

## Handout 1: Deriving least squares estimates of $B_0$ and $B_1$ in simple regression

Mission: Minimize the residual sum of squares (RSS) with respect to our model parameters,  $B_0$  and  $B_1$ . We use partial derivatives.

$$\begin{aligned} \sum_{i=1}^n E_i^2 &= \sum E_i^2 = \sum (Y_i - \hat{Y}_i)^2 = \sum (Y_i - (B_0 + B_1 X_{1i}))^2 = \\ (1) \quad &\sum (Y_i - B_0 - B_1 X_{1i})^2 \end{aligned}$$

1. Differentiate Eq. 1 with respect to  $B_0$

$$(2) \quad \frac{\partial \sum E_i^2}{\partial B_0} = \sum (-1)(2)(Y_i - B_0 - B_1 X_{1i}) = -2 \sum (Y_i - B_0 - B_1 X_{1i})$$

2. Differentiate Eq. 1 with respect to  $B_1$

$$\begin{aligned} \frac{\partial \sum E_i^2}{\partial B_1} &= \sum (-X_{1i})(2)(Y_i - B_0 - B_1 X_{1i}) = \\ (3) \quad &-2 \sum (X_{1i})(Y_i - B_0 - B_1 X_{1i}) \end{aligned}$$

3. Find the point on the function where the slope of the line tangent to the function (RSS) is zero. Thus, set Eqs. 2 and 3 to zero and solve simultaneously.

$$(4) \quad 0 = -2 \sum (Y_i - B_0 - B_1 X_{1i})$$

$$(5) \quad 0 = -2 \sum (X_{1i})(Y_i - B_0 - B_1 X_{1i})$$

4. Divide both sides of Eqs. 4 and 5 by -2

$$(4a) \quad 0 = \sum (Y_i - B_0 - B_1 X_{1i})$$

$$(5a) \quad 0 = \sum (X_{1i})(Y_i - B_0 - B_1 X_{1i})$$

5. Distribute summation operator in both equations and distribute  $X_i$  in Eq. 5a

$$(6) \quad 0 = \sum Y_i - nB_0 - B_1 \sum X_{1i}$$

$$(7) \quad 0 = \sum X_{1i}Y_i - B_0 \sum X_{1i} - B_1 \sum X_{1i}^2$$

6. Rearrange Eqs. 6 and 7, and we get the “least squares normal equations.”

$$(8) \quad \sum Y_i = nB_0 + B_1 \sum X_{1i}$$

$$(9) \quad \sum X_{1i}Y_i = B_0 \sum X_{1i} + B_1 \sum X_{1i}^2$$

7. Multiply Eq. 8 by  $-\sum X_{1i}$ ; multiply Eq. 9 by N

$$(10) \quad -\sum X_{1i} \sum Y_i = -nB_0 \sum X_{1i} - B_1 (\sum X_{1i})^2$$

$$(11) \quad n \sum X_{1i}Y_i = nB_0 \sum X_{1i} + nB_1 \sum X_{1i}^2$$

8. Add Eqs. 10 and 11 together.

$$(12) \quad n \sum X_{1i}Y_i - \sum X_{1i} \sum Y_i = nB_0 \sum X_{1i} - nB_0 \sum X_{1i} + nB_1 \sum X_{1i}^2 - B_1 (\sum X_{1i})^2$$

9. Reduce Eq. 12

$$n \sum X_{1i}Y_i - \sum X_{1i} \sum Y_i = nB_1 \sum X_{1i}^2 - B_1 (\sum X_{1i})^2$$

$$n \sum X_{1i}Y_i - \sum X_{1i} \sum Y_i = B_1 [n \sum X_{1i}^2 - (\sum X_{1i})^2]$$

10. Solve for  $B_1$

$$B_1 = \frac{n \sum X_{1i} Y_i - \sum X_{1i} \sum Y_i}{n \sum X_{1i}^2 - (\sum X_{1i})^2} =$$
$$\frac{\sum (X_{1i} - \bar{X})(Y_i - \bar{Y})}{\sum (X_{1i} - \bar{X})^2} = \frac{Cov(X_1, Y)}{Var(X_1)} = r_{Y, X_1} \frac{s_Y}{s_{X_1}}$$

11. Solve for  $B_0$ ; use Eq. 8 from the normal equations

$$\sum Y_i = nB_0 + B_1 \sum X_{1i}$$

$$\sum Y_i - B_1 \sum X_{1i} = nB_0$$

$$\frac{\sum Y_i}{n} - \frac{B_1 \sum X_{1i}}{n} = B_0$$

$$B_0 = \bar{Y} - B_1 \bar{X}$$