# Review of linear models

3 Sep 2023

## Basics

- assume $\mathbf{y} \sim \text{Normal}(\mathbf{X}\beta, \sigma)$[1]
- $\mathbf{X}$ 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
$$\widehat{\beta} = (\mathbf{X}^\top \mathbf{X})^{-1} \mathbf{X}^\top \mathbf{y}$$
is the BLUE (or MVUE).
- we'll embrace the assumptions (which are needed for inference!)

## 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. ...

## Inference

- $\sigma^2$ (residual variance) is RSS$/(n-p)$
- The covariance matrix is $\sigma^2(\mathbf{X}^\top \mathbf{X})^{-1}$.
- Individual coefficients are $t$-distributed
- Linear combinations of coefficients are $F$-distributed
- Wald and likelihood ratio test comparisons are equivalent (but need to be careful about marginality)

[1] Notation-abuse warning ...

## Model matrices

- model definition converted to **X** before we start
- **input variables** vs **predictor variables** (Schielzeth (2010), Gelman & Hill (2006), CV)
  - transformations
  - encoding of categorical variables: **contrasts**
  - interactions
  - basis expansions (e.g. polynomials)

## Wilkinson-Rogers formulas

- Wilkinson & Rogers (1973), updated by Chambers & Hastie (1991, ch. 2)
- operators: `+`, `*`, `:`, `/`, `-`, `^`
- `I()`

## Contrasts

## Marginality

- Venables (1998)
- 'type (X) sums of squares'
- scaling and centering (Schielzeth, 2010)

## Downstream methods

- prediction, effects plots
- uncertainty of predictions
- `emmeans`, `marginaleffects`, `effects`, `sjPlot` ...
- `tidy()`, `performance`, `insight`, etc. ...

## Diagnostics

- linearity,
- base R: `stats::plot.lm()`
- `performance::check_model()`
- `DHARMa` (`simulateResiduals(., plot = TRUE)`)

## References


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Gelman, A., & Hill, J. (2006). *Data Analysis Using Regression and Multilevel/Hierarchical Models*. Cambridge University Press.

Schielzeth, H. (2010). Simple means to improve the interpretability of regression coefficients: Interpretation of regression coefficients. *Methods in Ecology and Evolution*, *1*(2), 103–113. https://doi.org/10.1111/j.2041-210X.2010.00012.x

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