



UNIVERSITY OF
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Master of Science
in Analytics

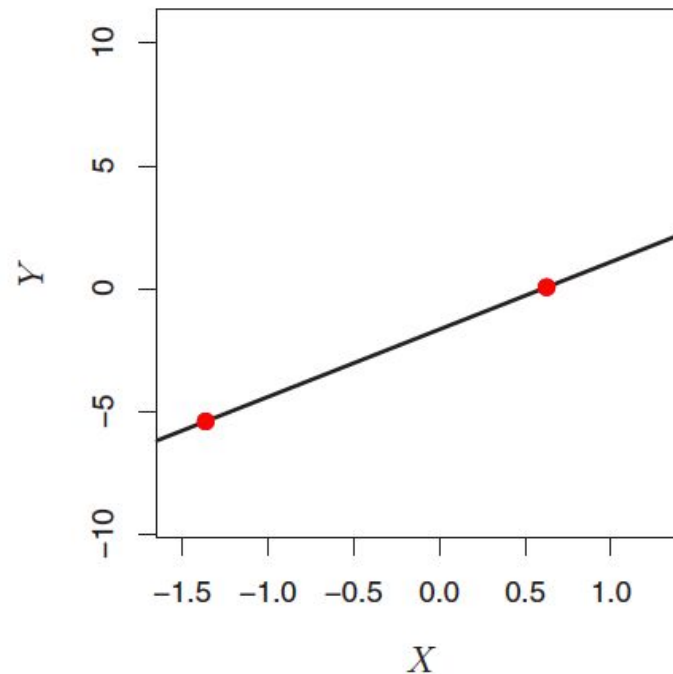
Machine Learning in High Dimensions

Machine Learning 1



High-dimensions; high problems

- How high is high?
 - Any time n is close to or smaller than count of features (p)
 - More common as time moves forward
- Simple Linear Regression fails
 - Example case: $n = 2$ & $p = 1$
 - $MSE_{\text{Train}} = 0$; $MSE_{\text{Test}} = ??$
 - All σ^2 -based techniques fail
 - C_p , AIC, BIC
 - *Adjusted R^2*





What could work

- Generally: fitting less flexible models
 - Any technique which (aggressively) avoids overfitting
 - Forward stepwise selection, ridge regression, the lasso, PCA
- Estimating test error
 - Directly: use validation set / cross validation
 - Indirectly: make adjustment to training error, account for overfitting



Conceptual Example

- Blood pressure
 - Assume model is predicting blood pressure
 - Features: height, weight, single-nucleotide polymorphism (SNP)
 - SNP is a variation in a single nucleotide in a specific genomic position
 - There could be ~ 500,000 SNPs
- Task:
 - Find the feature sets which predict high blood pressure
 - May use (for example) forward stepwise selection to create model
- If you find a solution, you may say
 - Your model is *one of possibly several* to predict the outcome
 - Your model should be validated on independent data sets