Causality Detection in Health Econometrics Using Big Data

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Digital foot print



https://www.teacherspayteachers.com/Product/What-is-my-digital-footprint-poster-3005468

Big data

What about Big Data?



https://www.google.com.au/search?q=what+is+big+data&source=Inms&tbm=isch&sa=X&ved=0ahUKEwimrrP575XcA hXFHpQKHZqqBCcQ_AUICyqC&biw=1920&bih=901#imgrc=LdhqI29DA0cq-M

Sources of big data

Sources	Some examples
Administrative data	 Electronic medical records Insurance records Tax records
Commercial transactions	 Bank transactions (inter-bank as well as personal) Credit card transactions Supermarket purchases Online purchases
Sensors and tracking devices	 Road and traffic sensors Climate sensors Equipment and infrastructure sensors Mobile phones Satellite/GPS devices
Online activities/social media	 Online search activities Online page views Blogs and posts and other authored and unauthored online content and social media activities Audio/images/videos

Source: ITU, adapted from UNSC (2013).

Sources of big data in health



https://tcf.org/content/report/strengthening-protection-patient-medical-data/?session=1

Conventional Econometrics and Big Data

- Conventional econometric techniques (regression) often work well, but there are issues unique to big datasets that may require different tools (Varian, 2014).
- Why?
 - Size of the data and p value
 - More potential predictors (variable selection)
 - Non linear relationship

Machine Learning

- Divide the data into training, testing and validation.
- Machine learning finds particular function (s) that provide a good prediction of y as a function of x.
- Out of sample performance (overfitting problem) minimizing sum of square residuals.
- Historically, machine learning deals cross-section data.
- The data may be "fat," (lots of predictors relative to the number of observations) or "tall" (lots of observations relative to the number of predictors).

Big Data Variety



Structured Data

- Automated Multiple-Pass Method (AMPM), developed by the Agricultural Research Service of the United States Department of Agriculture (USDA)
- ABS follow AMPM approach for Australian National Nutrition and Physical Activity Survey.
- Face to face and telephone interview

Bland and Altman plot for saturated fat intake for the two methods



Bland and Altman plot for sugar intake for the two methods



Supervised Learning: Neural Networks

- Most work in machine learning has involved cross-section data (Varian, 2014, 2016)
- The MLP Neural networks are composed of layers of elementary units, called neurones, linked to one another by weighted connections.





Effect size of the sources of disagreement of two measurements (fat and sugar intake)



Source of disagreement of two measurements for over reporting sugar value



Source of disagreement of two measurements for over reporting fat value



Unstructured data

- When big data is observational, generated from uncontrolled experiments/environment and often nonrandom.
- Often less expensive to collect.
- No statistical sampling (example: electronic health records (EHRs)).

Unstructured data: unsupervised and supervised learning (cross section)

- Unsupervised learning can be used for clustering, grouping, autonomous post stratification.
- Supervised learning could be used for testing theories by controlling the learning process.

Data extraction with wavelet analysis (big data time series)

Coherence



Data visualization



Neuronal Networks





Lüchters, G. 2017

G.L. Causality & Statistics Part 1 & 2

Neuronal Networks



Lüchters, G. 2017

G.L. Causality & Statistics Part 1 & 2

Neuronal Networks



Lüchters, G. 2017

G.L. Causality & Statistics Part 1 & 2

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