

Making a Case

HN

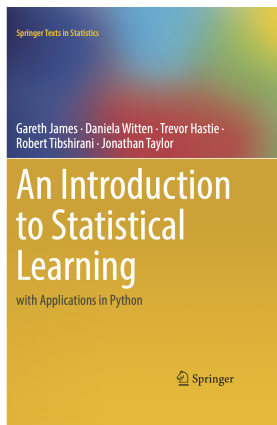
Washington State University

January 29, 2025



- Anything worth making needs patience and practice
- If you want mastery, you need to immerse yourself in it till it becomes your second nature, just like walking
- Top(?) two take aways from graduate school
 - 1 Try not to have a bias/prejudice. be playful Q
 - 2 When you are wrong, you are wrong ⚠

Source of This Notes



Notes from **An Introduction to Statistical Learning** with Applications in Python

- It's an easy read and as the name suggests, it's just Introduction. Good for intuition building
- Its PDF is available for free from the Authors' website.
- Sign-up & receive coupons (30%-50% off)

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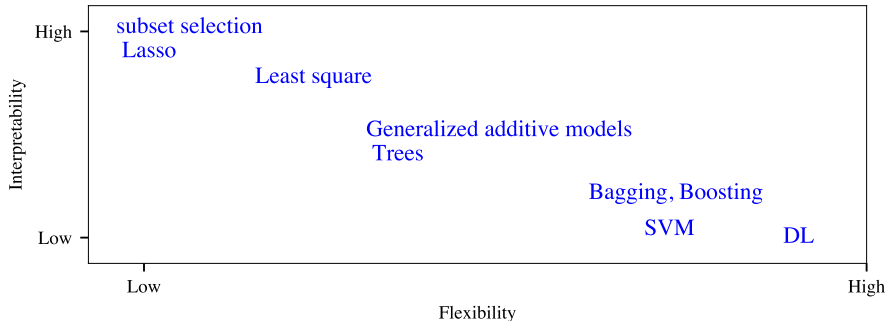
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- “...and an R^2 value well below 0.1 might be more realistic!” (page 79)

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More Resources

- Guesstimation: (Weinstein and Adam, [2008](#))
- First Course in Probability (Ross et al., [1976](#))
- Introduction to Linear Regression Analysis (Montgomery, Peck, and Vining, [2021](#))

Non-technical

- Excellent Sheep: The Miseducation of the American Elite and the Way to a Meaningful Life (Deresiewicz, [2014](#))
- The Culture Code: The Secrets of Highly Successful Groups (Coyle, [2018](#))

Tools






- Software Carpentry has lots of tutorials including [GitHub](#)!

Some Definition I

Definition (Variance of a method)

Variance refers to the amount by which \hat{f} would change if we estimated it using a different training data set.

Bibliography I

-  Coyle, D. (2018). *The Culture Code: The Secrets of Highly Successful Groups*. Bantam Books, an imprint of Random House, a division of Penguin Random House LLC New York.
-  Deresiewicz, W. (2014). *Excellent Sheep: The Miseducation of the American Elite and the Way to a Meaningful Life*. Free Press.
-  Montgomery, D. C., Peck, E. A., and Vining, G. G. (2021). *Introduction to Linear Regression Analysis*. John Wiley & Sons.
-  Ross, S. M. et al. (1976). *A first course in probability*. Vol. 2. Macmillan New York.
-  Weinstein, L. and Adam, J. A. (2008). *Guesstimation: Solving the world's problems on the back of a cocktail napkin*. Princeton University Press.