Introduction to learning, multiple and nonparametric regression

Machine Learning

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Course details

Basic info:

- **My email**: js.fi@cbs.dk or jonas.striaukas@gmail.com
- **Lecture time**: TBA
- **Auditorium**: TBA
- **Office hours**: TBA
- **Course website**: https://jstriaukas.github.io/ml_course

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...

Machine learning, computing, etc.

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

Richard Hamming

Topics

- Introduction to learning, multiple and nonparametric regression
  - BLAH BLAH
- High-dimensional linear regression
  - BLAH BLAH
- High-dimensional regression properties and generalized linear models (GAMs)
  - BLAH BLAH
- Prediction, loss functions and M-estimators
  - BLAH BLAH
- Introduction to deep learning
  - BLAH BLAH
- Introduction to causal machine learning
  - BLAH BLAH

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 econometrics... "Black-box" deep learning is not that black box after all...

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 US, euro area data);
- text-based data (e.g., newspaper articles, GDELT project; EC news data).

Problems associated with Big data:

- Data are collected from various sources and populations $\implies$ heterogeneity;
- typically large numbers of variables are collected $\implies$ some variables are heavy-tailed, i.e. have high kurtosis which is much higher than the normal distribution;
- incidental endogeneity due to high-dimensionality $\implies$ huge impact on model selection and statistical inference (Fan and Liao, 2014);
- computation/optimization of model parameters $\implies$ convexity so far is a way out to guarantee the stability of solutions;
- noise accumulation and spurious correlation has a large impact on model selection $\implies$ high-dimensional statistics methods.

For curious students: see Fan, Han, and Liu (2014) for an overview of how these features impacts the developments of big data analysis techniques.

Learning, multiple and nonparametric regression

Spurious correlations – examples

Learning, multiple and nonparametric regression

Spurious correlations – some explanation

According to Bickel (2008), the main goals of high dimensional inferences are:

- to construct a method as effective as possible to predict future observations and;
- to gain insight into the relationship between features and responses for scientific purposes, as well as, hopefully, to construct an improved prediction method.

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