

# Review of linear models

2 Sep 2023

## Basics

- assume  $\mathbf{y} \sim \text{Normal}(\mathbf{X}\beta, \sigma)$ <sup>1</sup>
- $\mathbf{X}$  is the *model matrix*, can be anything we want it to be
- the *Gauss-Markov theorem* ([Wikipedia](#)) makes weaker assumptions:  $\mathbf{y} = \mathbf{X}\beta + \epsilon$ ; as long as  $\epsilon$  is mean-zero, homoscedastic with finite variance, and uncorrelated ... then the OLS solution

$$\hat{\beta} = (\mathbf{X}^\top \mathbf{X})^{-1} \mathbf{X}^\top \mathbf{y}$$

is the BLUE (or MVUE).

- we'll embrace the assumptions (which are needed for inference!)

## Model matrices

- everything has to be converted to  $\mathbf{X}$  before we start
  - transformations
  - encoding of categorical variables: **contrasts**
  - interactions
  - basis expansions (e.g. polynomials)

## Wilkinson-Rogers formulas (Wilkinson and Rogers 1973)

- operators: +, \*, :, /, -, ^
- I()

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<sup>1</sup>Notation-abuse warning ...

## Contrasts

## Computation

- matrix decompositions (QR with pivoting)
- big problems: `biglm`, `speedglm`, `RcppEigen::fastLm`
  - optimized BLAS, kernel trick, etc.
  - memory vs speed vs robustness ...
  - $p$  vs.  $n$  vs. many-small-regressions vs. ...

## Diagnostics

## References

Wilkinson, G. N., and C. E. Rogers. 1973. "Symbolic Description of Factorial Models for Analysis of Variance." *Journal of the Royal Statistical Society. Series C (Applied Statistics)* 22 (3): 392–99. <https://doi.org/10.2307/2346786>.