# Introduction to statistical/machine learning and R programming

Applied Machine Learning for Economics and Finance

Jonas Striaukas



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#### Course details

#### Basic info:

My email: jost.fi@cbs.dk or jonas.striaukas@gmail.com

Lecture time: TBA

Auditorium: TBA

Office hours: TBA

Course website: https://jstriaukas.github.io/teaching ☐

#### Exam:

Structure: TBA

When: TBA

#### What I expect from you:

▶ Understand the concepts we learn in the class. In particular derivations of some simple theoretical results as well as full understanding of more complex theory.

▶ Be creative, active during class presentations and work hard! And try not to miss classes...

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Machine learning, computing, etc.

"The purpose of computing is insight, not numbers."

Richard Hamming

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#### Topics of the course

- Introduction to statistical/machine learning and R programming
  - ► Brief introduction on statistical learning, data, some definitions. Basic programming in R statistical software.
- Predictive linear regression
  - Revisit linear regression, its basic properties. Estimation, assessment of the parameter estimates. Discussion on best linear predictor. Construction of predictions and prediction confidence intervals.
- Multiple linear regression and regularization
  - Estimation.
- Loss function, classification. Logistic and quantile regression
  - ▶ BIAH BIAH
- Guest lecture
  - ▶ BLAH BLAH

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### Topics of the course

- Principal component analysis and factor models
  - $\blacktriangleright$
- Time series models
  - $\blacktriangleright$
- · Resampling methods for ML
  - ▶
- Optimization methods for ML
  - •
- Introduction to advanced topics in machine learning
  - ▶

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# Introduction to statistical/machine learning and R programming Big data

Nowadays, Big Data are ubiquitous: from the internet, biology and medicine to government, business, economics, finance, ...

#### Some quotes:

- "There were 5 exabytes of information created between the dawn of civilization through 2003, but that much information is now created every 2 days", according to Eric Schmidt, the CEO of Google,in 2010.
- "Big data is not about the data", according to Gary King of Harvard University.

Do we need ML or even AI to understand economics and/or finance data?

▶ Yes! ML is not that different from classical statistics/econometrics...

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# Introduction to statistical/machine learning and R programming

Big data - examples

Big data examples in economics and finance:

- ▶ high-frequency financial assets data (e.g., stocks, bonds, fx, derivatives, ...);
- ▶ large panels of economic data (e.g., 131 macroeconomics time series FRED MD database with monthly updates, McCracken and Ng (2016));
- ▶ spatial data (e.g., state-level data in the US, Euro area data);
- ▶ text-based data (e.g., newspaper articles, GDELT project; EC news data);

**...** .

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#### Notation

Let  $x_{i,j}$  represent the value of the  $j^{\text{th}}$  variable for the  $i^{\text{th}}$  observation, where  $i = \{1, 2, ..., n\} \triangleq [n]$  and  $j = \{1, 2, ..., p\} \triangleq [p]$ .

We let **X** (typically called the *design* matrix) denote an  $n \times p$  matrix whose (i,j)th element is  $x_{i,j}$ . That is,

$$\mathbf{X} = \begin{pmatrix} x_{11} & x_{12} & x_{13} & \dots & x_{1p} \\ x_{21} & x_{22} & x_{23} & \dots & x_{2p} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ x_{n1} & x_{n2} & x_{n3} & \dots & x_{np} \end{pmatrix}$$

At times we will instead be interested in the rows/columns of

Rows: 
$$x_i = \begin{pmatrix} x_{i1} \\ x_{i2} \\ \vdots \\ x_{ip} \end{pmatrix}$$
 Columns:  $\mathbf{x}_j = \begin{pmatrix} x_{1j} \\ x_{2j} \\ \vdots \\ x_{ni} \end{pmatrix}$  Design matrix:  $\mathbf{X} = (\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_p)$ 

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#### Supervised learning

Suppose that we observe a quantitative response  $\mathbf{Y}$  and p different predictors,

$$\boldsymbol{x}_1,\boldsymbol{x}_2,\dots,\boldsymbol{x}_p.$$

Assume that there is some relationship between **Y** and  $\mathbf{X} = (\mathbf{x}_i)_{i \in [p]}$ , which can be written in the very general form:

$$\mathbf{Y} = f(\mathbf{X}) + \mathbf{e}$$

#### Questions:

- ▶ What is f? Can we infer/estimate it from the data?
- How assumptions on f after the estimation and/or prediction?
- ► Can we generalize f such that the prediction of  $\mathbf{Y}_{new}$  is as accurate as possible, given that we have  $\mathbf{X}_{new}$ ?
- ► Assumptions on **e**, **X**, **Y**. Binary outcomes.
- ► Etc ...

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Supervised learning

Suppose that we observe a quantitative response  $\mathbf{Y} = (y_1, \dots, y_n)^{\top}$  and p different predictors,  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_p$ .

Assume that there is some relationship between **Y** and  $\mathbf{X} = (\mathbf{x}_i)_{i \in [p]}$ , which can be written in the very general form:

$$\mathbf{Y} = f(\mathbf{X}) + \mathbf{e}$$

Learning f(.) from the data, i.e., the function that links **Y** and **X** is called *supervised learning*.

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Supervised learning — data example

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Unsupervised learning

Suppose that we observe a matrix of data points, i.e.,  $\mathbf{X} = (\mathbf{x}_i)_{i \in [p]}$ . An example could be an image.

#### Questions:

- ▶ Suppose *p* is large. Can we summarize the data in compact way?
- Can we learn patterns in X?
- ➤ Can we summarize the data in compact way and use this compact information in a predictive model?

Learning patterns in **X** is called *unsupervised learning*.

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Unsupervised learning — data example

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MCCRACKEN, M. W., AND S. NG (2016): "FRED-MD: A monthly database for macroeconomic research," Journal of Business & Economic Statistics, 34(4), 574–589.