercept)
- slope (Estimate)
- legend explaining level of significance codes (α =0.05)



Reading a Regression Output

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- Overall model performance
- Standard error of the residuals (SSE)
- R² and the adjusted R²
- Use the adj. R2 since it is more conservative
- F statistic and the p-value (prob. of getting a bigger F)



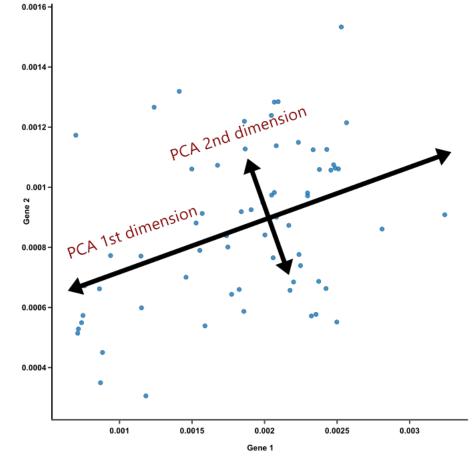
Multivariate

- $Im(y \sim v1 + v2 + v3...)$
- Assumptions and test the same as univariate linear regression
- One additional
 - Multicollinearity
 - Cannot have a high correlation between independent variables
 - > +/- 0.70 it typically the threshold
 - Correlation matrices are helpful
 - Can test this using a variance inflation factor (VIF)
 - >5



Principal Component Analysis (PCA)

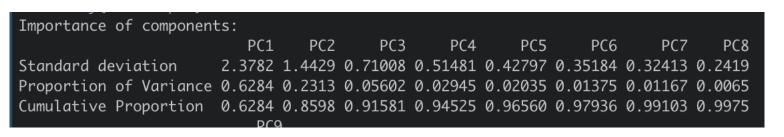
- Test:
 - Clustering
 - A way to summarize complex real-valued data with a single categorical variable
 - Dimensionality reduction
 - A way to simplify complex high-dimensional data
- Data: ratio of 5 cases: 1 variable, variables must be of similar magnitude, and must not have outliers
 - Scale and center
- In R prcomp and princomp are used

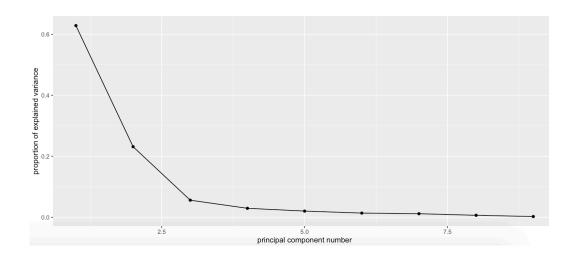




PCA Outputs

- Variance explained
 - Summary(pca1)
 - # of variables = # of PCs
- Scree plot
 - Displays how much variation each principal component captures from the data
 - Used to determine the number of factors to retain components in an analysis

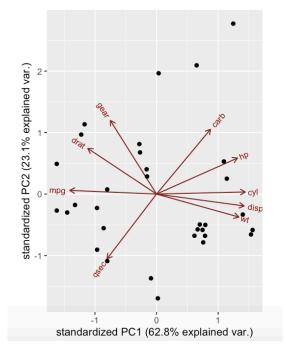


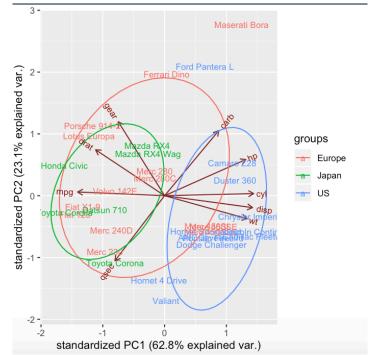




PCA Outputs

- Biplot
 - PCA biplot = PCA score plot (points) + loading plot (arrows)
 - Biplots are a type of exploratory graph used in statistics, a generalization of the simple two-variable scatterplot.
 - A loading plot shows how strongly each characteristic influences a principal component.
 - Shows how samples cluster based on their similarity
- Eigenvalues (scores)
- Eigenvectors (loadings)







Loadings (eigenvectors)

```
PC2
                                                         PC4
                                                                       PC5
                          0.03786529 -0.5932799 -0.089331673
                                                              0.040512600
length
diameter
                          0.06532324 -0.5853661 -0.008285814
                                                              0.008517628 -0.711025627
height
                          0.86683603
                                      0.3148764 -0.165564868 -0.027110424
whole_weight
               0.3906735 -0.23327117
shucked_weight 0.3781883 -0.34801069
                                      0.2315678 -0.496179039
gut_weight
                         -0.25290295
                                      0.2702527 -0.140972073
shell_weight
               0.3789217 -0.05837478
                                      0.1621047
                                                 0.834110000 -0.181668556
```

- Scores (eigenvalues)
- Scores can be mapped (to look for spatial clustering) and can be used in linear regression
 - Can be difficult to interpret



What is reproducible analysis?

- One particular form:
 - code transforms raw data and meta-data into processed data,
 - code runs analyses on the data, and
 - code incorporates analyses into a report
- Ideally, the process involves a one-click build
 - Knitr: an engine for dynamic report generation with R.
- Public sharing of document, code, and data is optional, but forms part of gold standard of scientific openness
 - Journals are now requiring this



intro_to_r_lecture2_excercise.R

kristineccles

2020-01-12

```
# Introduction to R
# Lecture 2- Statistics
# By: Kristin Eccles
# Written in R 3.6.2
# Install Libraries
# only need to run this once
#install.packages(c("psych", "car", "stats", "corrplot", "factoextra", "lmtest", "devtools"))
# Load Libraries
library(ggplot2)
library(psych) # describe and mutli.hist
## Attaching package: 'psych'
## The following objects are masked from 'package:ggplot2':
     %+%, alpha
library(car) #stats
## Loading required package: carData
## Attaching package: 'car'
## The following object is masked from 'package:psych':
##
```



Reproducible analysis in R

 Combine R and plain text file format to produce documents (e.g., pdfs, HTML documents, etc.)

Rmarkdown

- General workflow
 - Create an RMarkdown File
 - either use Rstudio File New File RMarkdown or just create a file with an empty text file with the .rmd extension
 - Optionally add a header
 - Options to specify output format (see RMarkdown cheat sheet)
 - Write the main document in Markdown
 - Embed R code chunks
 - R code chunks can be customised to control output





See RMarkdown Cheat Sheet for more info



Reproducible Science is good. Replicated Science is better.

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