



'Crossing the Chasm' – Bridging the Gap Between Machine Learning and Business Decision-Making

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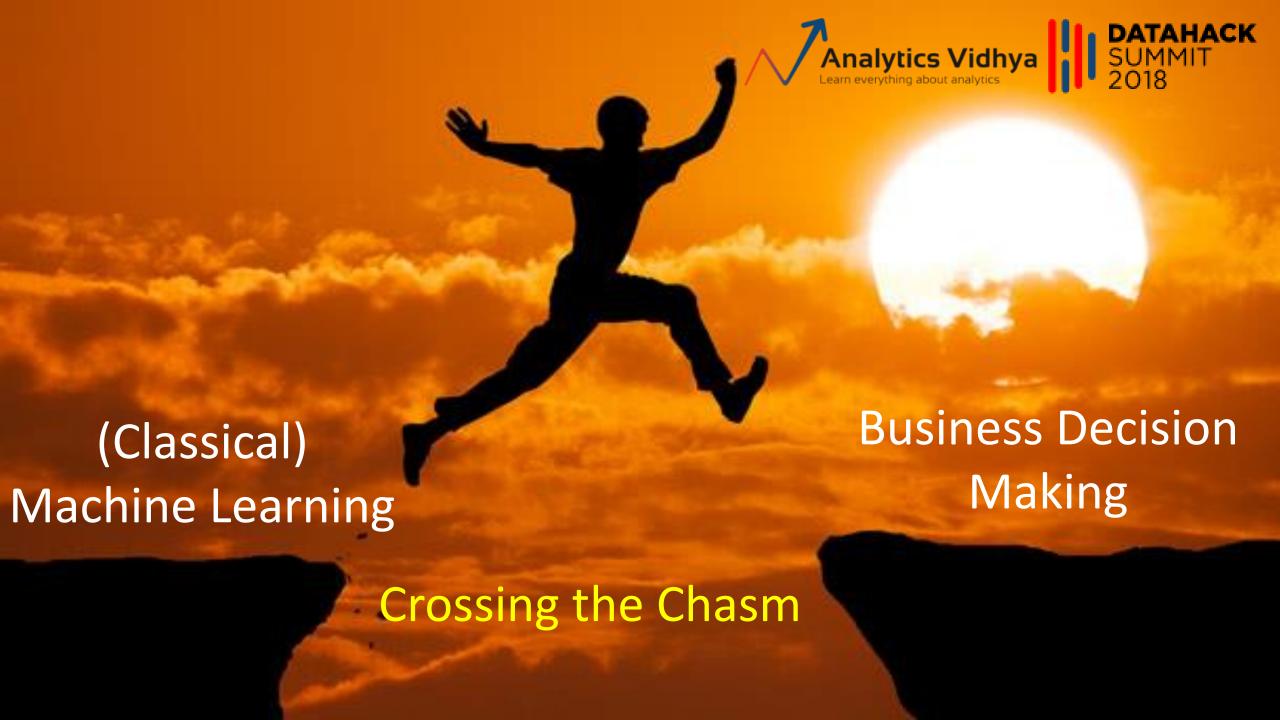


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### What is this talk about? (and what it is not!)

### What is this talk about:

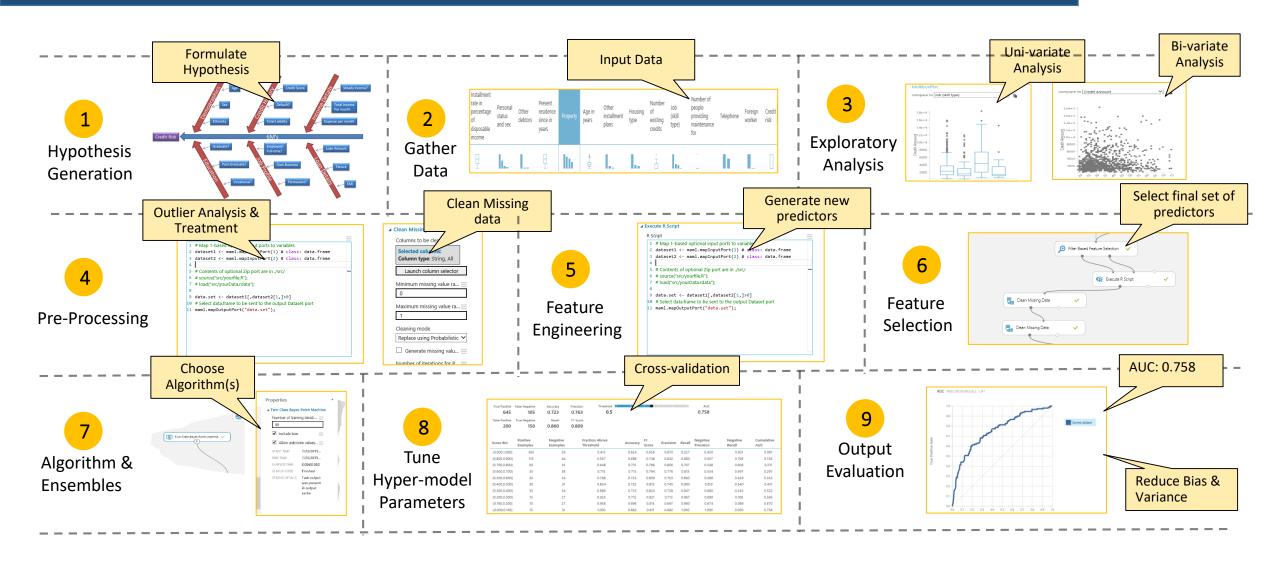
- ML / DL are fascinating and powerful tools...but
- Significant gap exists in utilizing ML / DL predictions for business decision making
- Introduction to Simulation Modeling
- Explore synergies between Simulation Modeling & Data science techniques

### What this talk is not:

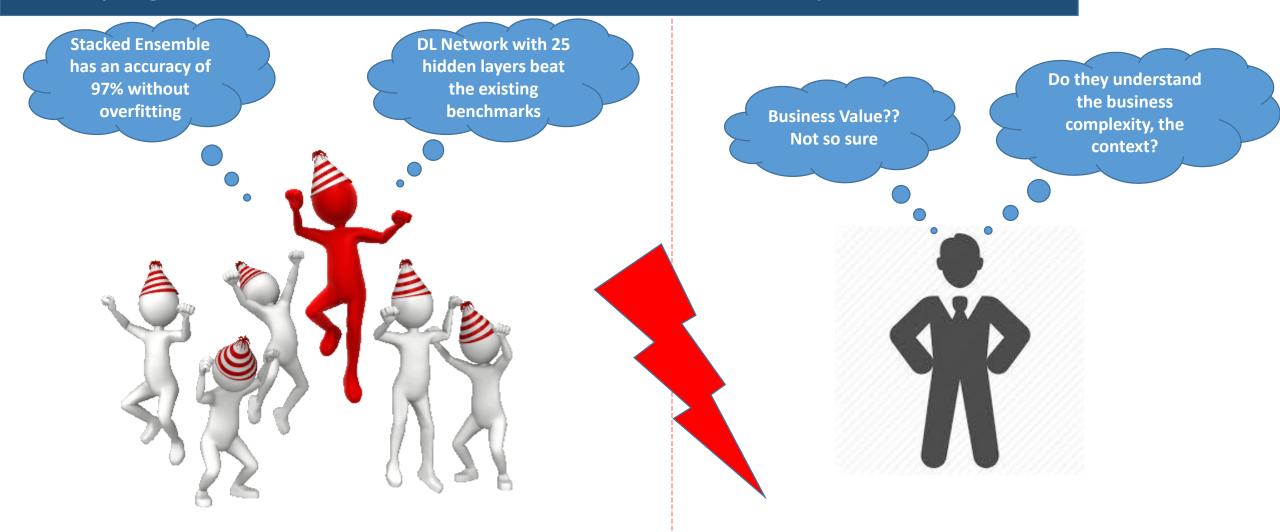
- Does not cover any specific ML / DL / AI techniques
- Does not cover any math / stats concepts
- Does not cover moral & ethical dimensions of data science / Al

GitHub Link: https://github.com/skkeyan-mlai/DHS2018

## (Classical) Supervised Machine Learning Pipeline



### 'Partying' Data Scientists vs 'So What' Business Guy

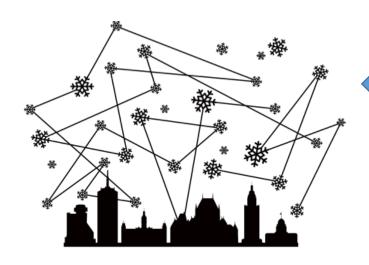


**Data Scientists** 

**Business Decision Maker** 

### Decision Making 101

# Real World (Business)



- Multiple Entities
- Complex Non-Linear Relationships
- Delayed Feedback
- 'Black Swan' Events



#### **EXPERTISE**

- ✓ Education
- ✓ Training
- ✓ Experience



Business Decision Maker

### **Approximation of the Real World**

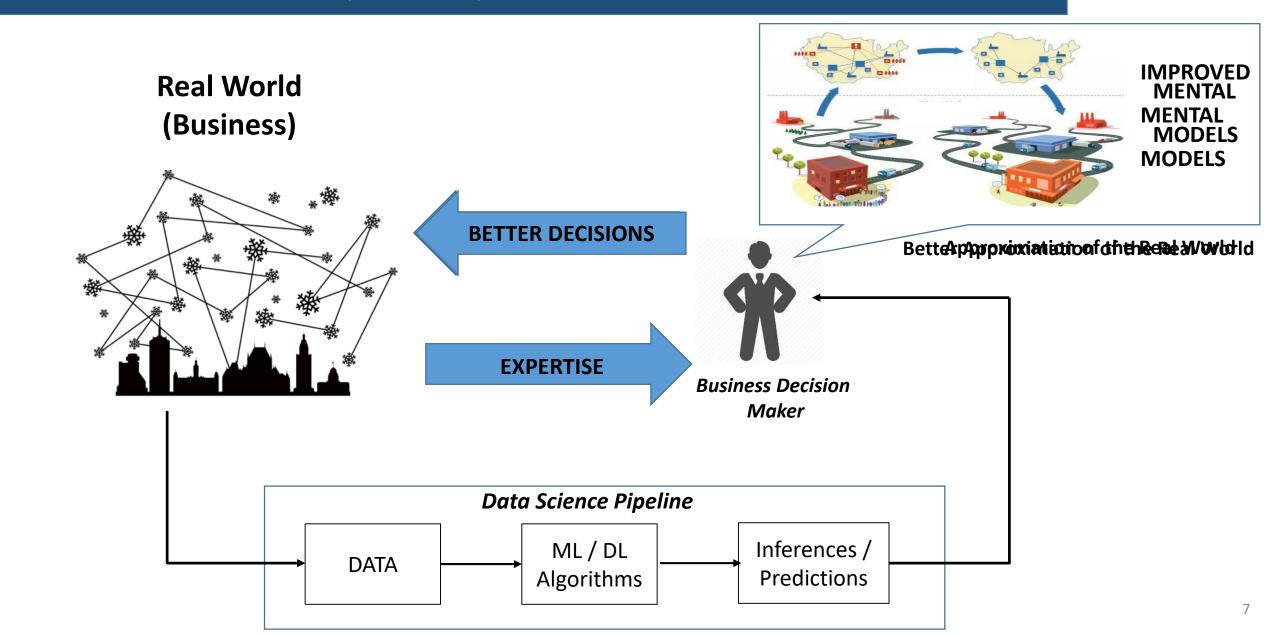
**MENTAL** 

**MODELS** 

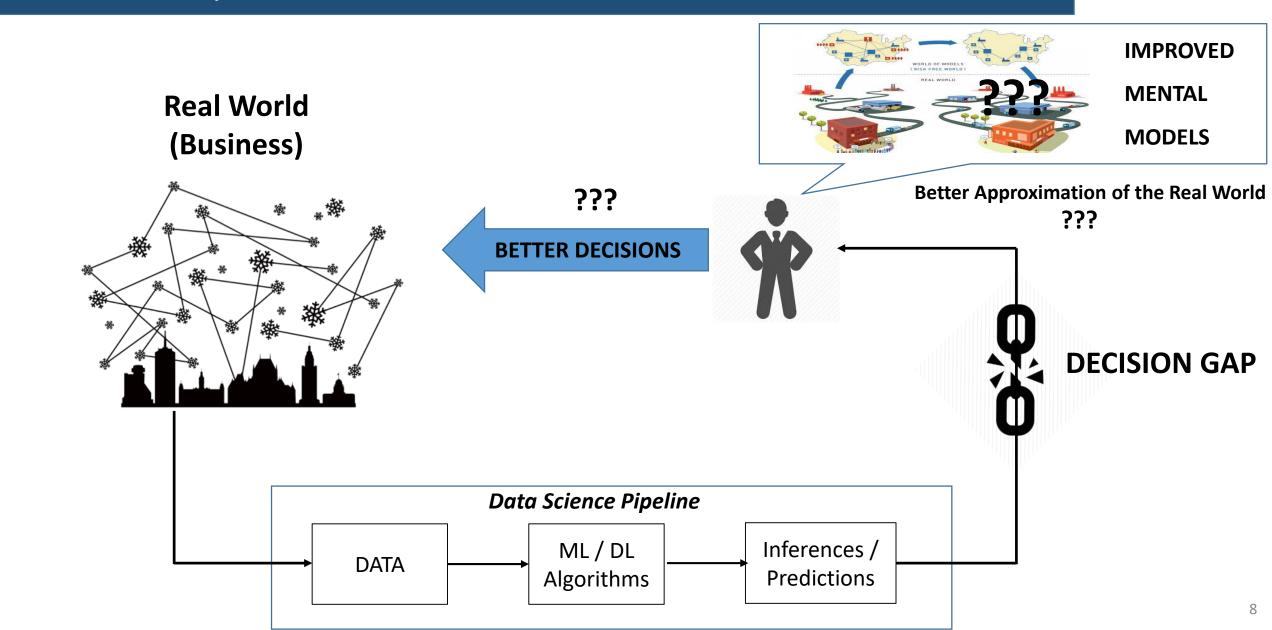
### **Characteristics of Mental Models**

- Process Centric
- Cause & Effect based
- Estimates risk & uncertainty
- Biased
- Rigid (Relatively difficult to change)

### Goal of Data Science (ML / DL)



## 'Decision Gap' – What is it?



# The Decision Gap – Why it exists?



Predictions from Machine Learning

1

Black Box Algorithms — Lack of Interpretability of black-box algorithms (GBMs, CNNs etc.) leading to unexplainable predictions at a local level (specific customer or region)

2

**No Quantification of Uncertainty** – ML models are obtained by minimizing a cost function or by maximum likelihood which results in point estimates with no quantification of uncertainty

3

Lack of Cause & Effect Relationships - ML models are based on correlations in data and does not provide "cause & effect" relationships. Business decisions are about making 'interventions' and that requires an understanding of causation

4

**No explicit formulation of constraints** - Business decision making is always under constraints (time, resources etc.). ML models does not typically take these constraints into account

5

**Does not account for Dynamic, Non-linear Interactions** - Hey, what do I do with predicted numbers when the business is dynamic, has multiple entities with non-linear interactions?



Business Decision Making

### Data Science Approach – To Bridge the Decision Gap



Predictions from Machine Learning

1

Machine Learning Interpretability (MLI) — Techniques (LIME, Shapley values etc.) that deal with explains predictions at a local / individual record level

2

**Probabilistic Programming** – Techniques that leverage Bayesian Inference to provide distribution of parameter values (not point estimates) thus quantifying uncertainty

3

Causal Machine Learning — Techniques like Bayes Nets, Structural Equation Models (SEM) etc. that helps determine causal interaction among variables with ability to represent complex interactions in a manageable compact form

4

**Optimization** – Well understood area with techniques like linear programming to help maximize or minimize objective functions along with decision variables and related constraints

5

**Reinforcement Learning** – Techniques that helps in finding optimal policies in an environment through Dynamic Programming, Q-Learning etc. But still very much restricted to "Grid Worlds" and not widely adopted in business setting



Business Decision Making

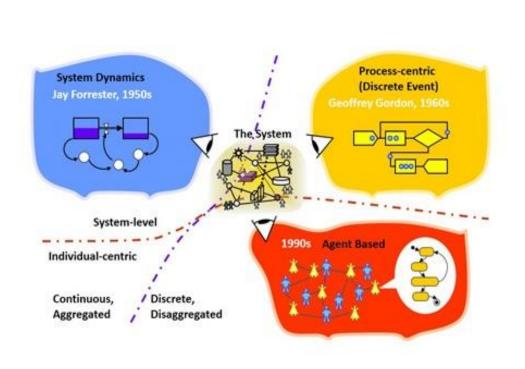
# ENTER THE WORLD OF SIMULATION MODELING

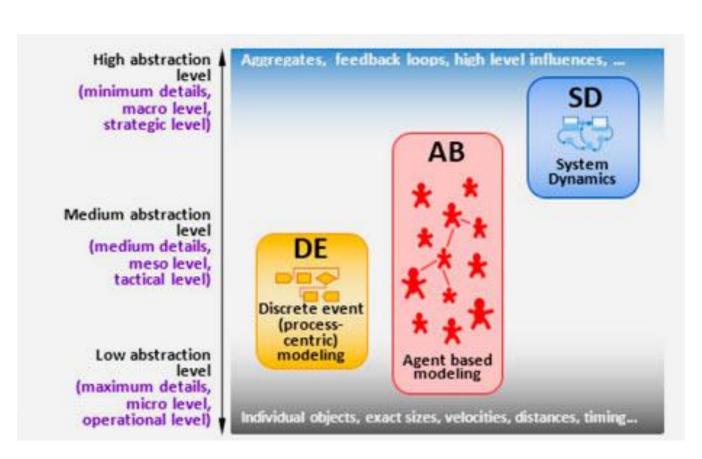
### What is Simulation Modeling?

Simulation Modeling (or Business Simulations) helps build realistic approximations of the real world and factors in business complexity

What is AnyLogic?

### **Techniques for Simulation Modeling**





Many Tools: AnyLogic, Vensim, Stella, Simio, Arena, Flexsim, Netlogo etc.

### Data Science + Simulation Modeling – A Powerful Combination

### **Simulation Modeling**

### **Data Science**

(ML/DL/AI)

1

Simulations are created based on the expert view on process relationships between entities

Could be unknown / biased / not realistic



Data Science by utilizing real world data and machine learning techniques can help correct the bias, increase precision, provide insights

Python Package PySD

2

Simulations can help generate good quality data from a representative model of the real world under different situations



Supervised Machine Learning needs good quality labelled data to generate reliable predictions

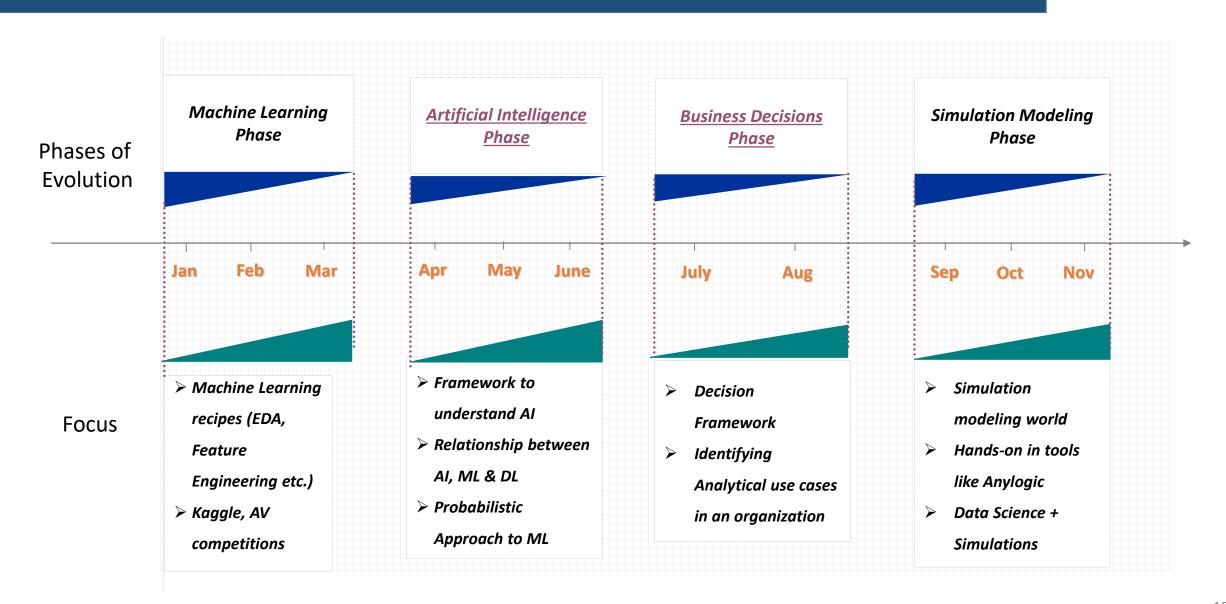
3

Simulation models can provide the environment in which agents can operate across multiple runs to learn the policy

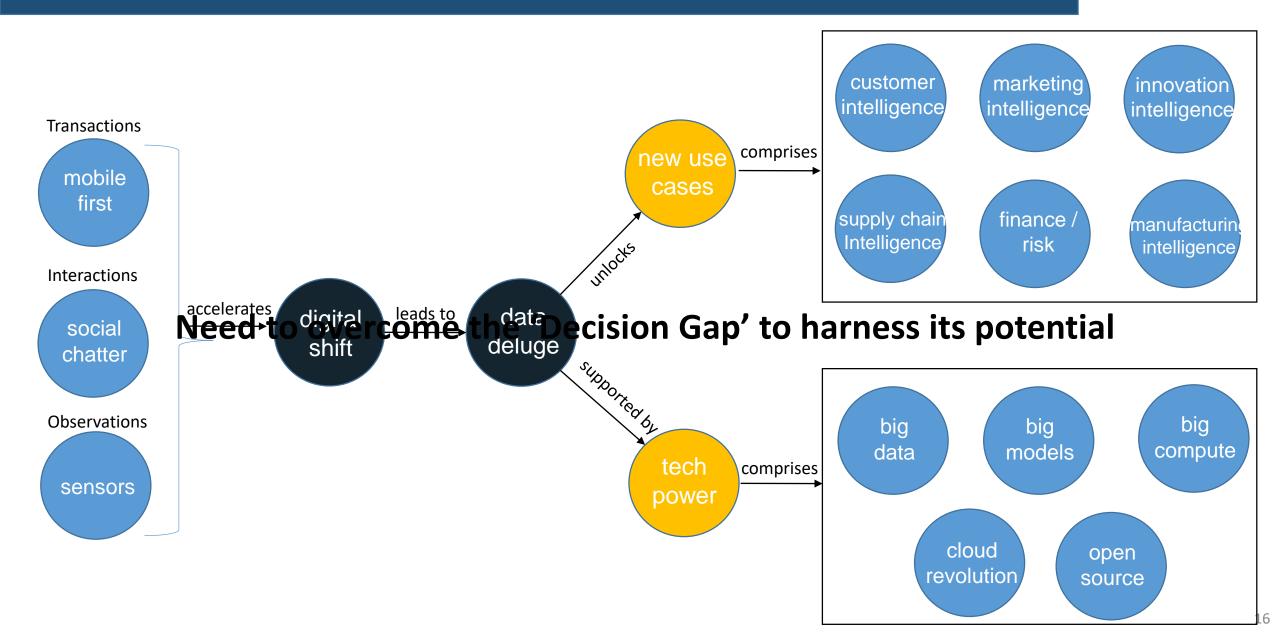


In Reinforcement Learning actions have to be performed in an environment multiple times to derive the appropriate policy (Q values)

# My personal 'Data Science' journey this year



### Data Science & Analytics – 'Clear & Present' Opportunity



### Thank You!



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- Tapchief <u>www.tapchief.com/karthik</u>
- GitHub <a href="https://github.com/skkeyan-mlai/DHS2018">https://github.com/skkeyan-mlai/DHS2018</a>

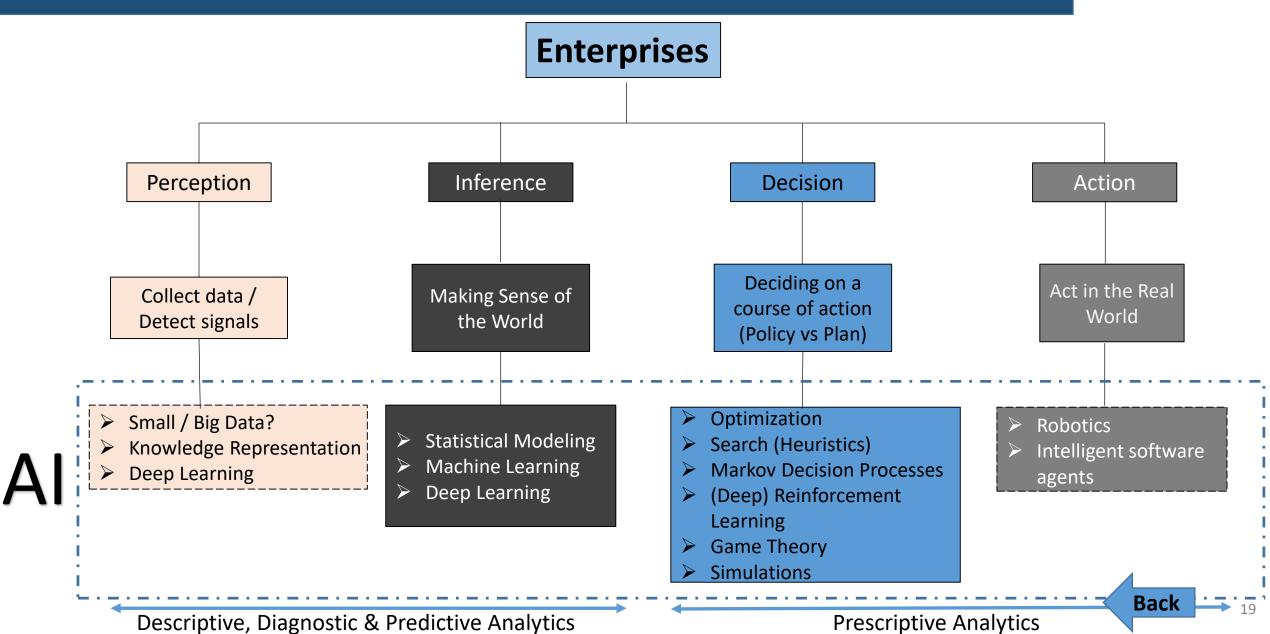
### Artificial Intelligence and Human Intelligence

Artificial Intelligence refers to the theory and development of computer systems & machines with the ability to perform tasks normally requiring <a href="https://example.com/html/>human intelligence">human intelligence</a>

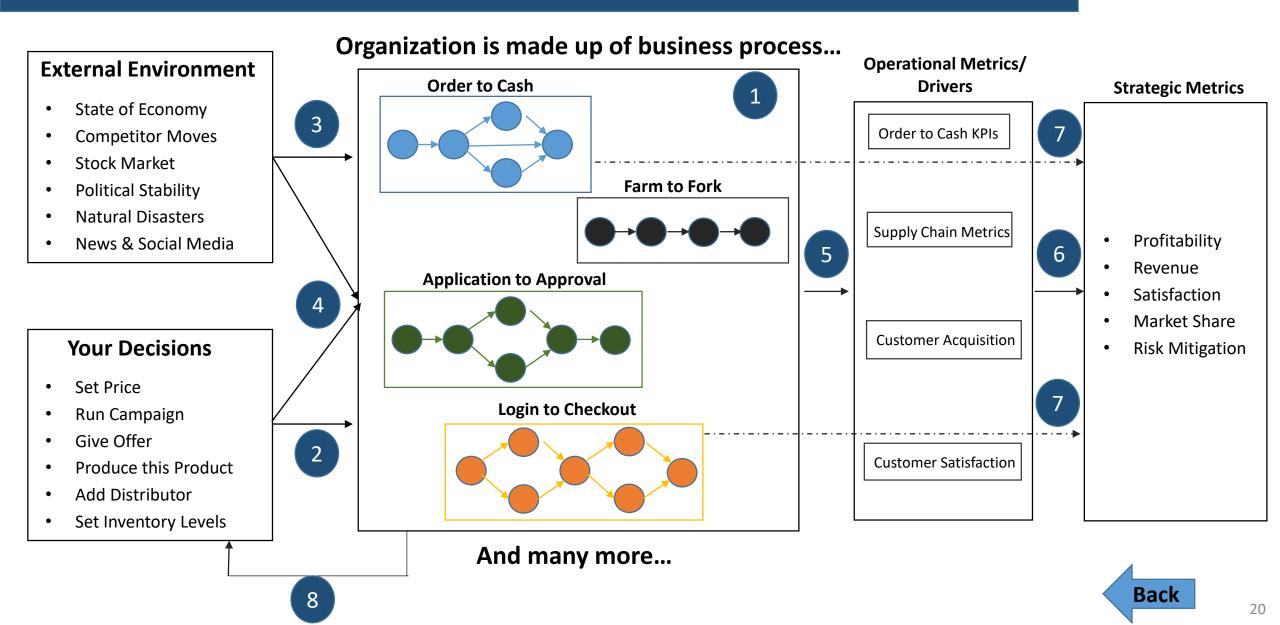


- Perceive the world, detect signals and collect data
- 2. Make sense of the world using data (Insights, <u>Inference</u>, Predictions etc.)
- Decide on the next course of action
- 4. <u>Act</u> in the Real World

### Al Techniques in Enterprises – Parallels to Human Intelligence



### **Business Decision Making Framework**



# My Analytics Mindmap

**Global Trends in Society** 

Macro-economy

**Business Fundamentals** 

**Specific Industry Domain** 

**Analytical use cases** 



Analytical
Platforms &
Techniques

**Data Management** 

**Reporting & Self-service** 

**Quantitative Techniques** 

**Performance Mgmt** 

**Insight Delivery** 

Analytics for Business Value

http://bit.ly/31KArT8

**Scan for New Products** 

**Evaluate Maturity** 



**Monitor Ecosystem** 

**Leverage Resources** 

