

# Example memory-imagery experiment: data for reference

How does mental imagery affect memory?

Two groups of subjects (say G1, G2) were asked to memorize pairs of words and then recall the pair given the first word as a cue after a 24 hours gap. G1 was instructed to actively use imagery to link the two words and given examples of such linking. G2 was just instructed to remember the pairs in whatever way they could. The number of pairs recalled was the response variable. The independent variable is the group (or treatment) and levels are G1-imagery, G2-control.

The results for 10 subjects with 5 subjects randomly assigned to each of G1, G2 are given below for 20 word pairs.

G1	G2
8	1
8	2
9	5
11	6
14	6

G1 mean=10; G2 mean=4; Grand mean=7.

# Exercise 1: Memory, imagery experiment I

- If we work out the details of exercise 1:  
 $SS_{\text{betweengroups}} = 90.0$ ,  $SS_{\text{withingroups}} = 48.0$ ,  
 $SS_{\text{total}} = 138.0 = 90.0 + 48.0$ . Also, verify that:  
 $SS_{\text{total}} = SS_{\text{betweengroups}} + SS_{\text{withingroups}}$ .
- $df_{\text{total}} = (\ell \times N) - 1 = 2 \times 5 - 1 = 9$ ,  
 $df_{\text{betweengroups}} = \ell - 1 = 2 - 1 = 1$ ,  
 $df_{\text{withingroups}} = \ell(N - 1) = 2 \times 4 = 8$ . Verify that  
 $df_{\text{total}} = df_{\text{betweengroups}} + df_{\text{withingroups}}$ .
- $s^2_{\text{betweengroups}} = \frac{SS_{\text{betweengroups}}}{df_{\text{betweengroups}}} = 90.0/1 = 90.0$
- $s^2_{\text{withingroups}} = \frac{SS_{\text{withingroups}}}{df_{\text{withingroups}}} = 48.0/8 = 6.0$

## Exercise 1: Memory, imagery experiment II

- $F = \frac{s_{\text{betweengroups}}^2}{s_{\text{withingroups}}^2} = 90.0/6.0 = 15.0$ . The corresponding  $F_{\text{critical}}$  values with  $\nu_1 = 1$ ,  $\nu_2 = 6$  are: at  $\alpha = 0.05$ ,  $F_{\text{critical}} = 5.987$ ; at  $\alpha = 0.01$ ,  $F_{\text{critical}} = 13.745$ . So,  $H_0$  is rejected at both  $\alpha$  levels.

## Detour: how are ANoVAs reported

<b>Source</b>	<b>df</b>	<b>SS</b>	<b>MS</b>	<b>F</b>
Between	1	90.0	90.0	15.0**
Within	8	48.0	6.0	
Total	9	138.0		

\*\*  $p < 0.01$

## $R^2_{Y.X}$ and $F$

- $R^2_{Y.X} = \frac{SS_{\text{betweengroups}}}{SS_{\text{total}}} = 90.0/138.0 \approx 0.6522.$
- Calculate  $F$ :

$$\begin{aligned} F &= \frac{R^2_{Y.X}}{1 - R^2_{Y.X}} \times \frac{df_{\text{withingroups}}}{df_{\text{betweengroups}}} \\ &= \frac{90/138}{48/138} \times \frac{8}{1} \\ &= 15 \end{aligned}$$

- We get exactly the same  $F$  value of 15.

# Connection between ANOVA and regression I

- ANOVA creates  $\ell$  groups based on the levels of a categorical factor.
- To see the connection between ANOVA and regression we need to convert the categorical factor into a numeric variable  $X$  that can be used to predict the value of the dependent variable  $Y$  via a regression equation.
- We use the following trick: Identify a group by its group mean  $\mu_I$  for the  $I^{th}$  group. So, instead of identifying a group by the value of the categorical variable we choose to identify it by the corresponding group mean  $\mu_I$ .

## Connection between ANOVA and regression II

- Simultaneously, the best predictor for any member of a group is the corresponding group mean  $\mu_I$ . So, one way to look at the X-value in a group is to use the corresponding  $\mu_I$  for each X-value. Note that in the ANOVA view there is no X-value - so this an invented X-value but it is the most logical choice.
- So  $X$  has been transformed into a numeric independent variable and we can now construct a regression equation using the data.

$$\hat{Y} = a_r + bX$$

- We have:  $a_r = \mu_Y - b \mu_X$  and  $b = \frac{SCP_{X \cdot Y}}{SS_X}$ .
- Our choice  $\mu_I$  for the group categorical variable and for the predictor values in a group means that  $\mu_Y = \mu_X = \mu_{\text{grand}}$ . Also,  $SCP_{X \cdot Y} = SS_X$ , so  $b = 1$  and  $a_r = 0$ . (Proved more formally later)

# Connection between ANOVA and regression III

- So,  $\hat{Y} = X = \mu_I$ . So, ANOVA and regression are essentially the same thing.

One way to see this is to look at the memory-imagery data as data for regression - (superimposed with the factor levels)

	$I = \text{control}$					$I = \text{imagery}$				
<b>X</b>	4	4	4	4	4	10	10	10	10	10
<b>Y</b>	1	2	5	6	6	8	8	9	11	14

We see exactly why  $\mu_X = \mu_Y = \mu_{\text{grand}}$ .

- The previous also means:

$$SS_{\text{total-regr}} = SS_{\text{total-anova}}$$

$$SS_{\text{regression}} = SS_{\text{betweengroups}} \quad \text{and}$$

$$SS_{\text{residual}} = SS_{\text{withingroups}}$$



# Connection between ANOVA and regression IV

- The degrees of freedom also have similar equivalences:

$$df_{\text{total-regr}} = df_{\text{total-anova}}$$

$$df_{\text{regression}} = df_{\text{betweengroups}}$$

$$df_{\text{residual}} = df_{\text{withingroups}}$$

- The quality of the regression and the intensity of the dependence also align.

$$r_{X \cdot Y}^2 = R_{Y \cdot X}^2$$

$$\frac{SS_{\text{regression}}}{SS_{\text{total-regr}}} = \frac{SS_{\text{betweengroups}}}{SS_{\text{total-anova}}}$$

# Comparison chart: Regression, ANOVA

## Regression

$Y$

$\hat{Y}$

$X$

$\mu_Y$

$\mu_X$

$SS_{\text{regression}}$

$SS_{\text{residual}}$

$R^2_{Y \cdot X}$  or  $r^2_{Y \cdot X}$

$N$  - total subjects

## ANOVA (I-group label, j-subject in group I)

$Y_{Ij}$

$\mu_I$

$\mu_I$

$\mu_{\text{grand}}$

$\mu_{\text{grand}}$

$SS_{\text{betweengroups}}$

$SS_{\text{withingroups}}$

$R^2_{Y \cdot X}$

$\ell \times N$ ,  $N$ =no. of subjects in a group

# The regression-ANoVA connection is more general I

- To see the connection between regression and ANoVA we identified each group by the corresponding ANoVA group mean  $\mu_I$  and for each X-value in the group we again used the same mean  $\mu_I$ .
- However, the choice of the group mean is the most intuitively natural one it is not necessary for the correspondence between regression and ANoVA. We can use any value in place of  $\mu_I$  to identify the group and for the X-values within a group provided the values for each group are distinct.
- using an arbitrary value instead of  $\mu_I$  changes the values of the regression coefficients  $a_r$  and  $b$  which were respectively 0 and 1 when we used  $\mu_I$ . But the principle and the other correspondences remain unchanged.

## The regression-ANoVA connection is more general II

- The reason for having two different ways of doing the same thing is historical. The psychological literature uses ANoVA while the clinical literature uses regression. ANoVA makes it easier to handle the analysis when more factors or independent variables are involved.

### Exercise 8

*Repeat the memory-imagery analysis by using +1 for the imagery group and -1 as the value for the control group. Repeat again with values 1 and 0. Observe how the values of  $a_r$  and  $b$  change based on how the groups are represented numerically. Also, verify via calculation that the correspondences remain unchanged.*

## Some proofs-1

For ANOVA we only have Y-values for each of  $\ell$  groups written  $Y_{lj}$  for the value of subject  $j$  in group  $l$ . Each group has  $N$  subjects.

$\mu_l$  is the mean of the  $l^{th}$  group  $\mu_l = \frac{\sum_{j=1}^N Y_{lj}}{N}$ , and  $\mu_{grand} = \frac{\sum_{l=1}^{\ell} \mu_l}{\ell}$  is the grand mean.

$$SS_{\text{betweengroups}} = \sum_{l=1}^{\ell} \sum_{j=1}^N (\mu_l - \mu_{grand})^2 = N \sum_{l=1}^{\ell} (\mu_l - \mu_{grand})^2,$$

$$SS_{\text{withingroups}} = \sum_{l=1}^{\ell} \sum_{j=1}^N (Y_{lj} - \mu_l)^2,$$

$$SS_{\text{total}} = \sum_{l=1}^{\ell} \sum_{j=1}^N (Y_{lj} - \mu_{grand})^2.$$

For regression we have the pairs:

$\{(X_{11}, Y_{11}), \dots, (X_{1N}, Y_{1N}), \dots, (X_{\ell 1}, Y_{\ell 1}), \dots, (X_{\ell N}, Y_{\ell N})\}$ . The sample size is  $N_r = \ell \times N$  (we use  $N_r$  for regression instead of  $N$  to avoid confusion). For each  $l = 1.. \ell$  and each  $j = 1..N$ ,  $X_{lj} = \mu_l$  so  $\mu_X = \mu_Y = \mu_{grand}$  (see the ANOVA formulae).

Let the regression equation be:  $\hat{Y} = a_r + bX$ . We know:

$a_r = \mu_Y - b\mu_X$  and  $b = \frac{SCP_{X \cdot Y}}{SS_X}$ . So we calculate  $SCP_{X \cdot Y}$  and  $SS_X$ .

## Some proofs-2

$$\begin{aligned} SCP_{X.Y} &= \sum_{l=1}^{\ell} \sum_{j=1}^N (X_{lj} - \mu_X)(Y_{lj} - \mu_Y) \\ &= \sum_{l=1}^{\ell} \sum_{j=1}^N (X_{lj} - \mu_{\text{grand}})(Y_{lj} - \mu_{\text{grand}}), \quad \mu_X = \mu_Y = \mu_{\text{grand}} \\ &= \sum_{l=1}^{\ell} \sum_{j=1}^N (\mu_l - \mu_{\text{grand}})(Y_{lj} - \mu_{\text{grand}}), \quad X_{lj} = \mu_l \\ &= \sum_{l=1}^{\ell} (\mu_l - \mu_{\text{grand}}) \sum_{j=1}^N (Y_{lj} - \mu_{\text{grand}}) \\ &= \sum_{l=1}^{\ell} (\mu_l - \mu_{\text{grand}}) \left[ \sum_{j=1}^N Y_{lj} - \mu_{\text{grand}} \sum_{j=1}^N 1 \right] \\ &= \sum_{l=1}^{\ell} (\mu_l - \mu_{\text{grand}}) [N\mu_l - N\mu_{\text{grand}}] \\ &= N \sum_{l=1}^{\ell} (\mu_l - \mu_{\text{grand}})^2 \end{aligned}$$

## Some proofs-3

Similarly,

$$\begin{aligned}SS_X &= \sum_{l=1}^{\ell} \sum_{j=1}^N (X_{lj} - \mu_{\text{grand}})^2 \\&= \sum_{l=1}^{\ell} \sum_{j=1}^N (\mu_l - \mu_{\text{grand}})^2, \quad X_{lj} = \mu_l \\&= \sum_{l=1}^{\ell} (\mu_l - \mu_{\text{grand}})^2 \sum_{j=1}^N 1 \\&= N \sum_{l=1}^{\ell} (\mu_l - \mu_{\text{grand}})^2\end{aligned}$$

So,  $SCP_{X.Y} = SS_X$  which implies  $b = 1$  and  $a_r = 0$ . This means  $\hat{Y} = X = \mu_l$ .

## Some proofs-4

We show  $SS_{\text{regression}} = SS_{\text{betweengroups}}$  and  
 $SS_{\text{residual}} = SS_{\text{withingroups}}$ .

$$SS_{\text{regression}} = \sum_{i=1}^{N_r = \ell \times N} (\hat{Y}_i - \mu_Y)^2, \quad \text{see regression lecture}$$

$$= \sum_{l=1}^{\ell} \sum_{j=1}^N (\hat{Y}_{lj} - \mu_{\text{grand}})^2, \quad \mu_Y = \mu_{\text{grand}}$$

$$= \sum_{l=1}^{\ell} \sum_{j=1}^N (\mu_l - \mu_{\text{grand}})^2, \quad \hat{Y}_{lj} = \mu_l$$

$$= SS_{\text{betweengroups}}$$

$$SS_{\text{residual}} = \sum_{i=1}^{N_r = \ell \times N} (Y_i - \hat{Y}_i)^2 \quad \text{see regression lecture}$$

$$= \sum_{l=1}^{\ell} \sum_{j=1}^N (Y_{lj} - \hat{Y}_{lj})^2$$

$$= \sum_{l=1}^{\ell} \sum_{j=1}^N (Y_{lj} - \mu_l)^2, \quad \hat{Y}_{lj} = \mu_l$$



# Exercise

## Exercise 9

*For the Romeo-Juliet experiment repeat the calculation by using the group mean to represent the groups and the individual X-values in each group as was done for the memory-imagery experiment and use regression analysis to confirm that you get the same results as the ANOVA.*