The correlation

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1 The problem

In 1997, Robert Clarke and collaborators conducted a meta-analysis to determine the quantitative importance of dietary fatty acids to blood concentrations of cholesterol. They report 6 studies with the same protocol applied to 14 subjects each. Here is their data (the study is real but the data is fake):

Saturated fat in diet	Blood cholesterol
% total calories	$\mathrm{mmol/L}$
25.7	5.9
5.1	5.9
13.0	4.8
20.5	5.8
7.0	6.3
27.9	6.1

In your R session, manually enter the data in 2 vectors of length 6 (\mathtt{diet} , \mathtt{blood}).

```
> diet <- c(25.7, 5.1, 13.0, 20.5, 7.0, 27.9);
> blood <- c(5.9, 5.9, 4.8, 5.8, 6.3, 6.1);</pre>
```

Exercise 1

Can we assume that the data is Gaussian? If not, can we assume that the response variable (blood cholesterol) is Gaussian given the diet? What is the null hypothesis?

Exercise 2

We would like to find a statistic that measures 'association' between the two variables. Can you suggest a score? What is the alternative hypothesis?

2 The test statistic

For this problem, R made our life particulary easy with the function $\operatorname{\mathtt{cor}}$ which computes just our test statistic r.

Exercise 3

Plot the data in the (x, y) plane. Add the regression line with abline (lm(blood ~ diet)). Display the coefficients of the line by calling lm(blood ~ diet). Also compute cor(blood, diet) * sd(blood) / sd(diet), what do you observe?

Exercise 4

Verify by the formula, or with some examples at the terminal that r is invariant by translation and scaling of any or both of the variables. What does that mean for the resampling of r?

Exercise 5

Resample r, finish the test, estimate the p-value, compare with the results of cor.test.

3 Properties of the correlation

The coefficient of correlation has a few properties worth mentioning, mostly because **they can be very dangerous**.

Exercise 6

What is the maximum value that the coefficient of correlation can take? What is the minimum?

Exercise 7

What does that mean if two variables have a coefficient of correlation of 1 in absolute value?

Exercise 8

What does that mean if two variables have a coefficient of correlation of 0? Assign to \mathbf{x} all the integer numbers between -5 and 5. Specify $\mathbf{y} \leftarrow \mathbf{x}^2$, plot the variables in the (x,y) plane and compute their correlation. Conclude.

Answer of Exercise 1

We can assume that given a certain diet, the distribution of blood cholesterol is Gaussian ('reaction norms' are often Gaussian). Like for the t test, the null hypothesis can be formulated as follows:

- 1. The blood cholesterol is sampled from a Gaussian distribution.
- 2. The parameters are unknown, but equal in all cases.
- 3. Sampling is IID.

Answer of Exercise 2

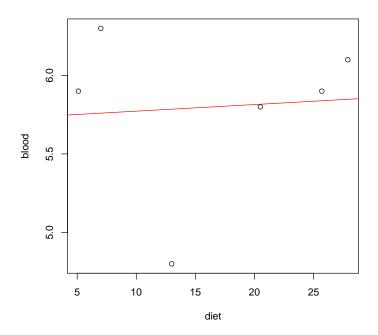
The coefficient of correlation r is a good score for the problem at hand. By definition it is the covariance of the two variables divided by he product of their standard deviations.

$$r = \frac{\sum_{i=1}^{6} (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum_{i=1}^{6} (x_i - \bar{x})^2 \sum_{i=1}^{6} (y_i - \bar{y})^2}}.$$

You can check that this is the same as computing the covariance (cov) between two vectors and dividing by the product of their standard deviations (sd). The n-1 terms cancel out in the numerator and the denominator.

In the alternative hypothesis, item 2 is replaced by 'The expected value of the response variable is a linear function of the independent variable (the diet here)'.

Answer of Exercise 3



The last term is the slope of the line. This is a property of the regression line, the slope is intimately linked to the correlation (by the formula used above).

Answer of Exercise 4

The proof is similar to that used for the effect size t. To verify it you can try different transformations.

```
> r.obs <- cor(blood, diet);
> r.obs;
[1] 0.07770512
> cor(blood+1, diet);
[1] 0.07770512
> cor(pi*blood+1, diet);
[1] 0.07770512
> cor(pi*blood+1, diet+2);
[1] 0.07770512
> cor(pi*blood+1, -1*diet+2);
[1] -0.07770512
```

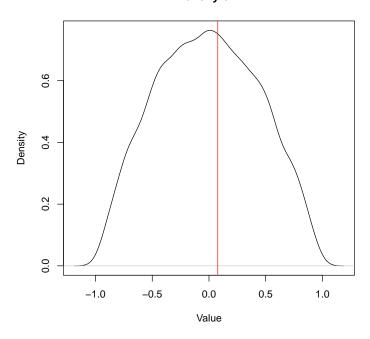
Note that the correlation changes sign but not value when one variable is multiplied by a negative number.

This means that we can resample r from standard Gaussian variables because all Gaussian variables have the same distribution of r under the null hypothesis.

Answer of Exercise 5

```
> r.smpl <- rep(NA, 10000);
> for (i in 1:10000) {
    r.smpl[i] <- cor(rnorm(6), diet);</pre>
> plot(density(r.smpl), main="Density of r", xlab="Value");
> abline(v=r.obs, col=2);
> quantile(abs(r.smpl), probs=0.95);
      95%
0.8178391
> mean(abs(r.smpl) > r.obs);
[1] 0.8826
> cor.test(blood, diet);
        Pearson's product-moment correlation
data: blood and diet
t = 0.1559, df = 4, p-value = 0.8837
alternative hypothesis: true correlation is not equal to {\tt O}
95 percent confidence interval:
 -0.7832498 0.8365138
sample estimates:
       cor
0.07770512
```

Density of r



Answer of Exercise 6

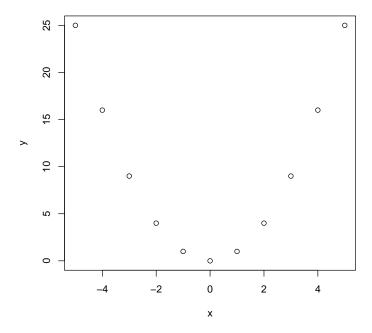
The coefficient of correlation is always between -1 and 1 (for real variables). This is a direct result of the Cauchy-Schwarz inequality, but it is rather difficult to prove if you don't know this result. You can verify by yourself by trying different inputs to cor. The highest you can find will be for cor(blood, blood) and the lowest for cor(blood, -blood).

Answer of Exercise 7

This means that one of the variables is a linear transformation of the other, *i.e.* a scaling and translation of the other. You can see it with cor(blood, 3*blood+4) etc.

Answer of Exercise 8

> x <- -5:5; > y <- x^2; > plot(x,y); > cor(x,y);



The coefficient of correlation is sometimes called the coefficient of **linear** correlation because it measures only linear trends between two variables. The coefficient of correlation should **never** be interpreted without a plot of y versus x, because many variables show a strong mutual dependency, but no linear correlation.