

Operational indicator frameworks

An operational indicator framework is just a list. % Can we 'enrich' these lists?

- net income
- sales
- hack attempts
- approval rating
- taxes avoided
- netwets
- reliability
- trustworthiness
- system downtime

% often: time-series,
regularly updated / added-to,
something we do low-level statistical
analysis to, e.g. correlation

They are used for:

1. optimization of processes
2. strategic decision-making
3. public (or internal) communication
4. ontology / simplification

Problem: ~~the~~ how do we take data from one indicator framework and relate it to data from another framework?

~~ie. data integration~~:

Eg. how do I illustrate the benefits / ^{extended} impact of my project to another aspect of my organization / domain?

Abstract indicator frameworks

Defn. a "Formula" from a causal model to a data model. (measurement/observation)

The philosophy: toward a type system for measuring and modeling complex systems

- math artifact, e.g.
- (relational) table of data
 - List of random variables
 - * - Hilbert space of random var.
 - observation tables of a regular language

- math artifact, e.g.
- * - graphs / DAGs
 - Bayesian networks
 - dynamical systems
 - finite automata

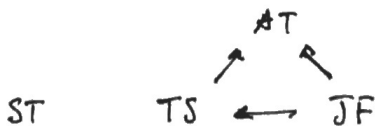
To start: fix one of each!

Ex. bus data from Nashville, TN

	actual travel time AT	scheduled travel time ST	jam factor JF	traffic speed TS
AT	1	.81	.15 ?	-.04
ST		1	.05	-.02
JF			1	<u>-.48</u>
TS				1

Correlation

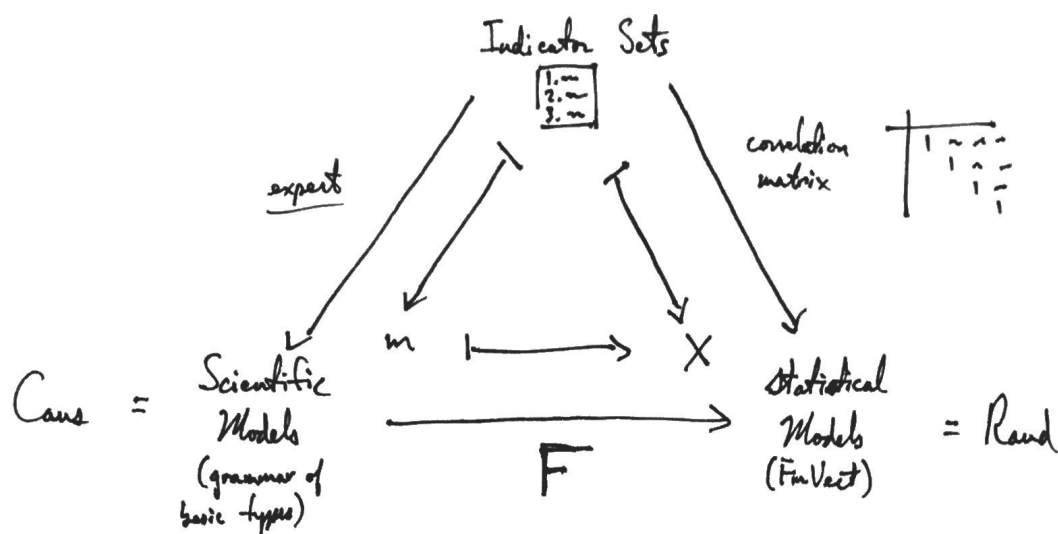
Why .15? ~~Basic~~ First step toward analysis: build a model! (basic, maybe false)



Causation

Dumb question: how do you "wrap" the two pieces of / perspectives on the system $\{AT, ST, JF, TS\}$ into one, consistent data structure?

How to construct an abstract indicator framework



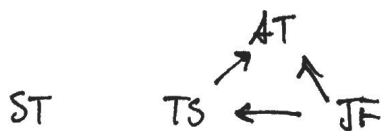
Step 1. Take as input a list of indicators, a correlation matrix on these indicators, and an "expert".

notation: indicator $i \in I$

C_I is a correlation matrix on I

$X = X_I$ is the f.d. Hilbert space w/ fixed basis $B_X = \{b_1, \dots, b_d\}$ s.t. $\langle b_i, b_j \rangle = C_I(i, j)$

Step 2. Define a causal diagram over I , e.g.



How to construct... continued

Step 3. Divide your model into "grammatical" types.

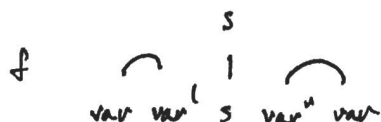
ST, TS, AT, JF \in var

$\longrightarrow \in \text{var}^l \text{ } s \text{ } \text{var}^r$ \leftarrow technically, this is a ^{formal} multiplication!

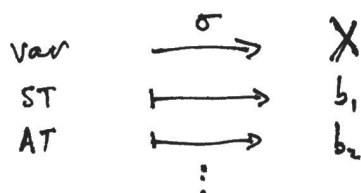
what's going on here? Basically, we are encoding " \longrightarrow " as a relation that requires a var on the left and a var on the right. "s" is a causal fact.

For now, we will just take a causal diagram as just a disjoint set of causal facts.

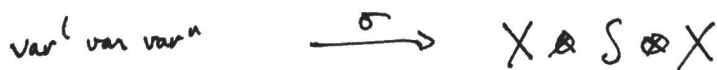
Step 4. Define a type reduction ~~map~~ $f: \text{var} \text{var}^l s \text{var}^r \text{var} \longrightarrow s$
or, diagrammatically,



Step 5. Define a map F from types (in Caus) to vector spaces (in Rand) which respects the multiplication.



note the " σ " here, which needs to be defined under F .



$$\begin{matrix} \text{" } \longrightarrow \text{"} \\ \longmapsto \end{matrix} \Psi = \sum_{i,j,k} c_{ijk} (n_i \otimes s_j \otimes n_k)$$

Done!

Using an abstract indicator framework

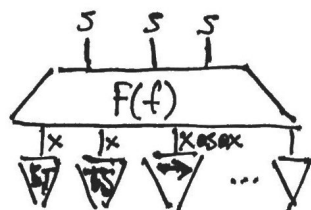
Take a causal diagram and plug it into F .
First, we get a vector

$$\sigma(ST) \otimes \sigma(TS) \otimes \sigma(\longrightarrow) \otimes \sigma(AT) \otimes \dots$$

"

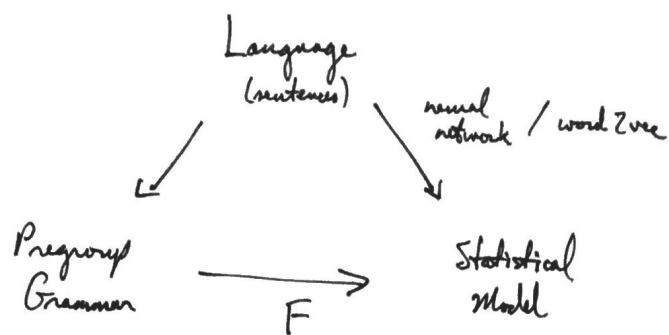
$$b_1 \otimes b_2 \otimes \Psi \otimes b_3 \otimes \dots$$

then we can apply our reduction $F(f)$ to it



in order to obtain a vector $s_1 \otimes s_2 \otimes s_3 \in S \otimes S \otimes S$ that represents the "meaning"
for validity of the model in X ,

An analogy: natural language



Idea: vector spaces can be used to assign meanings to words in a language;
 "the meaning of a word is its context of nearby words". Key thing: it's automatic.

Idea: pregroups can be used to assign grammatical structure to sentences

n: noun s: sentence v: verb

John Likes Mary
 n n's n' n → 1 s n' → 1 s 1 → s ✓

↓ F

$N \otimes N \otimes S \otimes N \otimes N \ni \vec{\text{John}} \otimes \vec{\text{Likes}} \otimes \vec{\text{Mary}}$

↓ F(f)

S

↓ F(f)

$$\begin{aligned}
 F(f) (\vec{\text{John}} \otimes \vec{\text{Likes}} \otimes \vec{\text{Mary}}) &= \\
 \sum_{ij} \langle \vec{\text{John}}, \vec{m}_i \rangle \vec{\text{Likes}}_{ij} \langle \vec{f}_j, \vec{\text{Mary}} \rangle &= \\
 \sum_{ij} \delta_{\vec{\text{John}}, \vec{m}_i} \vec{\text{Likes}}_{ij} \delta_{\vec{f}_j, \vec{\text{Mary}}} &= \\
 \vec{\text{Likes}}_{\vec{\text{John}}, \vec{\text{Mary}}}
 \end{aligned}$$