High-dimensional statistics

topics

- ▶ High-dimensional (generalized) linear models:
 Lasso and modifications
 → Book
- lacktriangle Group Lasso for group sparsity \longrightarrow Book
- Additive models and many smooth univariate functions
- ▶ De-biasing and de-sparsified Lasso → Additional Material

 \longrightarrow Book

- ► Stability selection → Book
- lacktriangle Multiple sample splitting \longrightarrow Book
- Hidden confounding and deconfounding methods
- lacksquare Undirected graphical modeling \longrightarrow Book

(Text-) Book:

Bühlmann, P. and van de Geer, S. (2011). Statistics for High-Dimensional Data: Methods, Theory and Applications. Springer.

Additional Material:

https://stat.ethz.ch/~buhlmann/teaching/desparsifiedLasso.pdf

James, Witten, Hastie and Tibshirani (2023). An Introduction to Statistical Learning with Applications in R/Python, 2nd ed. Springer

See also: https:

//stat.ethz.ch/~buhlmann/teaching/highdim-stats-HS2023.html

An example: Riboflavin production with Bacillus Subtilis (in collaboration with DSM (Switzerland))

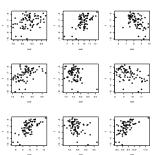
goal: improve riboflavin production rate of Bacillus Subtilis using clever genetic engineering

response variables $Y \in \mathbb{R}$: riboflavin (log-) production rate covariates $X \in \mathbb{R}^p$: expressions from p = 4088 genes sample size $n = 115, p \gg n$

Y versus 9 "reasonable" genes

gene expression data





High-dimensional data

general framework:

$$Z_1,\ldots,Z_n$$
 i.i.d., $\dim(Z_i)\gg n$

for example:

$$Z_i = (X_i, Y_i), \ X_i \in \mathbb{R}^p, \ Y_i \in \mathbb{R}$$
: regression with $p \gg n$ $Z_i = (X_i, Y_i), \ X_i \in \mathbb{R}^p, \ Y_i \in \{0, 1\}$: classification with $p \gg n$ $Z_i = (X_i), \ X_i \in \mathbb{R}^p$: graphical modeling with $p \gg n$

numerous applications:

biology, imaging, economy, environmental sciences, ...