# 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

# Learning, multiple and nonparametric regression

## Big data

Nowadays, Big Data are ubiquitous: from the internet, biology and medicine to government, business, economy, 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.
- "*Data are becoming the new raw material of business*", according to Craig Mundie, Senior Advisor to the CEO at Microsoft.
- "*Big data is not about the data*", according to Gary King of Harvard University.

# Learning, multiple and nonparametric regression

## Big data – examples

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 (McCracken and Ng, 2015) with FRED MD database with monthly updates);
- spatial data (e.g., state-level data in US, euro area data);
- text-based data (e.g., newspaper articles, GDELT project; EC news data).

# Learning, multiple and nonparametric regression

## Impact of Big Data & dimensionality

Problems associated with Big data:

- Data are collected from various sources and populations $\Longrightarrow$ heterogeneity;
- typically large numbers of variables are collected $\Longrightarrow$ some variables are heavy-tailed, i.e. have high kurtosis which is much higher than the normal distribution;
- incidental endogeneity due to high-dimensionality $\Longrightarrow$ huge impact on model selection and statistical inference (Fan and Liao, 2014);
- computation/optimization of model parameters $\Longrightarrow$ 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 $\Longrightarrow$ 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

# Learning, multiple and nonparametric regression

## Statistical learning theory

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.