

A Conversation with ... Ganesh Mani

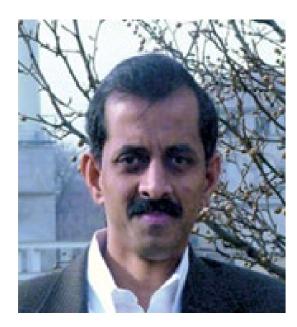
Managing the Data Supply Chain

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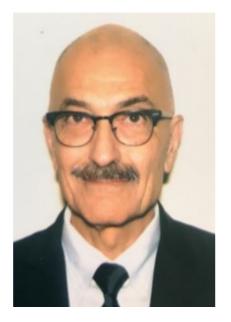
Agenda



Ganesh Mani Adj. Faculty Carnegie Mellon



Katherine Wilkens Sr. Curriculum Advisor FDPI



Mehrzad Mahdavi **Executive Director FDPI**





Mirjam Dekker PM **FDPI**

Webinar Series



sup-ply chain /sə'plī CHān/

the sequence of processes involve in the production and distribution of a commodity. Motivated by the above definition and noting that data is not necessarily a commodity (even though the comparison to oil has been made by many, as data

is thought of as fueling the modern information economy), we can offer up the following broad summary:

Data Supply Chain Management is the selection, collection, organization, streamlining and flow-control of data – including any pre-processing, repair and normalization steps - to make it usable, guided by domain knowledge, for the subsequent downstream process. Typically, the next step involves analysis

via traditional statistical or contemporary machine learning tools. The end goal of the exercise is to generate insights that can imply customer value, inform revenue or pricing metrics, optimize costs and help gain a competitive advantage in the marketplace.

Why is the Data Supply Chain important? The outcome of analytical engines in general and machine learning in particular can be highly dependent on the quality and other attributes of the ingested data. The exponential increase in the quantity and variety of data provides opportunities as well as challenges with regards to the sourcing, selection, and preparation (cleansing, aggregating, and normalizing) of the data. Another dimension of the data supply chain is its integrity as well as compliance with many evolving regulations regarding privacy, security and ethical use. Trustworthy data is key to trusted outcomes.

The moniker 'Big Data' has been popularized over the last decade. In the following section, we describe its nuances, especially as it relates to performing due diligence on the data supply chain.

Key data dimensions:

These dimensions are considered standard across many domains.

• Volume: The amount of data collected and stored via records, transactions, tables, files, etc.

· Velocity: The speed at which data is sent or received. Data can be streamed; or, received in batch mode, real-time or near-real-time.

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Managing the Data Supply Chain Ganesh Mani and Mehrzad Mahdavi

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• Variety: Data often comes from a variety of sources and in various formats be it structured (such as SQL tables or CSV / Excel numeric files), semi-structured (such as documents in XML, HTML or JSON format), or unstructured such as a blog post or video).

g data, especially the recently-dubbe Iternative variety, in investment anagement can be further classified ased on how it was generated: produced y individuals (such as social media posts erated through business processes uch as e-commerce or credit card transactions data), or generated by sensors (e.g., via radar or satellite imagery, sensors installed in industrial facilities or on equipment, drones). Datasets generated by individuals are often in the structured textual format and only require natural language ocessing. Sensor-generated data can be produced in both structured and ictured formats. Many business generated datasets, such as credit card transactions, and company 'exhaust' data may need to comport with existing and emerging legal and privacy consideratio In addition to the source of data, collect niques can be passive or active, including proactively seeking additional lements. Additional data elements n ving a cost-benefit analysis - e.g., when it is estimated that

spending \$X on additional data collect and processing may result in a benefit exceeding \$2X! Other attributes of data (besides the preceding classification) may be importa from an intended use standpoint. Investment professionals may want to map a dataset to an asset class or investment style, after considering its quality, technical specifications and alpha potential. The following related issues come to mind and will be discussed:

 Asset classes. Most ready-available datasets are still focused on equities and modities. There is relatively little alternative data on interest rates and urrencies, making such data sets more valuable to select investors

· Investment styles. Much of the data is sector- and stock-specific, germane to equity long-short investors. There is also a significant amount of data relevant

macro investors (such as consumer credit, China economic activity and shipping volumes). Certain alternative datasets can be used to substitute traditional metrics of market risk, and

FDP Leadership Team Dr. Mahdavi, Ph.D., Executive Director FDP Institute Dr. Mahdavi is a technology entre with focus on eakthrough digita

ansformation for the intech and the energy sectors. He is a recognized expert and frequent keynote speaker on application (Al, IoT, and Cloud computing in financial sector and industries

Dr. Mahdavi managed major global businesses in the energy sector. He is currently the executive director of the Financial Data Professionals Institute (FDPI), a non-profit organization founded with CAIA. Mehrzad holds a PhD in Nuclear Science and Technology from the University of Michigan a Bachelor of Science in Electrical and Electronics Engineering from the University of Illinoi: at Urbana-Champaign.

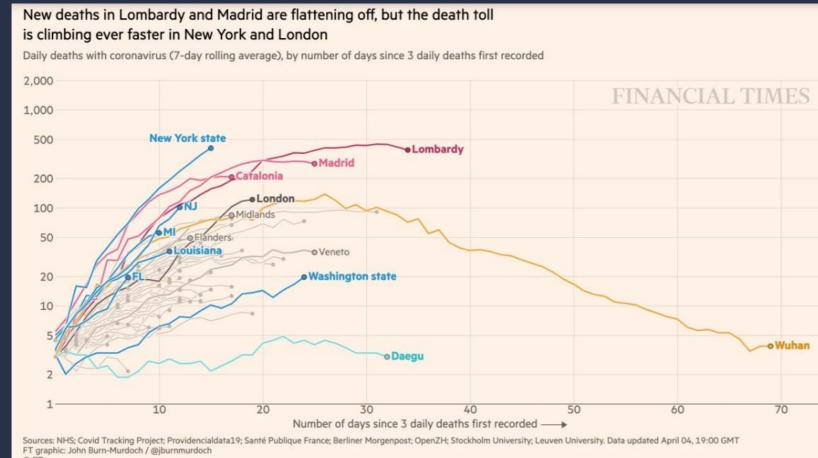


ought leader in th management and areas of investmen Al / FinTech. He has been a pioneer in applying innovative techniques to mul asset classes and has worked with large asset management firms incl. hedge fund after having sold one of the earliest Al/ML-based investment management boutiques into SSgA (in the late nineties nucleating the Advanced Research Center there.

Ganesh has been featured on ABC Nightline and in a Barron's cover story titled "New Brains on the Block", Mr. Man has an MBA in Finance and a PhD in Artificial Intelligence from the University of Wisconsin-Madison, as well as an undergraduate degree in Computer Science from the Indian Institute of Technology, Bombay. Ganesh is a charter member of TiE (www.TiE.org), was an early member of the Chicago Quantitative alliance and is on the advisory board of the Journal of Financial Data Science.

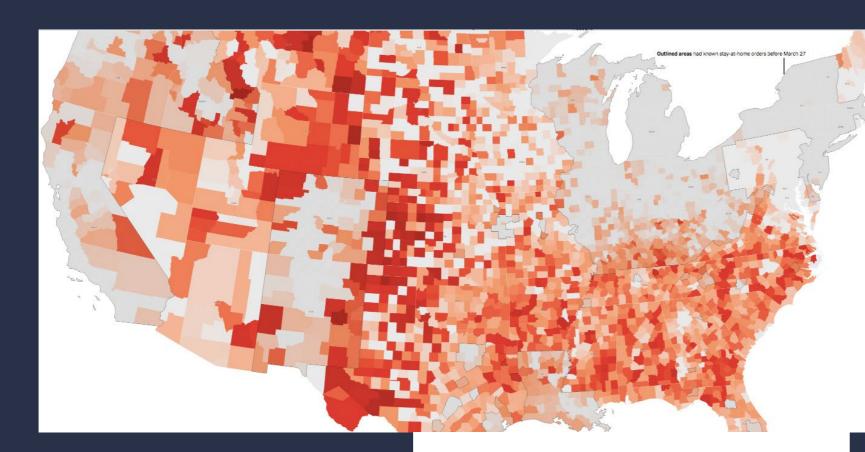


By Region View (as of late April 4th)



Source: Financial Times

Social Distancing Compliance

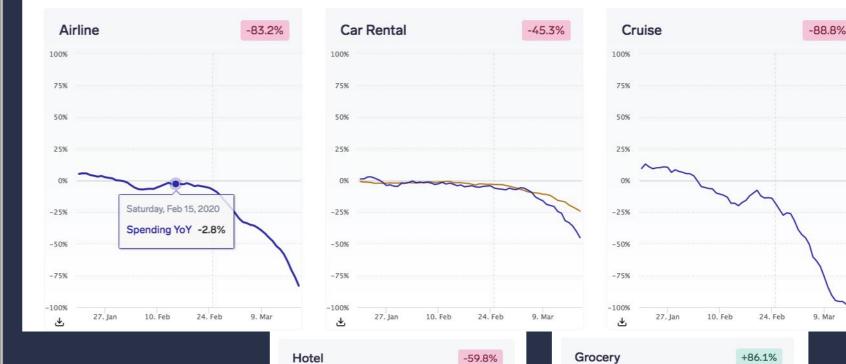


Where people were still traveling

Percent change in average travel for the week of March 23, compared with travel before the coronavirus outbreak.

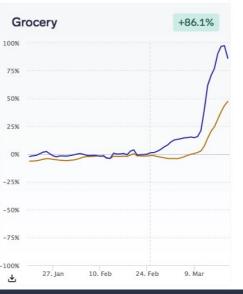
Normal travel

Closer to normal travel ---



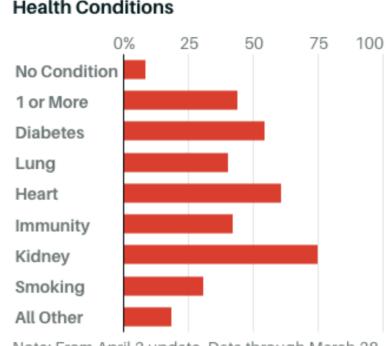
US Demand: Travel vs. Grocery (Late Jan – Early Mar)





Source: Exabel

Health

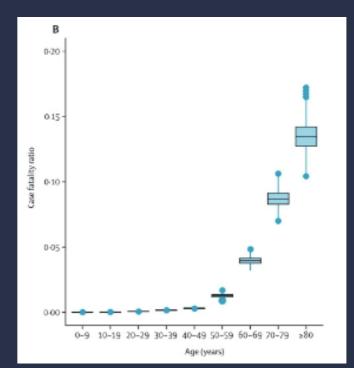


CDC Hospitalization Data By Underlying Health Conditions

Note: From April 3 update. Data through March 38. Study based on 7, 162 cases with completed information.

Source: Centers for Disease Control and Prevention

Age category (binary), years				
	<60	194	30 763	0·631% (0·545– 0·726)
	≥60	829	13 909	5·96% (5·57– 6·37)



Chinese fatality data (by age) Source: (Mar 30, 2020) Lancet, Verity et al.

Source: Barron's







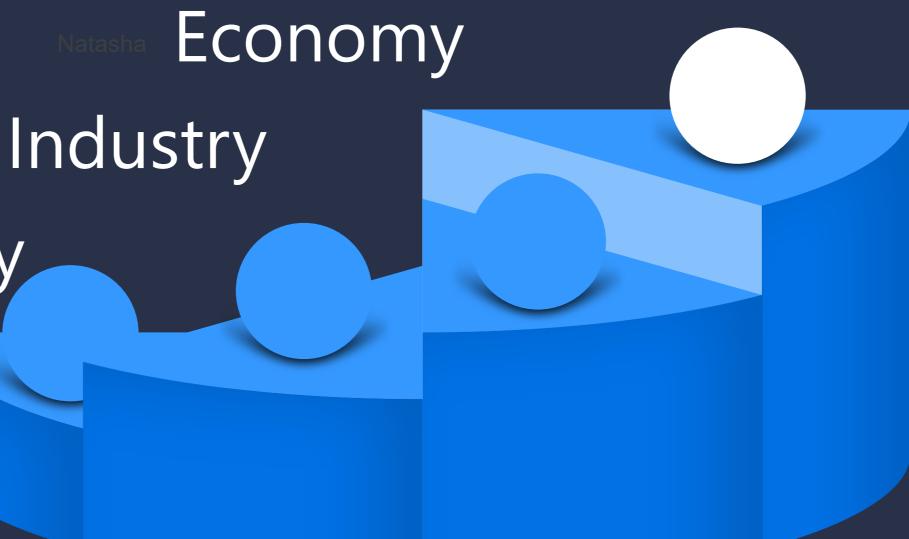
Carnegie Mellon University Ganesh Mani, PhD, MBA Adjunct Faculty

Ganesh has designed and developed many innovative alpha-generation frameworks based on disparate data (both traditional and alternative) and AI/ML techniques. Clients include leading asset management firms (incl. hedge funds) and plan sponsors.

Company

AI / ML: Useful for Analysis as well as for Data scrubbing / repair.

Alternative











Reduce Complexity AI / ML provides a rich toolbox





Insights Actionable for PMs and traders



Rules or other summaries

Interpretability trade-offs

Intermediate Representations

Features (e.g., Temporal patterns, Cross-asset relationships, Snippets w/ affect)



Data Large amounts often w/ noise







Data

Volume, Velocity & Variety Provenance, Vintage & Noise-level Traditional vs. Alternative

Augmented Intelligence

Man with bicycle analogy

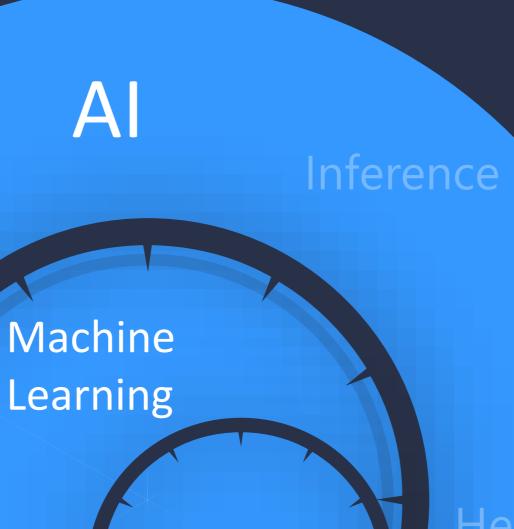
Opinion vs. Fact

AI / ML provides strong opinions or hints

Search

Machine Translation

Planning



Deep Learning

Heuristic Simulation



Diving for insights Making information actionable

Models

Choice should be dictated by data characteristics and goals

Game Theory

Optimization

Beware of overfitting

Careful use of cross-validation and out-ofsample data







Test and Deploy

Validate



Time-series data may pose additional challenges

Drift?

V

Monitor errors and type! Is new data similar to training data?

How often?

Retrain

Many factors (e.g., lookback) go into this decision

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Multiple models / solutions possible

Try to be creative!

Whether you use a simpler technique or a more complex one, try to expose insights.







Q & A



WEBINAR SERIES A Conversation With...

Rick Roche, CAIA Man. Dir. Little Harbor Advisors "Evolution of Machine Learning in Investment Mgmt."

> April 22, 2020 <mark>1 pm EST</mark>



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Kind reminders of upcoming webinars as we go through the Q & A.



In Closing

>New curriculum available at the end of April

Learn more about the FDP Institute at www.fdpinstitute.org



- ➢ Registration for the October 26 − November 8th exam opens May 10th
- > For a recent candidate webinar go to <u>www.fdpinstitute.org/webinars</u>