

“Predictive Analytics”:

Understanding and Addressing The Power and Limits of Machines, and What We Should do about it

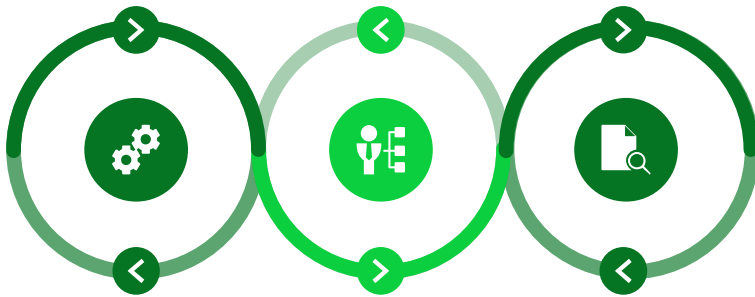
Daniel T. Maxwell, Ph.D.
President, KaDSci LLC



Models are used to:

- 01. Explain, account for, or describe a phenomenon (Diagnostic)
- 02. Predict, forecast, or estimate. (Predictive)
- 03. Recommend a course of action (Prescriptive)

Analysis uses models to



Synthesis

Tell the Story
"Narrative"

Structured Process

Analysis

What are the Pieces?
How do they work?



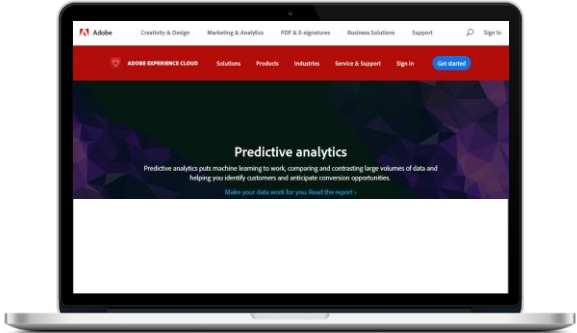
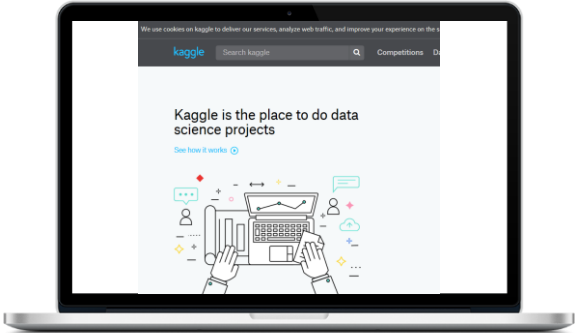
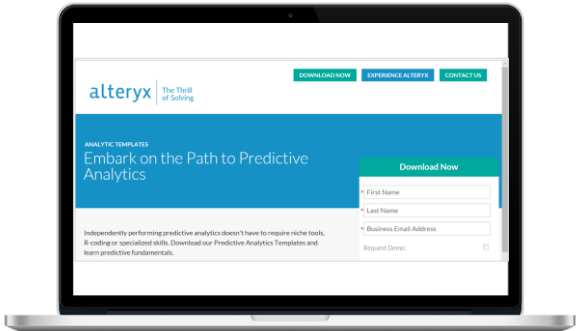
1 & 2. above are data focused – "Issues of fact"

3 above is also predictive & includes decision maker (or decision making system)

Goals and preferences

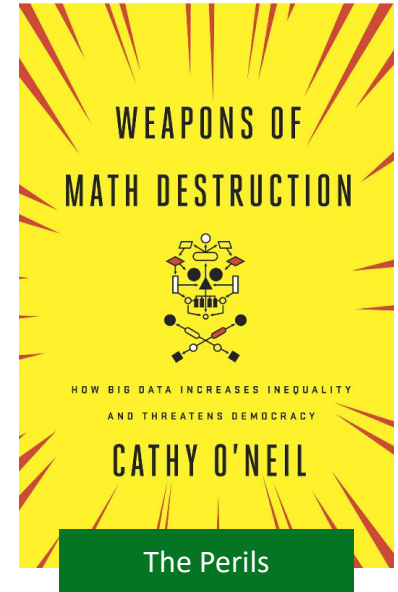
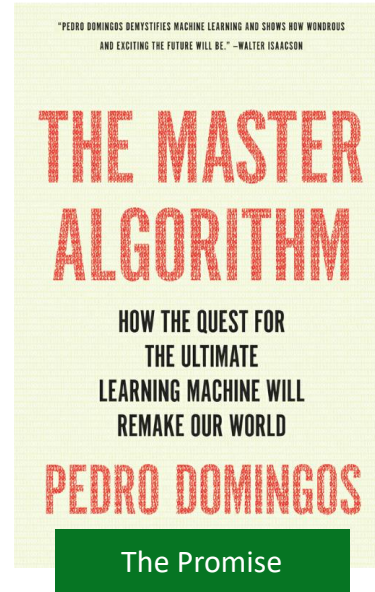
- > Objectives
- > Risk tolerance

BLUF– You can't forget the thinking part!!!



 **ALGORITHMS!!**

- **Regression**
With all sorts of new and fancy names
- **Bayesian Learning**
- **Neural Nets**
- **Case Based Reasoning**
- **Influence Diagrams**
-



All algorithms (and by extension AI tools) rely on data.



Noisy → Accurate



Uncertain → Certain



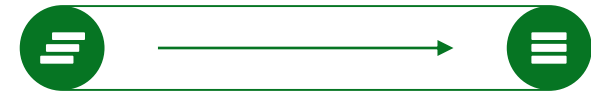
Sparse → Dense



Small → Big



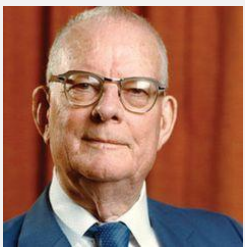
Perishable → Persistent



Unstructured → Structured



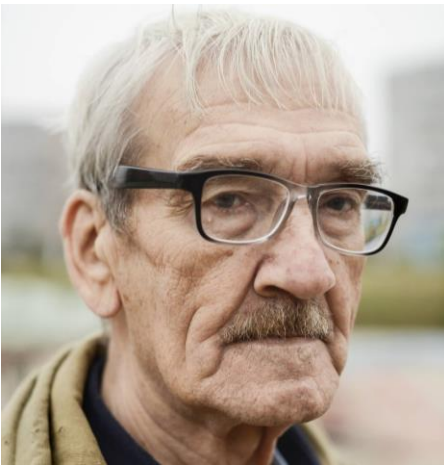
Coarse → Precise



"Without data you're just another person with an opinion"
W. Edwards Deming



Data Science and Data Engineering are different things – The latter is necessary but not sufficient for providing effective analytics.



Time – September 26, 1983,
Three weeks after KAL 007
was shot down.

He disobeyed those orders
and declared the alert a false
alarm

Lt. Col. Stanislav Petrov
(Russian Air Force) observed
sensor alerts indicating the
US launched an ICBM,
followed by five more.

He was neither praised or
punished.

Russian standing orders
called for an immediate
counter strike against the US
and NATO allies.

The cause – A rare alignment
of clouds, sunlight, and
Russian Satellites that
watched North Dakota

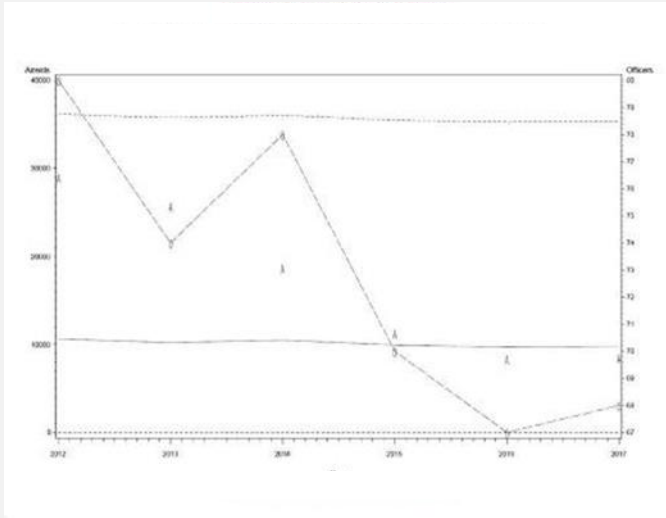
Why did he not report it:

- Five missiles were inconsistent with his understanding of how the US would attack
- Alert system was new
- Alert passed through 30 layers of verification too quickly
- Lack of corroborative evidence
- His civilian experience helped him to make that judgment



He believed if one of his pure military colleagues had been on duty, the outcome could have been very different.

Could be inexpensive and effective



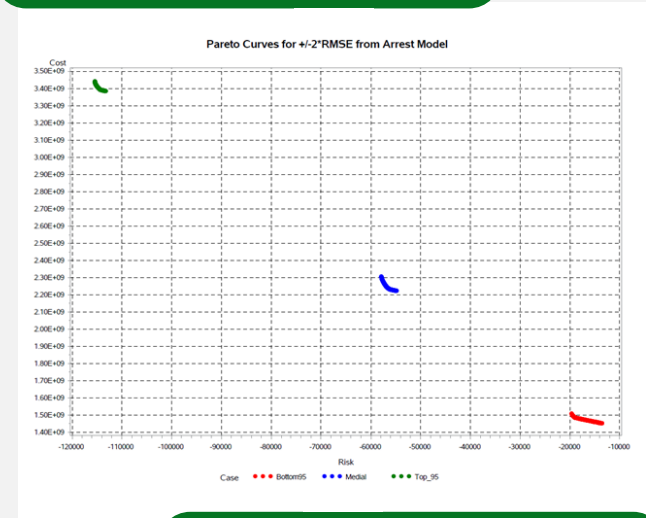
Hiring Strategy



Mean Squared Error is 10X bigger than the coefficient



Officer Coefficient ~ 100
Mean Squared Error ~ 1000
Y Intercept ~ -100



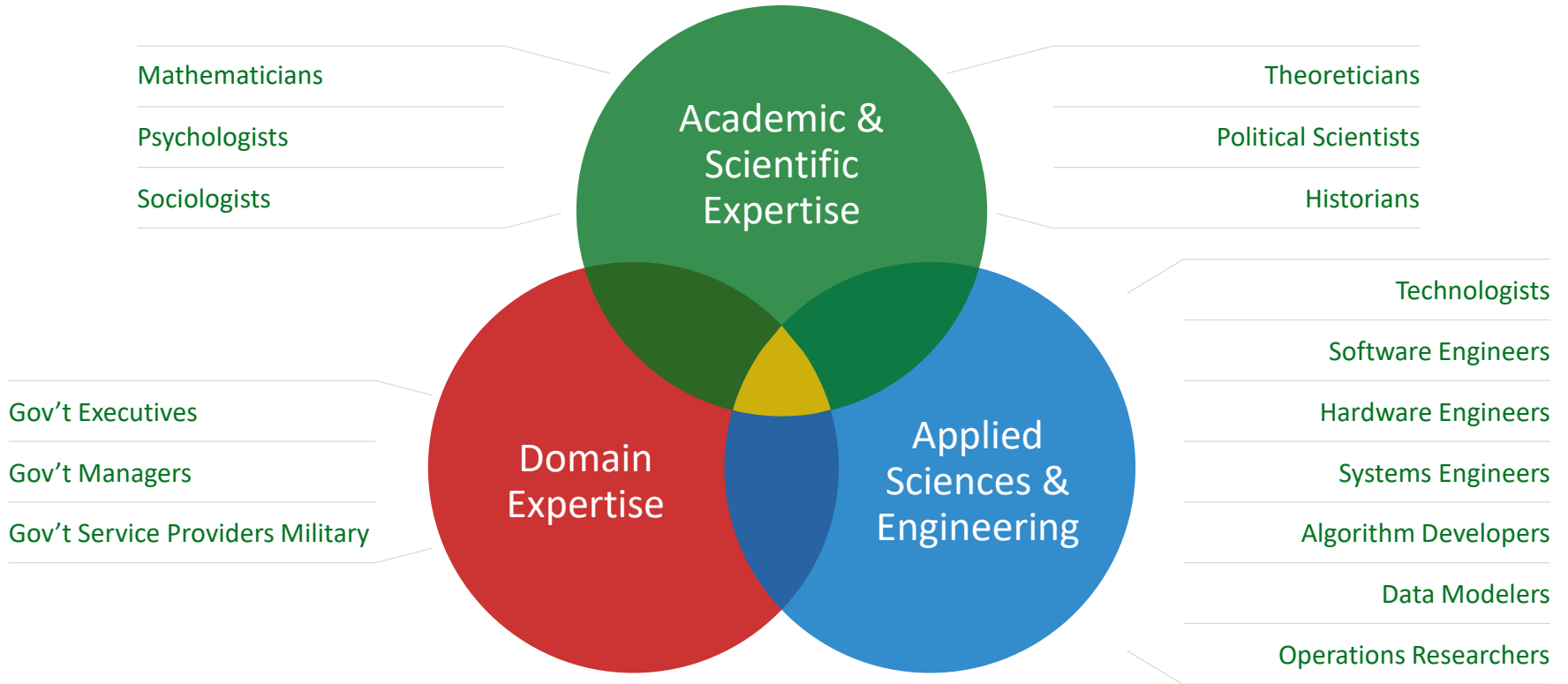
Could be expensive and ineffective



Real Situation

Notional Numbers

Anonymized Organization



Good Analytics

Provides the foundation for useful models and tools

Is multi-disciplinary, drawing on expertise from across the spectrum of science and engineering

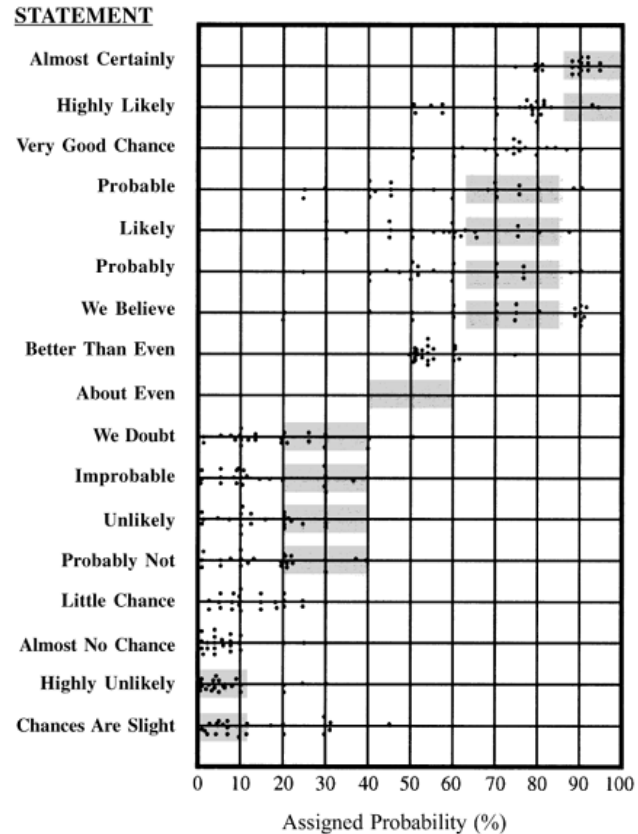
Sherman Kent, CIA

Note

the large number of dots outside the selected regions

100% Certainty		
<i>The General Area of Possibility</i>		
93%	give or take about 6%	Almost certain
75%	give or take about 12%	Probable
50%	give or take about 10%	Chances about even
30%	give or take about 10%	Probably not
7%	give or take about 5%	Almost certainly not
0% Impossibility		

Figure 18: Measuring Perceptions of Uncertainty



Representing uncertainty is messy.
Ambiguity exists even in epistemic uncertainty



Assertion

Data is nothing more (or less) than evidence

Facts:

There is science behind evidence (*Schum, 1994*)

It has been applied (loosely) by the IC.

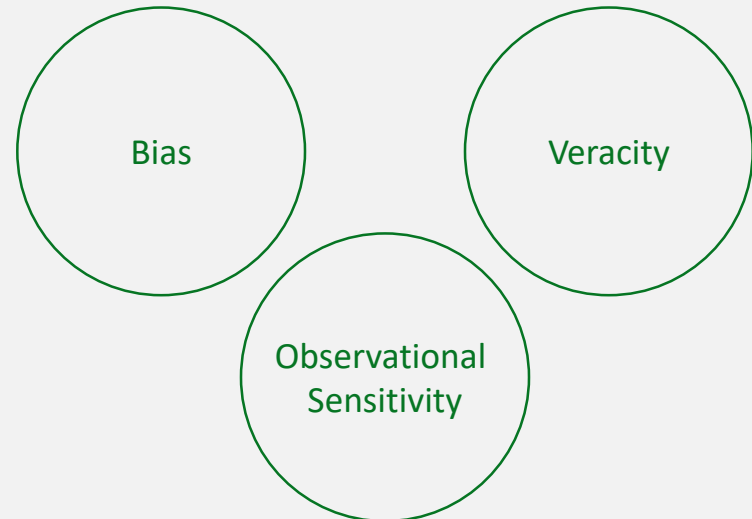
Assertion:

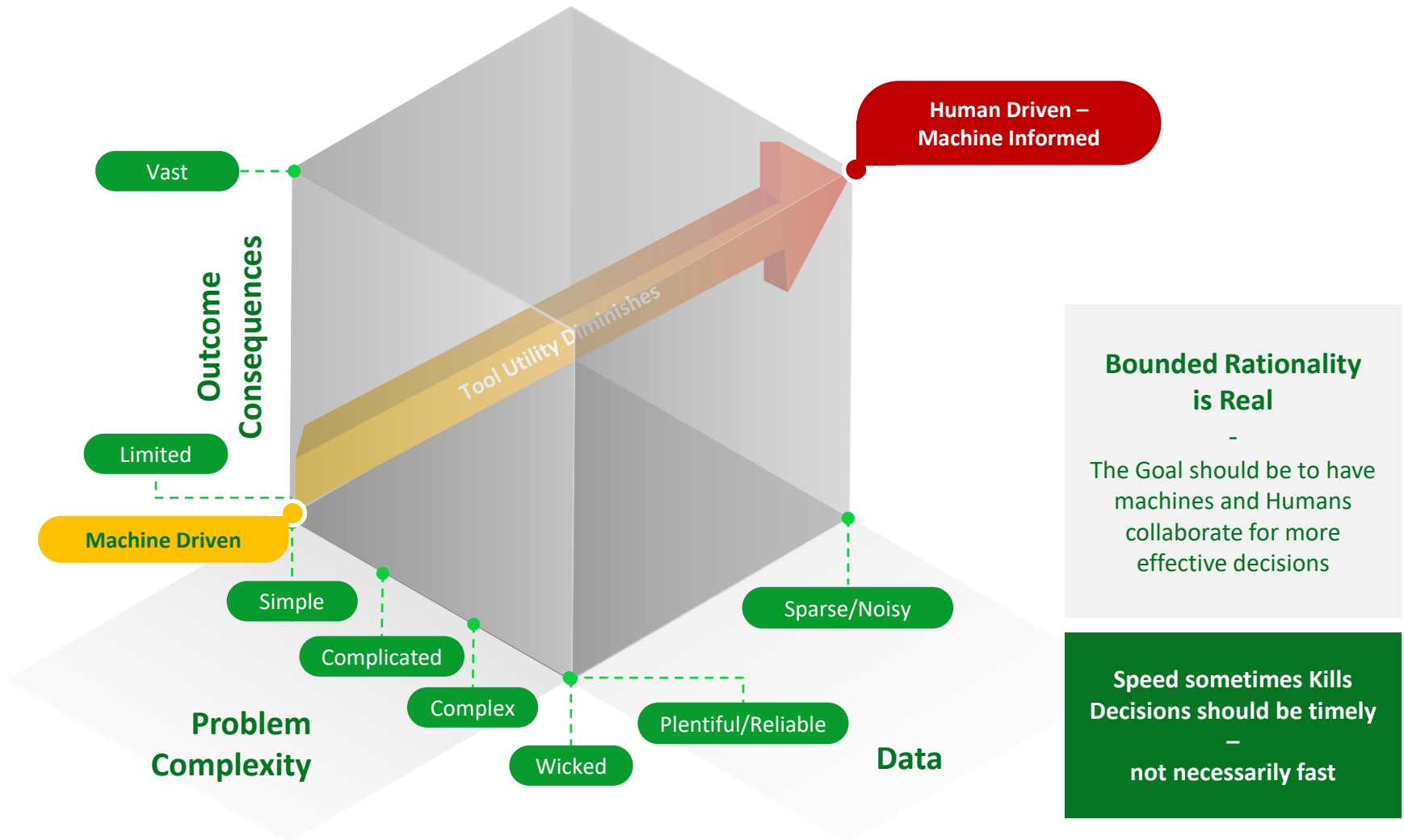
Failure to account for these evidential factors undermines the quality of machine reasoning.

These factors can and should be systematically addressed (and reported)

The weight (value) of evidence is based on quality criteria.

Things like:





The importance of human in the loop increases with situational complexity



Numbers and fluidity of objectives
(wickedness of the problem)



Impact and size of irreducible uncertainty
on the situation

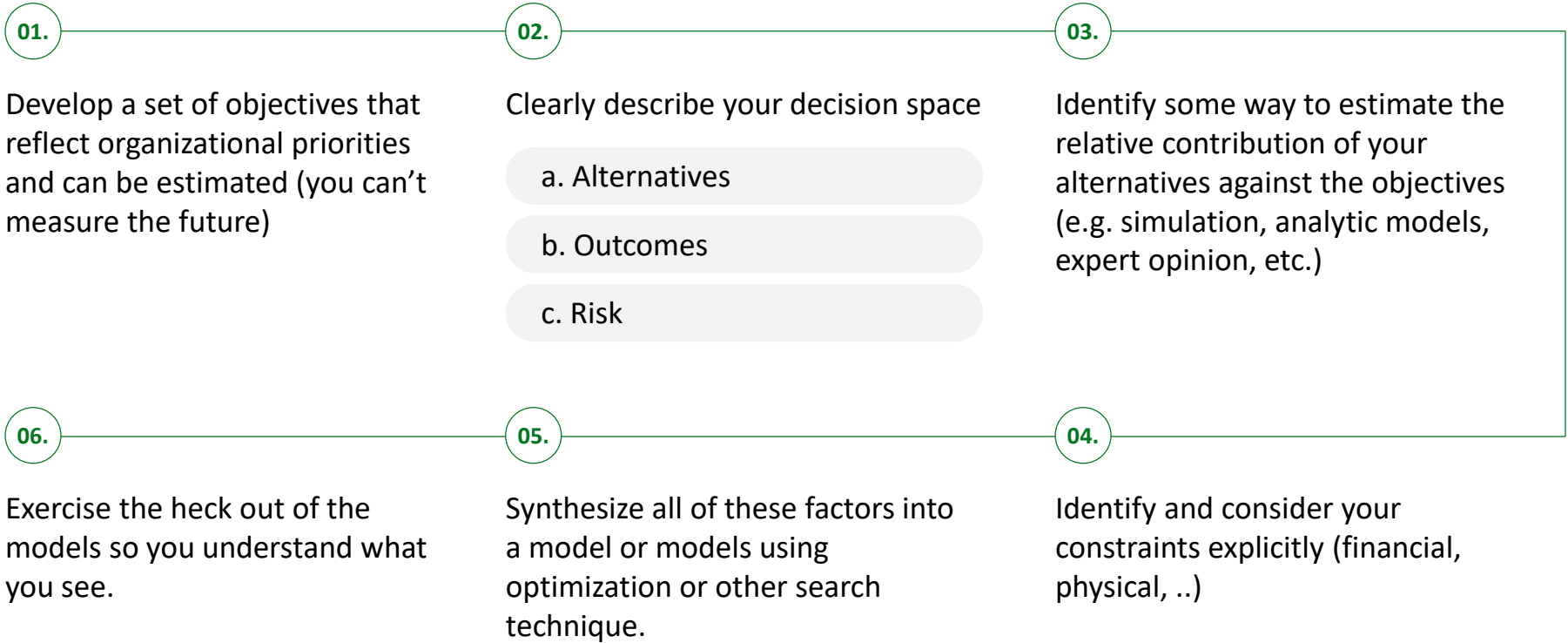
There is settled science we can draw upon to
improve predictive (inferential) performance

It appears the solution lies in improving Man –
Machine Collaboration

Nugget – IARPA is already working the Problem
(Hybrid Forecasting Program)

*“There is no substitute
for knowledgeable
human eyes on
the data”*

Professor Loerch GMU



Then AND ONLY THEN are we ready to:



Trust predictions



Make recommendations



Automate decisions (implement as AI)

**You can't
forget the
thinking
part!!!**

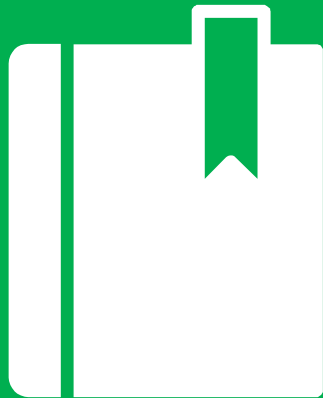


Questions???



Comments

**You can't
forget the
thinking
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Hammond, K.R. (1996) Human Judgment and Social Policy, Irreducible Uncertainty, Inevitable Error, Unavoidable Injustice. Oxford University Press: New York.

Kent, S. (1964) Words of estimative probability. *Studies in Intelligence*, 8, pp. 49-65.

Schum, D. (1994) *The Evidential Foundations of Probabilistic Reasoning*, John Wiley and Sons, New York.

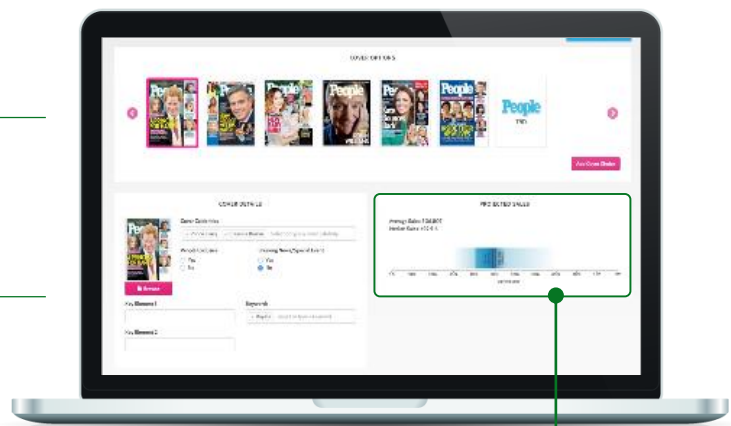
Decision Support for Magazine Cover Design

Document variables for 257 Issues

Issue	Issue Type	SEASON	Exclusive	Salestv of	2012 Norm	2103 Norm	2104 Norm	2015 Norm	2016 Normalized
2011-02	Special	Wint_1	TRUE	1,245,054	1083196.98	942381.37	815871.79	713288.46	620560.96
2011-03	Regular	Wint_1	TRUE	925,372	805073.64	700414.07	609360.24	530143.41	461224.76
2011-04	Regular	Wint_1	TRUE	1,172,343	1019938.41	887346.42	771991.38	671632.50	584320.28
2011-05	Regular	Wint_1	FALSE	998,184	868420.08	755225.47	657307.16	571857.23	497515.78
2011-06	Regular	Wint_1	FALSE	983,681	855802.47	744548.15	647756.89	563548.49	490287.11
2011-07	Regular	Wint_1	FALSE	910,056	791748.72	688821.39	599274.61	521368.91	453590.95
2011-08	Regular	Wint_1	TRUE	1,147,807	998192.09	868775.12	754834.35	657575.89	572091.02
2011-09	Regular	Wint_1	TRUE	950,569	834825.03	726297.78	611879.07	540734.79	478209.26
2011-10	Regular	Wint_1	FALSE	1,064,172	925829.64	805471.79	700760.45	609661.60	530405.98
2011-11	Special	Wint_1	FALSE	1,261,884	1097839.08	955120.00	830954.40	722930.33	628949.39
2011-12	Regular	Wint_1	FALSE	867,825	755007.75	656856.74	571465.37	497174.87	432542.14

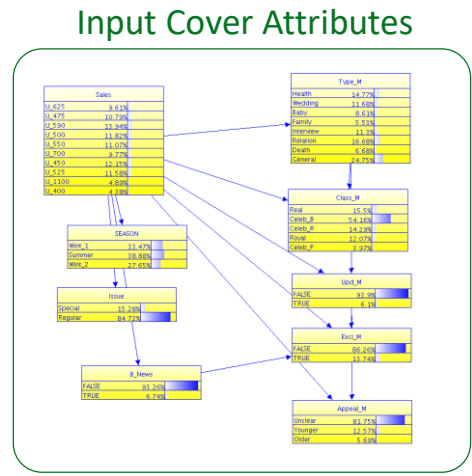
Normalize sales to 2016 levels

Next Step

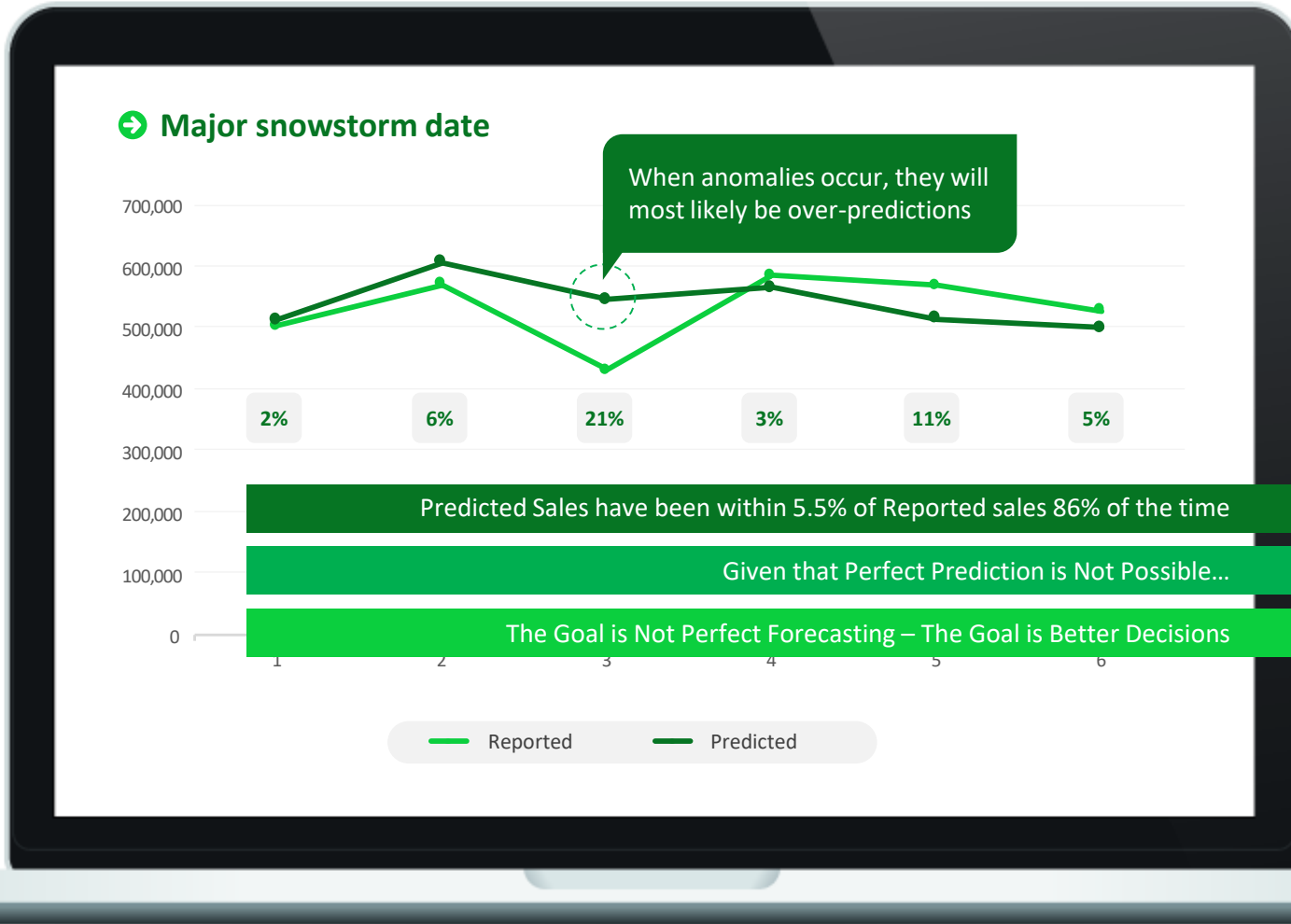


Generate synthetic data that Expands 257 issues to 6,000+

Learn the Forecasting Model Machine Learned Bayesian Network

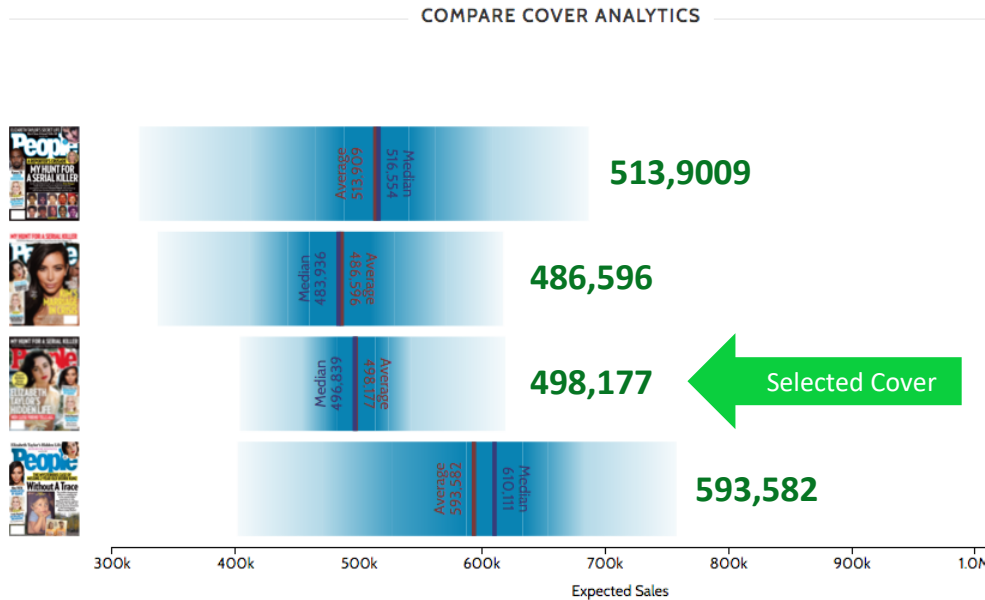


Estimated sales/ With variance



“It’s tough to make predictions, especially about the future” – Yogi Berra

Forecasted Sales March 7th 2016 cover options



If we can believe the model & assuming a \$3.00 profit per unit sold at the newsstand the opportunity cost of this decision is ~ \$285K



Current Model is at the summary level

Social Media Analysis is indicating that opinions about celebrities vary by location

For example Adelle is popular in the northeast US, weak negative perception in Pacific Northwest



Sales geographically



Celebrity Appeal Geographically

