Big Data , GeoAnalytics and Systems Analysis

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ESCOLA DE ADMINISTRAÇÃ DE EMPRESAS DE SÃO PAULO

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Apresentação



Professor of the TDS (Technology and Data Science) Department at **FGV EAESP**. He teaches disciplines related to GeoAnalytics, Spatial Statistics and Data Science. Bachelor in Computer Science from IME-USP and Master and PhD in Business Administration from FGV EAESP. Deputy Coordinator of the Graduate Course in Business Administration at FGV EAESP. He is a visiting researcher at the Spatial Information Research Center at the University of Otago, New Zealand.

He is CKO (Chief Knowledge Officer) of startup Meia Bandeirada and founding partner of GisBI, a study and promotion group for Geotechnologies and Business Intelligence.



He has been working in the GIS, Geomarketing and Business Intelligence markets since 1994. He was technology manager of the GIS project and Market Planning project and responsible for the development of Marketing Strategies and Market Research Planning at AES Eletropaulo for 13 years. Participates as a speaker in various Geomarketing, BI, GIS and Utilities events. He is a columnist for GV Executivo and InfoGEO magazines, and a consultant in Spatial Statistics and Predictive Models for Credit and Real Estate. Member of the **Fundação SEADE** Board of Trustees.

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Greasing the Wheels Capital expenditures for 2018 \$0 billion \$5 \$10 \$15 \$20 Google* Amazon** **Royal Dutch Shell** Exxon Mobil BP Microsoft Facebook Chevron

*excludes Other Bets ** includes capital leases Source: company data, FactSet



uhuhu cisco

Data is the new oil.

\$25

We see in data the same transformative, wealth-creating power that 19th-century visionaries once sensed in the crude black ooze trapped underground.

If "crude" data can be extracted, refined, and piped to where it can impact decisions in real time, its value will soar. And if data can be properly shared across an entire ecosystem and made accessible in the places where analytics are most useful, then it will become a true game changer, altering the way we live, work, learn, and play.





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The

Economist

Regulating the internet giants The world's most valuable resource is no longer oil, but data

The data economy demands a new approach to antitrust rules



R Print edition | Leaders > May 6th 2017

A NEW commodity spawns a lucrative, fast-growing industry, prompting antitrust regulators to step in to restrain those who control its flow. A century ago, the resource in question was oil. Now similar concerns are being raised by the giants that deal in data, the oil of the digital era. These

DATA & AI LANDSCAPE 2019

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"Big Data" are <u>data and processes</u> whose scale, distribution, diversity and / or creation speed requires the use of new storage and analysis technologies to allow the capture of the value inserted in them (FRANCISCO, 2016, adapted from EMC, 2013)

1. Data Volume

Billions of lines x billions of columns
Increase of 44 times from 2009 to 2020 (0,9ZB to 35ZB)

2. Processing Complexity

Data structures are constantly changingNeed to analyze such data in real time

3. Data Structure

- Large variety (80-90% unstructured) to be analyzed
- These characteristics make it necessary to use parallel and parallel mass computing (MPP) systems





Source: EMC (2013)









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Big Picture

Monitoramento e controle



Dashboards





:::gisbi

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Challenges for Systems Analysis in the Big Data Era

- Big Data, Data Science, Analytics ...
- Challenge in the integration of analytical techniques
 - AI, Neuroscience applied to marketing and business, spatial statistics, behavioral models
 - Models to handle stakeholders' relationships
- Computational Challenge
- Challenge in the adoption of AI and Data Science by companies and public organizations
 - Adoption of forceps, apart from core management
 - Analytical Sandbox vs. IT Policy
- Cultural challenge new skills of analytical teams and managers

Use of alternative information (nonstructured "Big" data) in the generation of indicators

- Open Data API or Web Scrapping
- Sectoral Reports, Management Reports, Integrated Reports - Web Scrapping
- Interpretation of Images
- Data Enrichment by Geo-Analysis and Spatial Statistics



Serious Implications for Scientific Research in Applied Social Sciences

100

Classic (Inferential) Statistics	Machine Learning	
Strong hypotheses	Flexible approach	
Theoretical justification	Empirical Efficacy	
Exact solution	Approximate solution	
Explanation/Interpretation	Prediction	

Source: OLLION, 2018



Framework: Adaptive Business Intelligence and Big Data



Source: LETOUZÉ, 2017; MICHALEWICZ et al., 2006 ; GisBI, 2015

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Challenges

Inferring socio-demographic indicators



Scientific Prize and Ethics Mention: Construction of socio-demographic indicators with digital breadcrumbs

F. Bruckschen (1), T. Schmid (2), T. Zbiranski (1)

We show that socio-demographic indicators such as population, age, literacy, poverty, religion, ethnicity, electricity supply and others can be estimated in unprecedented detail and virtually ad-hoc using antennato antenna traffic data only. We offer a uniform approach that can be easily extended to other variables. Results are tested for spatio-temporal robustness and visualized as heat maps.

(1) Humboldt Universität Berlin, Germany - (2) Freie Universität Berlin, Germany



Source: LETOUZÉ, 2017

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OBJ: Analyze the relationship between household income and electricity consumption

Create an income indicator based on electricity

Prediction Model:

Electricity Consumption Household Income

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Traditional OLS Regression:

$$\hat{y} = \beta_0 + \beta_1 x$$
 R² = 86,6%

SAR (*Spatial Auto-Regressive*):

$$\hat{y} = \beta_0 + \beta_1 x + \rho W y$$
 R² = 94,4%

GWR (Geographically Weighted Regression):

$$\hat{y}_i = \beta_0(u_i, v_i) + \beta_1(u_i, v_i) x_i$$
 R² = 96,8%



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Microcredit Score and Socio-Economic Indicators based on Electric Energy





Source: FRANCISCO, 2011

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 Development of an Indicator of Propensity to Energy Commercial Losses using Geospatial Statistical Techniques and Socio-Economic Data: the Case of AES Eletropaulo

FRANCISCO, E., FAGUNDES, E., PONCHIO, M., ZAMBALDI, F. - EnANPAD 2009

• Use of geospatial techniques for the design and operation of a fraudprone indicator, and analysis (through GWR) of the association between this indicator and customer satisfaction



Source: FRANCISCO et al., 2009



Managerial Implications

Socio-Economic Indicators based on Electricity Consumption

- Should be published widely by the electricity distributors
- Useful for strategy formulation and decision support
- Support for Customer Relationship Characterization and Management
- Support for Public Policies and Systems Analyses





Concentric circles (progressive radius of 125 m)



___ Quadricules (1 square kilometer)



Household Income Density

EGV



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Health and Longevity Research and Study Platform



https://youtu.be/1ET4glwLAe0





Thank you! ¡Gracias! Obrigado!

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