

NOMINAL TESTS



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2 Analyses

- Binomial Test
- McNemar
- Cochran's Q
- Chi-Squared

3 Our Data

- Choice Of Variables
- Research Questions

Introduction

These approaches only allow for the use of **categorical** (nominal) **variables**!

Prominent nominal tests include:

- **Binomial Test**
- **McNemar**
- Fisher's Exact
- **Cochran's Q**
- **Chi-Squared**
- ...

The `table()` function

In reality, you will need to convert your data to fit the various nominal test specifications. To do so, you may wish to enlist the help of the `table()` function of base R which converts nominal records into count data.

```
Samples <- c("A", "B")
set.seed(42)
counts <- sample(Samples, size = 1000, replace = TRUE)
table(counts)

## counts
##      A      B
## 499 501
```

Purpose And Assumptions

Binomial Test

`binom.test()` in base R

Purpose: To test whether the observed distribution of data values of a binomial variable differ from what was expected.

H_0 *The observed binomial data proportions do not differ significantly from the expected proportions.*

Assumptions:

- Variable values are binomial.
- The population is significantly larger than the sample.
- The sample accurately represents the population.
- Sampled values are independent (one value does not influence another).

Minimal Working Example

We feed the `binom.test()` function a 1/4 (`c(200, 800)`) data set whilst expecting the distribution to be 1 to 1 ($p = 0.5$).

```
binom.test(c(200, 800), p = 0.5)

##
## Exact binomial test
##
## data: c(200, 800)
## number of successes = 200, number of trials = 1000,
## p-value <2e-16
## alternative hypothesis: true probability of success is not equal to 0.5
## 95 percent confidence interval:
##  0.18 0.23
## sample estimates:
## probability of success
##                0.2
```

The result is significant ($p \approx 0$).

Purpose And Assumptions

McNemar

`mcnemar.test()` in base R

Purpose: To test whether there is a change in proportion of paired data.

H_0 *The observed binomial data proportions do not differ significantly between treatments/paired sets.*

Assumptions:

- Variable values are binomial.
- The population is significantly larger than the sample.
- The sample accurately represents the population.

Minimal Working Example

We feed the `mcnemar.test()` function a 1 to 1 (`c(500, 500)`) as well as a 1 to 4 (`c(200, 800)`) data set for the paired data sets.

```
Performance <- matrix(c(500, 500, 200, 800), nrow = 2)
```

```
Performance
```

```
##      [,1] [,2]
```

```
## [1,]  500  200
```

```
## [2,]  500  800
```

```
mcnemar.test(Performance)
```

```
##
```

```
## McNemar's Chi-squared test with continuity
```

```
## correction
```

```
##
```

```
## data: Performance
```

```
## McNemar's chi-squared = 128, df = 1, p-value <2e-16
```

With a p-value of ≈ 0 the test concludes significantly.

Purpose And Assumptions

Cochran's Q

`cochran.q()` in the `nonpar` package

Purpose:

To test whether there are differences in matched sets of three or more frequencies or proportions.

H_0

The observed proportions of data values in treatments is equal among the matched sets.

Assumptions:

- The first (dependent/response) variable is binomial.
- The second (independent/predictor) variable is nominal/categorical with three values.
- The population is significantly larger than the sample.
- The sample accurately represents the population.

Minimal Working Example

We feed the `cochrans.q()` function a 6 by 4 matrix of binomial values. The first column represents our dependent, binomial variable. The remaining columns represent our independent variable on three levels expressed as binomial values.

```
CochranMatrix <- matrix(c(1, 1, 1, 1, 1, 1, 1, 1, 0,
  1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 1, 1), 6,
  4)
cochrans.q(CochranMatrix)

##
##  Cochran's Q Test
##
##  H0: There is no difference in the effectiveness of treatments.
##  HA: There is a difference in the effectiveness of treatments.
##
##  Q = 9.31578947368421
##
##  Degrees of Freedom = 3
##
##  Significance Level = 0.05
##  The p-value is 0.0253739987887868
##  There is enough evidence to conclude that the effectiveness of at least two treatments differ.
##
```

Purpose And Assumptions

Chi-Squared

`chisq.test()` in base R

Purpose:

To tests whether distributions of categorical variables differ from one another thus identifying whether they are related.

H_0

The distributions of nominal variables are equal.

Assumptions:

- Variable values are nominal/categorical.
- The population is significantly larger than the sample.
- The sample accurately represents the population.
- Sampled values are independent (one value does not influence another).

Minimal Working Example - One Sample Situation I

We feed the `chisq.test()` function an unbiased nominal set of three levels:

```
set.seed(42)
ChiMat1 <- table(sample(c("A", "B", "C"), 1000, replace = TRUE))
ChiMat1

##
##   A   B   C
## 329 360 311
chisq.test(ChiMat1)

##
##  Chi-squared test for given probabilities
##
## data:  ChiMat1
## X-squared = 4, df = 2, p-value = 0.2
```

Obviously, the observed distribution does not differ from our expectation of equally distributed proportions and so the test concludes non-significantly.

Minimal Working Example - One Sample Situation II

We feed the `chisq.test()` function a skewed (towards "A") nominal set of three levels:

```
set.seed(42)
ChiMat2 <- table(sample(c("A", "B", "C"), 1000, replace = TRUE,
  prob = c(0.8, 0.1, 0.1)))
ChiMat2

##
##   A   B   C
## 804 111  85
chisq.test(ChiMat2)

##
## Chi-squared test for given probabilities
##
## data:  ChiMat2
## X-squared = 998, df = 2, p-value <2e-16
```

Obviously, the observed distribution does differ from our expectation of equally distributed proportions and so the test concludes significantly.

Minimal Working Example - Two Sample Situation

We feed the `chisq.test()` our unbiased as well as our skewed (towards "A") nominal set of three levels to see whether their distributions differ significantly.

```
ChiMatrix <- cbind(ChiMat1, ChiMat2)
```

```
ChiMatrix
```

```
##      ChiMat1 ChiMat2
```

```
## A         329      804
```

```
## B         360      111
```

```
## C         311       85
```

```
chisq.test(ChiMatrix)
```

```
##
```

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

```
##
```

```
## data: ChiMatrix
```

```
## X-squared = 460, df = 2, p-value <2e-16
```

Clearly, they do differ significantly.

Variables We Can Use

Which variables in our *Passer domesticus* data set are nominal?

- Site Index
- Climate
- Population Status
- Colour
- Sex
- Nesting Site
- Flock
- Home Range
- Predator Presence
- Predator Type

All of these are nominal but some are binomial.

Research Questions And Hypotheses

So which of our major research questions (seminar 6) can we answer?

Binomial Test

- *Sexual Dimorphism*: Are the sexes represented in equal proportions?
- *Predation*: Are our sites dominated by predators or not?

McNemar

- *Sexual Dimorphism*: Compare sex ratio over time (we need to generate some new data for this using the sample function).

Cochran's Q

- *Sexual Dimorphism*: Are colours related to sex?
- *Predation*: Are colours related to predators?

Chi-Squared

- *Sexual Dimorphism*: Are colours related to sex?
- *Predation*: Are colours or nesting sites related to predators?