



Irving Fisher Committee on
Central Bank Statistics

BANK FOR INTERNATIONAL SETTLEMENTS

IFC – Bank Indonesia International Workshop and Seminar on “*Big Data for Central Bank Policies / Building Pathways for Policy Making with Big Data*”

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Promise: measuring from inflation to discrimination¹

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¹ This presentation was prepared for the meeting. The views expressed are those of the author and do not necessarily reflect the views of the BIS, the IFC or the central banks and other institutions represented at the meeting.

PROMISE:

Measuring From Inflation to Discrimination

July 2018

1

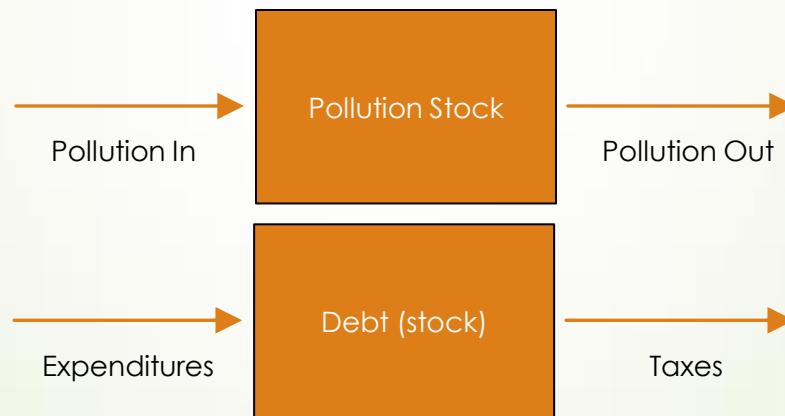
Roberto Rigobon
MIT, NBER, CNStat

Dimensions of Social Wellbeing



Principles

	P	R	O	M	I	S	E
Principles	Balance	Trust	Future Value	BB-NN	Consistency	SP	EE



Values

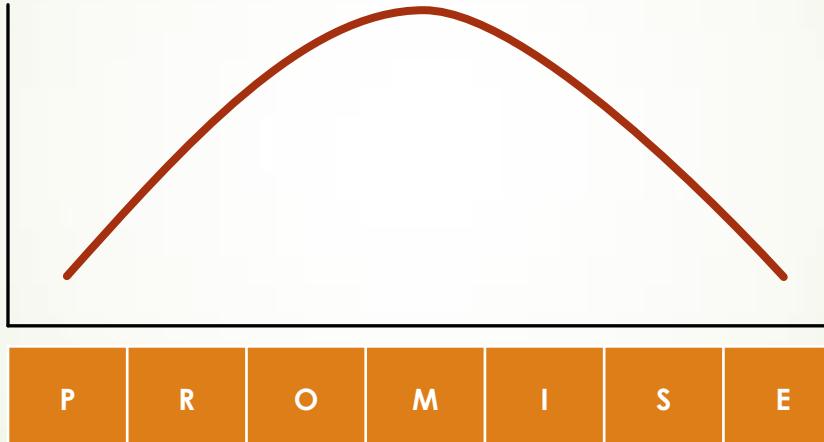
	P	R	O	M	I	S	E
Principles	Balance	Trust	Future Value	BB-NN	Consistency	SP	EE
Values	Weights on each dimension						

- How much do you care about each dimension?

Measurement

Measurement Quality

Measurement Quality

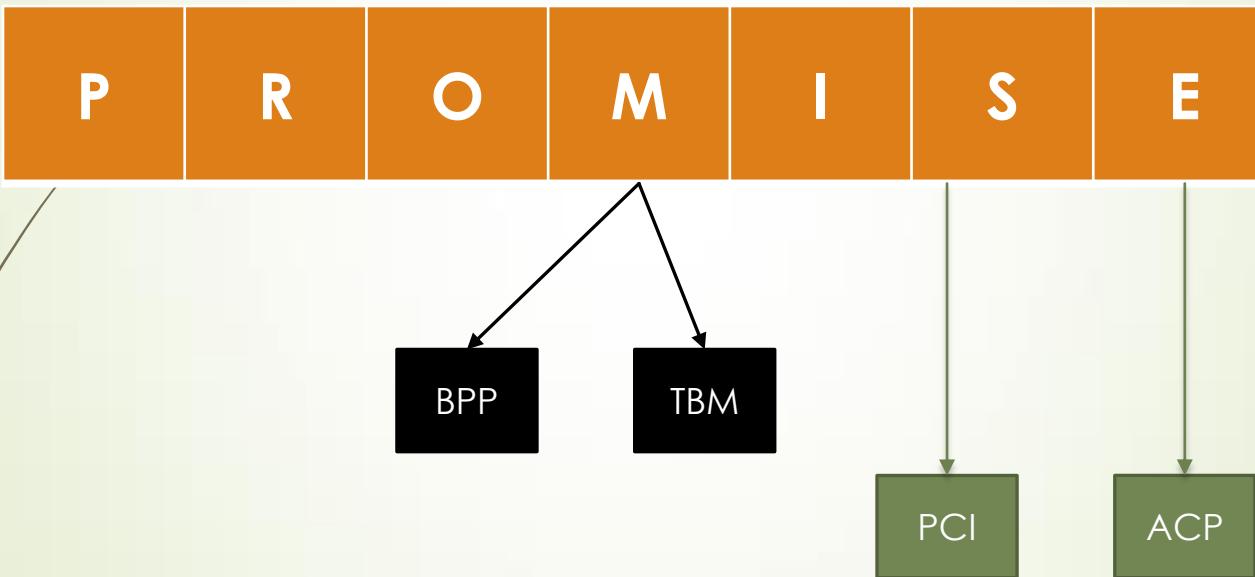


What you measure changes what you do!

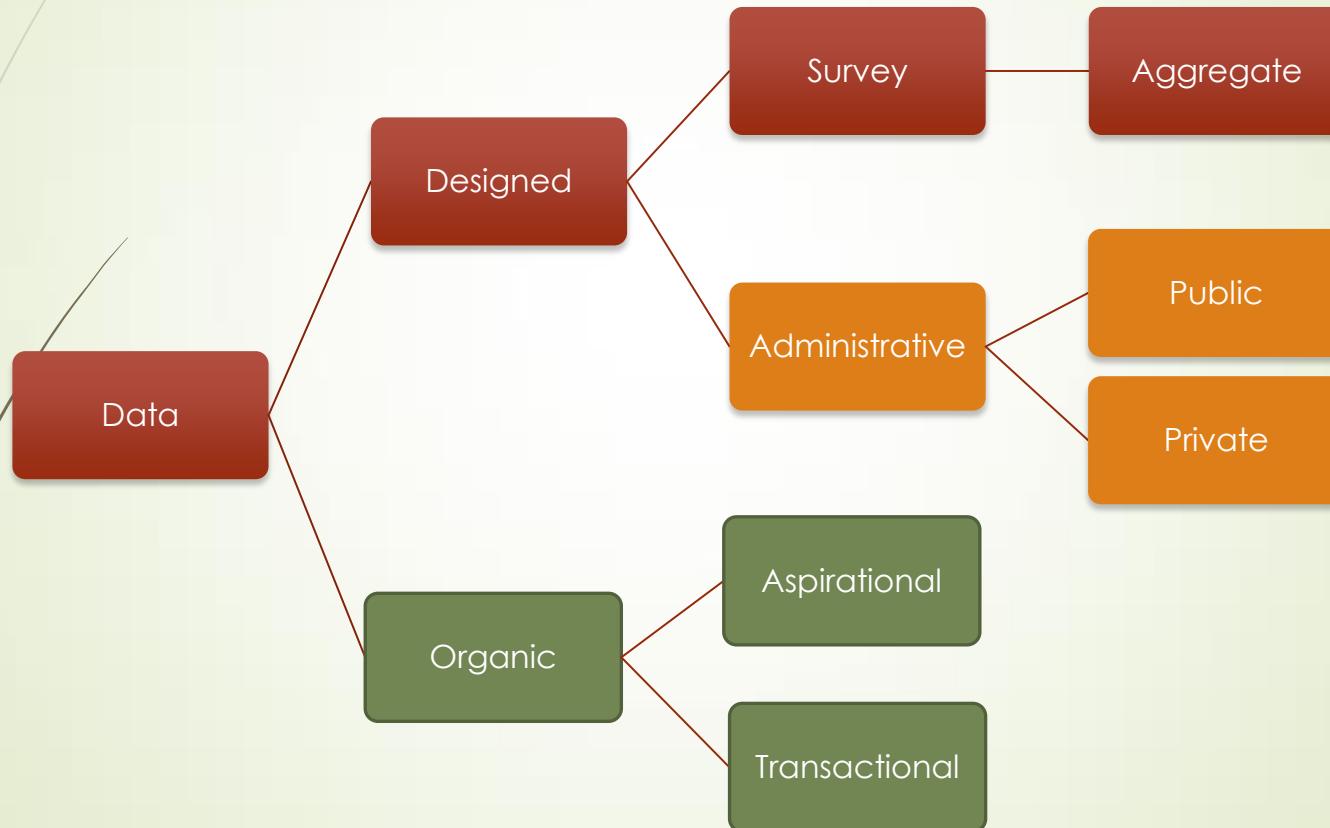
- Incentives aligned with measurement
- Actions aligned with measurement



My Research



Evolution of Data Sources

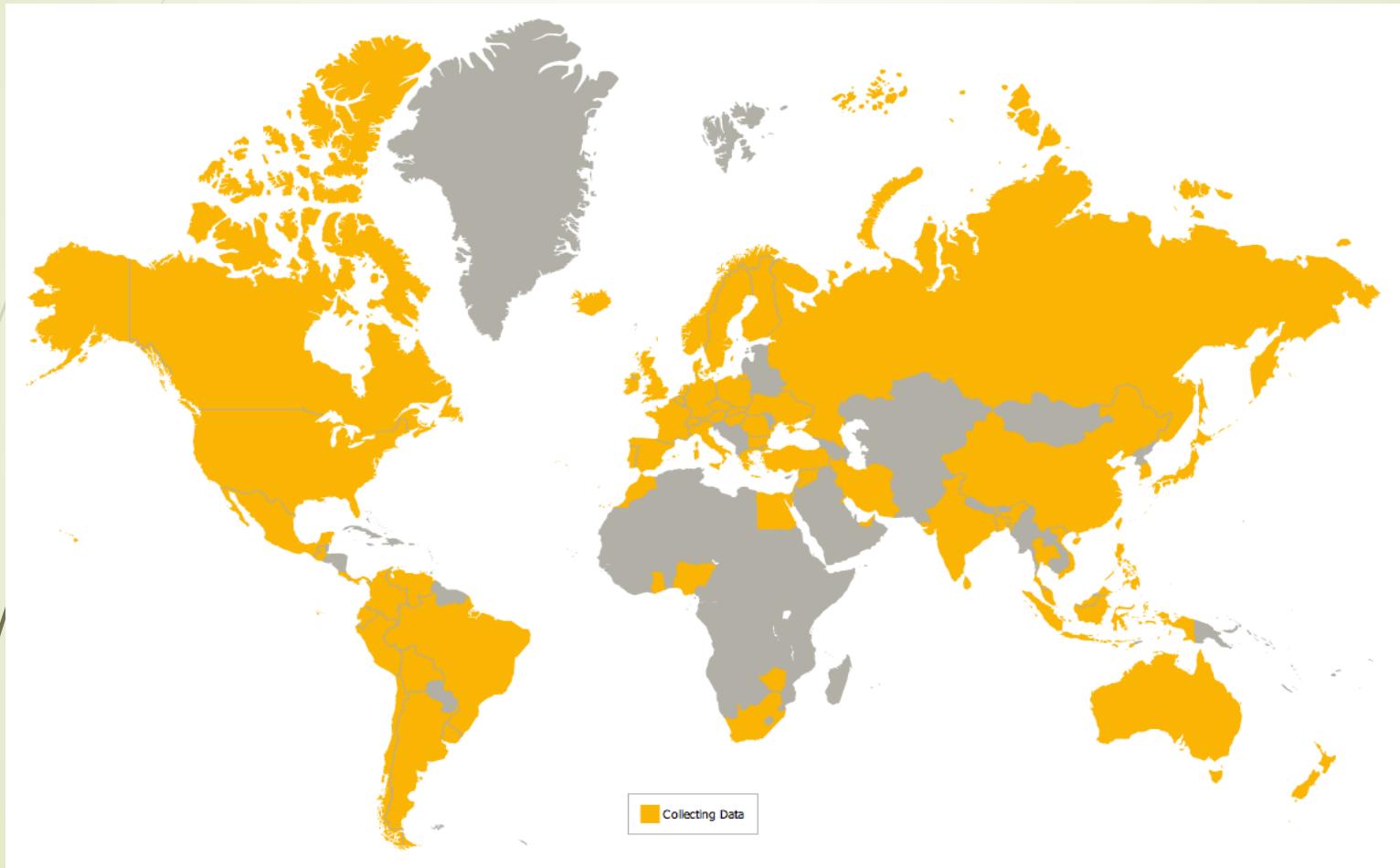


Advantages / Disadvantages

	Designed	Organic
Representative	Yes	No
Sample Selection	Response Rates Deteriorating	Extreme
Intrusive	Extremely Intrusive	Non-intrusive
Cost	Large	Small
Curation	Well-studied	Unclear
Structure	Geography and Socio-Economic	Behavior
Privacy	Well protected	Large Violations of Privacy

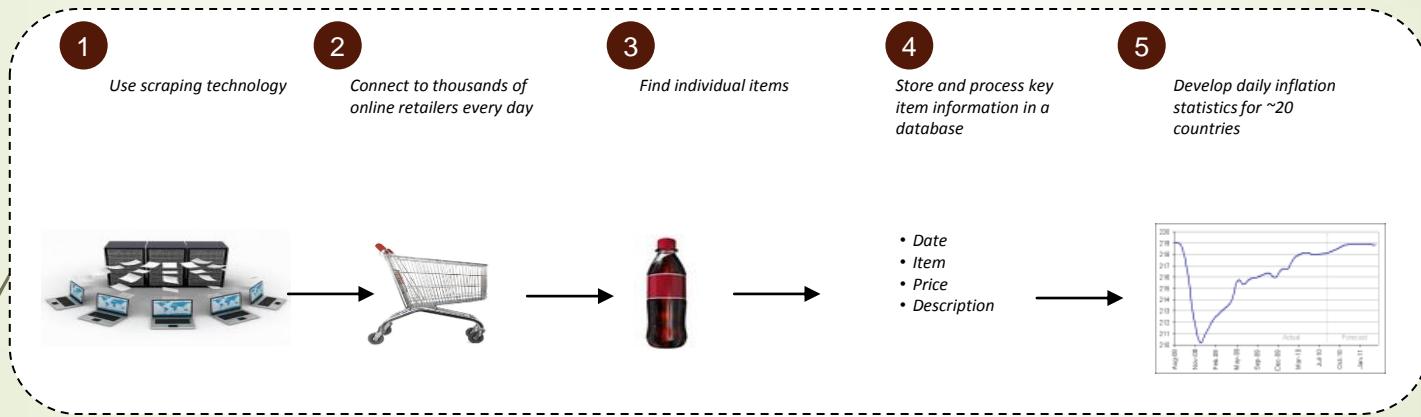
Hybrid approach to macro measurement

BPP: Countries covered



Online Information and Indexes

Our Approach to Daily Inflation Statistics

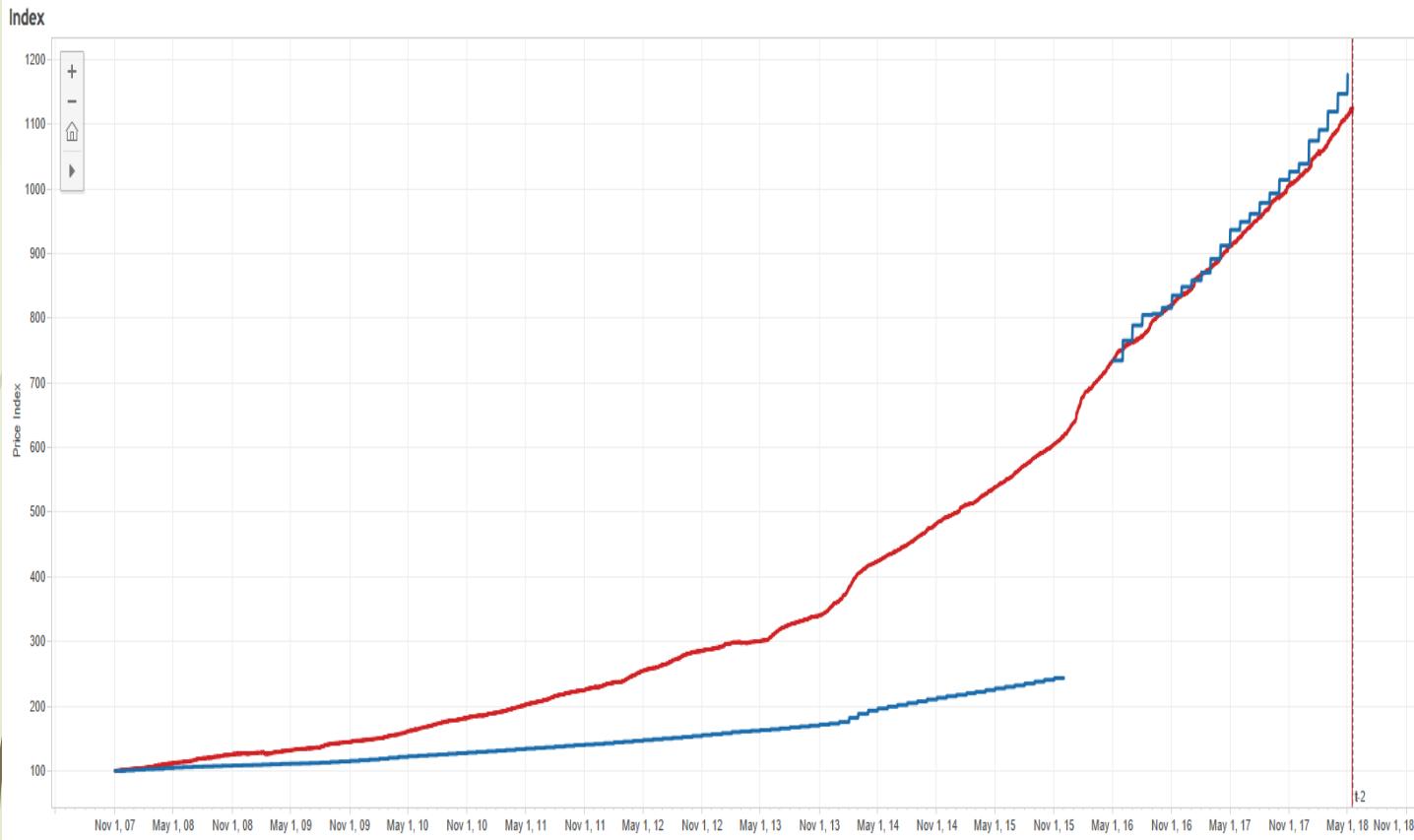


Online versus Offline?

Preference to buy online versus in-store

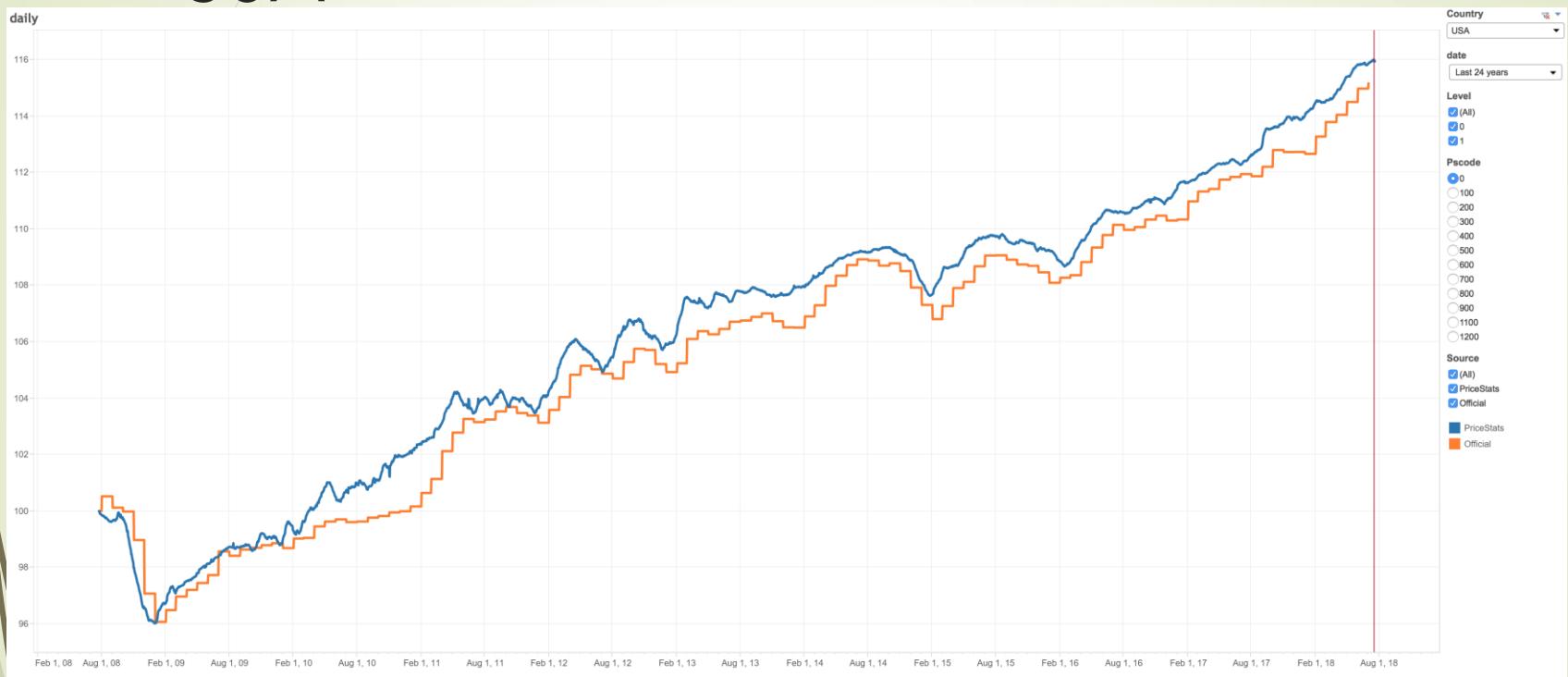
<i>Online</i>		<i>In-store</i>
60%	Books, music, movies & video games	28%
39%	Toys	37%
43%	Consumer electronics & computers	51%
36%	Sports equipment/outdoor	44%
37%	Health & beauty (cosmetics)	47%
40%	Clothing & footwear	51%
32%	Jewelry/watches	49%
33%	Household appliances	56%
30%	DIY/home improvements	52%
30%	Furniture & homeware	59%
23%	Grocery	70%

Argentina



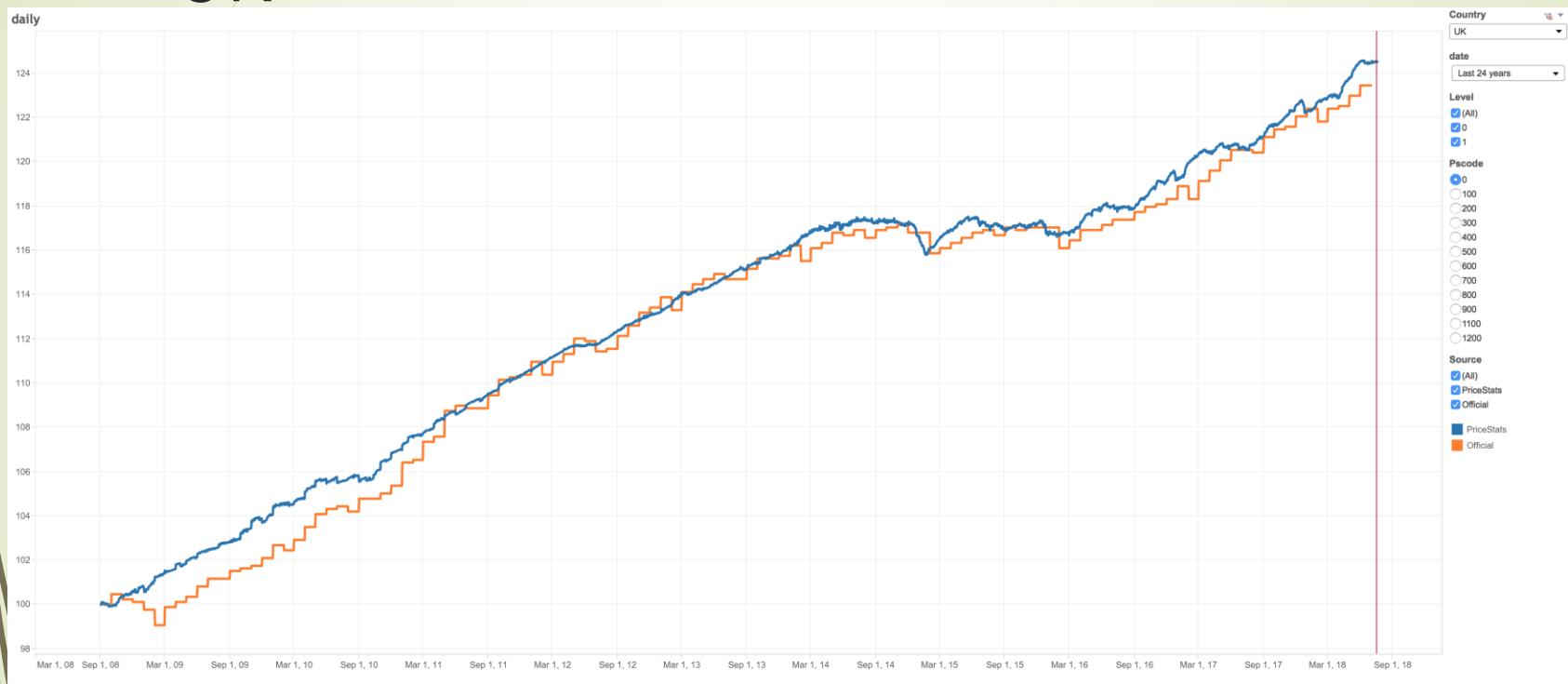
14

USA



15

UK



Inflation y-o-y



USA



UK

Thousands Big Mac's Project



Compare prices for a bottle of Coke across countries

Apply similar approach to hundreds of products on a daily basis

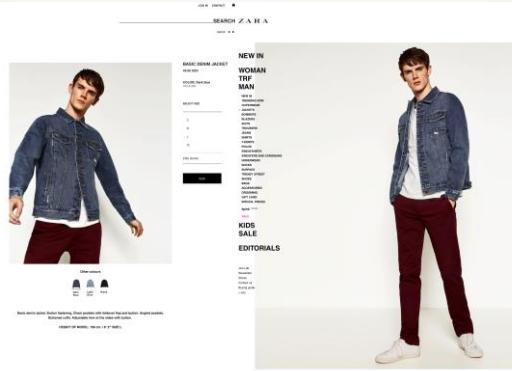
PPP Indices

- Online prices represent an effective tool to measure PPP fluctuations
 - Identical items sold around the world
 - Detailed descriptions to achieve a nearly perfect matching
 - Daily Prices
- PPP indices:
 - More than 300 narrow product categories
 - With thousands individually matched items
 - In food, fuel, and electronics: we are missing clothing, personal care, household products.
 - Cars we will never match

Relative Prices

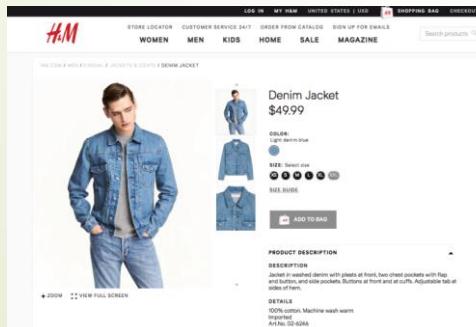


Price: 49.90
Product: 4081762



Price: 29.99
Product: 4081762

$$\frac{49.90}{29.99} = 1.664$$

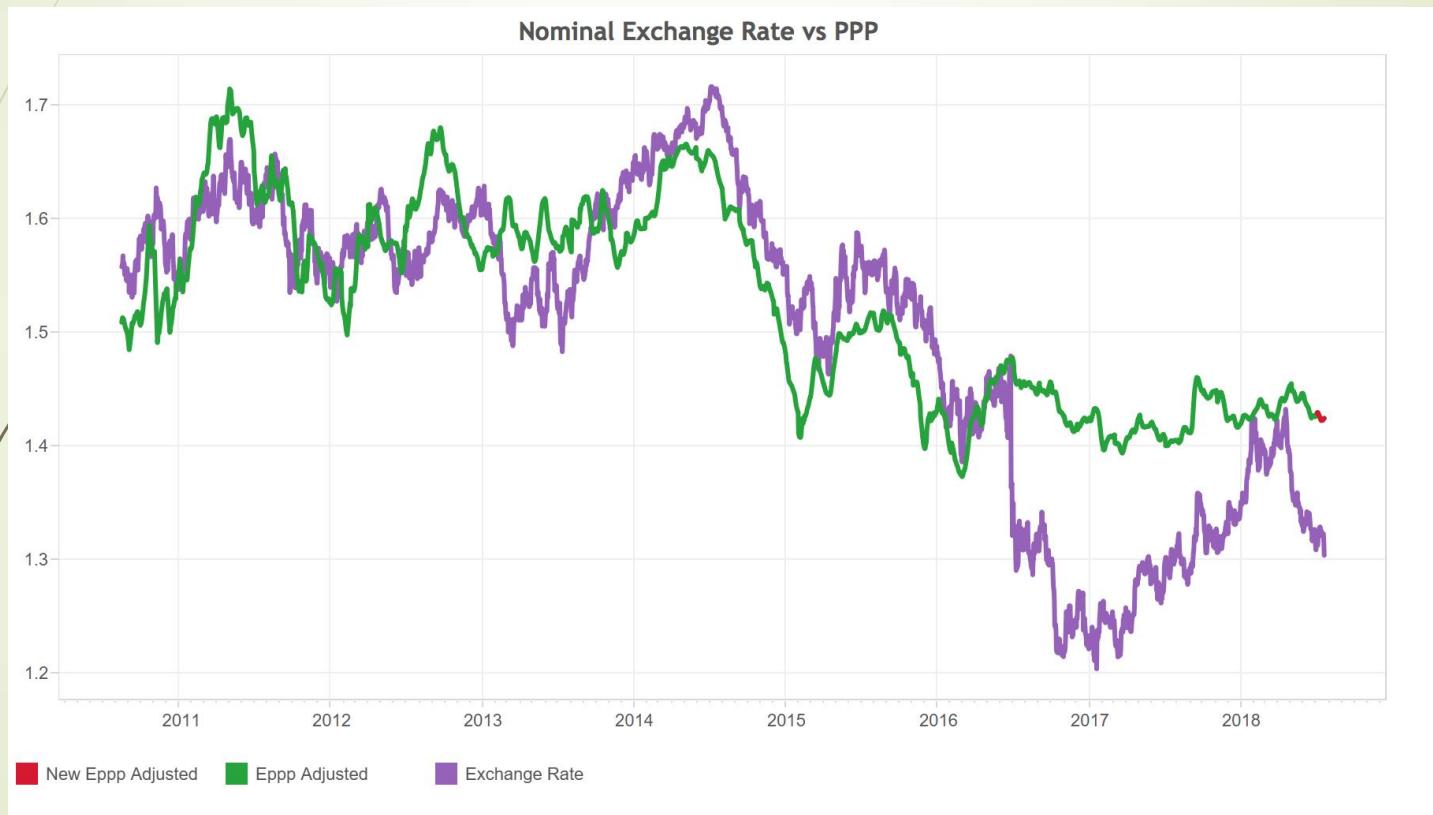


Price: 49.99
Product: 70136

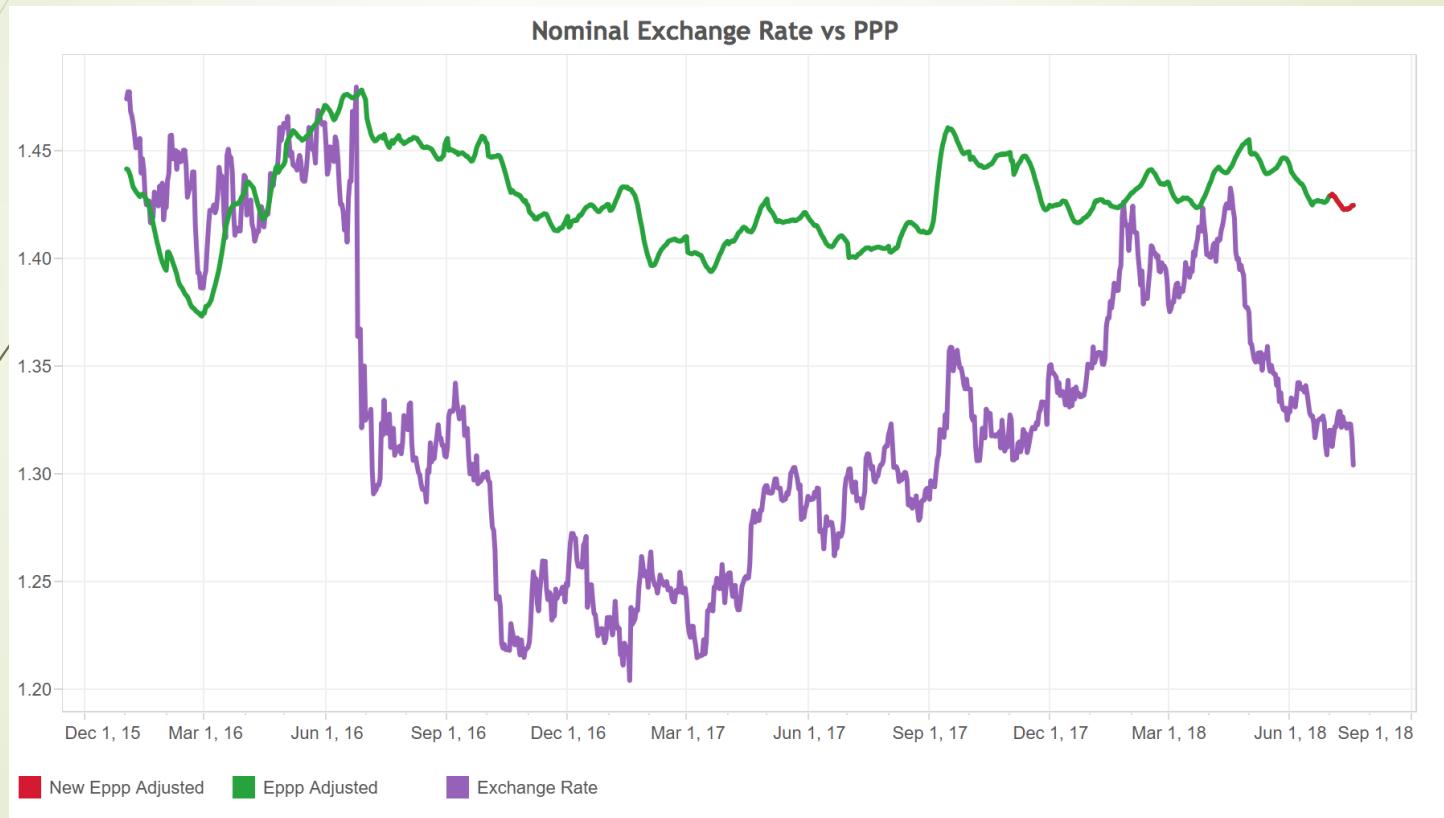


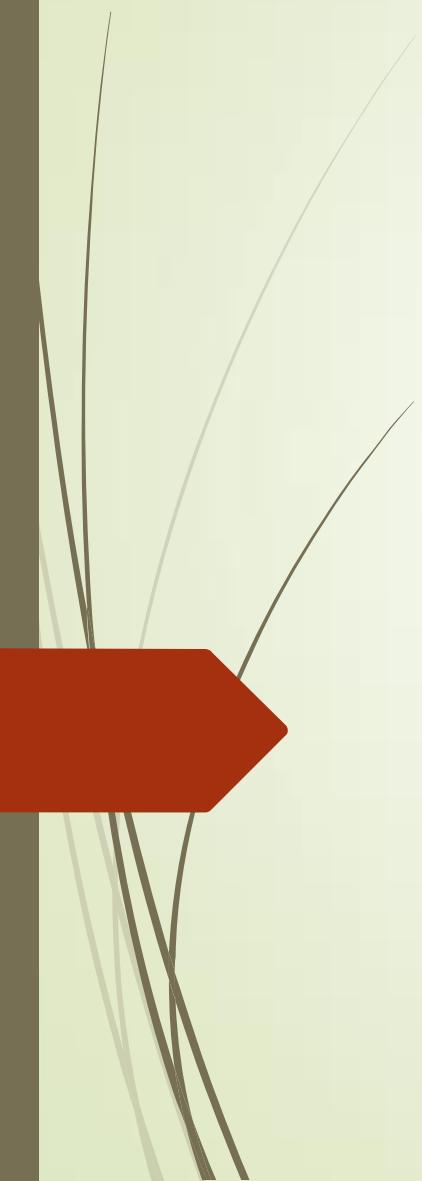
Price: 29.99
Product: 70136

$$\frac{49.99}{29.99} = 1.667$$



UK (3 years)





Aggregate Confusion

Solving the confusion of ESG ratings

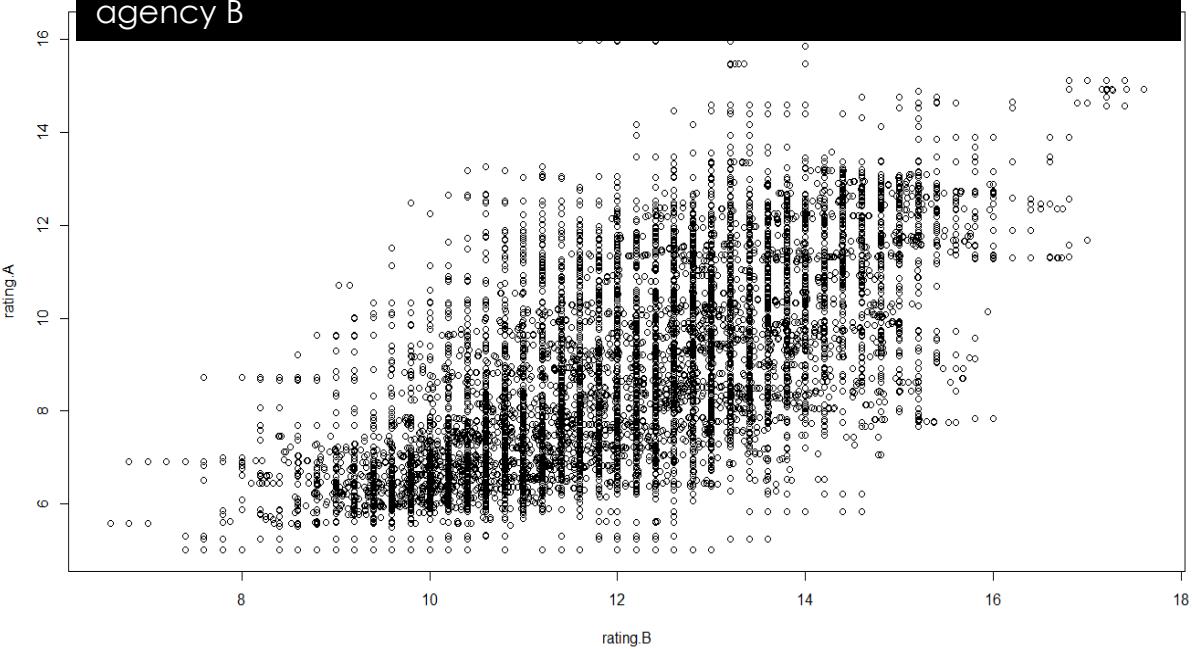
Julian Koelbel, Florian Berg, Roberto Rigobon

July, 2017



ESG Rating Agencies' Scores diverge

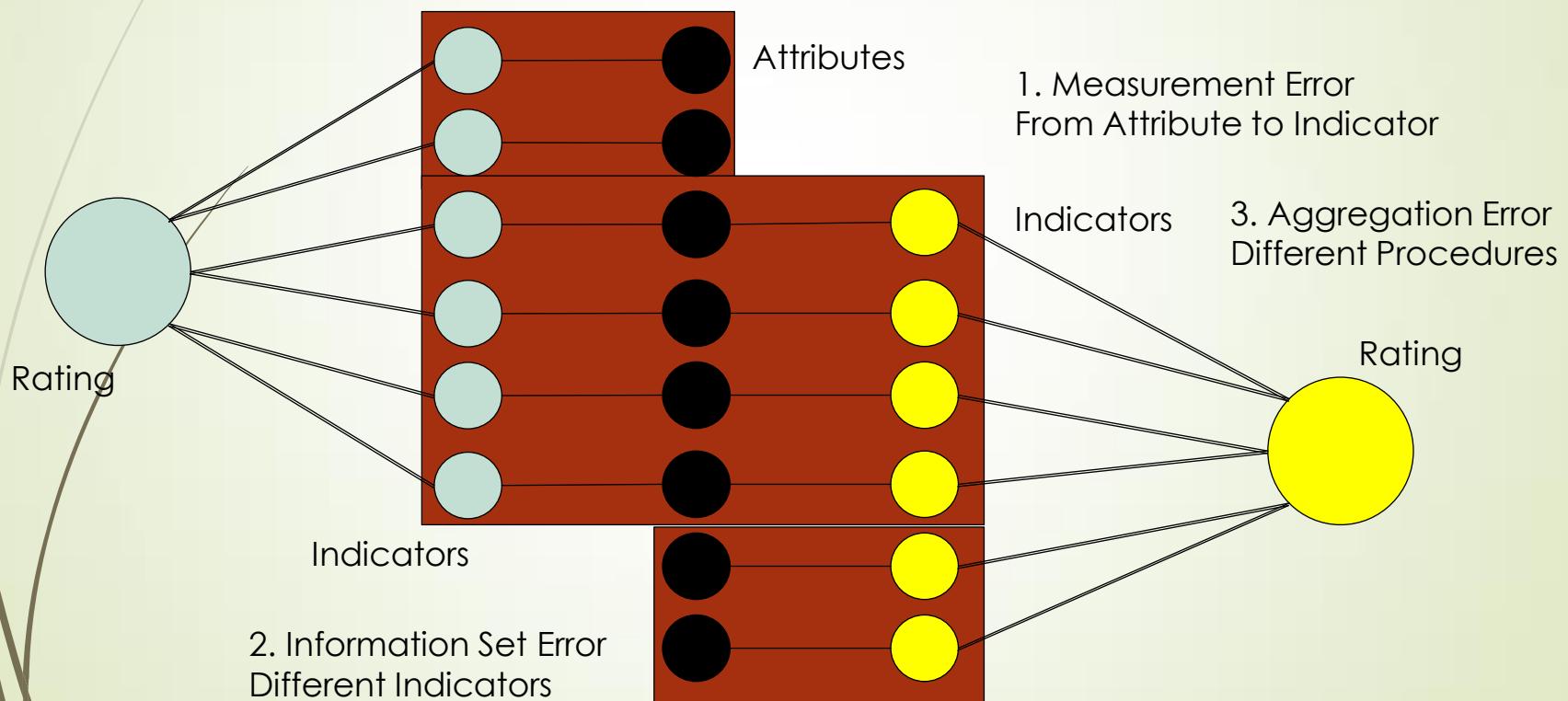
Comparison of rating scores from agency A versus rating scores from agency B



Things that are more correlated than the ratings!

- The number of people that downed every year in swimming pools in the US is correlated with the number of Nicholas Cage movies: 66 percent
- 72 percent of people who dislike licorice understand HTML
- 75 percent of people who can't type without looking at the keyboard prefer thin-crust pizza to deep-dish
- Per capita consumption of mozzarella and PhD's in civil engineering are correlated in 96 percent

Sources of errors



Preliminary Results

	Aggregation	Discrepancy	Information and Unmeasured	Straight Explanation
	VI	Measurement	VI	
MSCI/KLD	3.3%	48.6%	48.1%	19.47%
RS	1.7%	87.7%	10.6%	49.14%
SA	3.8%	22.5%	73.7%	50.38%
	RS			
MSCI/KLD	0.0%	54.9%	45.1%	18.71%
VI	0.3%	82.0%	17.7%	49.14%
SA	3.5%	34.3%	62.1%	43.52%
	SA			
MSCI/KLD	1.5%	29.8%	68.7%	22.89%
VI	6.9%	56.3%	36.8%	50.38%
RS	5.4%	76.3%	18.3%	43.52%
	MSCI/KLD			
VI	0.0%	73.1%	26.9%	19.47%
RS	0.0%	84.1%	15.8%	18.71%
SA	1.5%	43.3%	55.2%	22.89%

58%

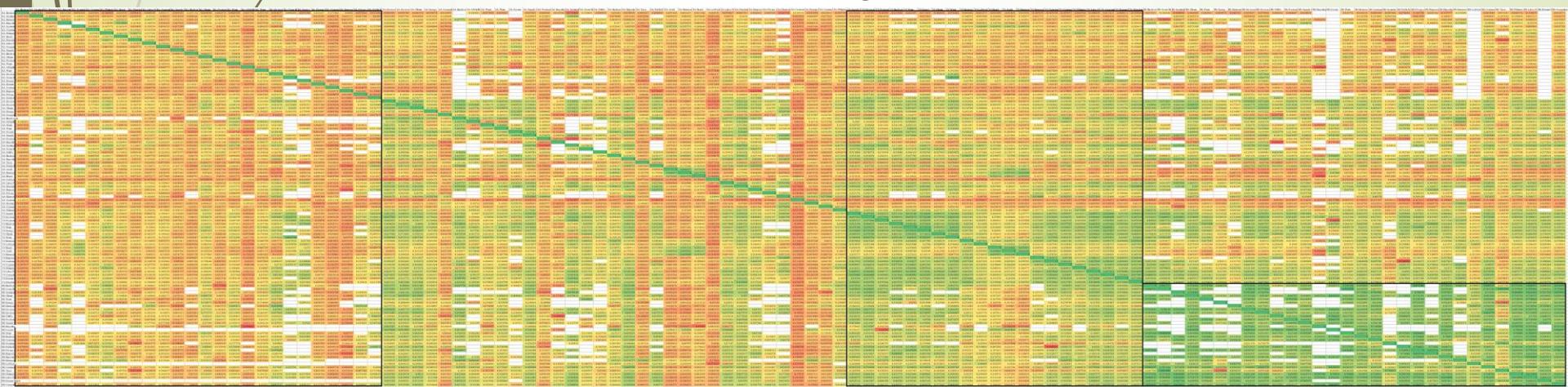
40%

Preliminary results

- ▶ Reverse Engineering implies quasi-linear rules
 - ▶ Get more than 98+ percent explanatory power within sample
- ▶ Variance Decomposition of Discrepancy
 - ▶ 58 percent is measurement
 - ▶ 40 percent is information set
 - ▶ 2 percent is from aggregation rules

Attribute measurement errors

- ▶ At the attribute level things are worse!
 - ▶ Some attributes are negatively correlated across rating agencies!
 - ▶ Lobbying
 - ▶ Responsible Marketing Policies



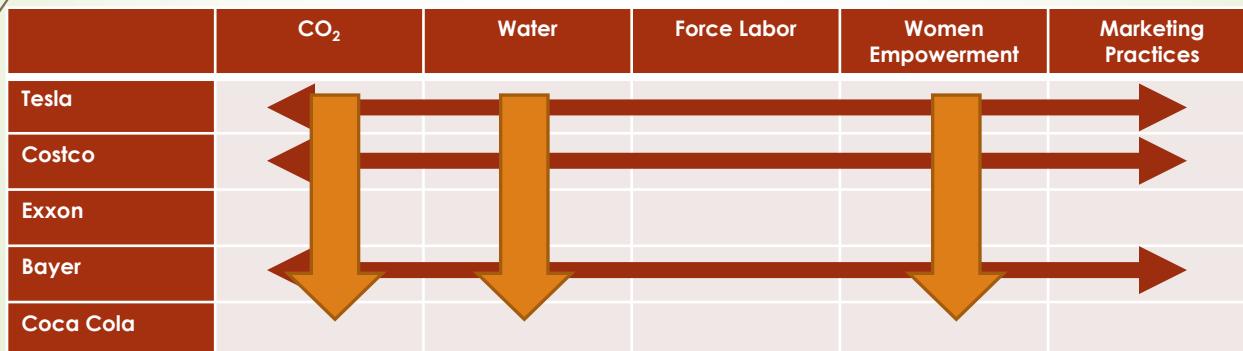


Three Domains for Improvement

- ▶ Organizational Structure
- ▶ Aggregation Procedure
- ▶ Data Source

Organizational Bias

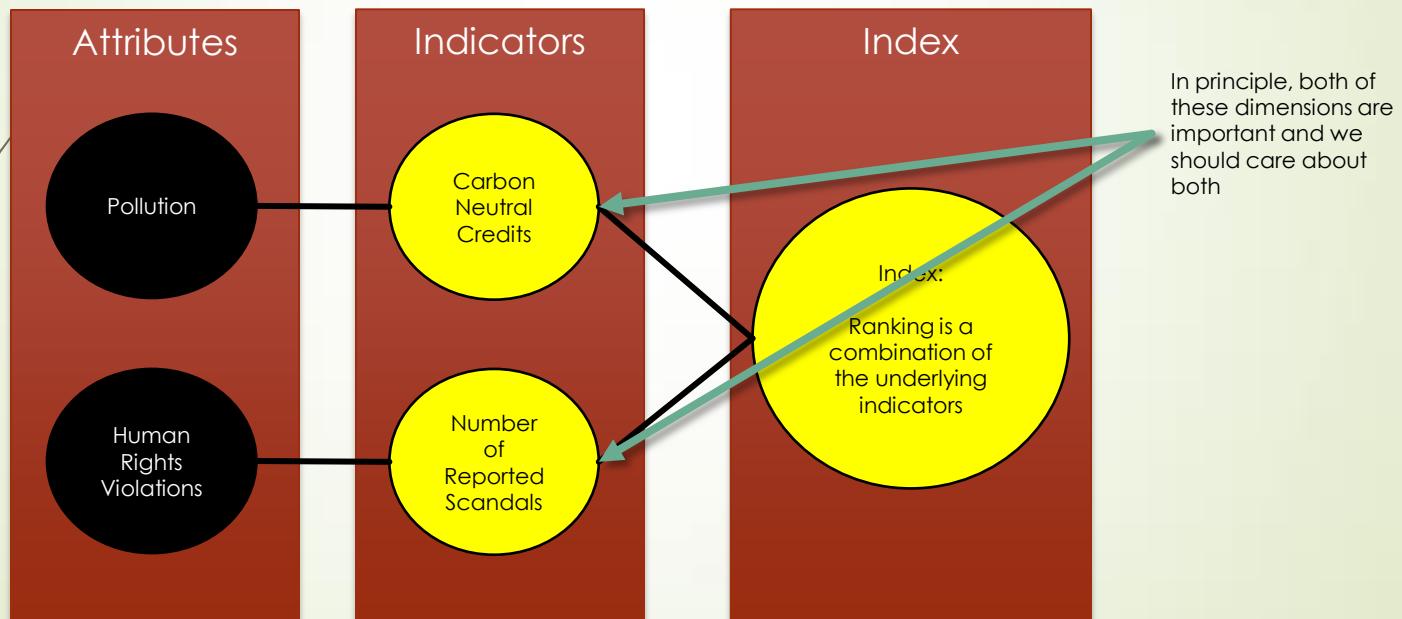
- ▶ Organizational Structure
 - ▶ Analysts are organized through firms and not across attributes (rows as opposed to columns)
 - ▶ Bias is firm (analyst) specific



- ▶ Florian's research (still preliminary)

Quasi-linear Rules

- Ranking Implies that Dimensions that should NOT be substitutable become substitutes





Data Source Bias

- ▶ Data mostly comes from Three sources
 - ▶ **Statements**: Self-reported or non-mandatory surveys
 - ▶ **Outcomes**: Extreme Events
 - ▶ Accidents or scandals measured in public documents, courts, or regulator's reports
 - ▶ **Perception** : Reporting on extreme events
 - ▶ Accidents or scandals as reported by the media
- ▶ Taxonomy might need another dimension

New Measures

- ▶ How to measure?
 - ▶ Women empowerment
 - ▶ Household Stress
 - ▶ Mansplaining and Hidden Networks
(Team Chemistry and/or Attribution Problem)
 - ▶ Human Trafficking
 - ▶ Opioid Crises
 - ▶ Job Quality and Job Satisfaction
 - ▶ Water management practices

Example

- ▶ Women Empowerment
 - ▶ How we measure today?
 - ▶ Counting Only relevant at the extremes
 - ▶ Scandals Only relevant if media agrees
 - ▶ Wage gap Discrimination on regressors
 - ▶ Characteristics?
 - ▶ Too **late**
 - ▶ Too concentrated on **extreme events**
 - ▶ Too **infrequent**
 - ▶ Too based on **perception** (news outlets)
 - ▶ Too fixated in the **wrong statistic**
- ▶ How are we evaluating materiality?



Pillars of the new measures

1. Continuous Measurement of process
 - ▶ Timely measures
2. Non-intrusive
 - ▶ Can't rely on surveys – needs electronic forms of data collection
3. Open source
 - ▶ Many could adopt the methodologies
4. Privacy protecting
 - ▶ Violations of privacy can be significantly harmful, especially when estimating hidden behavior that is morally questionable
5. Imperfect Measurement
 - ▶ To guarantee the previous 4 characteristics the measures need to be noisy.

Example

- ▶ Women Empowerment
 - ▶ Measure interruptions
 - ▶ How frequently men interrupt women and other men?
 - ▶ Airtime: how long each one talks?
 - ▶ Mansplaining: assertiveness and apologies
 - ▶ Culture, hierarchy, language, media matter – but why gender and minority status would?

How much of the unconscious biases are unmeasured biases?

Data sources of organic data

- ▶ Digital documentation of transactions with a service or manufacturing process
 - ▶ Credit card transaction, retail sales scanner data
- ▶ Data from social network communication
 - ▶ Facebook, Twitter, Instagram
- ▶ Data transmitted from software agents within mobile devices
 - ▶ GPS
- ▶ Data from the internet of things
 - ▶ e-commerce: Web pages, price aggregators
 - ▶ Utility meter data, sensor data for traffic, air, water, soil quality
- ▶ Biometric data
 - ▶ DNA
- ▶ Human communication digital data
 - ▶ Emails, blogs, text
- ▶ Digital video data

Weaknesses depend on use

	Survey	Estimation	Forecasting	Measurement
Representativeness				✓
Selection Bias		✓		✓
Reliability and Consistency	✓		✓	✓
Transparency on Data Collection and Treatment	✓	✓		✓
Errors-in-variables		✓	✓	✓
Aspirational (Transactional)	✓	✓		✓
Private (as opposed to public)	✓		✓	✓
Model Uncertainty and Behavioral Changes			✓	✓

Warnings

- ▶ The Promiscuous Pursuit of Data
 - ▶ Big Data reduces variance of the estimates, not their bias
 - ▶ Little sample uncertainty but Large model uncertainty
 - ▶ Do not fall in love with the Data: Transactional versus Aspirational
- ▶ The Human Element of the Data
 - ▶ Problem of identification
 - ▶ Who drowns more regularly at sea?
 - ▶ Happiness Index: Correlation and Causation
 - ▶ Measurement not Forecasting
- ▶ The Potential for Irrelevance
 - ▶ Problem of representativeness
 - ▶ Restaurant Reviews: Sampling Bias
 - ▶ Parental Guidance through Facebook
- ▶ Privacy is a First Order Problem
 - ▶ Aggregation does not solve the problem of privacy

The future of NSI's and CB's Data Collection

- ▶ Hybrid Approach: Change in National Statistics
 - ▶ Relevance and Timeliness
 - ▶ Demand driven (not supply driven)
 - ▶ Cost, Granularity, and Reducing Response Rates
 - ▶ Organic Data based indicators
- ▶ Privacy
 - ▶ Aggregation is not enough
- ▶ Organizational Paradigm
 - ▶ Geographical and Socio-economically organized; versus network and behaviorally organized.