

Review of linear models

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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 ϵ 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!)

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

- σ^2 (residual variance) is $\text{RSS}/(n - p)$
- 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)

¹Notation-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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- 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>
- Venables, W. N. (1998). *Exegeses on Linear Models*. <http://www.stats.ox.ac.uk/pub/MASS3/Exegeses.pdf>
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