Basic statistical review

Chi-square test, t-test, correlations

Recap

Basic statistical review

Chi-square test, t-test, correlations

Lung cancer epidemiology

- 1922: 617 deaths from lung cancer in the UK
- 1947: 9287 deaths from lung cancer in the UK



Doll and Hill's study

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SMOKING AND CARCINOMA OF THE LUNG

BY

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Member of the Statistical Research Unit of the Medical Research Council

A. BRADFORD HILL, Ph.D., D.Sc.

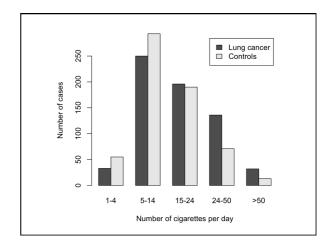
Professor of Medical Statistics, London School of Hygiene and Tropical Medicine: Honorary Director of the Statistical

Doll and Hill's study

- April 1948 to October 1949 2370 cases reported
- 150 over 75 and 80 incorrectly diagnosed
- 408 could not be interviewed for various reasons

Doll and Hill's study

- 1723 patients with carcinoma interviewed
- 743 general medical or surgical patients interviewed as controls for lung cancer patients
- Some carcinoma patients later proved to be misdiagnosed



Contingency table

Number of cigs	1-4	5-14	15-24	25-49	50+
Lung cancer	33	250	196	136	32
Control	55	293	190	71	13

Contingency table

Number of cigs	1-4	5-14	15-24	25-49	50+	Total
Lung cancer	33	250	196	136	32	647
Control	55	293	190	71	13	622
Total	88	543	386	207	45	1269

Expected value

For each cell in the table, the expected value is:

Column total X Row total Grand total

\subset	Con	ting	gen	cy t	abl	е	
Number of cigs	1-4	5-14	15-24	25-49	50+	Total	
Lung cancer	33	250	196	136	32	647	
Control	55	293	190	71	13	622	
Total	88	543	386	207	45	1269	
Expected value for highlighted cell is:							
88×647 1269 44.87							

Deviation from expected

33-44.87 = -11.87

We can calculate this for each cell in the table and get an indication of the extent by which the observed values deviate from the expected ones

Deviation from expected

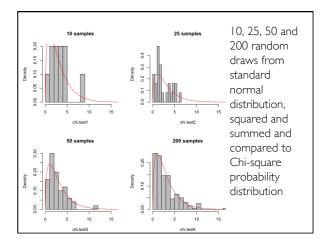
Number of cigs	1-4	5-14	15-24	25-49	50+
Lung cancer	-11.87	-26.84	-0.8	30.46	9.06
Control	11.87	26.84	0.8	-30.46	-9.06

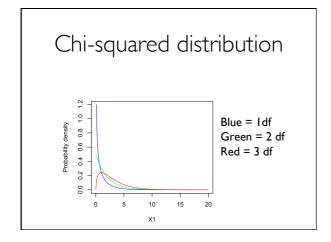
Deviation from the expected

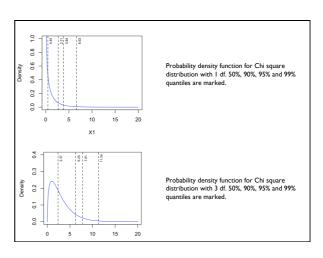
For each cell, we calculate:

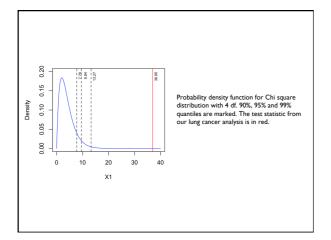
(observed-expected)² expected

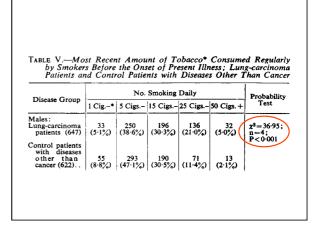
The sum of these is 36.95











In R: use chisq.test > lungs<matrix(data=c(33,250,196,136,32,55,293,190,71,13), byrow=TRUE, nrow=2) > chisq.test(lungs) Pearson's Chi-squared test data: lungs X-squared = 36.953, df = 4, p-value = 1.842e-07

Basic statistical review

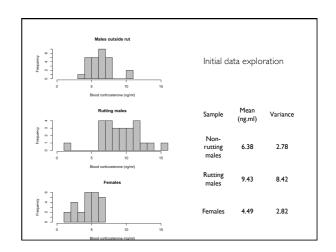
Chi-square test, **t-test**, correlations

A test of corticosterone levels in rutting stags

Study to investigate whether rutting causes elevated corticosterone levels in stags, and what levels of corticosterone are found in does

Stags darted and blood samples taken before and during the rut, from the same 25 individuals

Does darted and blood samples taken during the rut



Initial data exploration

All samples are roughly normally distributed
Rutting males>non-rutting males>females
Is the difference in means between rutting and non-rutting males statistically significant?

Testing the differences between males

We have sampled each male twice, before and during the nut

Therefore, we can express the change in corticosterone titre for each male as the second measurement minus the

Testing the differences between males

```
> deer$Males.in.rut-deer$Males
[1] 0.123 3.815 -0.403 8.179 6.266 5.849
0.263 4.492 4.894 2.734 0.733 7.589 5.034
4.179
[15] -1.338 -3.132 -5.721 5.125 1.555 9.207
7.221 0.390 0.746 0.862 8.742
> mean(deer$Males.in.rut-deer$Males)
[1] 3.09616
> sqrt(var(deer$Males.in.rut-deer$Males))
[1] 3.850616
```

Statistical testing recap

- We're trying to calculate the probability that our value for the mean could arise by random error when sampling from a population with a mean of zero
- Null Hypothesis.
- There is no difference between samples. The differences between samples are drawn from a population with a mean of zero
- Alternative hypothesis:
- There is a difference between samples. The differences between samples are drawn from a population with a mean not equal to zero

Calculating t

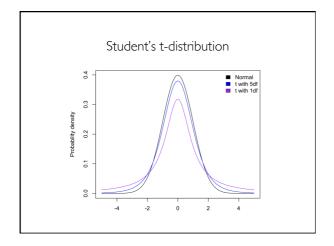
- We can do our statistical test if we calculate a value called "t", which is defined as the difference in means divided by the standard error
- The mean difference between our two samples is 3.10, s=3.85 so

t= $\frac{X-\mu}{s/\sqrt{N}}$ t= $\frac{3.10-0}{3.85/\sqrt{25}}$ t= $\frac{3.10/0.77}{4.02}$

Testing our mean

Using the estimate of the standard deviation causes problems because it will lead to systematic underestimation of σ

This is solved by comparing our value of t with *Student's t distribution*, which takes account of this



Testing our means

Our value of is t 4.02 with 24 df

The critical value of t for a 2-tailed test at 24 df is 2.064

Therefore we reject H_0 and accept H_1

Rutting stags have significantly higher corticosterone titres than the same stags sampled before the rut

In R:

use t.test

> t.test(deer\$Males.in.rut,deer\$Males, paired=TRUE)

Paired t-test

data: deer\$Males.in.rut and deer\$Males
t = 4.0203, df = 24, p-value = 0.0005005
alternative hypothesis: true difference in means is
not equal to 0
95 percent confidence interval:
1.506704 4.685616
sample estimates:
mean of the differences
3.09616

Basic statistical review

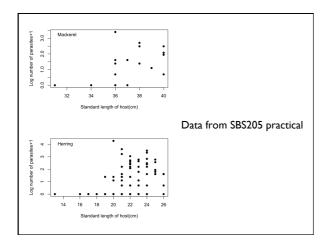
Chi-square test, t-test, correlations

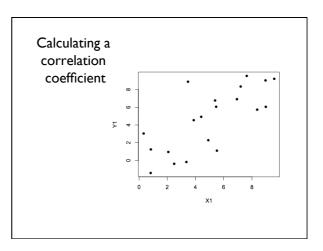
2 types of statistical test

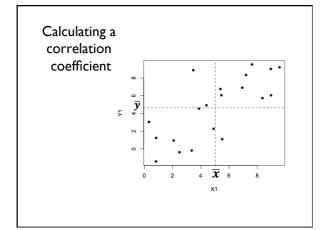
- Comparisons
- Relationships

Correlation

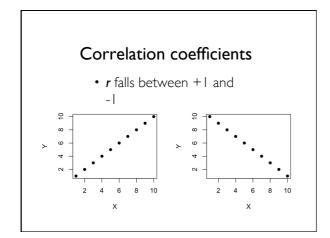
- Relationships or associations between continuous variables
- Can be positive or negative
- Shows the strength and significance of the relationship between 2 variables







Calculating a correlation coefficient
$$r = \sum \frac{(x-\overline{x})(y-\overline{y})}{s_x s_y} / \frac{1}{n-1}$$



Correlation coefficients: statistical significance

Ho: the two variables are unrelated Calculate a *t*-statistic and test at n-2df:

$$=\frac{r\sqrt{n-2}}{\sqrt{1-r^2}}$$

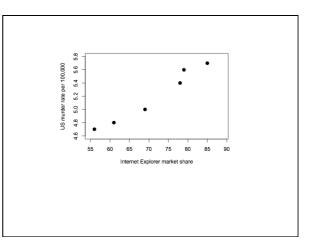
In R: use cor.test

Correlation coefficients:

- r varies between +1 and -1. The closer to 1 or -1, the stronger the correlation
- We can calculate a p-value associated with r
 to allow us to test for a statistically
 significant correlation
- Coefficient of determination, I^2 , is an estimate of the % variability in one variable explained by the other variable

Note: correlation does NOT mean causality

• If two variables are strongly and significantly correlated it does not mean one is the cause of the other



> cor.test(IE,murder)

Pearson's product-moment correlation

data: IE and murder
t = 10.1718, df = 4, p-value =
0.0005261
alternative hypothesis: true
correlation is not equal to 0
95 percent confidence interval:
 0.8329100 0.9980292
sample estimates:
 cor
0.981213

