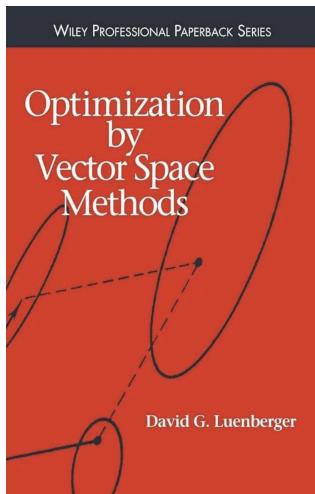


# The Western Gappy Canon

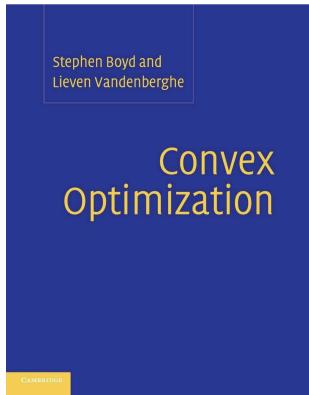
Here are some semi-random book recommendations I have given on X, both in public and private messages, with a short explanation of why. They are highly idiosyncratic. I spent time with these books.

## Optimization, Approximation and Basic Analysis



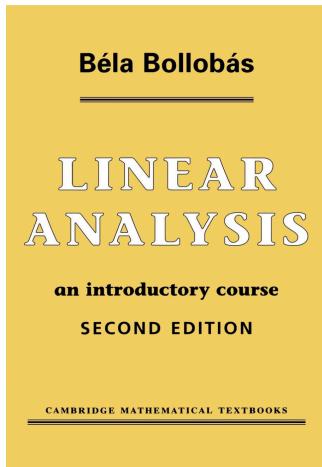
D. Luenberger, *Optimization by Vector Space Methods* ([hardcopy](#), [kindle](#) [much cheaper])

*This was an outgrowth of Luenberger's PhD thesis. It is a very friendly introduction to basic concepts of functional analysis, and a treatment of optimization in abstract spaces. A masterful example of great exposition, and extremely useful.*



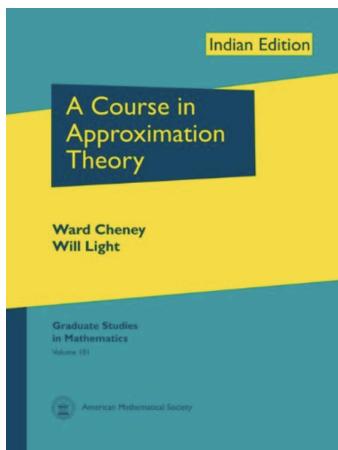
S. Boyd and L. Vandenberghe, Convex Optimization ([hardcopy/kindle](#), [free copy](#))

*Friendly, free, complete. It focuses on the theory of optimization, in a discursive but rigorous way. Can be read back to back for self-instruction. Very good reference.*



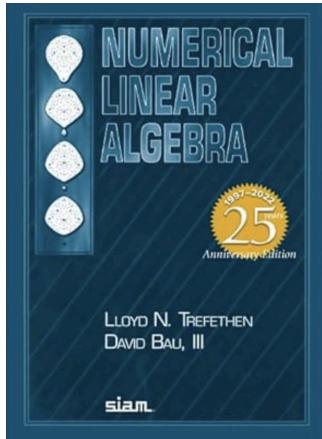
B. Bollobas, Linear Analysis, 2nd ed. ([hardcopy](#))

*Undergraduate course material on functional analysis covering all the bases for the working analyst. The easiest treatment I know of, written by another master expositor.*



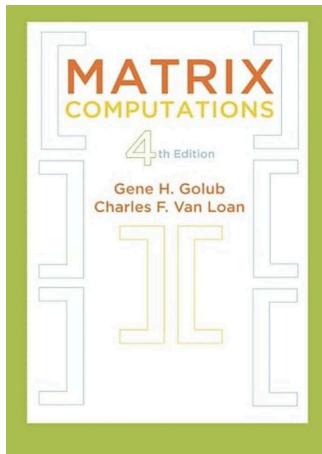
E. W. Cheney, W. Light, A Course In Approximation Theory ([Hardcopy](#))  
*You need to know approximation theory, especially in the era of overcomplete models. This is a book about the theory of AT. It will open your mind to new concepts.*

# Linear Algebra



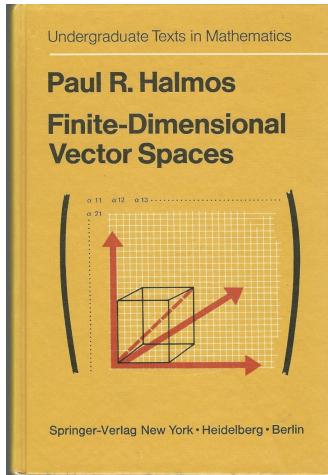
L. N. Trefethen, Bau, Numerical Linear Algebra ([paperback](#))

*The best way to learn linear algebra is to learn numerical linear algebra. Great introduction to matrix decompositions and computation of projections.*



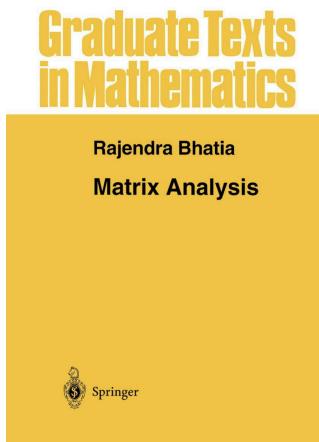
G. Golub, C. Van Loan, Matrix Computations, 4th ed. ([hard/paper/soft](#))

*The bible of numerical linear algebra. So clear and useful. Mostly useful as a reference*



P. R. Halmos, Finite Dimensional Vector Spaces ([hardcopy](#))

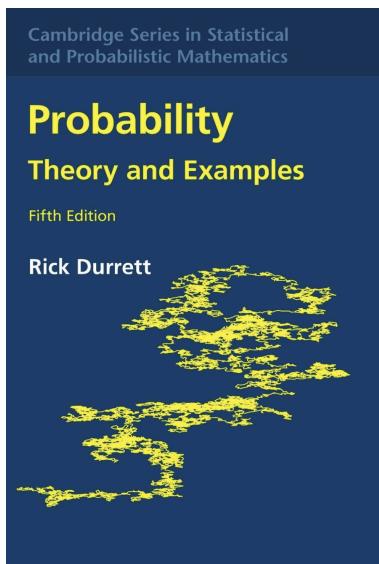
*How to get intuition about linear algebra. Undergraduate-level but not dumb*



R. Bhatia, Matrix Analysis ([hard/paper](#))

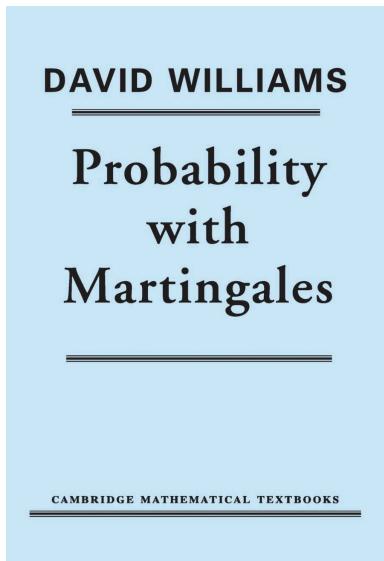
*This is in equal parts linear algebra and functional analysis in finite spaces. It is advanced and synthetic. If you know the contents of this book, you know linear algebra.*

# Probability



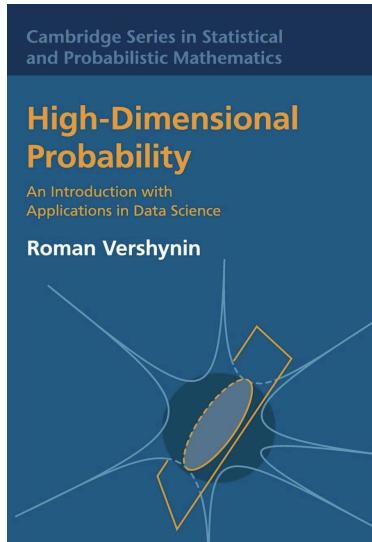
R. Durrett, Probability, 5th ed. ([hard/softcover](#))

*I learned graduate probability on this book. Not easy. It has thoroughly corrected typos. Self-contained. Interesting examples.*

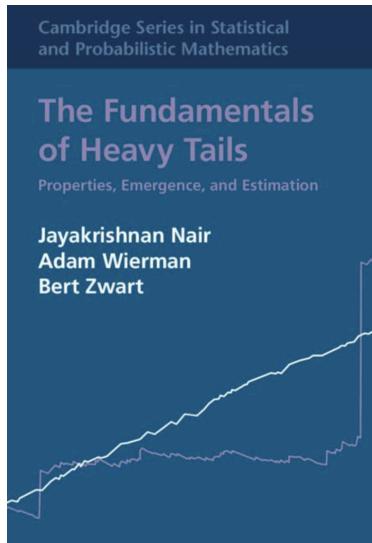


D. Williams, Probability with Martingales ([softcover/ebook](#))

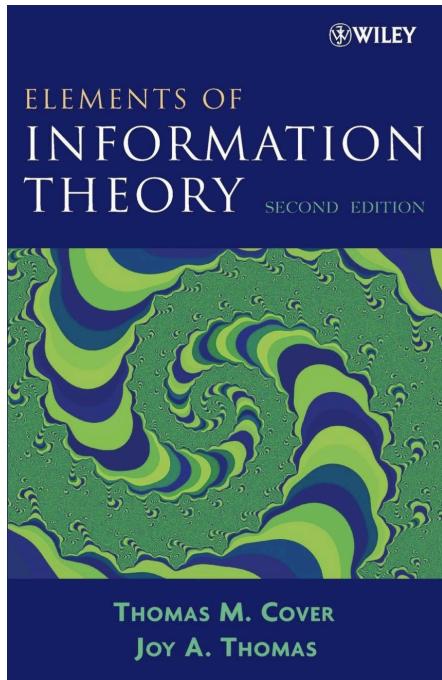
*Breezy reading, the shortest book to quickly learn measure-theoretic probability. It's funny, it's enlightening. It changed my life because it made me fall in love with probability.*



R. Vershynin, High-Dimensional Probability ([hardcover/ebook](#), [free softcopy](#))  
*Covers concepts that are relevant to machine learning, PAC bounds, concentration results, geometry of convex bodies/norms in high dimensions. Very well-written and friendly.*



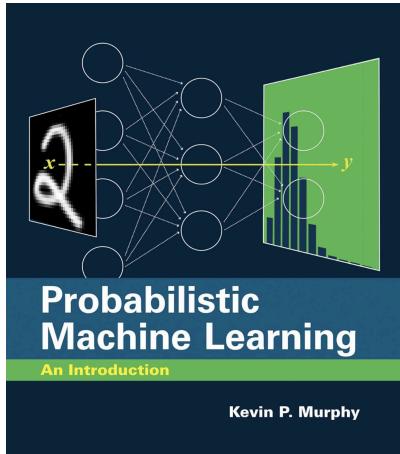
J. Nair, A. Wierman, B. Zwart, The Fundamentals of Heavy ([hardcover/kindle](#))  
*This is the best short introduction to heavy-tailed phenomena. Relevant to finance people.*



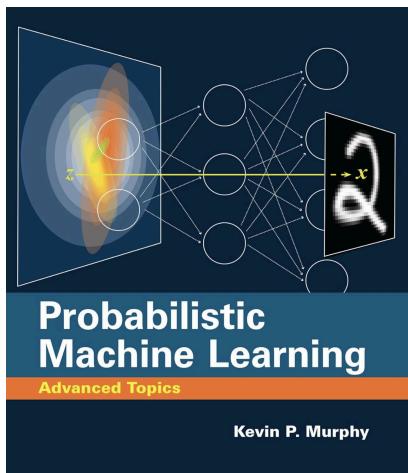
T. Cover, J. Thomas, Elements of Information Theory, 2nd ed.  
[hardcover/paperback/kindle](#)

*A classic of exposition. Not really probability, I know. I use it mostly as a reference.*

# Machine Learning/Statistics



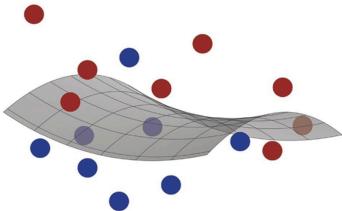
K. P. Murphy, Probabilistic Machine Learning: an Introduction ([Kindle/hardcover](#))



K. P. Murphy, Probabilistic Machine Learning: Advanced Topics ([Kindle/hardcover](#))

*These two books are some of the best theoretical (but with an eye to applications) intro to ML.*

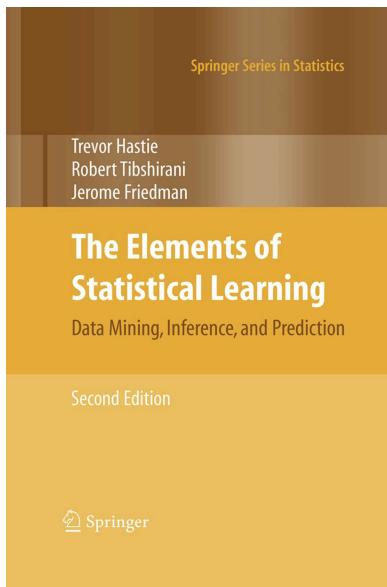
Foundations of  
Machine Learning second edition



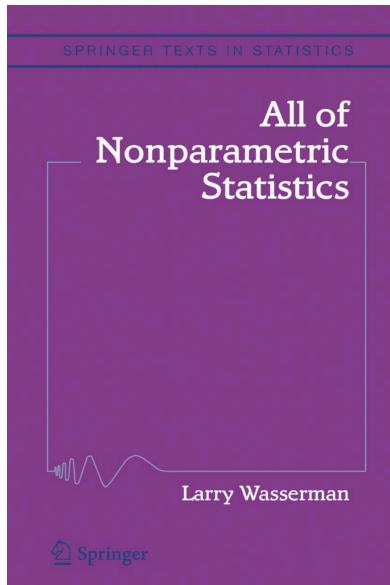
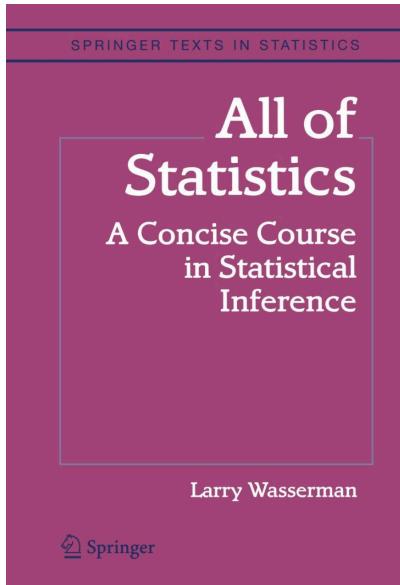
Mehryar Mohri,  
Afshin Rostamizadeh,  
and Ameet Talwalkar

**M. Mohri, A. Rostamizadeh, A. Talwalkar, Foundations of Machine Learning, 2nd ed.**  
[Kindle/hardcover](#)

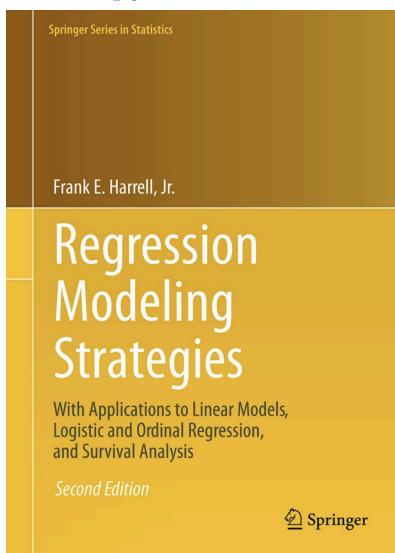
*I love this book because it is short, selective, elegant. It makes you understand the concepts behind ML.*



**T. Hastie, R. Tibshirani, J. Friedman, The Elements of Statistical Learning, 2nd ed** ([hard/softcover/kindle](#), [free copy](#))  
*Dated, with a poor treatment of neural networks, and questionable ideas about model selection. But unsupervised learning, PCA, ensemble methods are still masterfully explained.*

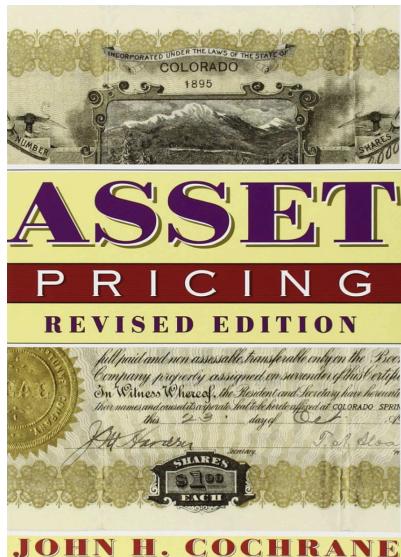


L. Wasserman, All of Statistics ([hardcopy/kindle](#)), All Nonparametric Statistics ([hardcopy/kindle](#))



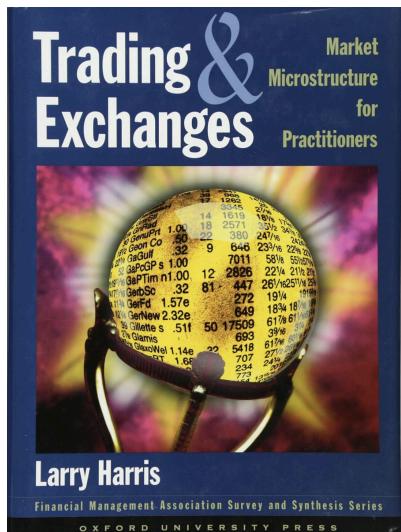
F. E. Harrell Jr., Regression Modeling Strategies ([hard/softcover/etextbook](#))

# Finance



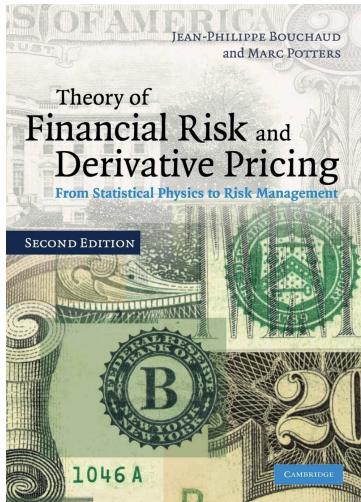
J. Cochrane, Asset Pricing ([hardcover/kindle](#))

*It has a very unique style: Cochrane uses "I" and "you" a lot. He focuses on concepts, and explains factor models, empirical pricing and Generalized Method of Methods like no one else.*



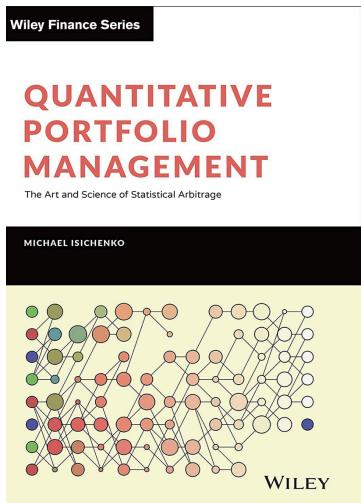
L. Harris, Trading and Exchanges ([hardcover/paperback/kindle](#))

*Overlong but essential treatment of the institutional details of market structure. A bit dated.*



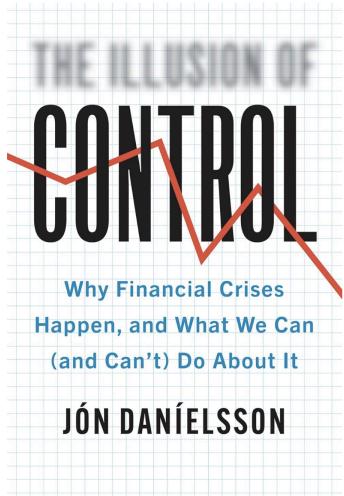
J.P. Bouchaud, M. Potters, Theory of Financial Risk and Derivative Pricing ([hard](#)/[soft](#)/[kindle](#))

*Another old book, written when Bouchaud was relatively new to the game. It's full of ideas and has aged well*

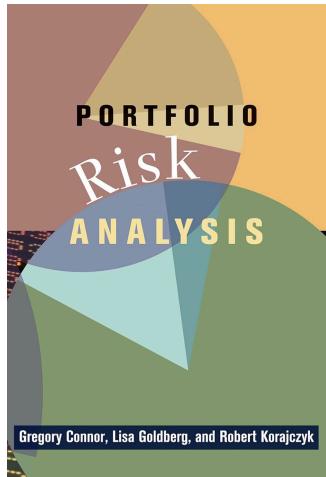


M. Isichenko, Quantitative Portfolio Management ([hard](#)/[kindle](#))

*This is the best available book on quantitative investing written by an actual practitioner. It's scattershot and in parts generic, but still worth reading.*



J. Danielsson, The Illusion of Control ([various formats](#))  
*The best (most intelligent, well written) book on broad, macro-level risk management available. Very accessible, just a bit too long.*



G. Connor, L. R. Goldberg, R. A. Korajczyk, Portfolio Risk Analysis ([hardcover/etextbook](#))  
*Connor worked as a consultant to Barra, and Goldberg worked at Barra for many years. This book is clear and a good introduction to factor models*