



Fulfilling the Promise of AI – Real Strategies to Drive Personalization and Value

Andrew Hannah, Co-Founder – Othot, a Liaison Company

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Andy Hannah

Chief Partnership Officer

- **Co-Founder of Othot, Inc. (a Liaison Company) – A Predictive Analytics Company**
 - awhannah@othot.com
- **Adjunct Professor of Business Analytics at the Katz School of Business at the University of Pittsburgh**
 - awhannah@pitt.edu
- **Senior Advisor and Faculty Member, International Institute of Analytics**
- **Co-Author of Othot’s “Demographic Cliff Research Report”**

Othot's advanced analytics guide colleges and universities to make informed decisions throughout the entire student-to-alumni lifecycle by understanding each individual better.

Topics

- Overview of AI, ML, and Advanced Analytics
- Data Before Analytics but Most Importantly, Insights
- ML in Action – University Advancement
- Wrap Up – What You Can Do “Back Home”

Overview of AI, ML, and Advanced Analytics

Patterns of
Behavior Repeat



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We are always
trying to predict



We can change a
person's likelihood



Using AI/ML IS About Better Decision Making

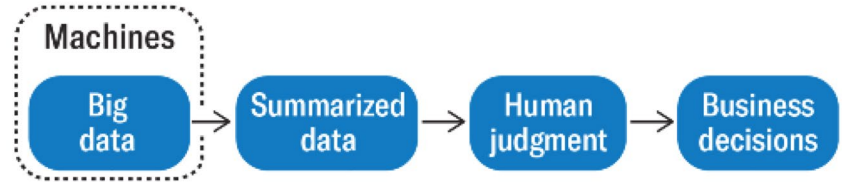
A Decision-Making Model Based on Human Judgment



Source: Eric Colson



A Decision-Making Model That Utilizes Summarized Data

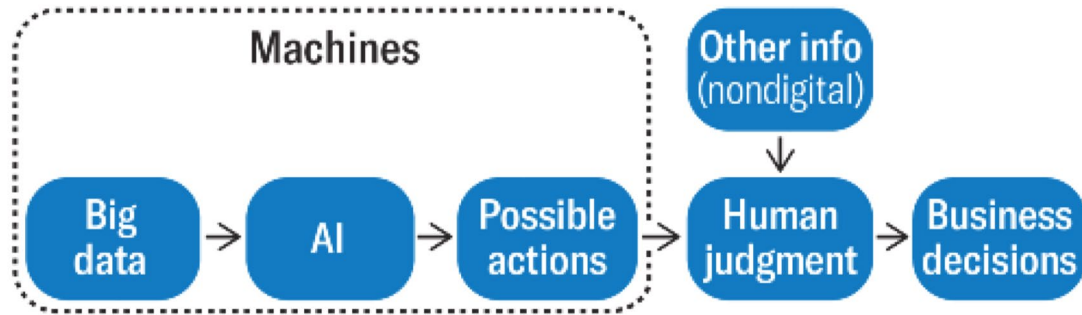


Source: Eric Colson



Using AI/ML IS About Better Decision Making

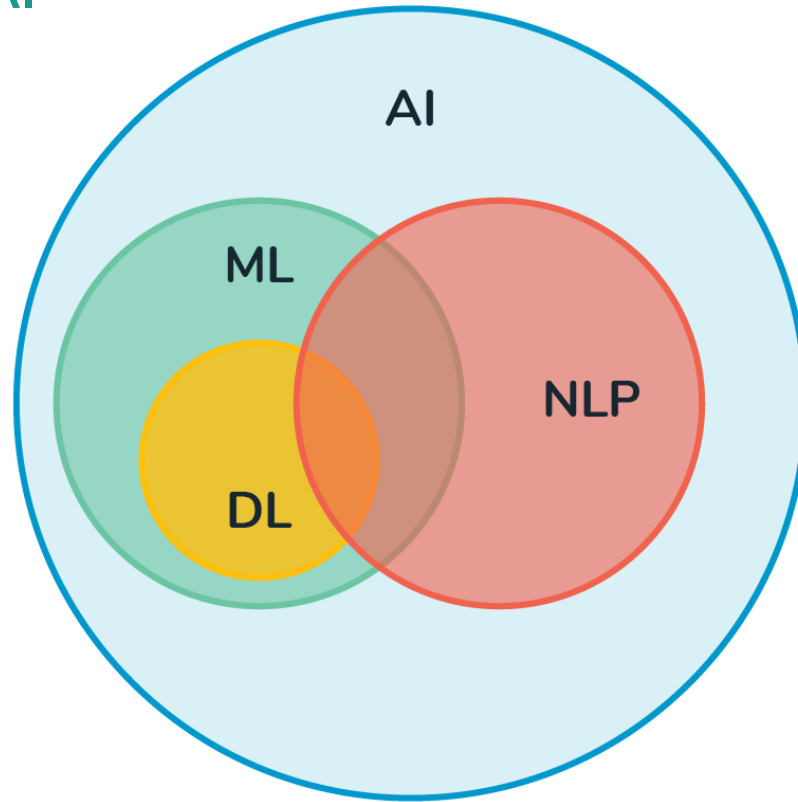
A Decision-Making Model That Combines the Power of AI and Human Judgment



Source: Eric Colson

HBR

An “Aerial View” of AI



- Artificial intelligence
- Machine learning
- Language Processing
- Deep learning

Digging a Little Deeper

Artificial Intelligence

Machine Learning

Bayesian
Techniques

Predictive
Analytics

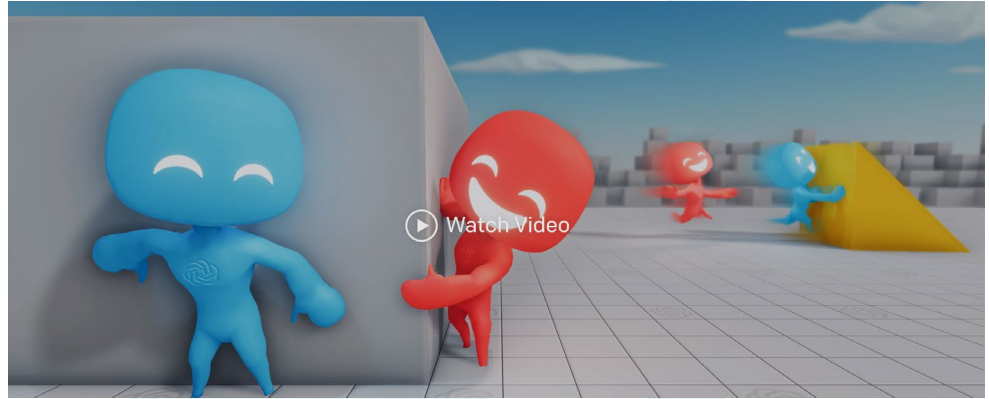
Prescriptive
Analytics

Big(?) Data
Unstructured
Data

Machine Learning and Multi-Agent Interaction



Deep Mind: Atari



<https://openai.com/blog/emergent-tool-use/>

The Root of Near-Term Success

Commercial Leaders:

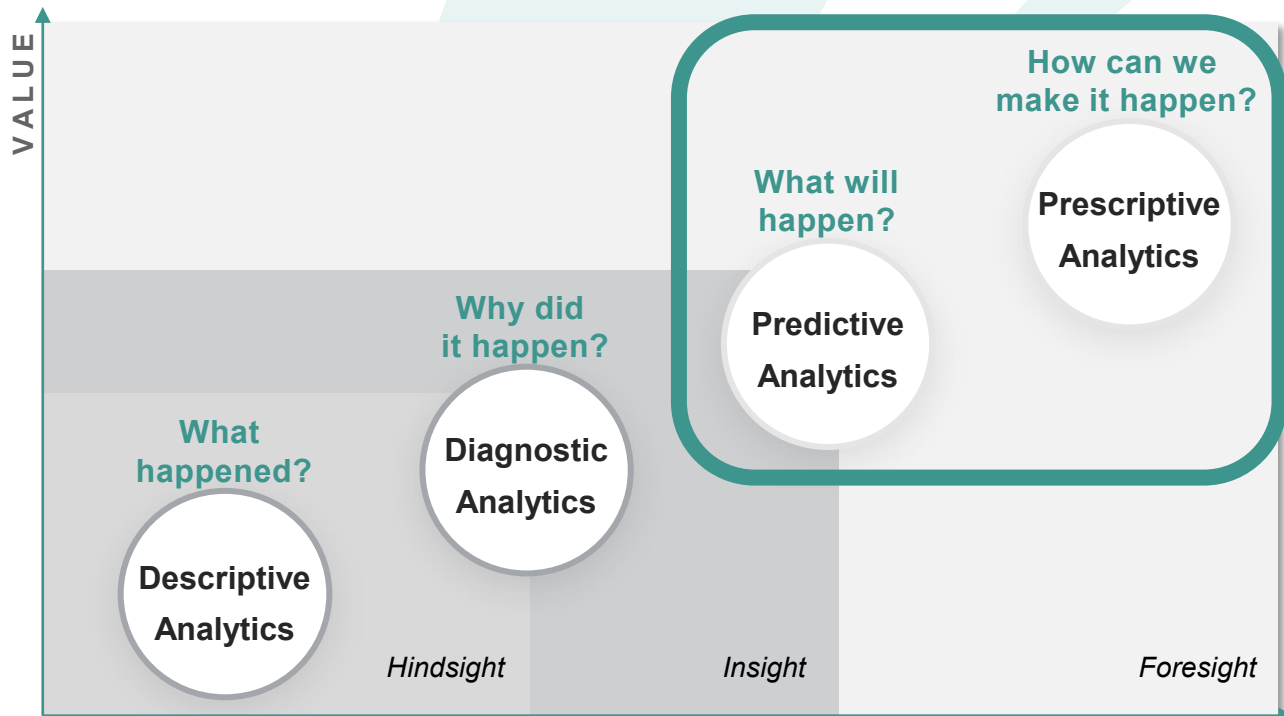


NETFLIX

amazon.com

Google

Capital One



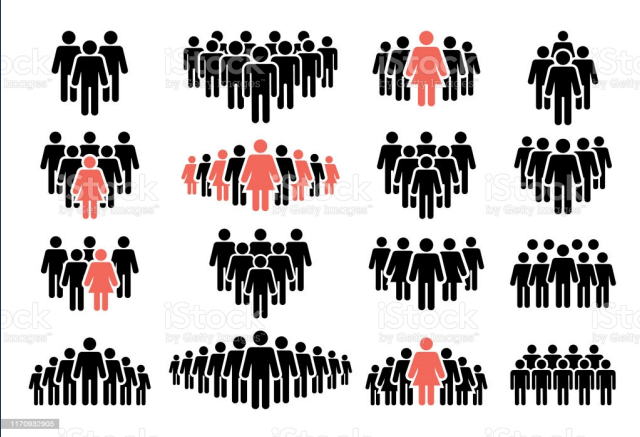
Gartner Analytics Value Escalator
www.gartner.com

DIFFICULTY



Learn

Predict Behavior



20% Likely

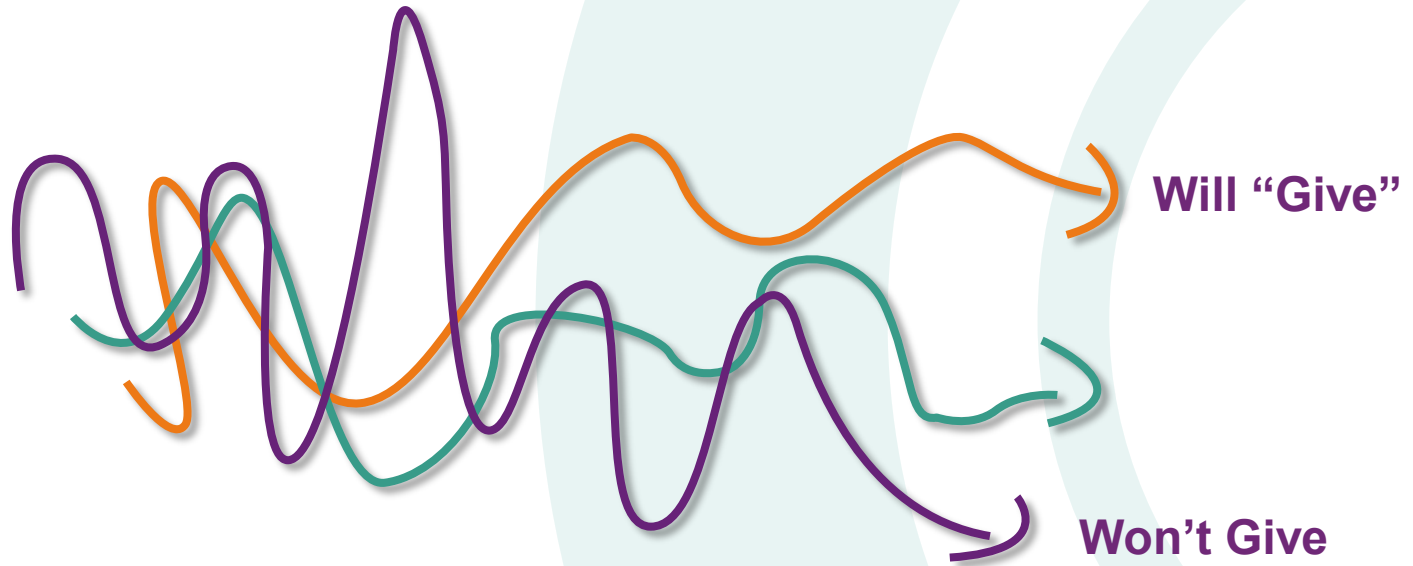
Prescribe



80% Likely

HIQ

Data and analytics reveal the path to insights

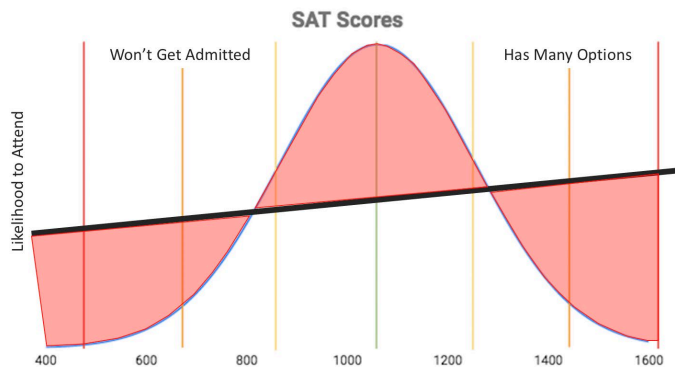


The Individual Replaces the Persona

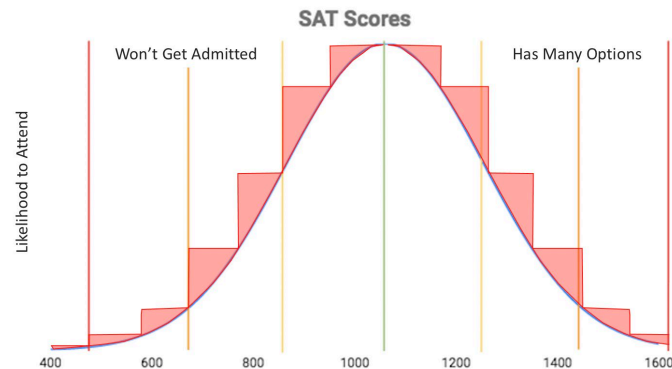
Everyone is Unique So We Need **Non-Linear Thinking**

Linear regression cannot capture this interaction between variables – **but machine learning can**

Linear



Nonlinear



Data Before Analytics but Most
Importantly, Insights

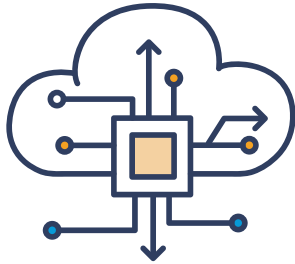
What are the Key Building Blocks for Success?

1



Leadership

2



Flexible &
Dynamic IT
Architectures

3



Data
Mastery

4



Product Thinking
Applied to
Analytics

5



Innovation-
Oriented
Cultures

How Should We Think About Data?

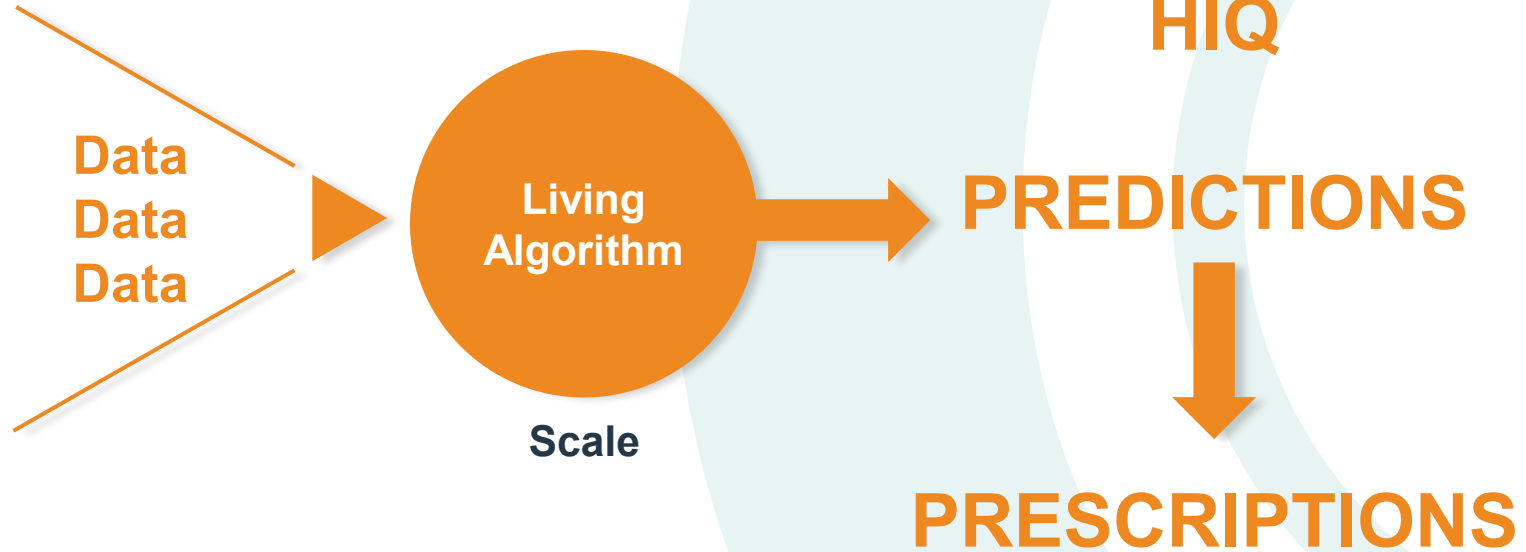
What data?

- a. Descriptive
- b. Behavioral (leads to prescriptive)

What's important?

- a. Money
- b. Status
- c. Relationship
- d. Excitement (marketing)

A Simple Model



HIQ's Are The Trees

What is the likelihood that the ultimate objective will occur?

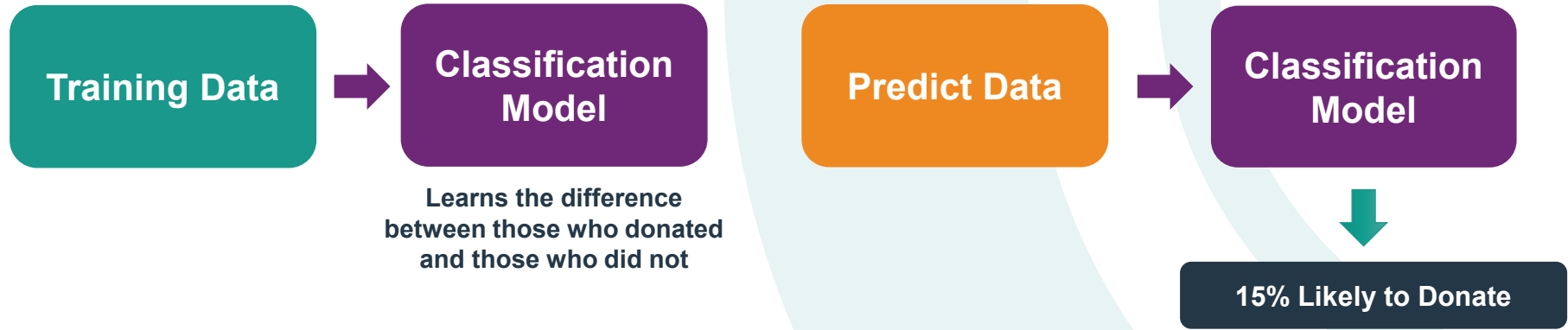


Insights are the Fruit

- How to prioritize resources
- What future actions are most impactful?
- What are the tactics to the strategy

Example: How Predictive Modeling Using Machine Learning Works

HIQ: What is the likelihood that a alumni will donate?



Non-Linear Modeling at Work

Predict Data

No Visit
No Call
No Digital Marketing



Classification Model

Variable	Impact
Visit	10
Call	5
Digital Marketing	1



Original Prediction

15%



Prescriptions A & B

50%

Visit
Call

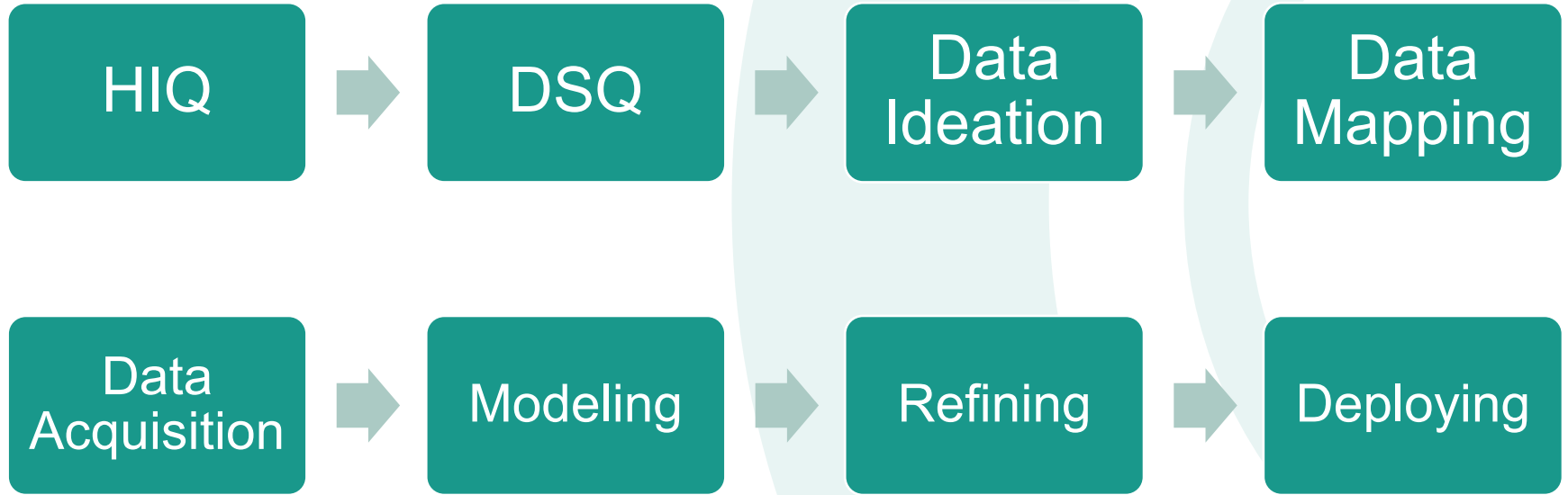


Prescriptions C

52%

Visit
Call
Digital Marketing

The Engineered, Advanced Analytics Process to Insights



An Othot Inc. developed process

ML In Action – University Advancement

“WHAT IS THE LIKELIHOOD OF AN ALUM TO DONATE?”

“What is the likelihood that an **ALUM** who **GRADUATED** between 1959 and 2009 will **DONATE** at least \$1,000 to the **XYZ FUND** in the current fiscal year?”

Two distinct models predict the likelihood that an individual who graduated between 1959 and 2009 will donate



Give at least \$1000



To a target fund



During the next fiscal year



For two populations separately:

3 Year Lapsed & Non-Donors
Recent Donors

Models for Two Populations Allow Us to Better Understand Individual Behavior

3 Year Lapsed and Non-Donors

- **Included: demographic, geographic, co-curricular, alumni association and behavioral data**

Top Importances

Alumi Association Status	42%	Current Age	4%
Gift Capacity Estimate	12%	Field of Work	2%
Yrs Since Joining Alum. Assoc.	6%	Most Common Income Bracket	2%
Alumni Association Type	4%	Graduate Degree Level	2%
Yrs Since Last Degr.	4%	Distance	2%

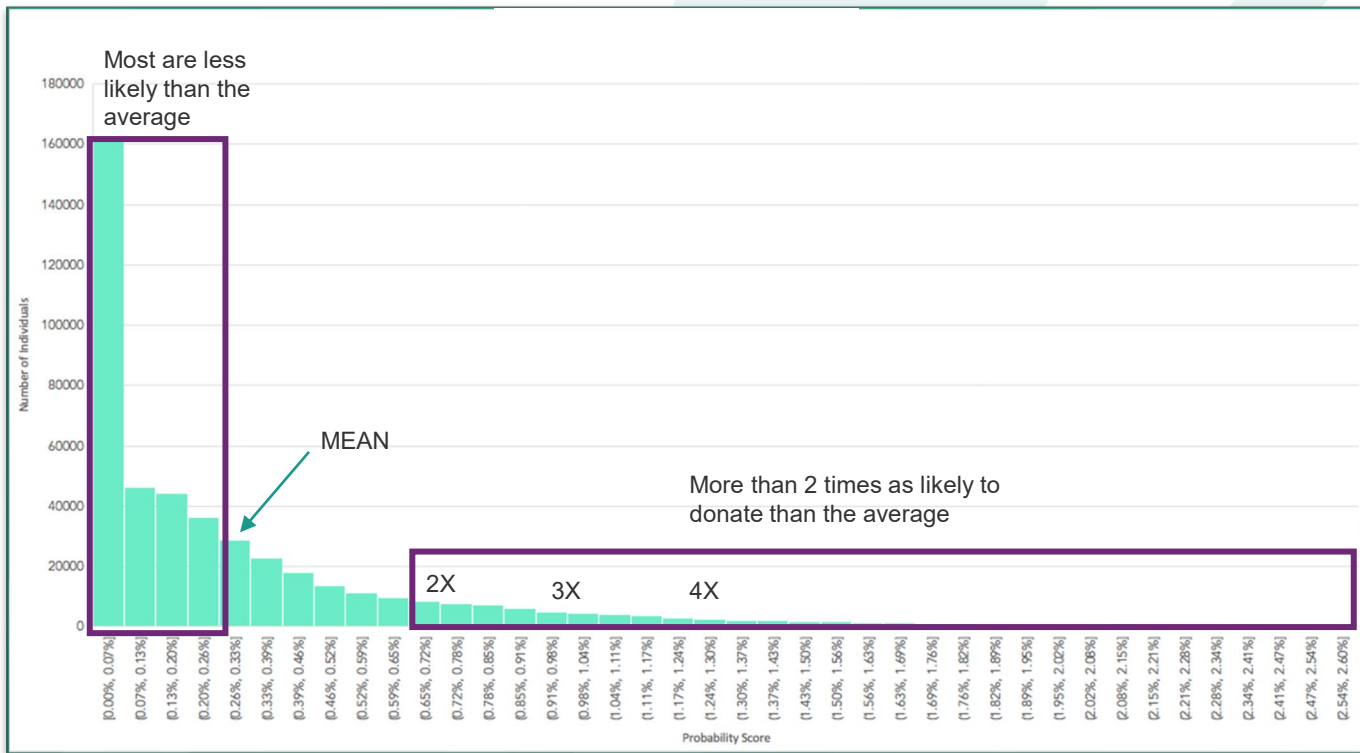
Recent Donors

- **Include: demographic, geographic, co-curricular, alumni association, and behavioral data as well as donation history data**

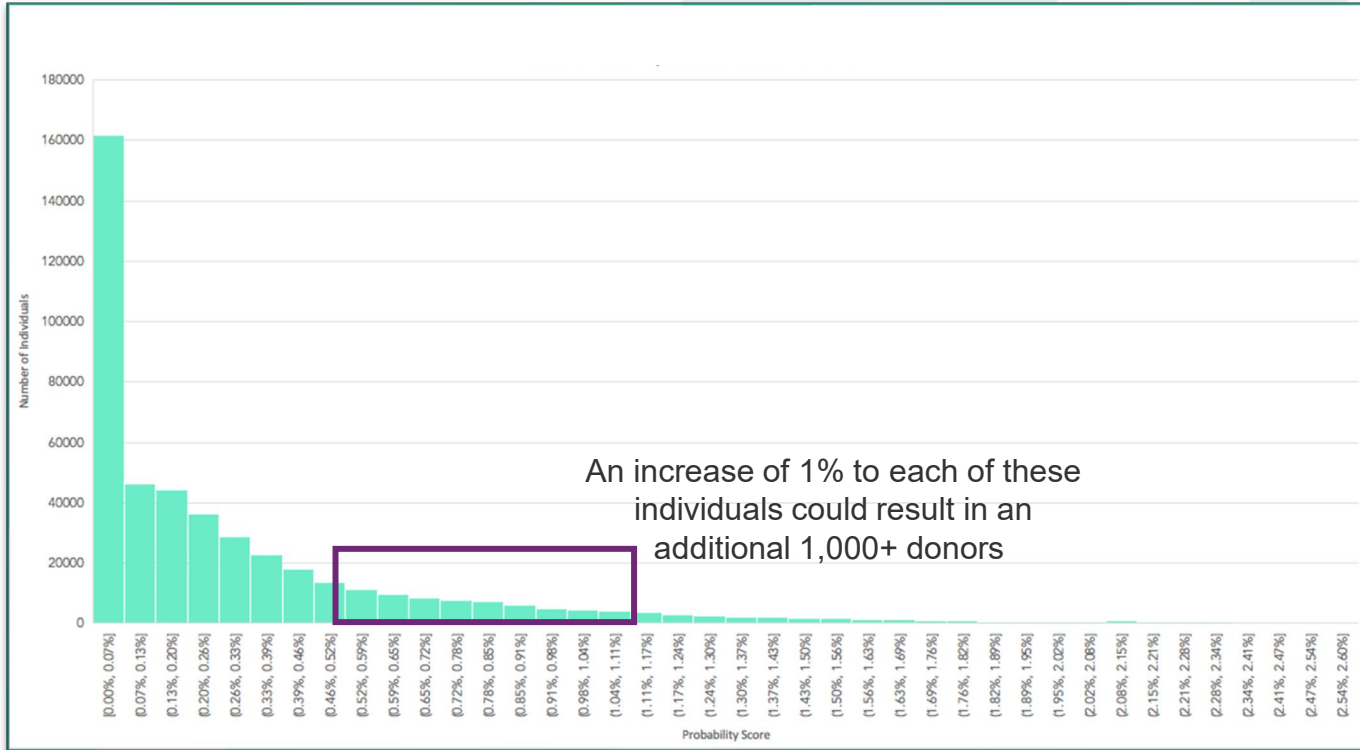
Top Importances

Sum Annual Giving Gifts: Previous FY	66%	Sum Student Fund Gifts: Cumulative 3 Years	1%
Sum Any Gifts: Two FY Ago	6%	Count Annual Giving Gifts: Cumulative 3 Years	1%
Sum Any Gifts: Three FY Ago	3%	Median Age	<1%
Avg Amt Any Gifts: Previous FY	2%	Usual Hours/Week Worked	<1%
Distance	1%	Sum Non Annual Giving Gifts: Two FY Ago	<1%

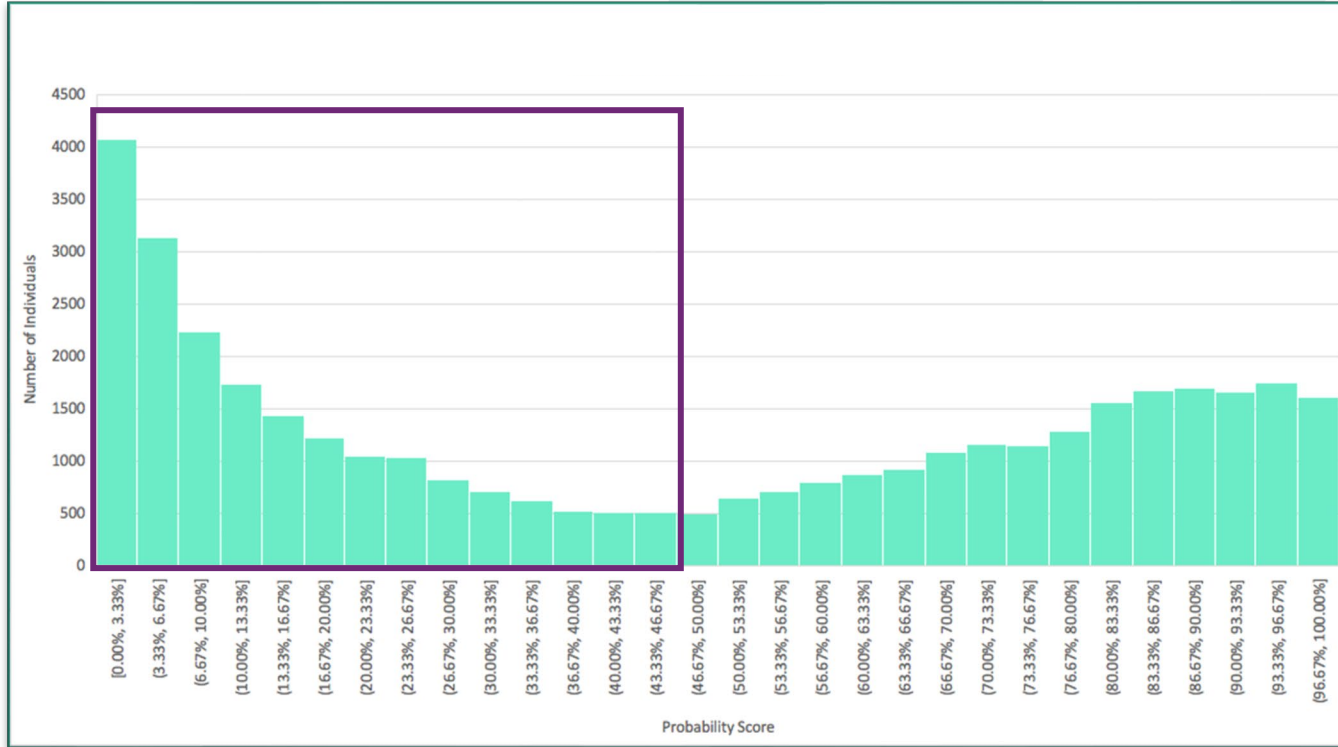
Even though probabilities are small for lapsed and non-donors, some are much more likely to donate than others



Slight increases in probability could have a big impact

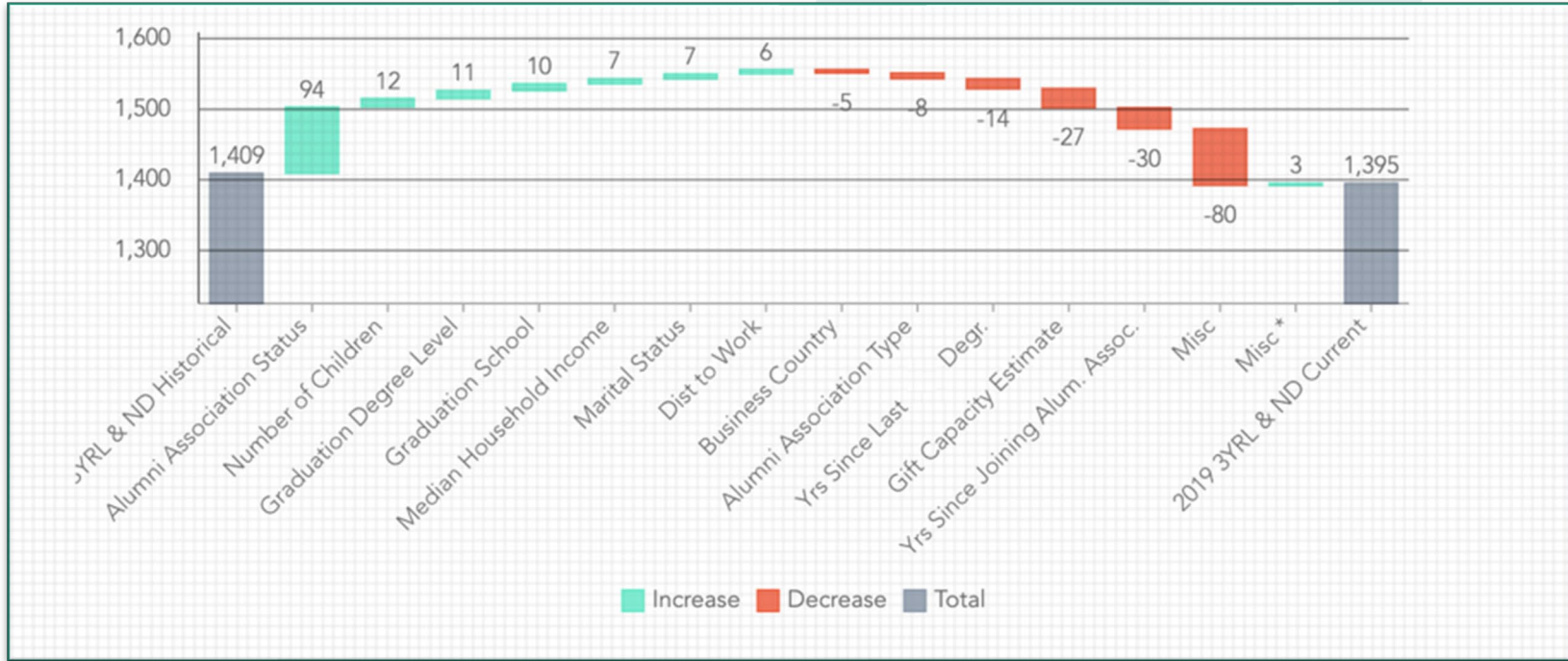


Focus on recent donors with low likelihood to identify donors who are at risk to churn



This Is Explainable AI

The impact of distinct variables on an individual's likelihood to give



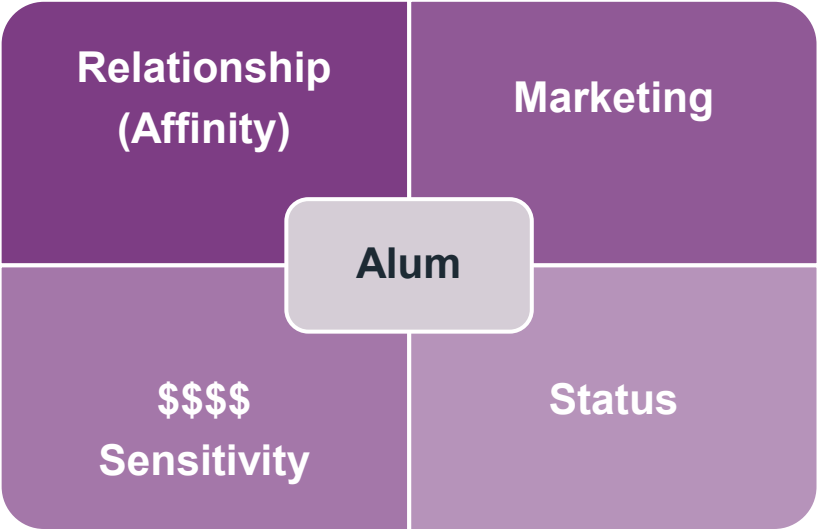
Looking Beyond Correlations

Total Average of Likelihood Score	
Capacity to Give	
R1 \$10M+	0.83%
R2 \$1M - \$9.9M	0.79%
R3 \$250K - \$999K	0.70%
R4 \$100K - \$249,999	0.59%
R5 \$25K - \$99K	0.50%
R6 \$10K - \$24K	0.42%
R7 \$2,500 - \$9,999	0.36%
R8 LESS THAN \$2,500	0.42%
R9 UNABLE TO RATE	0.01%
(blank)	0.10%
Grand Total	0.30%

Capacity to Give	Average of Likelihood Score Alumni Association Status	
	ACTIVE	INACTIVE
R1 \$10M+	1.00%	0.25%
R2 \$1M - \$9.9M	0.96%	0.30%
R3 \$250K - \$999K	1.04%	0.39%
R4 \$100K - \$249,999	0.95%	0.33%
R5 \$25K - \$99K	1.02%	0.31%
R6 \$10K - \$24K	1.00%	0.27%
R7 \$2,500 - \$9,999	1.02%	0.26%
R8 LESS THAN \$2,500	1.02%	0.31%
R9 UNABLE TO RATE	0.15%	0.00%
(blank)	0.86%	0.11%
Grand Total	0.95%	0.19%

Wrap up – What You Can Do “Back Home”

Data to Insights to Action



HIQ to Target
Personalized
Interaction

Next
Best Action

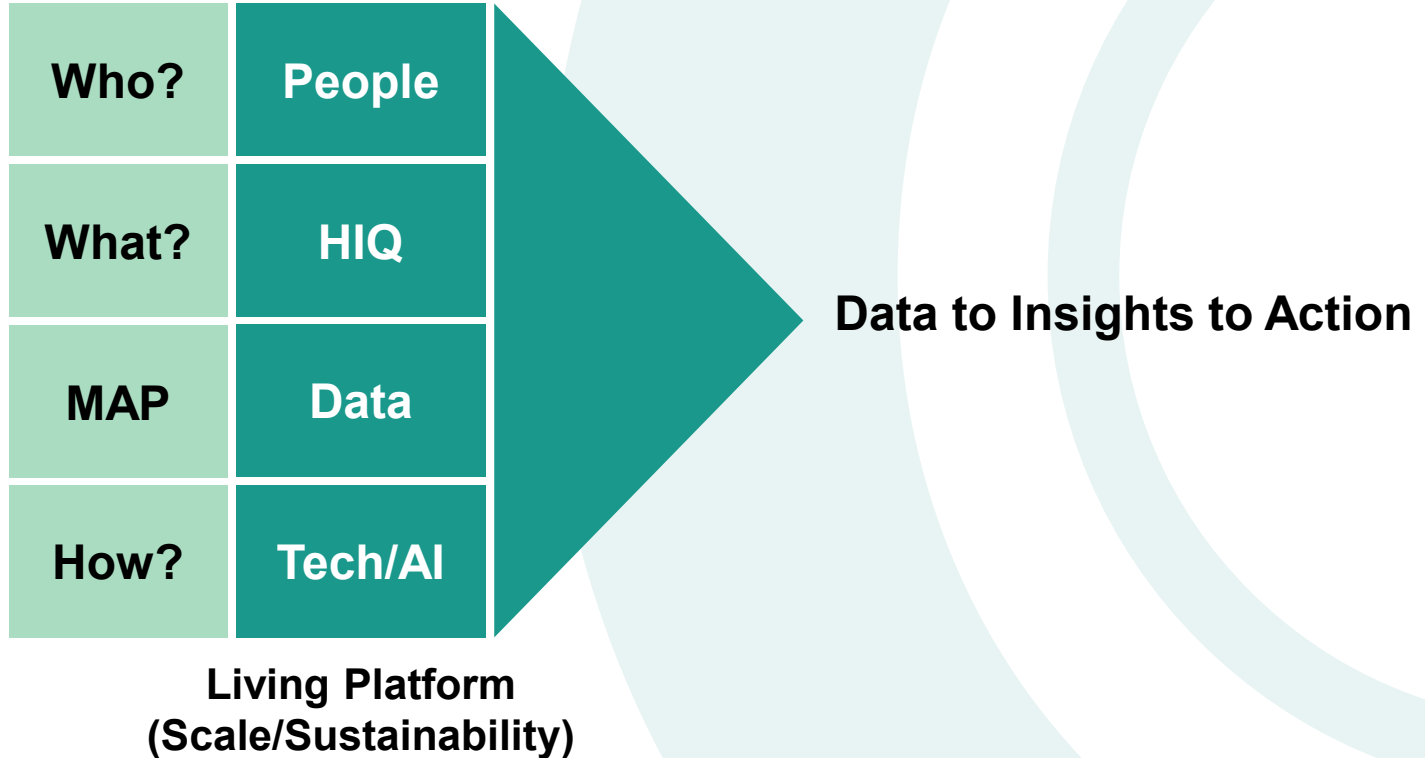


Behavior



Next Best Action

Insights to Action: Can you fill in the boxes for your organization?





Any Questions?

othotSM

A LIAISON  COMPANY

Higher Intelligence for Higher Education[®]

We are the future of higher ed advanced analytics - providing a clear vision of outcomes to make the best decisions possible for your students and your institution