

ANEXOS

Introducción a la Ciencia de Datos

Chi-square independence test

- For 2-way tables you can use `chisq.test()` to test independence of the row and column variable.
- By default, the **p**-value is calculated from the asymptotic chi-squared distribution of the test statistic.
- Optionally, the **p**-value can be derived via Monte Carlo simulation.

```
> (HairEye <- margin.table(HairEyeColor, c(1, 2)))
```

```
Eye
```

```
Hair Brown Hazel Green Blue
```

```
Black 68 15 5 20
```

```
Brown 119 54 29 84
```

```
Red 26 14 14 17
```

```
Blond 7 10 16 94
```

```
> chisq.test(HairEye)
```

```
Pearson's Chi-squared test
```

```
data: HairEye
```

```
X-squared = 138.29, df = 9, p-value < 2.2e-16
```

Fisher Exact Test of independence

- X must be a two-way contingency table in table form.
- Another form, `fisher.test(X, Y)` takes two categorical vectors of the same length. For tables larger than 22 the method can be computationally intensive (or can fail) if the frequencies are not small.

```
> fisher.test(GSStab)
Fisher's Exact Test for Count Data
data: GSStab
p-value = 0.03115
alternative hypothesis: two.sided
```

Testing for Normality: Shapiro-Wilks test

- To determine whether your data sample is normally distributed use the `shapiro.test()` function:

```
> shapiro.test(x)
Shapiro-Wilk normality test
data: x
W = 0.9651, p-value = 0.4151
9.13
```

- The large **p** -value suggests the underlying population could be normally distributed.
- a small **p** -value suggest that it is unlikely that this sample came from a normal population:

Testing for Normality: Shapiro-Wilks test

- When you choose a test, you may be more interested in the normality in each sample. You can test both samples in one line using the `tapply()` function, like this:

```
> with(beaver, tapply(temp, activ, shapiro.test))
```

- The large **p** -value suggests the underlying population could be normally distributed.
- a small **p** -value suggest that it is unlikely that this sample came from a normal population:

Performing Robust ANOVA (Kruskal–Wallis Test)

- Your data is divided into groups.
- Are there significant differences between these groups?

```
my_data <- PlantGrowth
```

- Is any significant difference between the average weights of plants in the 3 experimental conditions?

```
kruskal.test(weight ~ group, data = my_data)
```

Kruskal-Wallis rank sum test

data: weight by group

Kruskal-Wallis chi-squared = 7.9882, df = 2, p-value = 0.01842

P-value is less than the significance level 0.05, we can conclude that there are significant differences between the treatment groups.

Multiple pairwise-comparison between groups

- There is a significant difference between groups, but which pairs of groups are different?.
- The function `pairwise.wilcox.test()` calculate pairwise comparisons between group levels with corrections for multiple testing.

```
pairwise.wilcox.test(PlantGrowth$weight, PlantGrowth$group,  
p.adjust.method = "BH")
```

```
Pairwise comparisons using Wilcoxon rank sum test data: PlantGrowth  
$weight and PlantGrowth$group
```

```
ctrl trt1
```

```
trt1 0.199 -
```

```
trt2 0.095 0.027
```

```
P value adjustment method: BH
```

only trt1 and trt2 are significantly different ($p < 0.05$)..

Dealing with non normality

Data transformation

- if you are looking at relationships between variables (e.g., regression) it is alright just to transform the problematic variable
- if you are looking at differences within variables (e.g., change in a variable over time) then you need to transform all levels of those variables.

Data transformation	Can correct for
<i>log transformation</i>	Positive skew, unequal variances
<i>square root transformation</i>	Positive skew, unequal variances
<i>reciprocal transformation $1/(\text{variable}+1)$</i>	Positive skew, unequal variances
<i>reverse score transformation</i>	negative skew

Comparison of Statistical Analysis Tools for Normally and Non-Normally Distributed Data

Tools for Normally Distributed Data	R functions	Equivalent Tools for Non-Normally Distributed Data	R functions
T-test	<code>t.test()</code>	Mann-Whitney test; Mood's median test; Kruskal-Wallis test	<code>wilcox.test()</code> ; <code>mod.medtest()</code> ; <code>kruskal.test()</code>
ANOVA	<code>aov()</code>	Mood's median test; Kruskal-Wallis test	<code>mod.medtest()</code> ; <code>kruskal.test()</code>
Paired t-test	<code>t.test(x, y, paired = TRUE)</code>	One-sample sign test	<code>SIGN.test()</code>
F-test; Bartlett's test	<code>var.test()</code> ; <code>bartlett .test()</code>	Levene's test	<code>Levene.test()</code>

Gracias...

