

# Implementation of `lm.beta`

Stefan Behrendt

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The package `lm.beta` is based on equation (1) to estimate the standardized regression coefficients.

$$\hat{\beta}_i = \hat{b}_i \cdot \frac{\sqrt{\sum_j (X_{i,j} - \bar{X}_i \cdot I)^2}}{\sqrt{\sum_j (Y_j - \bar{Y} \cdot I)^2}} \quad (1)$$

with

- $\hat{\beta}_i$  the  $i$ -th standardized regression coefficient
- $\hat{b}_i$  the  $i$ -th unstandardized regression coefficient
- $I = \begin{cases} 0/1 & \text{for models without intercept*} \\ 1 & \text{for models with intercept} \end{cases}$ 
  - \* argument `complete.standardization` chooses the factor: `complete.standardization = FALSE`  $\Rightarrow I = 0$  / `complete.standardization = TRUE`  $\Rightarrow I = 1$
  - \* IBM<sup>®</sup> SPSS Statistics<sup>®</sup>, e.g., always uses  $I = 0$  for models without intercept
  - \* see e.g. [https://online.stat.psu.edu/~ajw13/stat501/SpecialTopics/Reg\\_thru\\_origin.pdf](https://online.stat.psu.edu/~ajw13/stat501/SpecialTopics/Reg_thru_origin.pdf)<sup>1</sup> for further information on which  $I$  to choose
- $\bar{A}$  the arithmetic mean of  $A$
- $Y$  the dependent variable
- $X_i$  the  $i$ -th independent variable

A simplification for  $I = 1$  is shown in equation (2) and for  $I = 0$  in equation (3).

$$\hat{\beta}_i = \hat{b}_i \cdot \frac{s_{X_i}}{s_Y} \quad (2)$$

$$\hat{\beta}_i = \hat{b}_i \cdot \frac{\sigma_{X_i}}{\sigma_Y} \quad (3)$$

with (additionally to above)

- $s_A$  the standard deviation of  $A$  (\*)
- $\sigma_A = \sqrt{\sum_j A_j^2}$  an estimate of the uncentered second moment of  $A$  (\*)
- \* The sample size—and the different methods for correcting it—doesn't have to be considered when estimating the moments, because the factors would be similar in numerator and denominator, and therefore would be reduced.

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<sup>1</sup>Eisenhauer J.G. (2003). Regression through the Origin. *Teaching Statistics*, 25(3), p. 76-80.