

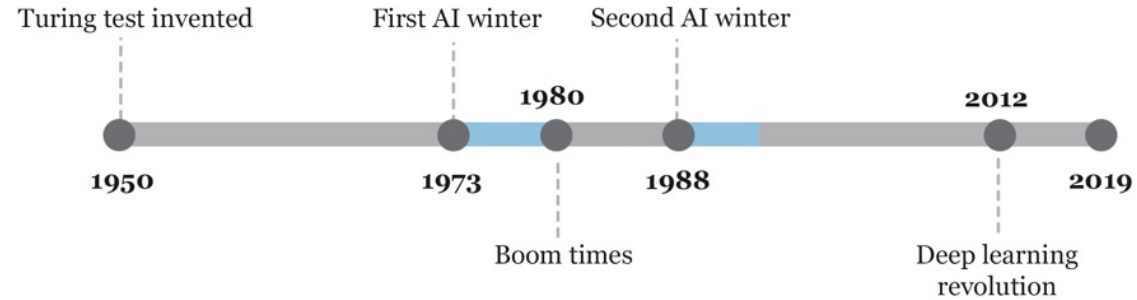
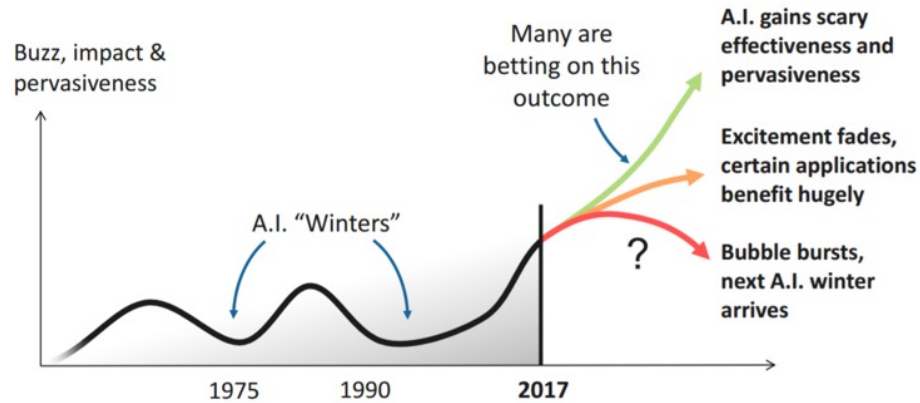
From Data to Decisions and Back

Babak Hodjat
CTO-AI

May 24th, 2022

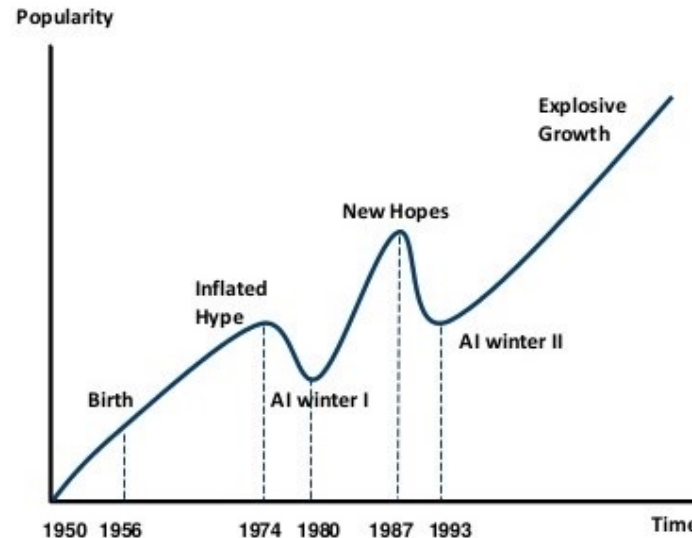
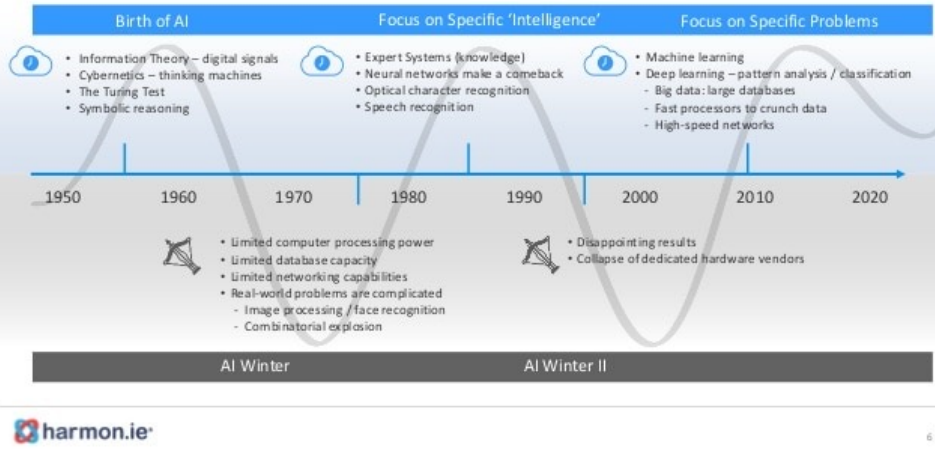
Another AI Winter?

AI is enjoying significant hype and investment



AI HAS A LONG HISTORY OF BEING "THE NEXT BIG THING" ...

An AI Timeline



Timeline of AI Development

- **1950s-1960s:** First AI boom - the age of reasoning, prototype AI developed
- **1970s:** AI winter I
- **1980s-1990s:** Second AI boom: the age of Knowledge representation (appearance of expert systems capable of reproducing human decision-making)
- **1990s:** AI winter II
- **1997:** Deep Blue beats Gary Kasparov
- **2006:** University of Toronto develops Deep Learning
- **2011:** IBM's Watson won Jeopardy
- **2016:** Go software based on Deep Learning beats world's champions

There is Always Room For AI

Science

Ideology

Known

Unknown

Software Engineering

AI

Programable

ML + Search + Heuristics

What Should we Expect from an AI System?

AI Should Tell Us:

What to do

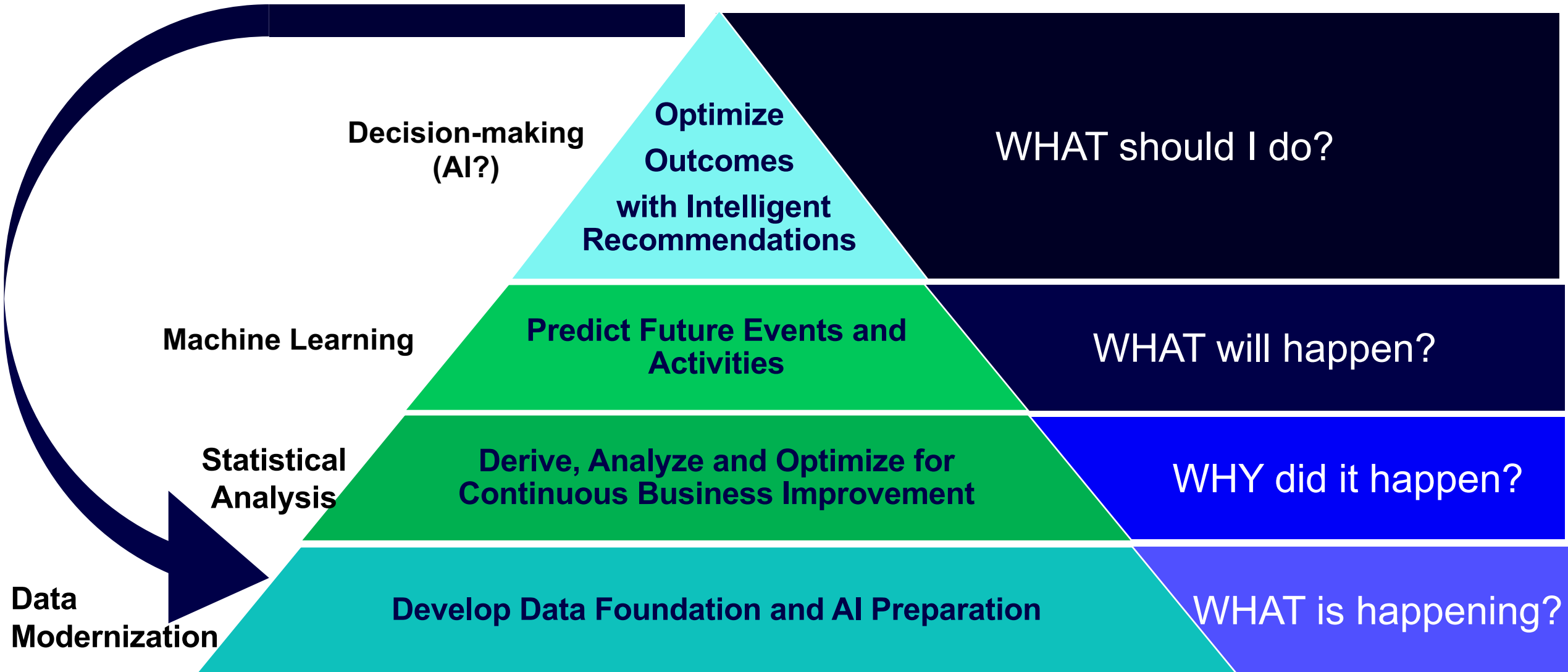
When to do it

What would happen if we do it

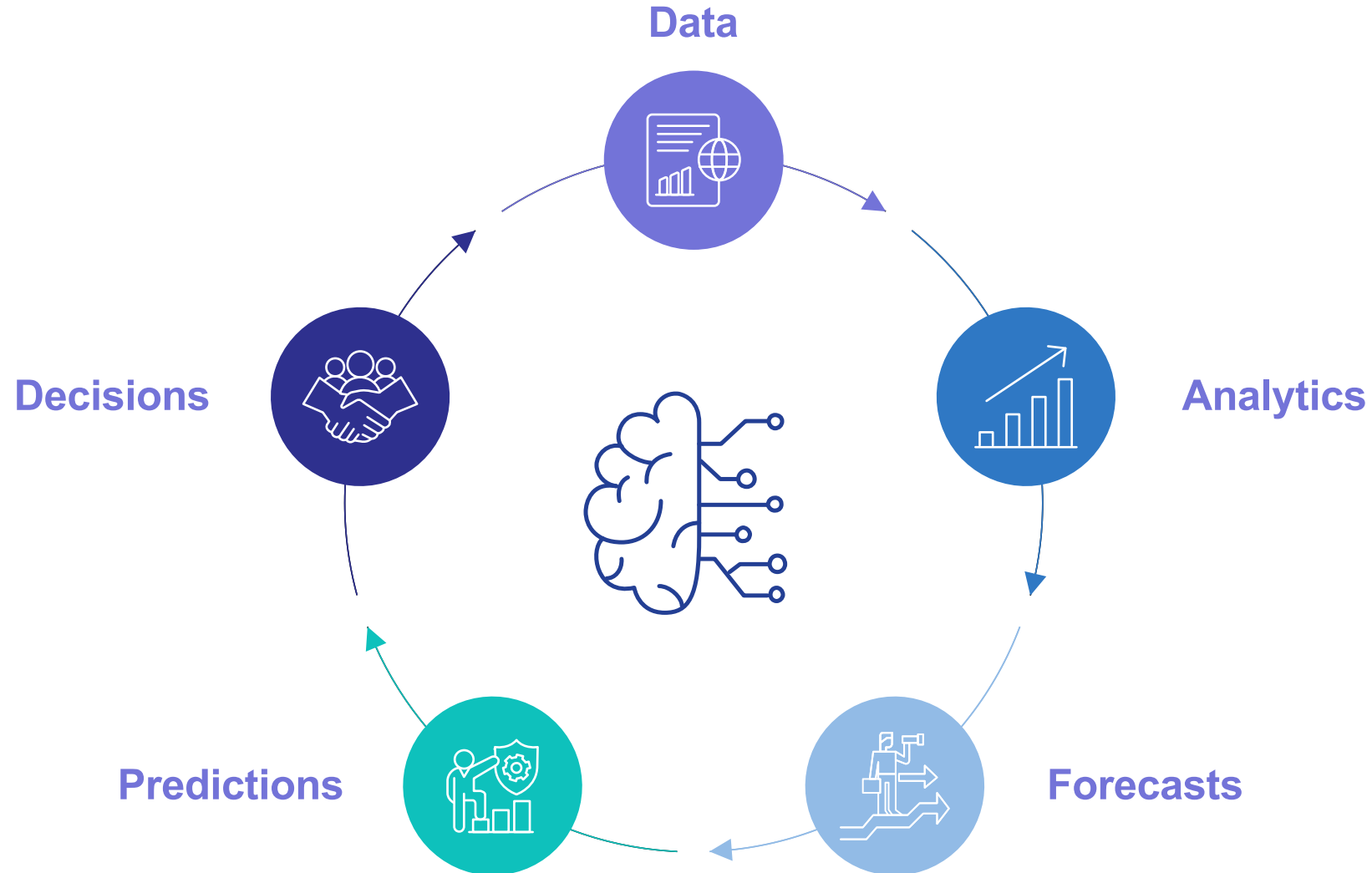
How much should we trust it

What would happen if we do something else...

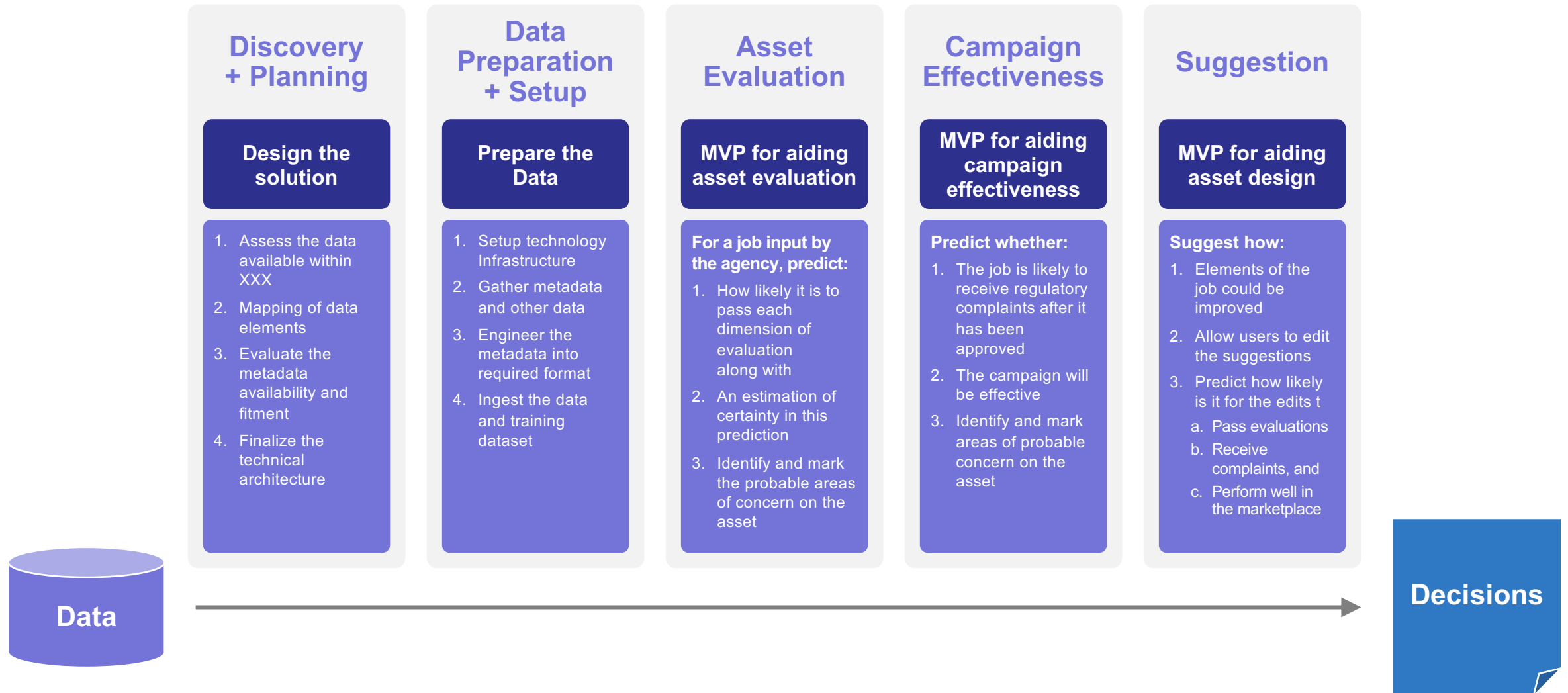
The AI Journey Should come Full Circle



The Scope of AI Enablement: Step-by-step vs Wholistic



What is wrong with this picture?



AI Hesitancy

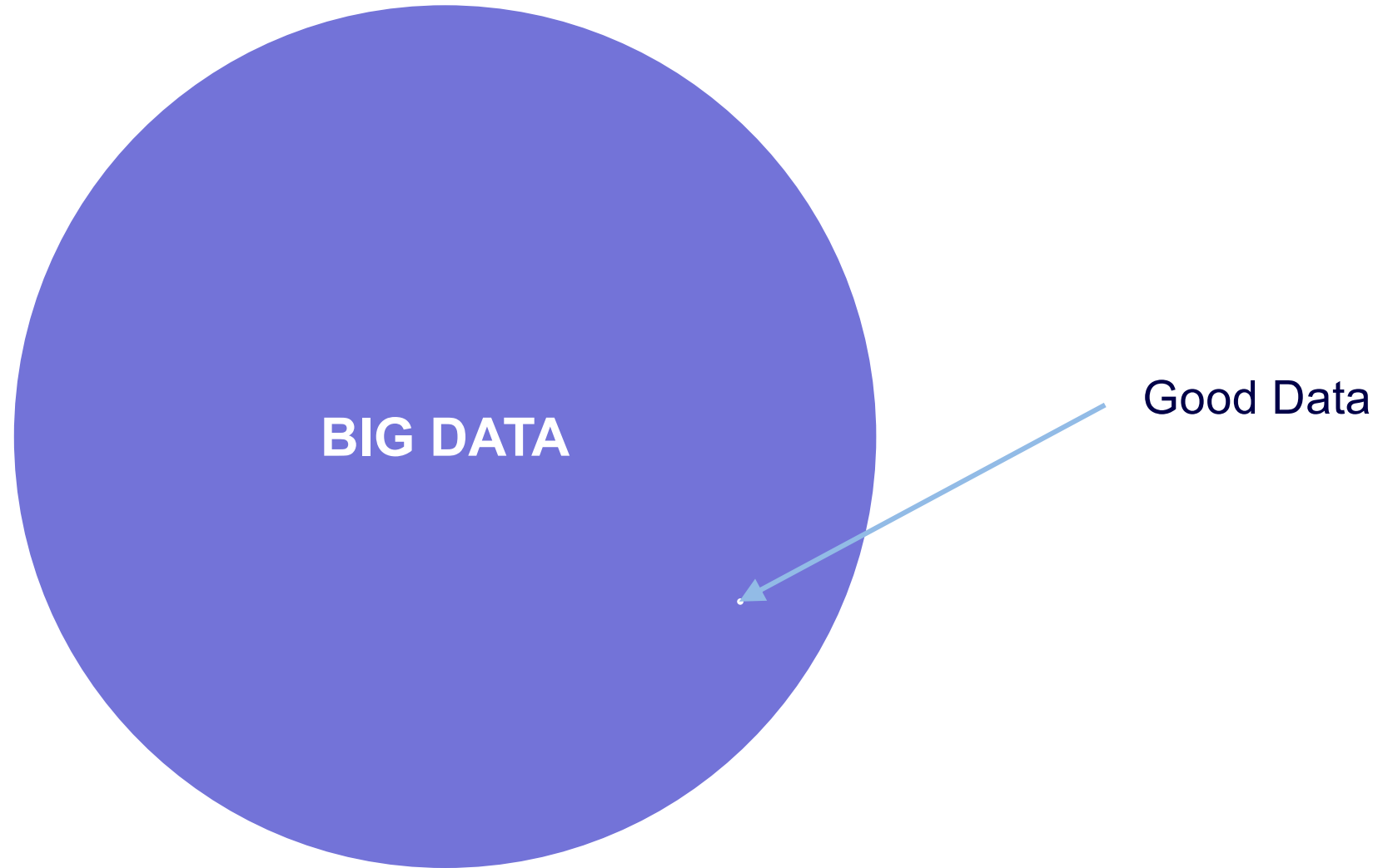


Typical excuses for deferring
or rejecting AI-enablement

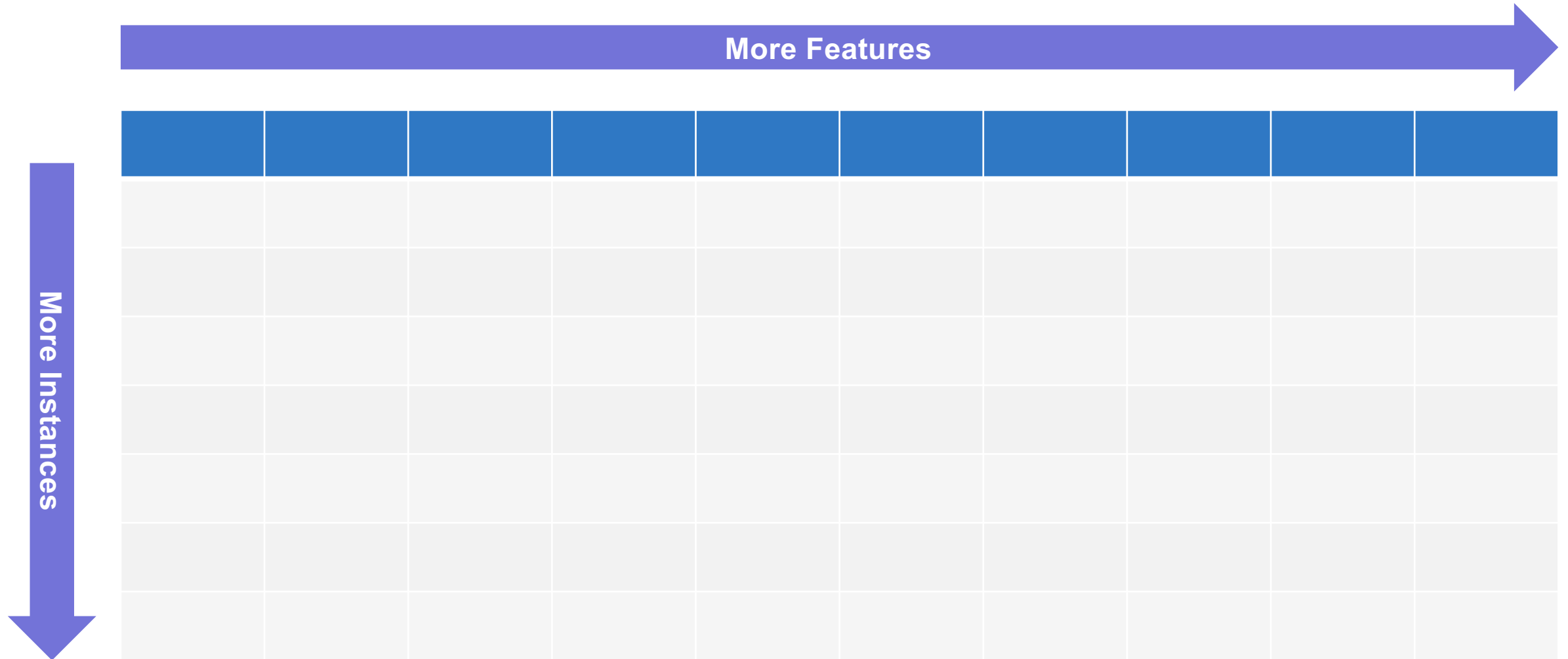
Our data is a mess/not ready

Timestamp	Rank 1	Rank 2	Rank 3	Rank 4	Rank 5
12/5/2016 13:18:46	Chris Stapleton, Golden Age	Kaytranada, 99.9%	Michael Kiwanuka, Love & Hate	Arkells, Morning Report	Hannah Georgas, For Evelyn
12/5/2016 13:19:50	Radiohead, A Moon Shaped Heart	James Blake, The Colour in White	Bon Iver, 22, A Million	Chance the Rapper, Coloring Book	Anderson .Paak, Malibu
12/5/2016 13:19:51	Frightened Rabbit, Painting of a Dove	Wilco, Schmilo	Weezer, White Album	Drive-By Truckers, American	Band of Horses, Why Are You
12/5/2016 13:19:55	Childish Gambino, "Awaken, My Love"	Basia Bulat, Good Advice	Kanye West, The Life of Pablo	Frank Ocean, Blonde	Tacocat, Lost Time
12/5/2016 13:20:09	Nada Surf, You Know Who You Are	The Jayhawks, Paging Mr. Phyllis	The Posies, Solid States	Purusa, Soundtrack	
12/5/2016 13:20:10	William Tyler, Modern Country	Hiss Golden Messenger, Hea	BJ Barnham, Rockingham	Mandolin Orange, Blindfaller	The Avett Brothers, True Sad
12/5/2016 13:20:15	22 A Million, Bon Iver	Coloring Book, Chance the R	We've Got It From Here, Thx	Moon Shape, Pop, Radiohe	The Colour in Anything, Jame
12/5/2016 13:20:19	The Posies, Solid States				
12/5/2016 13:20:28					
12/5/2016 13:20:31	ZAYN, Mind of Mine	Ariana Grande, Dangerous W	Solange, A Seat at the Table	Chance the Rapper, Coloring	dvsn, SEPT. 5TH
12/5/2016 13:21:01					
12/5/2016 13:21:07	BTS, Wings	Agust D, Agust D	Lindsey Stirling, Brave Enough	Babymetal, Metal Resistance	Red Velvet, Russian Roulette
12/5/2016 13:21:11	Hope Sandoval, Until the Hur	Madness, Can't Touch Us No	Suzanne Vega, Lover, Belove	Polica, United Crushers	Underworld, Barbara Barbara
12/5/2016 13:21:40	Beyonce, Lemonade	Radiohead, A Moon Shaped	Parquet Courts, Human Perfe	Cate Le Bon, Crab Day	Nels Cline, Lovers
12/5/2016 13:21:43	Zayn, Mind of Mine	Banks, The Altar	Beyonce, Formation	Rihanna, Anti	Jojo, Mad Love.
12/5/2016 13:21:59	Josh Ritter, Sermon on the R	Lori McKenna, The Bird & The	Sara Watkins, Young In All the Wrong Ways		
12/5/2016 13:23:06	Tancred, Out of the Garden	Jimmy Eat World, Integrity Bl	Sam Beam & Jesse Hooper, Lo	The Lumineers, Sloopatra	Sara Watkins, Young In All Th
12/5/2016 13:23:21	Anderson .Paak, Malibu	Toro y moi, Live from Mon	Kendrick Lamar, untitled un	Frank Ocean, Blonde	Chance the Rapper, Coloring
12/5/2016 13:23:59	Weezer, Weezer	Death Grips, Bottomless Pit	Jon Bellion, The Human Con	David Bowie, Blackstar	Explosions in the Sky, The W
12/5/2016 13:24:09	RVRB, FADED	Views, Drake	Latigo, Wilderado	TLOP, Kanye West	22 a million, Bon Iver
12/5/2016 13:24:16	Teenage Fanclub, Here	The Posies, Solid States	Phantogram, Three	Wilco, Shmilco	Monkees, good times
12/5/2016 13:25:10	RVRB, FADED	RVRB, FADED	RVRB, FADED	RVRB, FADED	RVRB, FADED
12/5/2016 13:25:11	AJJ, The Bible 2	Jeff Rosenstock, WORRY	Eric Bachmann, S/T	Joyce Manor, Cody	Kanye West, Life of Pablo
12/5/2016 13:25:26	Care Le Bon, Crab Day	David Bowie, Blackstar	Weaves, Weaves	Trentmøller, Fixion	Solange, a Seat at te Table
12/5/2016 13:25:26	RVRB, FADED	RVRB, FADED	RVRB, FADED	RVRB, FADED	RVRB, FADED
12/5/2016 13:25:28	Carseat Headrest, Teens of C	Beyonce, Lemonade	Radiohead, Moon Shaped Po	Margaret Glaspy, Emotions a	Dawes, We're All Gonna Die

We don't have enough Data

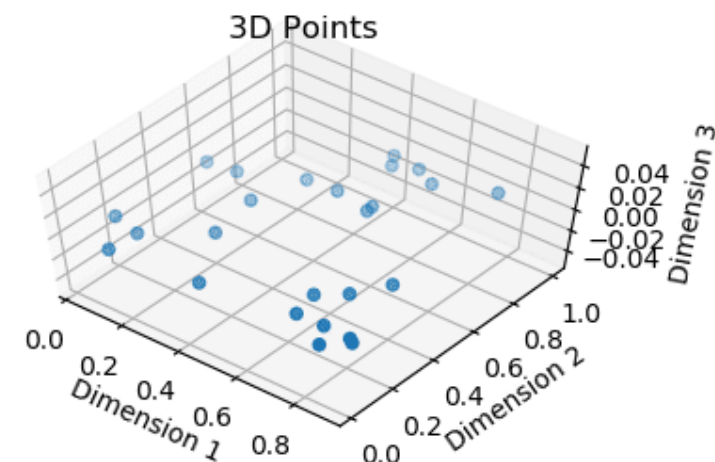
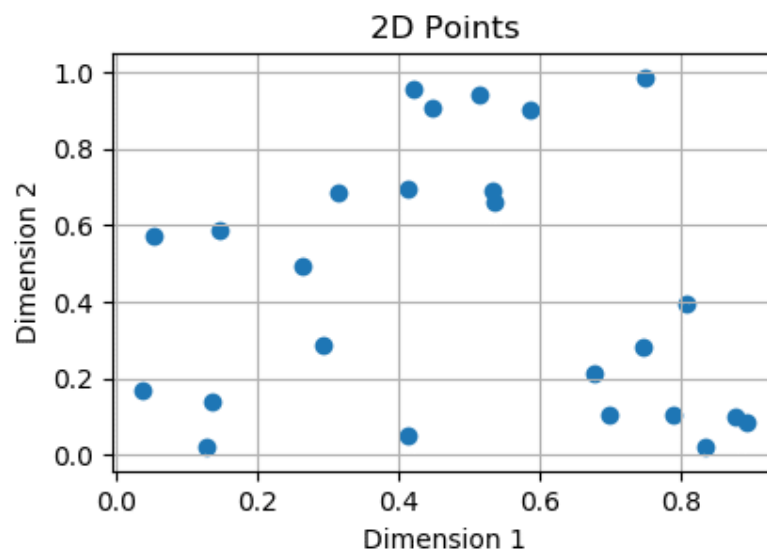
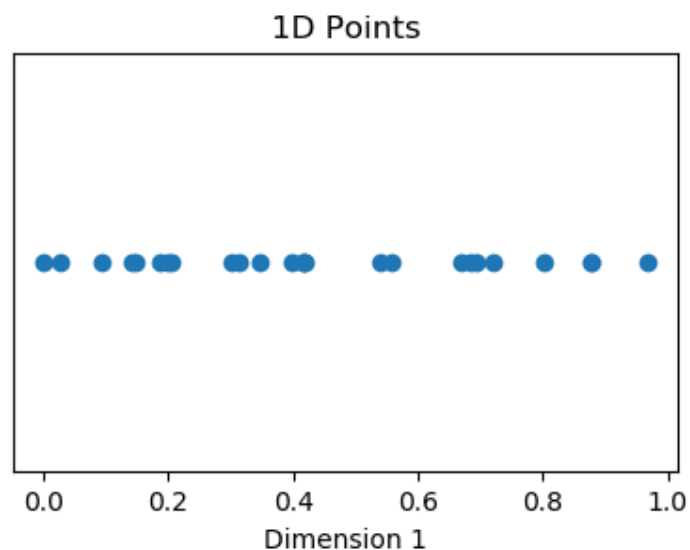


More data does not mean, more features!



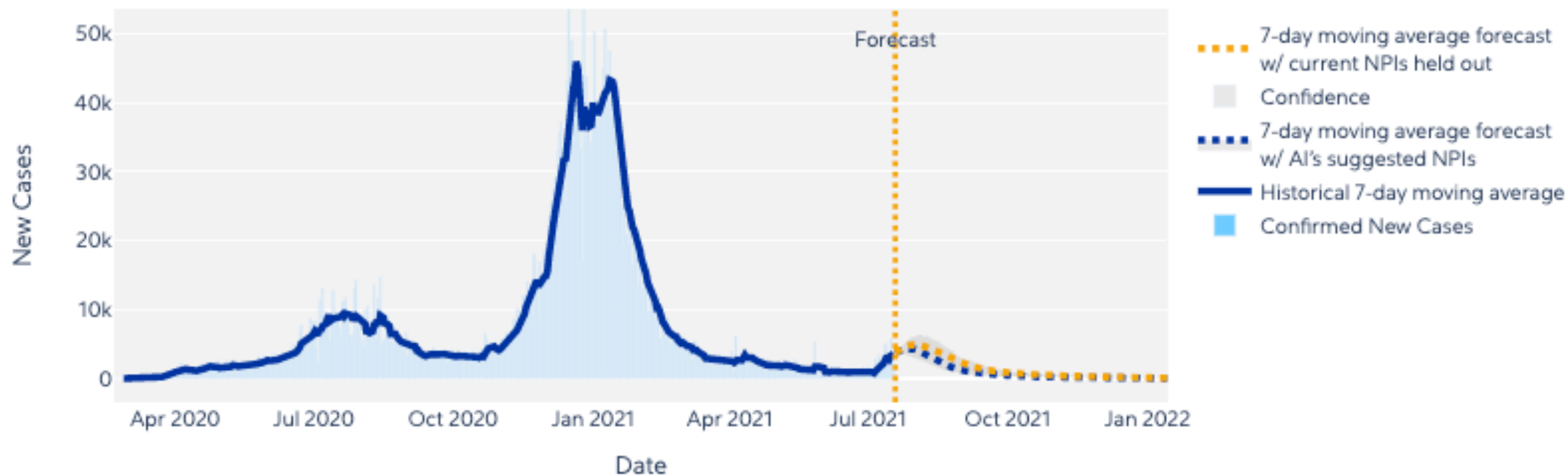
The Curse of Dimensionality

More features means exponentially more observations to avoid overfitting



Outcomes Timeseries Captures Effect of Missing Features

AI's Predicted New Cases per Day in United States / California

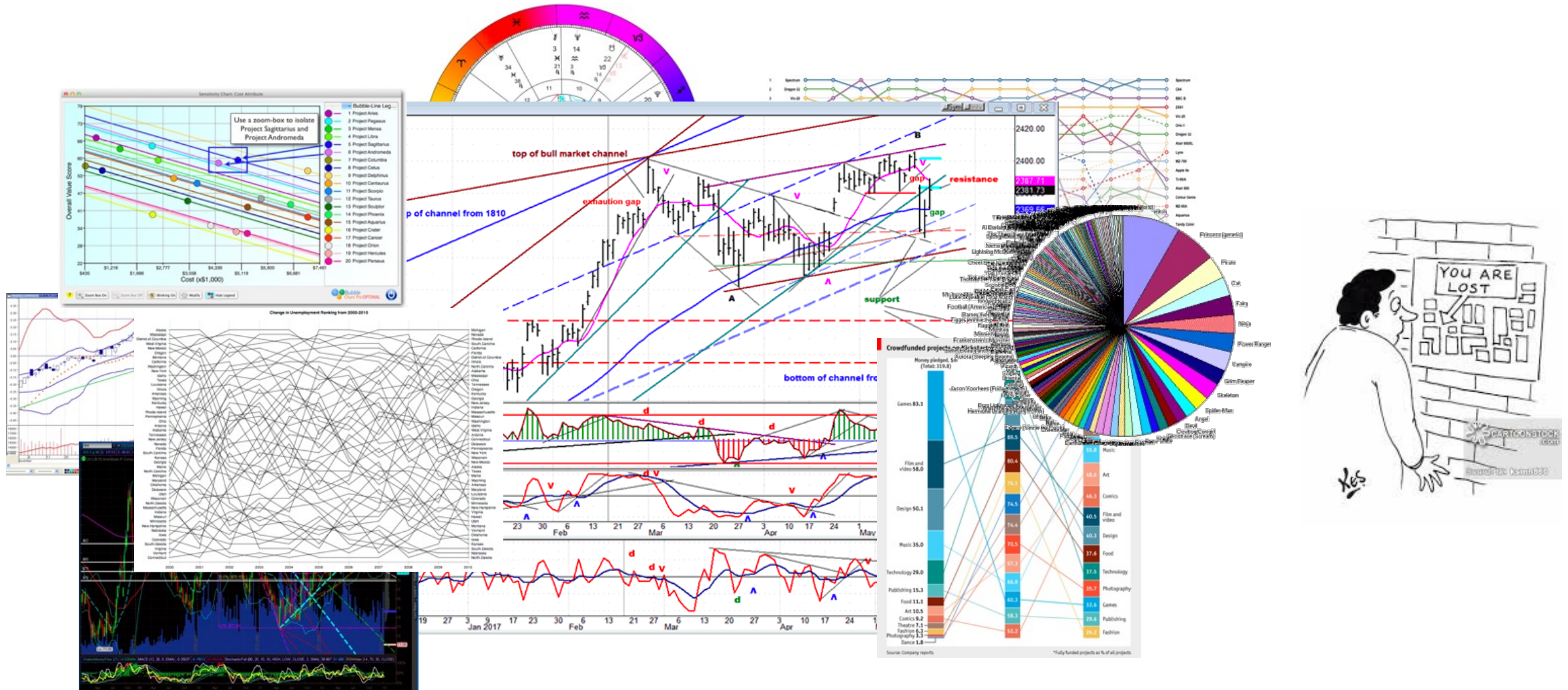


No One-Size-Fits-All AI

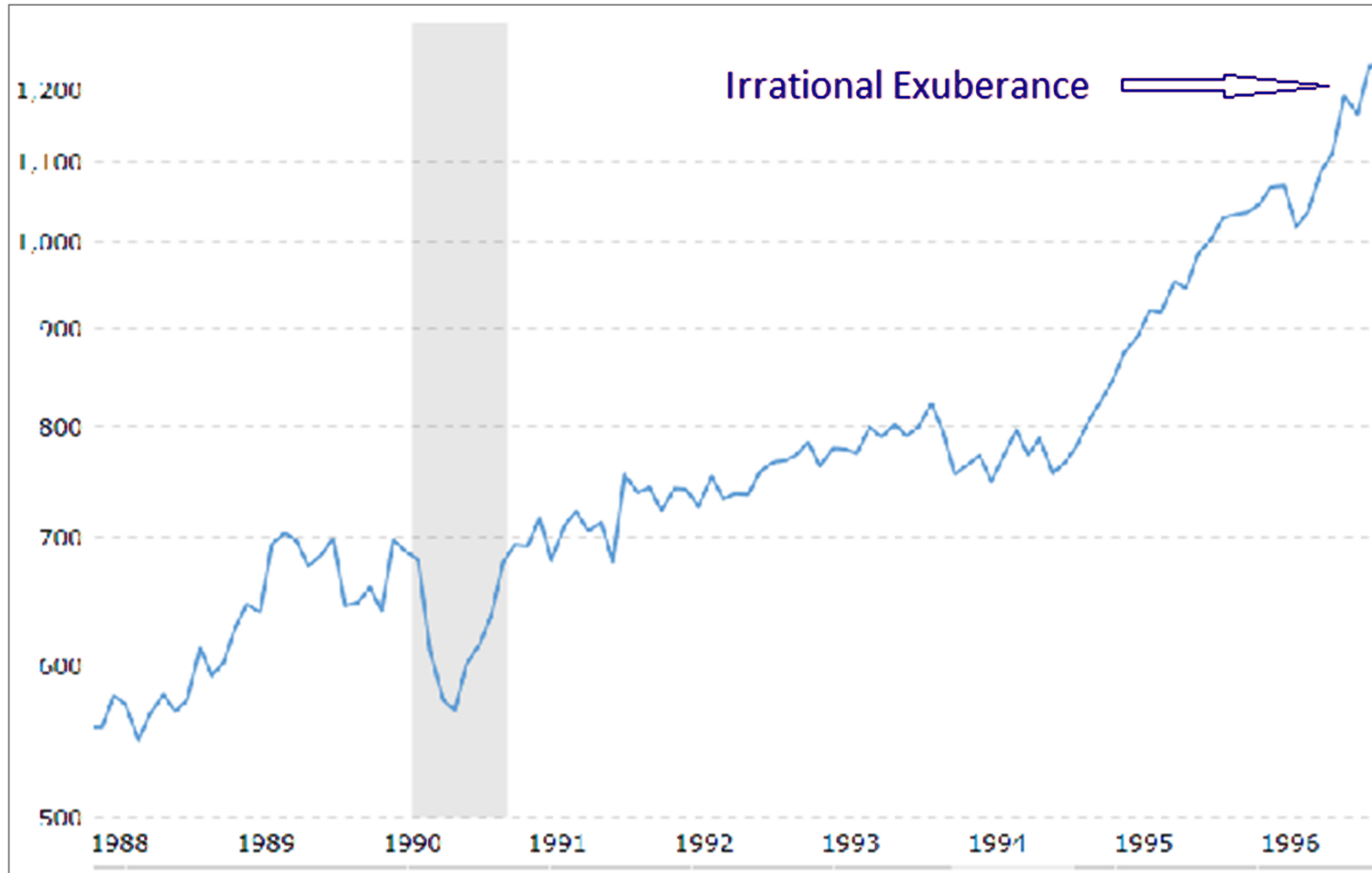
Domain Specific Models are the new IP



Need to get Insights First



What if Insights Show we are Good Already?



Human Decision-Making is Better



All we need is better insights - we can make the decisions

Projections Show we're Missing Revenue Targets!



Should we cut costs?



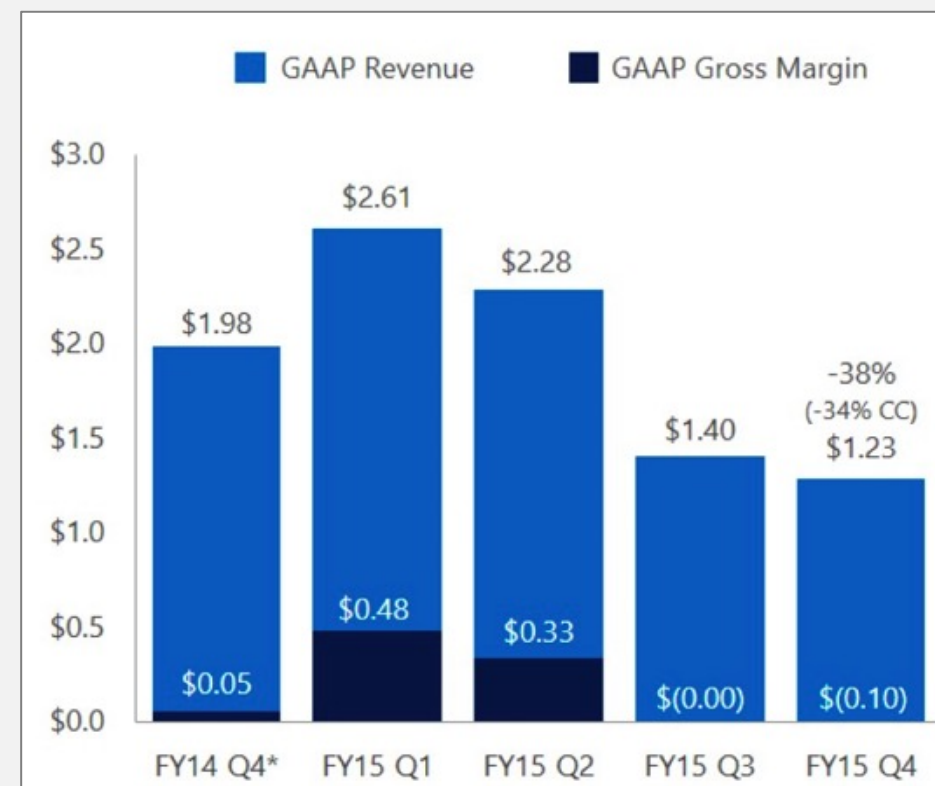
Sell at a discount?



Retarget sales?



Invest in growth?



We care about sales – Cost is for a different department

There is never a single objective!

We often want to:

- Maximize revenue, and
- Maximize sales, while
- Minimizing cost, and,
- Minimizing risk



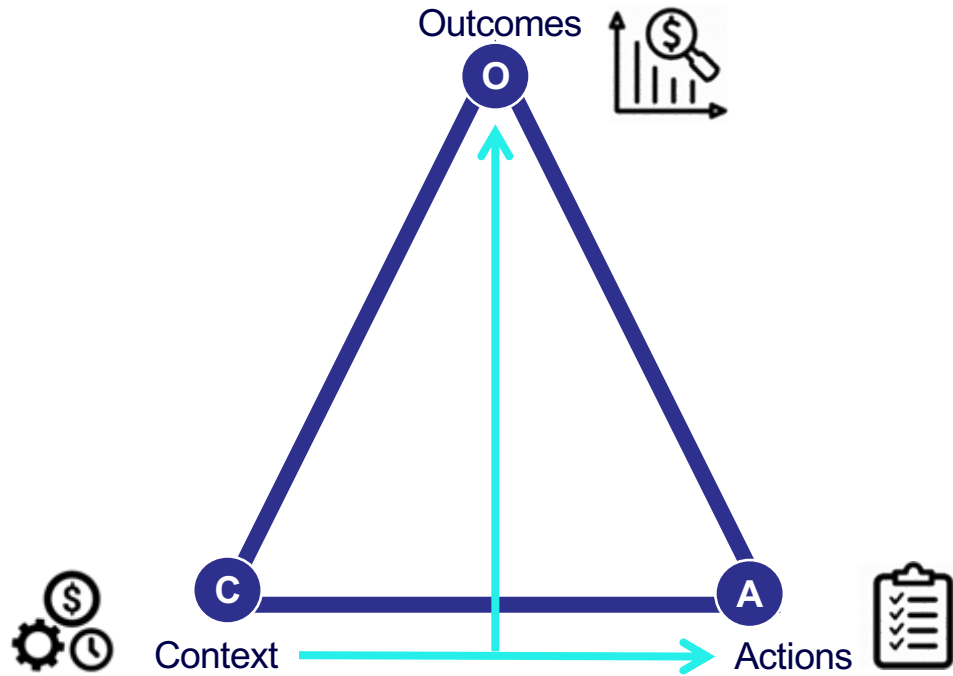
Worried about bias and ethics of using AI

© MARK ANDERSON, WWW.ANDERSTOONS.COM



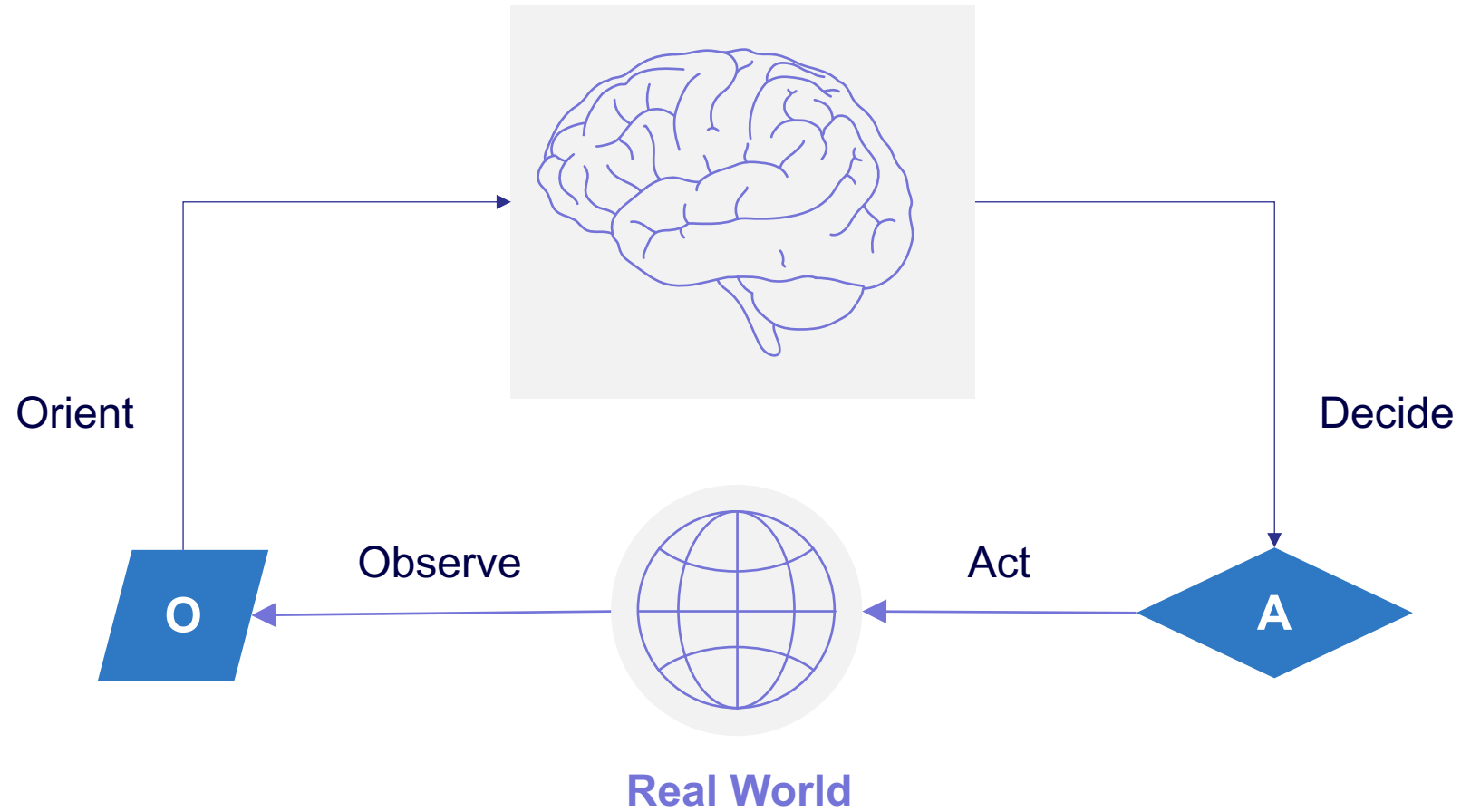
"And finally this is Phil Goodwin. He's part of the problem."

We make Decisions all the time

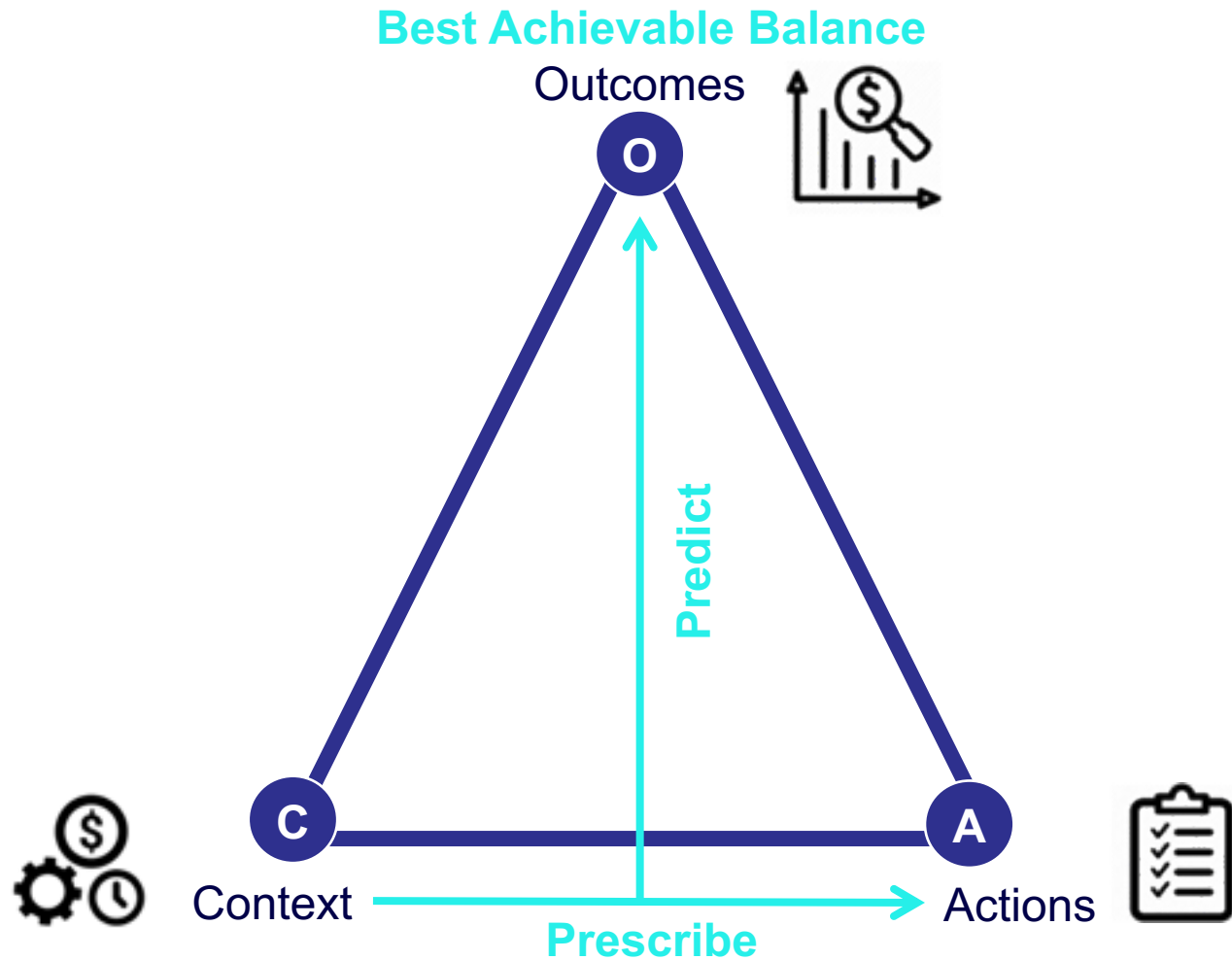


- Given a **context** (C)
- Which **actions** should I take (A)
- To optimize some **outcomes** (O)?

Adaptive Decision-Making: The OODA Loop



How do we make decisions?

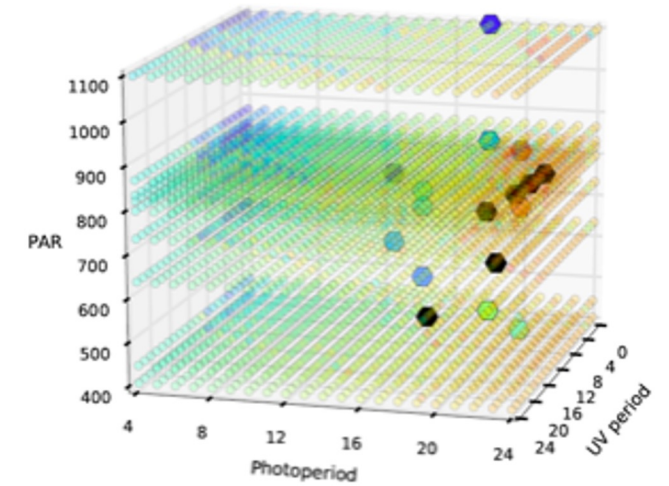
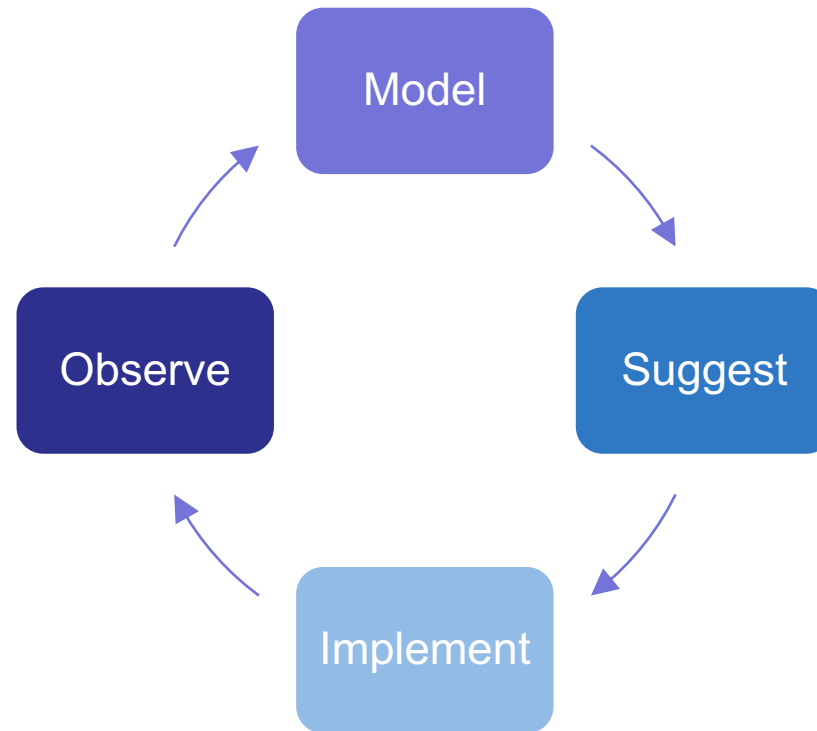


- Given a **context** (C)
- Which **actions** should I take (A)
- To optimize some **outcomes** (O)?

What we can't change

What we can change

AI-Enabled Decisioning (and Data Collection)



Important Questions AI can Answer

Predictor:

What will happen?

Prescriptor:

What should we do?

Uncertainty:

How much can we trust it?

Scratchpad:

What-if we do something else?

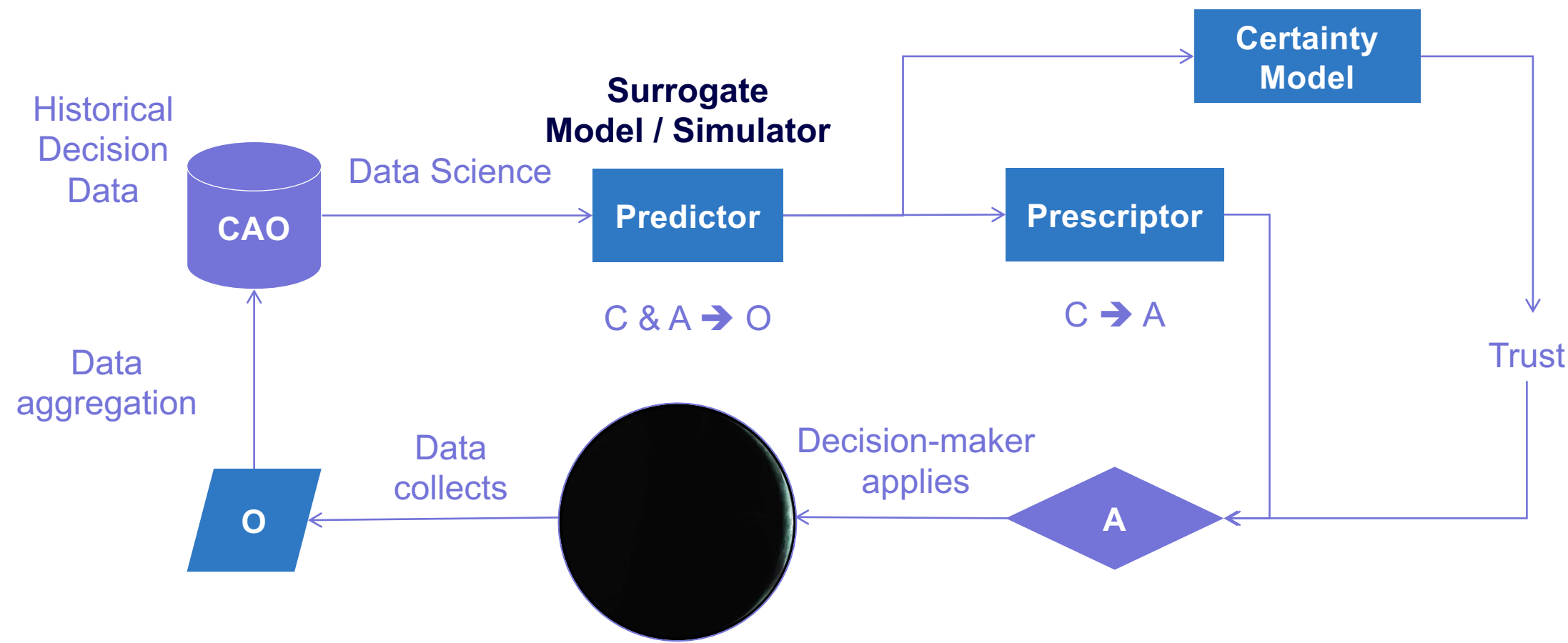
Impact Estimation (Retrospective):

What if we had done something else?

Multiple-Objectives:

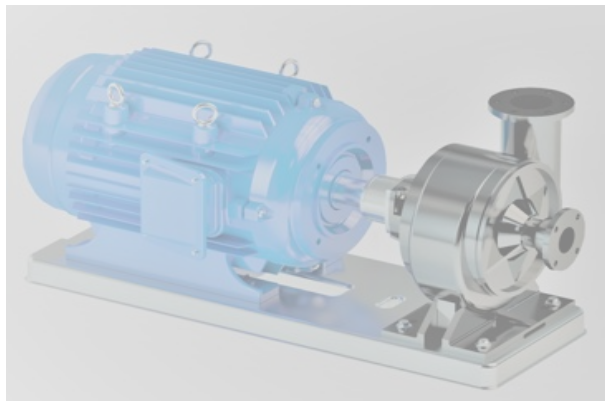
What if we want a different balance
of outcomes?

How it all comes together



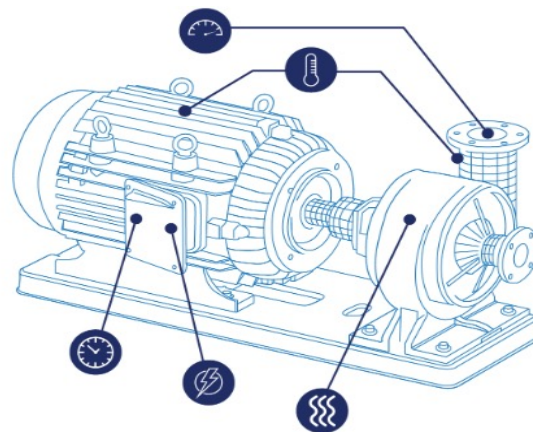
Digital Twins

Physical environment



- Design Systems
- Enterprise Data
- Sensor Data
- Historical data
- Image data
- Acoustic data
-

Digital Twin Instance



- Vibration
- Heating Temperature
- Flow Speed
- Viscosity
- Vapor pressure
- Power Consumption monitor
- Chamber humidity
- Current and voltage
-

Prediction and Optimization

- Simulation
- What-if scenarios
- Prediction and prescription
- Optimization
- Predictive maintenance
- Remaining useful life

Rigid Simulations

Make Assumptions

- May be unavailable
- May be wrong
- May be inaccurate
- May be insufficient

Create Model

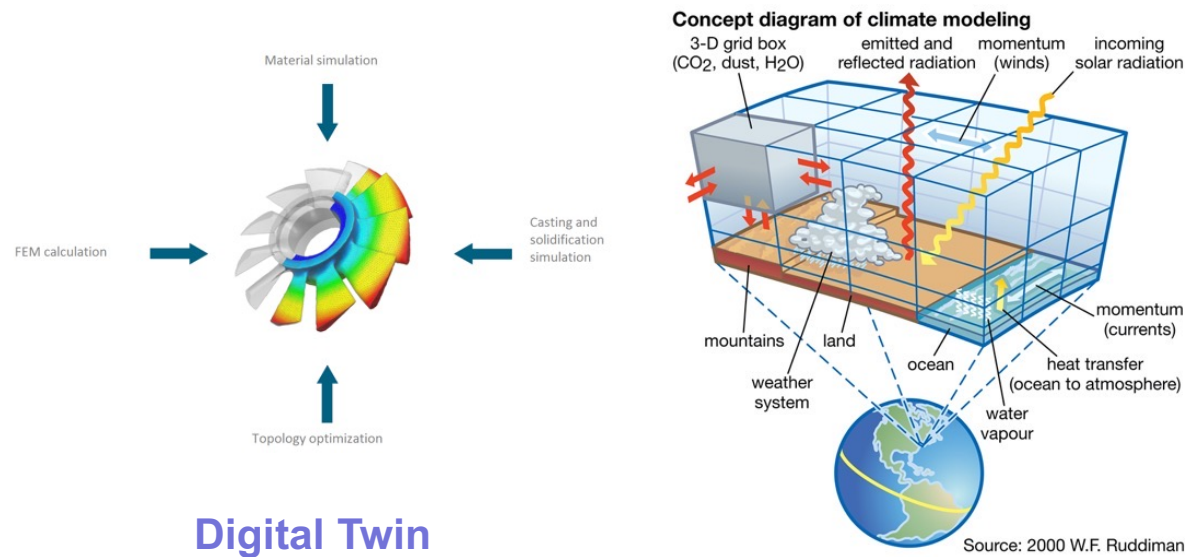
- Requires knowledge of mechanisms and dynamics

Run Model

- Often Expensive at scale
- Often inaccurate in miniature

Interrogate Model

- Rigid



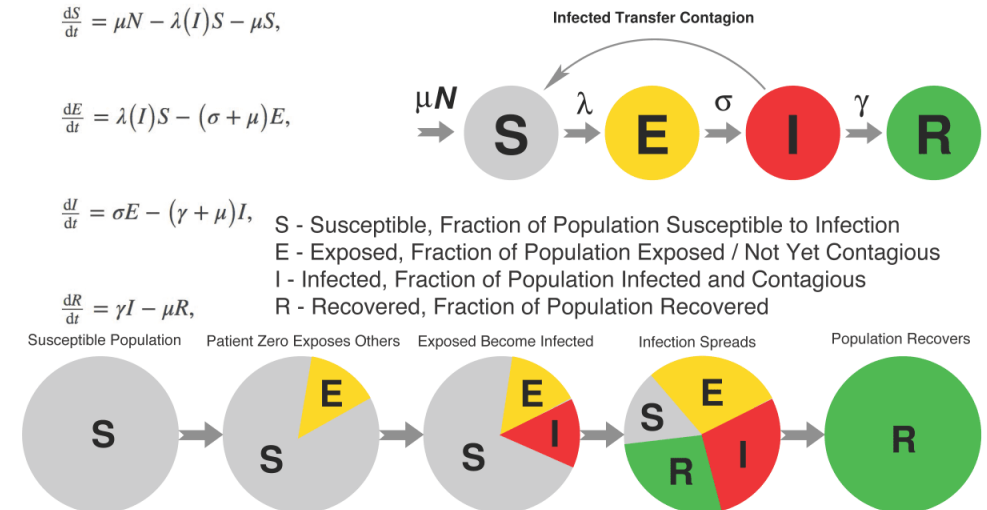
SEIR: Susceptible Exposed Infected Recovered Compartment Model

$$\frac{dS}{dt} = \mu N - \lambda(I)S - \mu S,$$

$$\frac{dE}{dt} = \lambda(I)S - (\sigma + \mu)E,$$

$$\frac{dI}{dt} = \sigma E - (\gamma + \mu)I,$$

$$\frac{dR}{dt} = \gamma I - \mu R,$$



ML Models

Robust

Data Driven

Does not require (as much) knowledge of the domain

Can be applied at scale

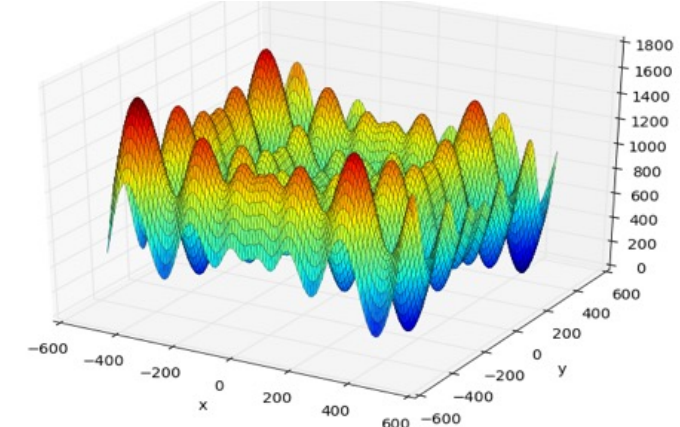
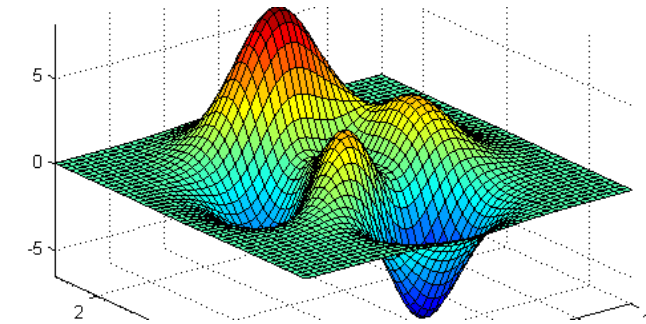
BUT not enough

Data-hungry

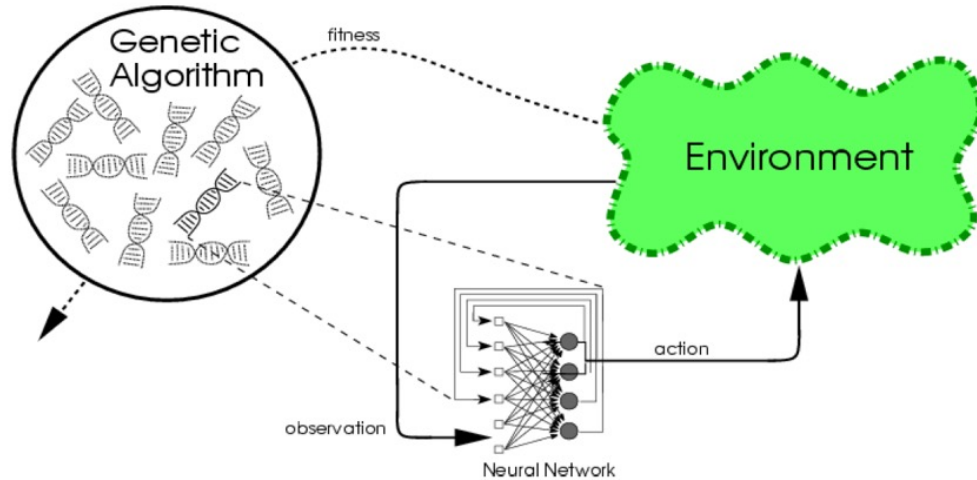
Training could be expensive/time consuming

Not adaptive

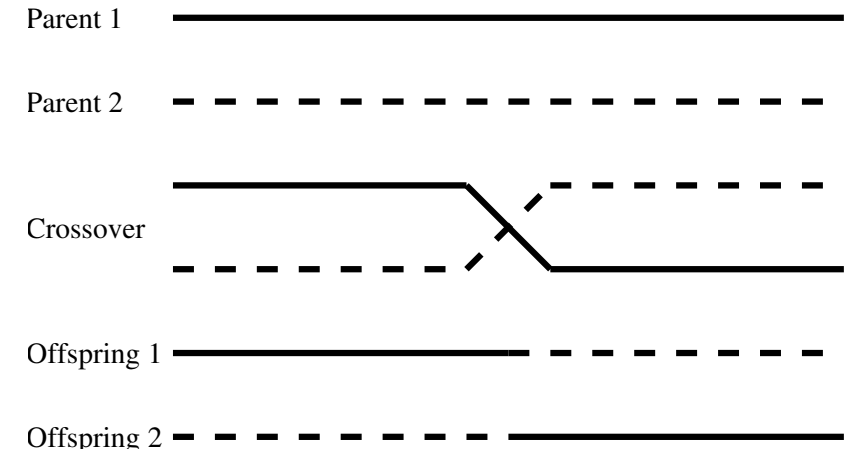
Gradient Descent is not good at large, deceptive, multi-objective search spaces



Why use Evolutionary AI?



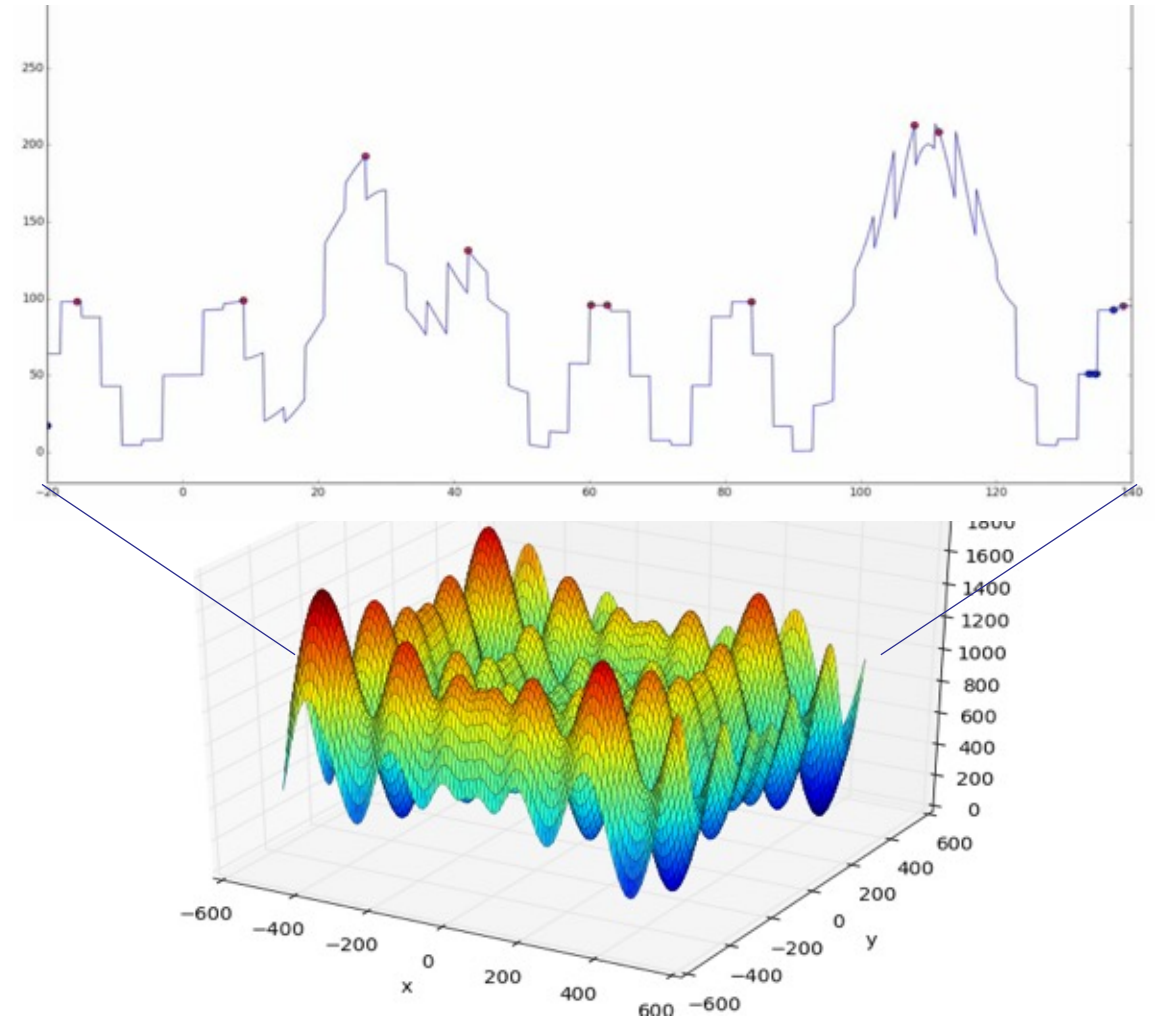
- A population of encodings (e.g., lists or trees)
- Decoded into individuals that are tested
- Good individuals retained, bad thrown away



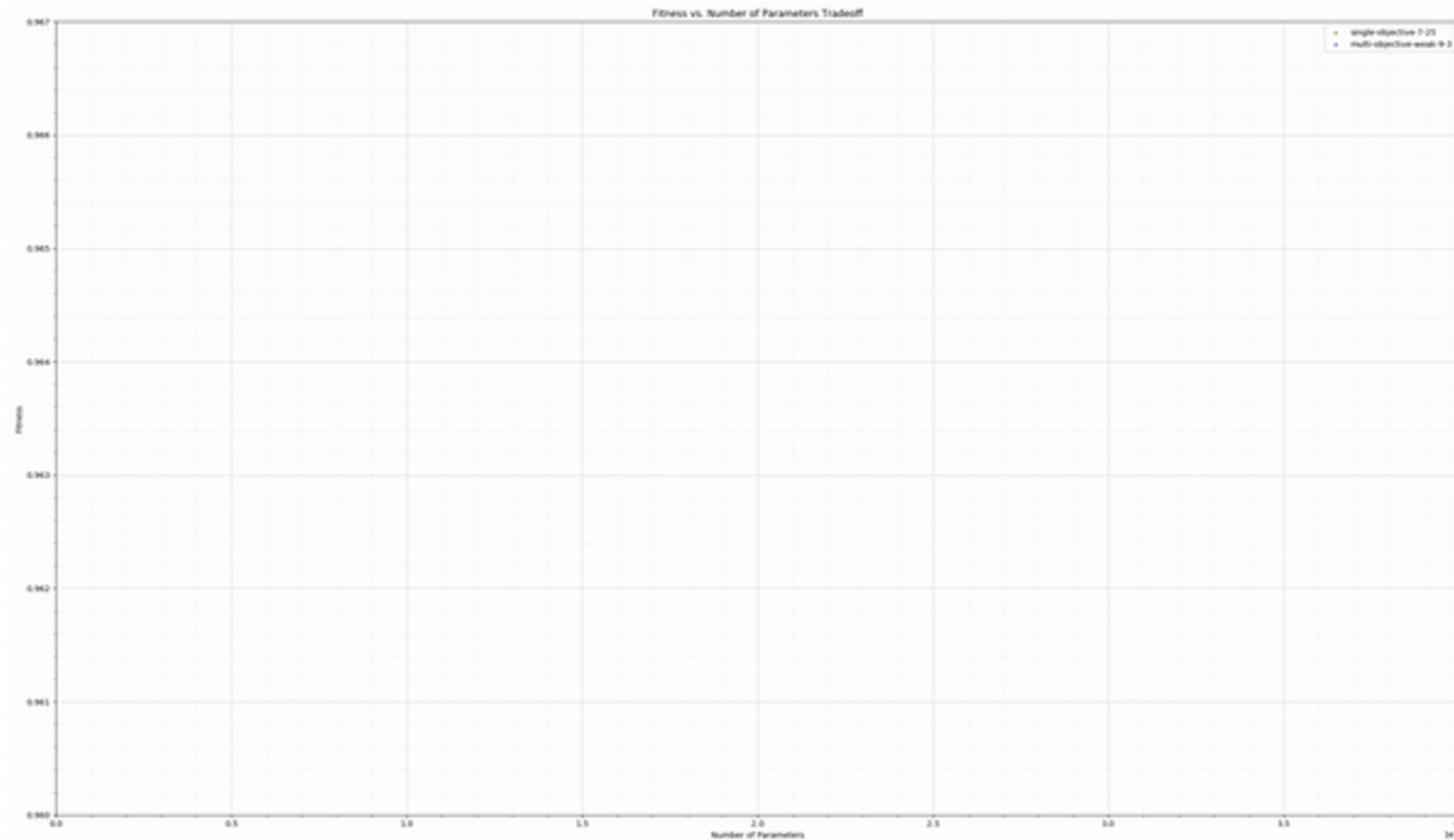
- New individuals generated from the parent encodings
 - Crossover: Combine building blocks from two parents
 - Mutation: Create new building blocks

Evolution is Efficient

- Works in large spaces (e.g., 2^{2^70})
- Multiple variables optimized at once (e.g., up to 1B)
- Multiple objectives and novelty get around deception

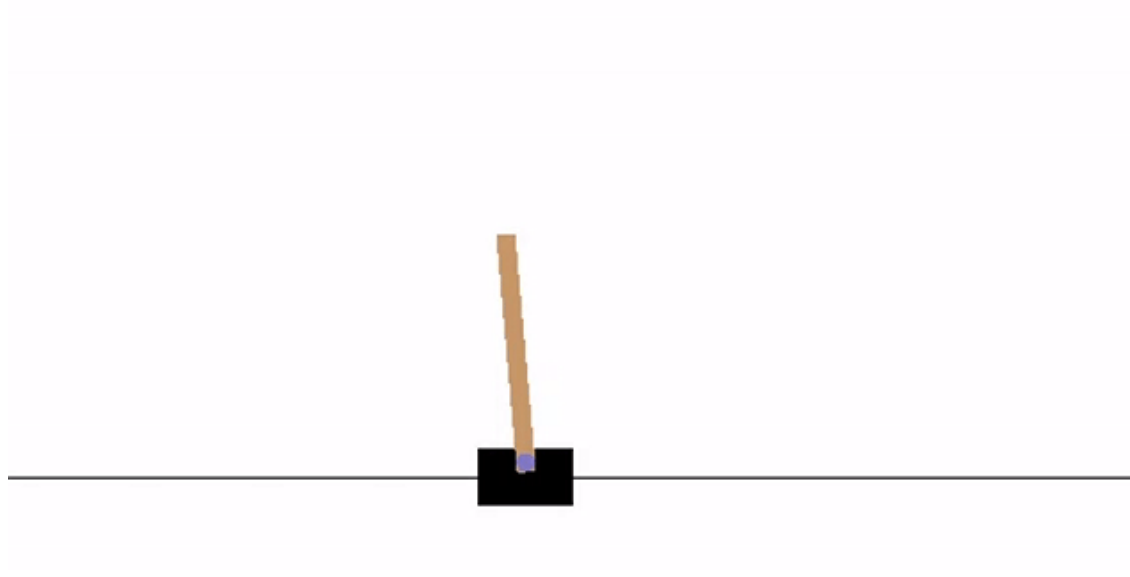


Evolution can handle Multiple Objectives

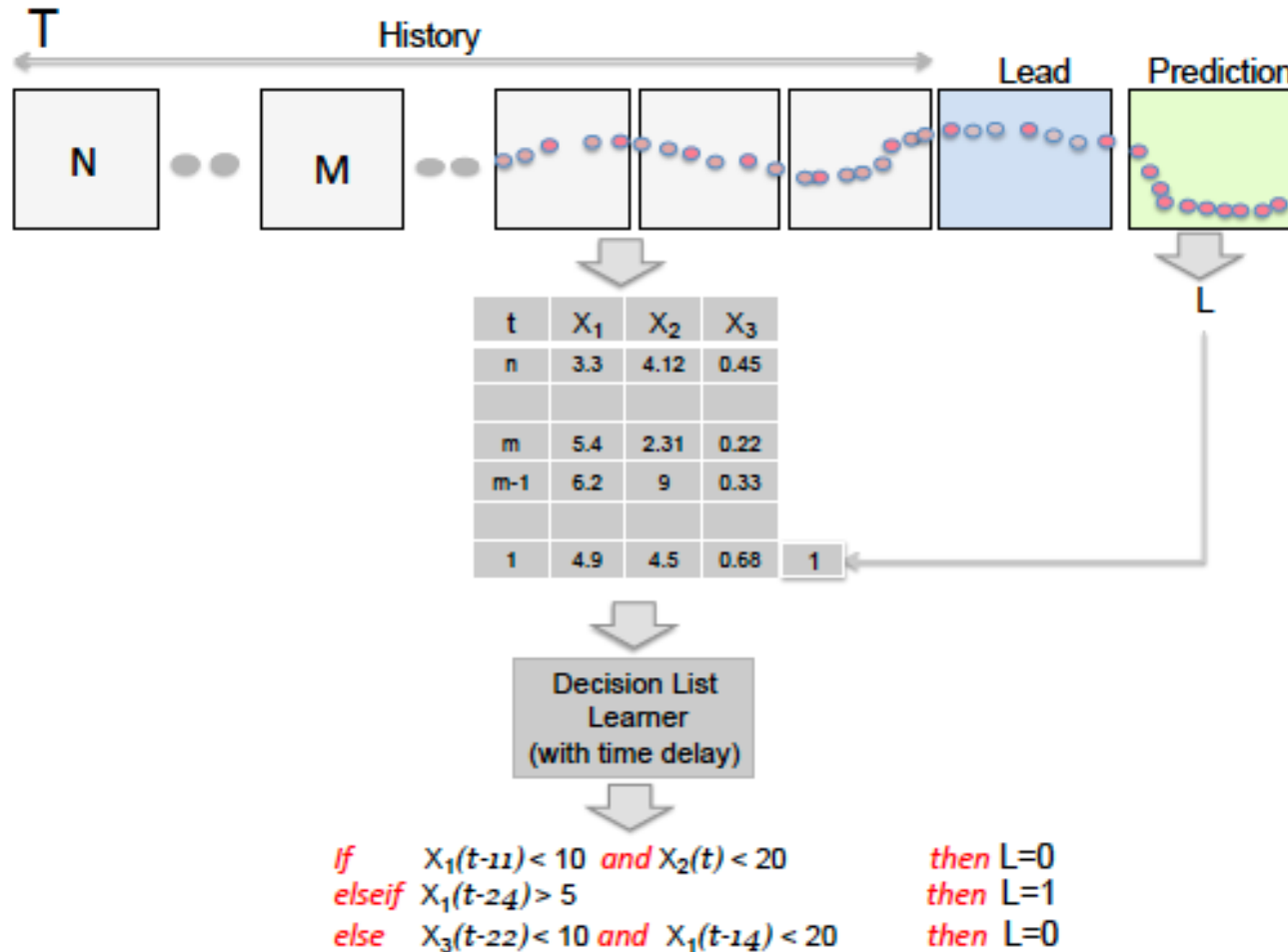


- Animation: Pareto front by generation for single-objective (green) vs. multi-objective (blue)
- Single-objective focuses on improving **largest networks**
- Multi-objective focuses on improving the **entire curve**
- Result: Multi-objective finds much smaller models for the majority of performance values.

Evolution Can Create Explainable Solutions



- 3_53 Rules agent:
- <22978> $(0.11 * S_1^3 < 0.87 * S_2)$ ----> **LEFT**
- <23022> Default Action: **RIGHT**



Each rule is a variable length conjunction of conditions with an associated class prediction. In the evaluation each condition compares a lagged value or the current value of the time series to a threshold (decision boundary). Decision lists in LEAF have a variable number of rules and conjunctive clauses in each rule.

This representation is different from many other classifiers e.g. Decision Trees, simple Decision Lists, Support Vector Machines and Logistic Regression, which requires every time lagged value or an aggregate to be explicitly presented as a different feature.

Diabetes Rules

1. $(\text{diag_1_injury} \leq 2.2 * \text{diag_1_respiratory}) \text{ and } (\text{diag_2_neoplasms} > 33) \rightarrow 18$
2. $(\text{age_}[60-70) \geq 1.2 * \text{age_}[10-20)) \rightarrow 7$
3. $(\text{diag_2_diabetes} \geq 0.82 * \text{diag_1_digestive}) \rightarrow 17$
4. $(\text{admission_source_id_Court/Law Enforcement} \geq 1.045 * \text{diag_3_circulatory}) \rightarrow 9$
5. $(\text{age_}[80-90) \geq 2.29 * \text{admission_type_id_Newborn}) \rightarrow 14$
6. $(\text{age_}[30-40) \leq 0.47 * \text{age_}[50-60)) \text{ and } (\text{admission_type_id_Newborn} \leq 0.47 * \text{age_}[50-60)) \text{ and } (\text{diag_2_respiratory} \geq 1.86 * \text{diag_3_musculoskeletal}) \text{ and } (\text{diag_1_diabetes} \geq 1.42 * \text{age_}[30-40)) \text{ and } (0.18 * \text{gender} \geq 0.38 * \text{race_Asian}) \text{ and } (0.12 * \text{race_AfricanAmerican} \geq 0.15 * \text{admission_source_id_Court/Law Enforcement}) \text{ and } (0.03 * \text{race_Hispanic} \leq 0.76 * \text{admission_source_id_Emergency Room}) \rightarrow 15$
7. Default 8

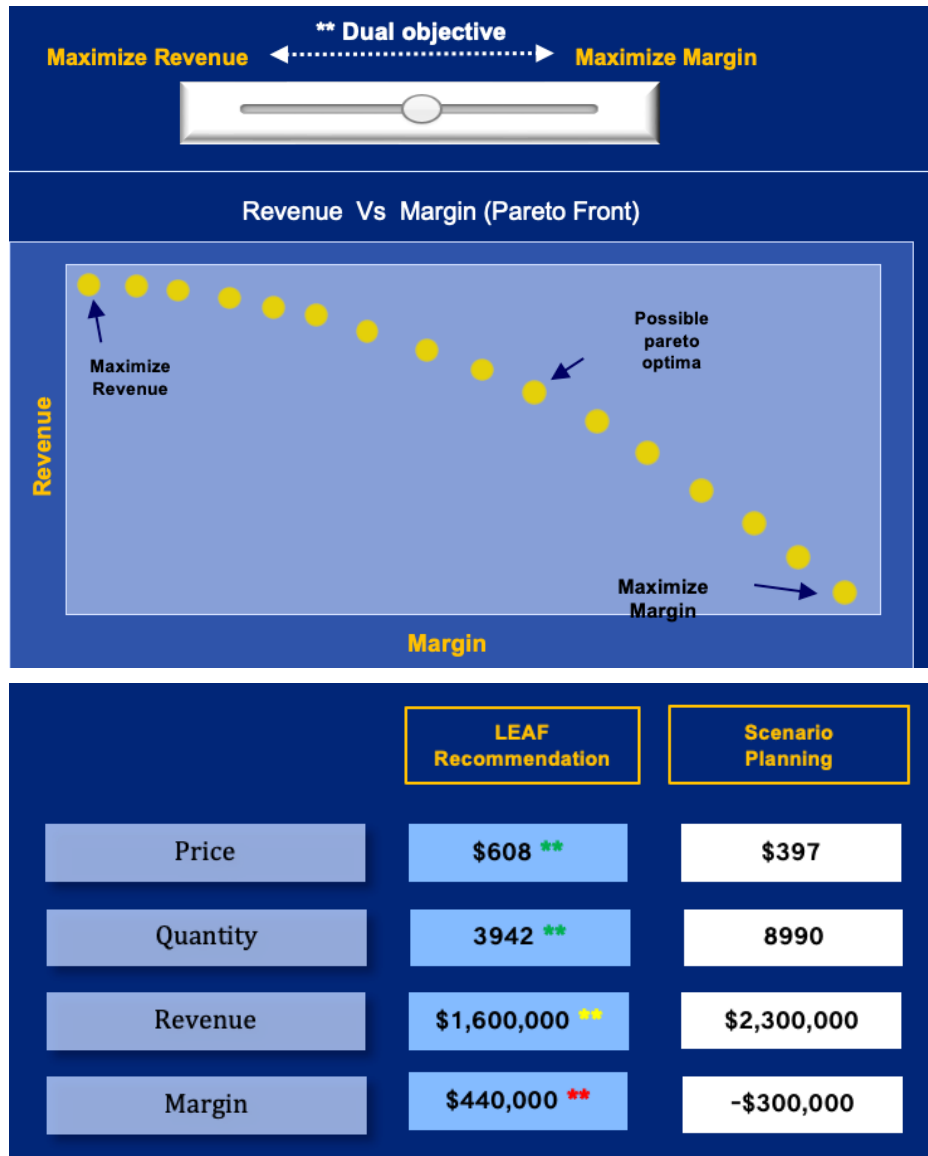
Diabetes Rules Interpretation

1. If Patient has a respiratory problem but not a neoplasms problem prescribe **metformin-pioglitazone**
2. If the person any age except 60-70 prescribe **glyburide**
3. If the person has Diabetic Diagnostic result as positive prescribe **glipizide-metformin**
4. If the person has a positive circulatory diagnosis prescribe **pioglitazone**
5. If the admission type is Newborn prescribe **tolazamide**
6. If the person is between 30 and 60 years old and has a positive musculoskeletal diagnosis and is Asian and was sent by court/law enforcement and was admitted in emergency, prescribe **insulin**
7. If none of the above, prescribe **glipizide**



Some Examples

Pricing Optimization



Improving pricing strategy for a leading home appliance manufacturer

- Using historical data, build an adaptive digital twin of the market
- With evolutionary optimization, maximize conflicting objectives of revenue and margin

Result: A Pareto front of the best possible tradeoffs

- The user can select a desired tradeoff from it
- RIO indicates uncertainty in the predictions

Scratchpad allows adjusting the price by hand

- The digital twin predicts the resulting revenue and margin

Insurance Underwriting

Underwriter Policy Prescriptor



John Smith
Underwriter

Policy Details

Policy Number / Insured Name

Policy Number
P7274347

Pending renewal

Policy Effective Date
01/01/2018

Policy Expiry Date
01/04/2019

Age
15

Number of Locations
2

Area (in Sq. ft.)
2000

Construction Type
Combustible

Technical Sub-Product
Real Factory

Channel
Broker

Market Segment
Corporate

Additional Policy Details

Issuing Country
United States

Issuing Sub-Region
01/04/2019

Product Tower
Property

Storey
2



View Coverage Details

LEAF Prescription

Can insure

\$80K - 90K
Prescribed
Renewal Premium

85%
Confidence in
Recommendation

90%
Probability for
customer to Buy

20%
Loss Ratio Prediction

LOW



Risk Prediction

Approve Prescribed Premium

Scratch Pad

\$50,000

→

70%
Loss Ratio
Prediction

LOW



Risk Prediction

Objectives Management

0.7

Win-Loss Ratio

0.3

Loss-Ratio

0.42

Cost

Save Objectives

Reset Values

Improving pricing strategy for a leading property insurance company

- Using historical data, build an adaptive digital twin of the properties
- With evolutionary optimization, maximize conflicting objectives of win/loss, risk, and cost

Result: A Pareto front of the best possible tradeoffs

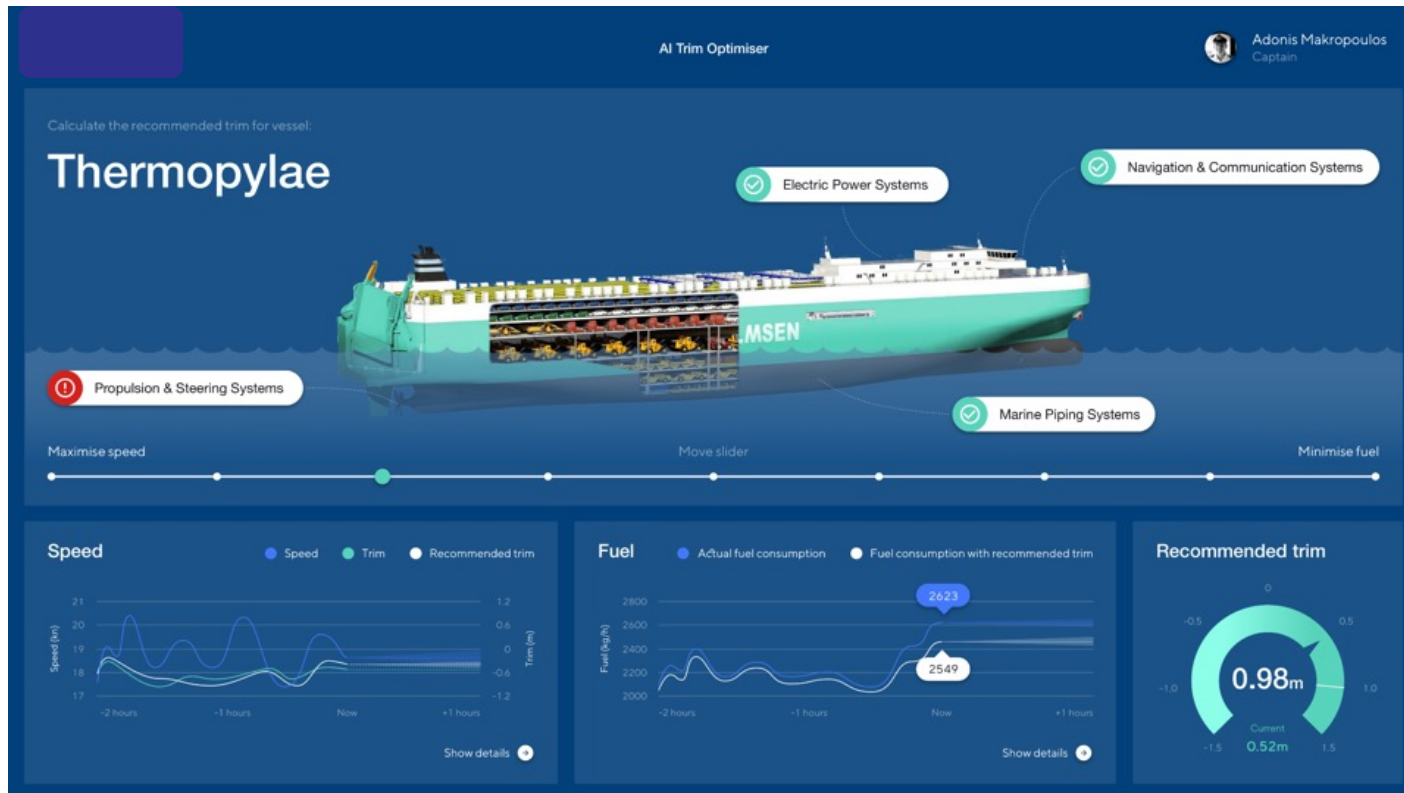
- The user can select a desired tradeoff from it
- RIO indicates uncertainty in the predictions

Scratchpad allows adjusting the premiums by hand

- The digital twin predicts the resulting risk and win/loss

Ship Trim Optimization

Improving ship speed/fuel efficiency for a shipping company



- Using historical data, build an adaptive digital twin of the ship
- With evolutionary optimization, maximize conflicting objectives of speed and fuel consumption

Result: A Pareto front of the best possible tradeoffs

- The user can select a desired tradeoff from it
- RIO indicates uncertainty in the predictions

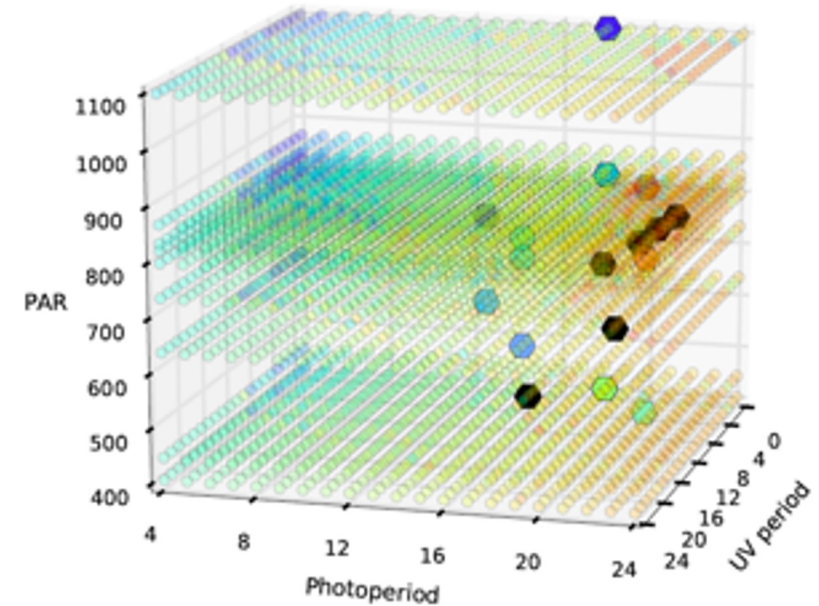
Scratchpad allows adjusting the premiums by hand

- The digital twin predicts the resulting speed and fuel consumption

Optimizing Growth Recipes for Agriculture



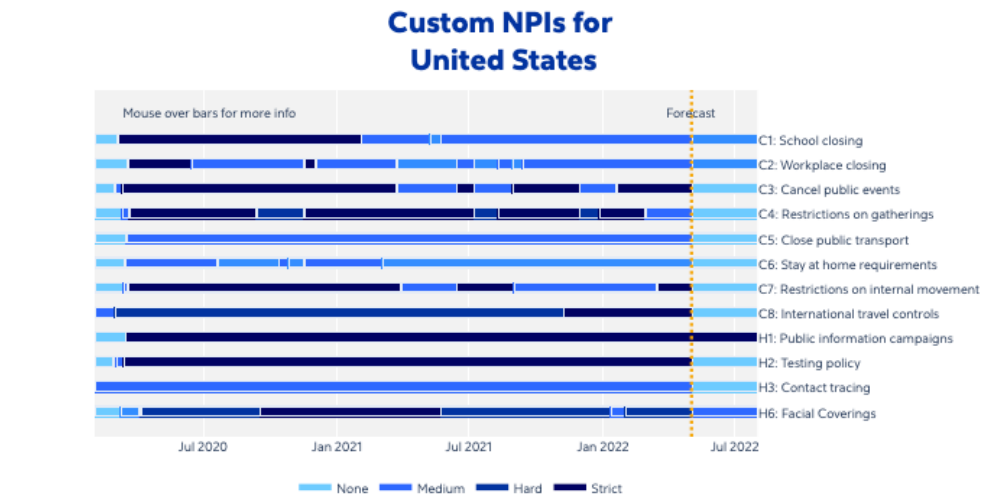
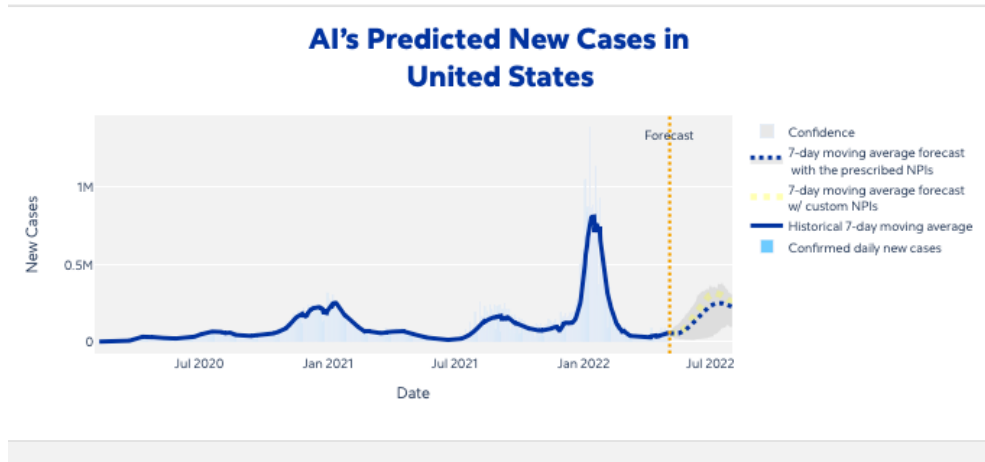
Environments at the OpenAg Lab of MIT Media Lab



Recipes discovered for light (darker is better)

- Build an adaptive digital twin of how plants grow in hydroponic growth environments
 - Light, temperature, water, nutrients etc. computer controlled
- Evolve growth recipes against the twin
 - Millions new recipes created and evaluated; only a few hundred actually planted
- Discovered that basil does not need to sleep!
 - Grows best when lights are on 24hrs/day
 - An insight that surprised domain experts

Optimizing COVID-19 Response



Build an adaptive digital twin of COVID-19 cases over time

- Predict number of cases in different countries
- In response to non-pharmaceutical interventions (NPIs)

Discover best interventions against the twin

- Resulting in smallest number of cases
- With minimal economic cost

Not just what will happen, but what we should do about it!

- Scratchpad, uncertainty to help explore alternatives

Retrained continuously since May 2020

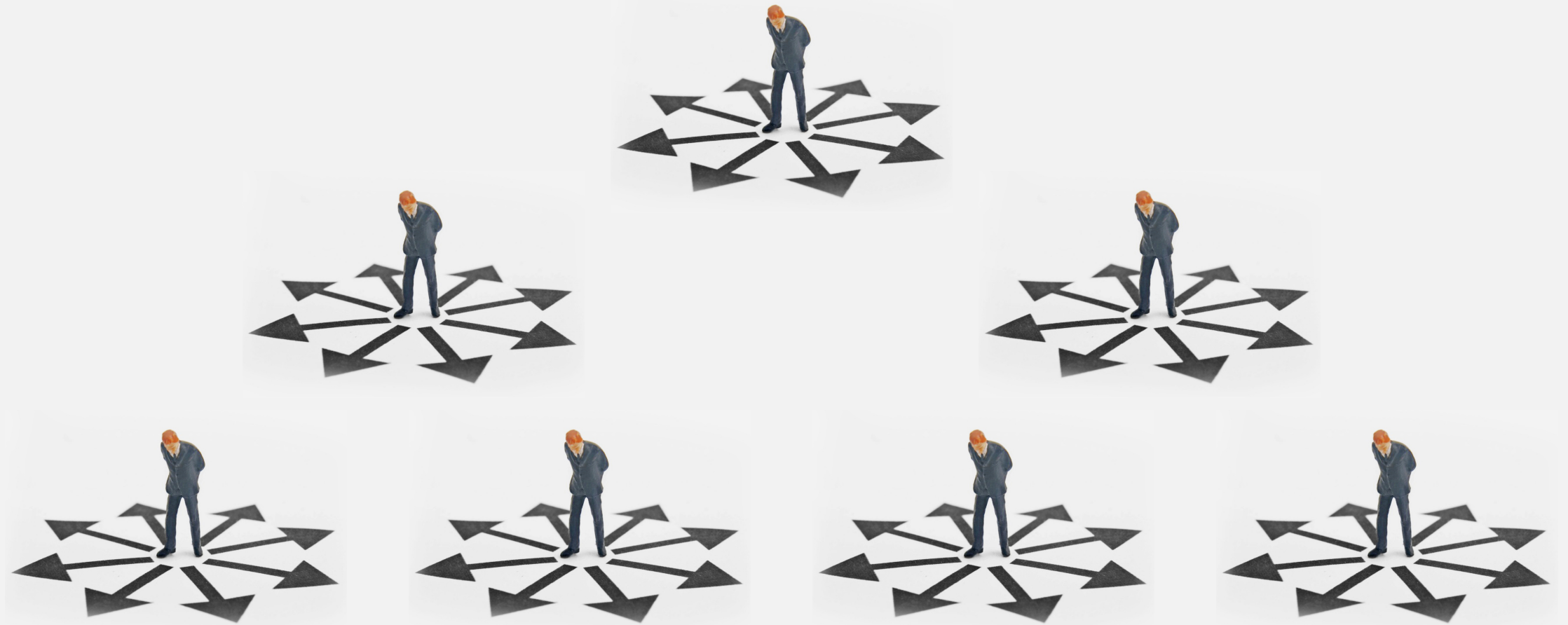
- Adapting to the different stages of the pandemic
- Generalizing from experiences across the world

Recommendations about two weeks in advance, e.g.

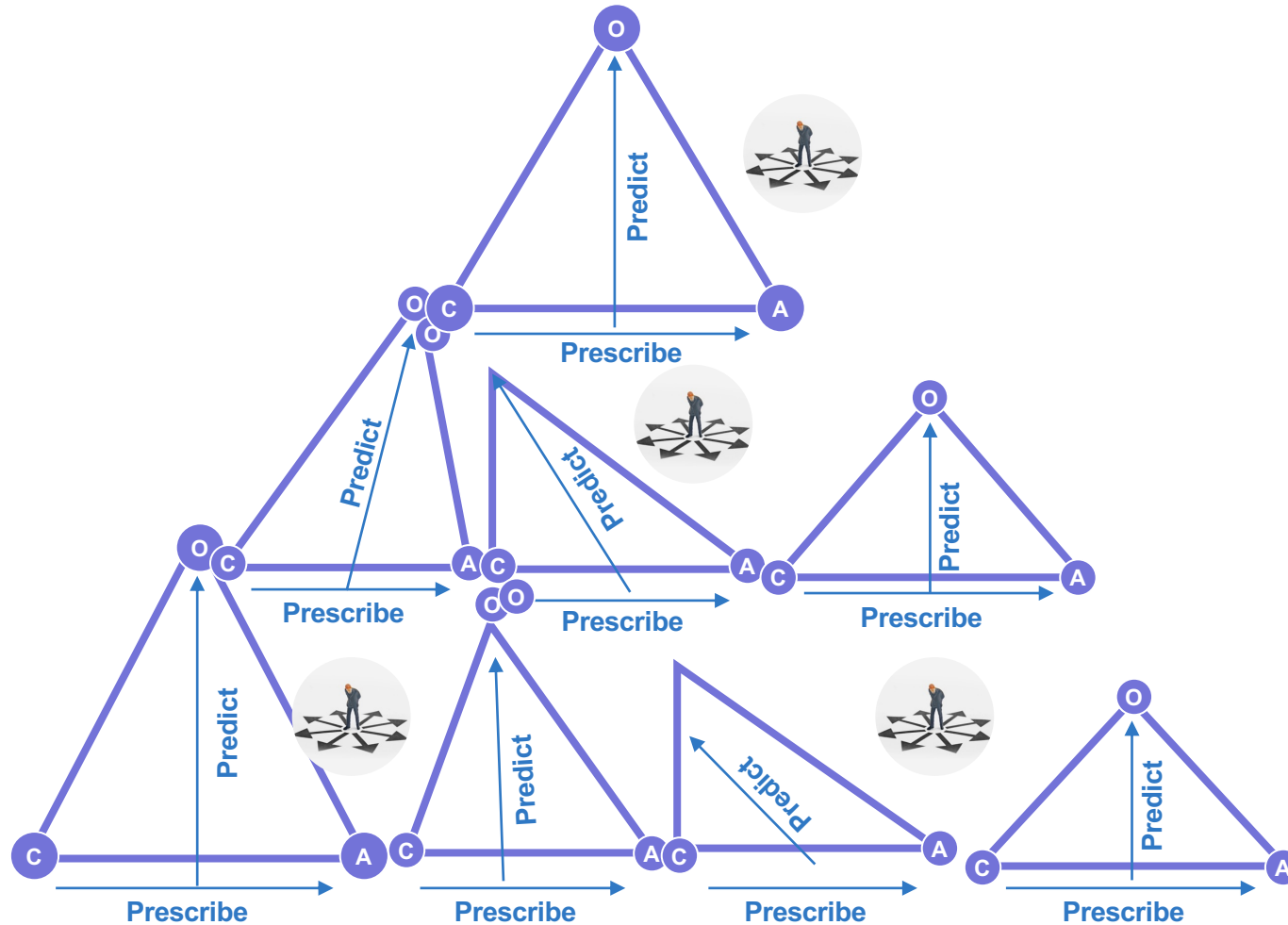
- May 2020: Focus on schools and workplaces (i.e. indoors)
- Sept 2020: Focus on gatherings, travel restrictions
- March 2021: India lockdown
- July 2021: Delta surge on countries with low rates so far
- March 2022: Masking to avoid a second Omicron surge

Interactive demo: <https://evolution.ml/demos/npidashboard>

An Organization Consists of Many Decision Makers



Scale in Entangled Decision-Loops



- More Accurate Decision
- More Efficient Decision-making
- Better Outcomes

Thank You