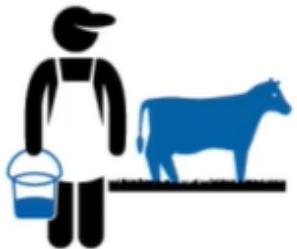




Applied Business Analytics

Week 1 – Introduction to Business Analytics

A Motivating Example : Dairy Farming in Kenya in 2011



Problems in dairy farming: Uncertain yields from cattle assets (both in milk and money), due to three main reasons:

- Biological cycles such as cattle gestation periods (which affect the milk yield)
- Cattle feed (quantity and quality), diseases
- Volatile market prices

Further problems (specific to Kenya):

- Farmers are small and dispersed over vast areas
- Markets are small and localised



The silver lining:

Mobile penetration was high in Kenya

A Motivating Example : Dairy Farming in Kenya in 2011

In this scenario, Su Kahumbu Stephanou started a subscription information service named iCow in Kenya



Image source: Image by Bret Hartman for <https://fellowsblog.ted.com/meet-su-kahumbu-the-kenyan-agriculturalist-helping-farmers-keep-their-herds-healthy-4b926b4a6d7b>

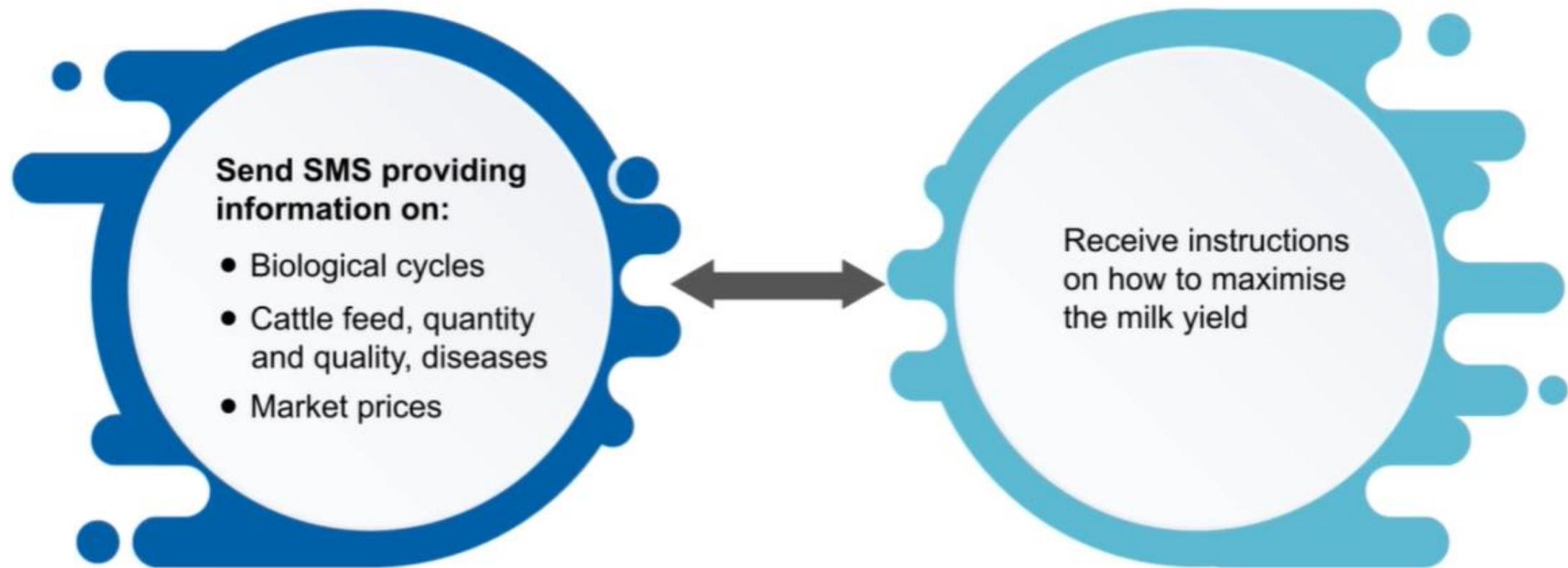
iCow



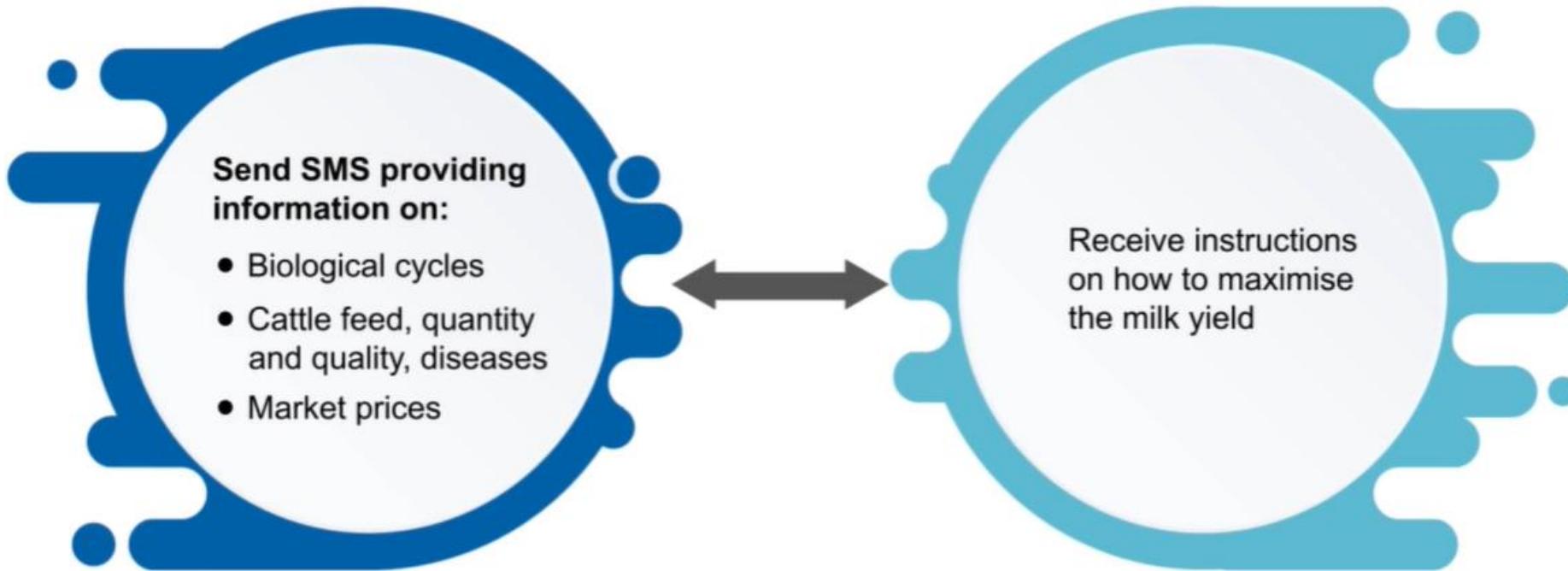
Image source: <https://icow.co.ke/>

The iCow Story

How iCow worked



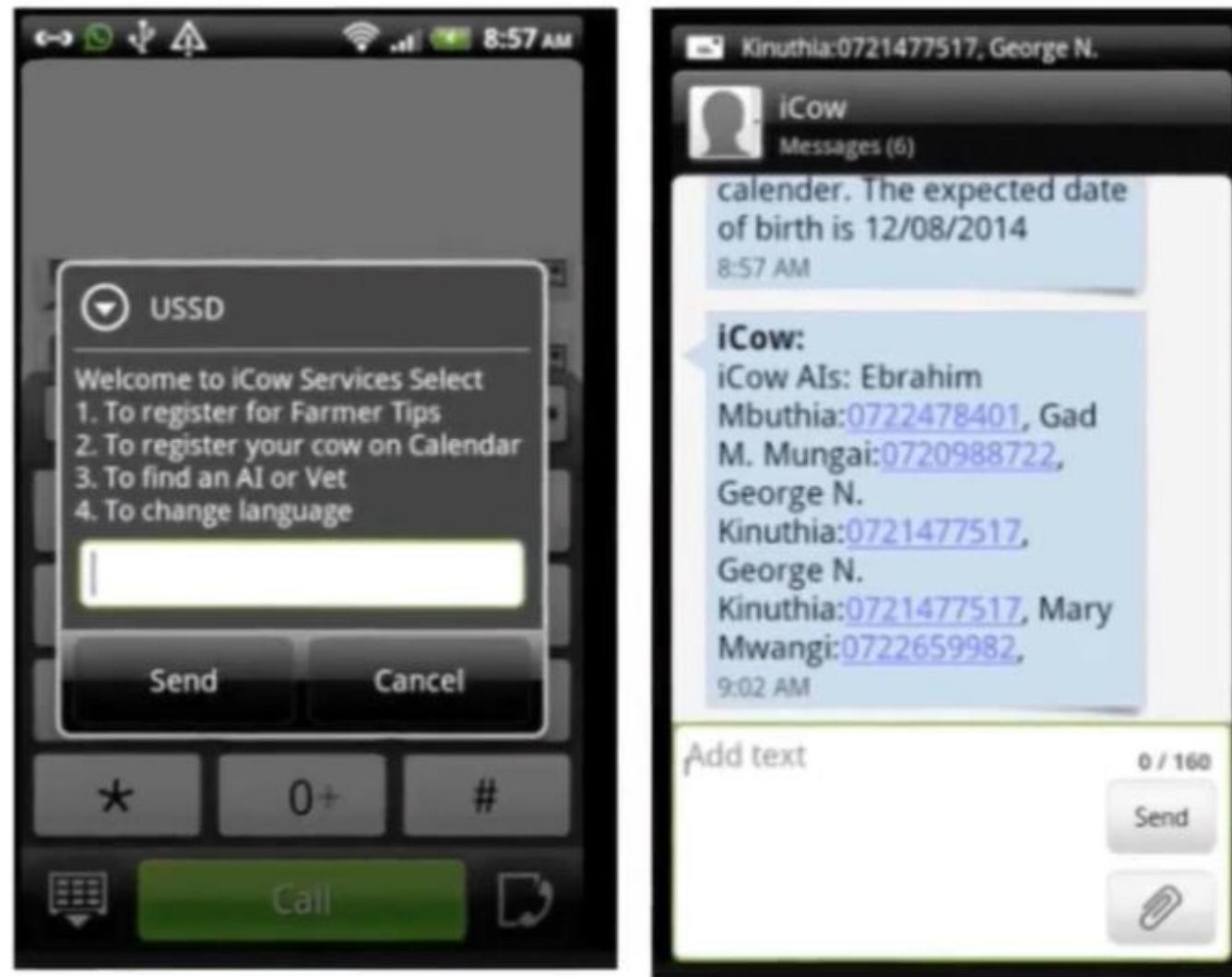
The iCow Story



- In its first year in 2011, 42,000 people signed up
- This in turn helped Kahumbu build a data repository containing livestock data, health and yield data, quantity and quality data on dairy yields and milk production nationwide

The iCow Story

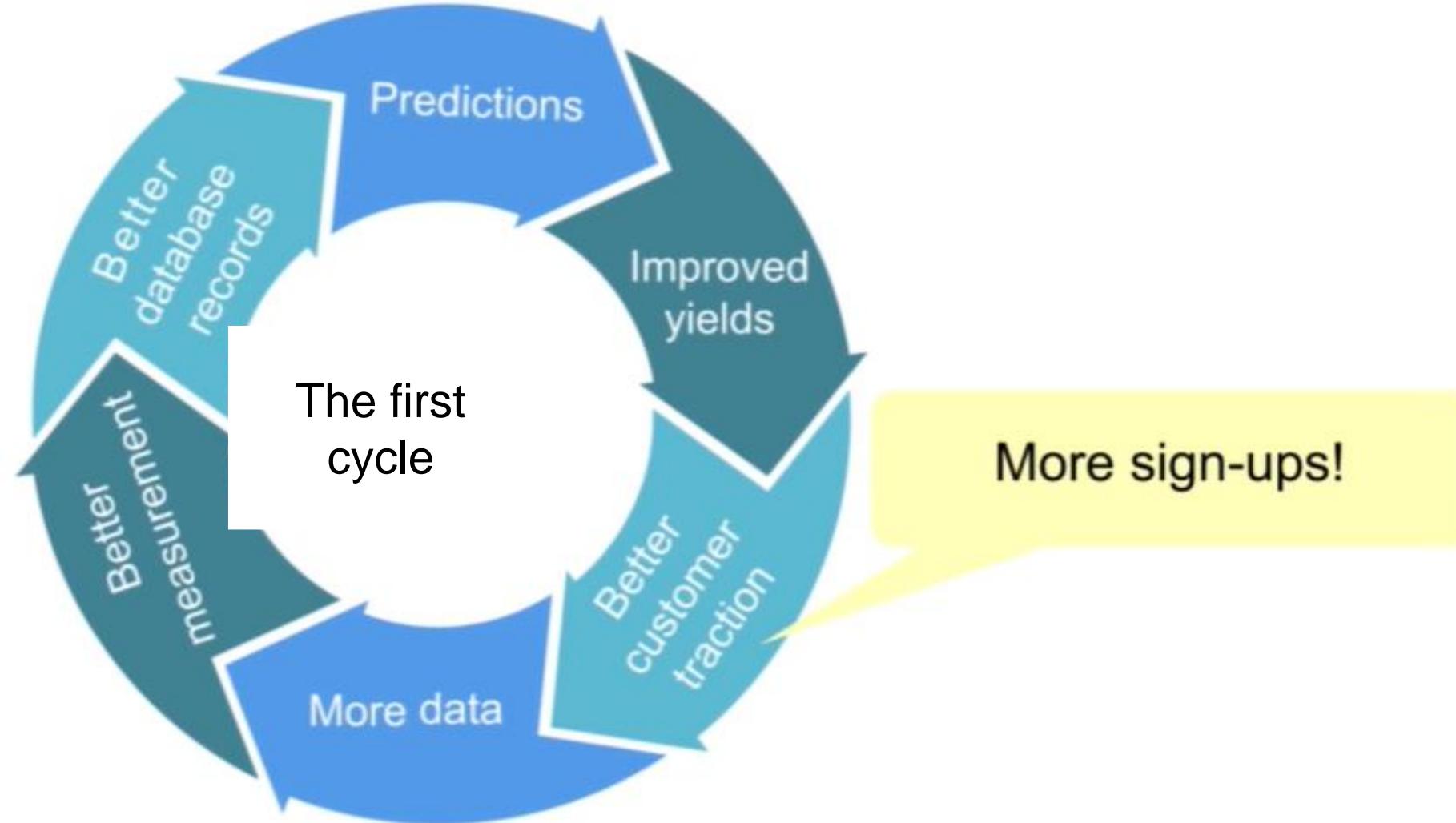
Registration of a cow on iCow:



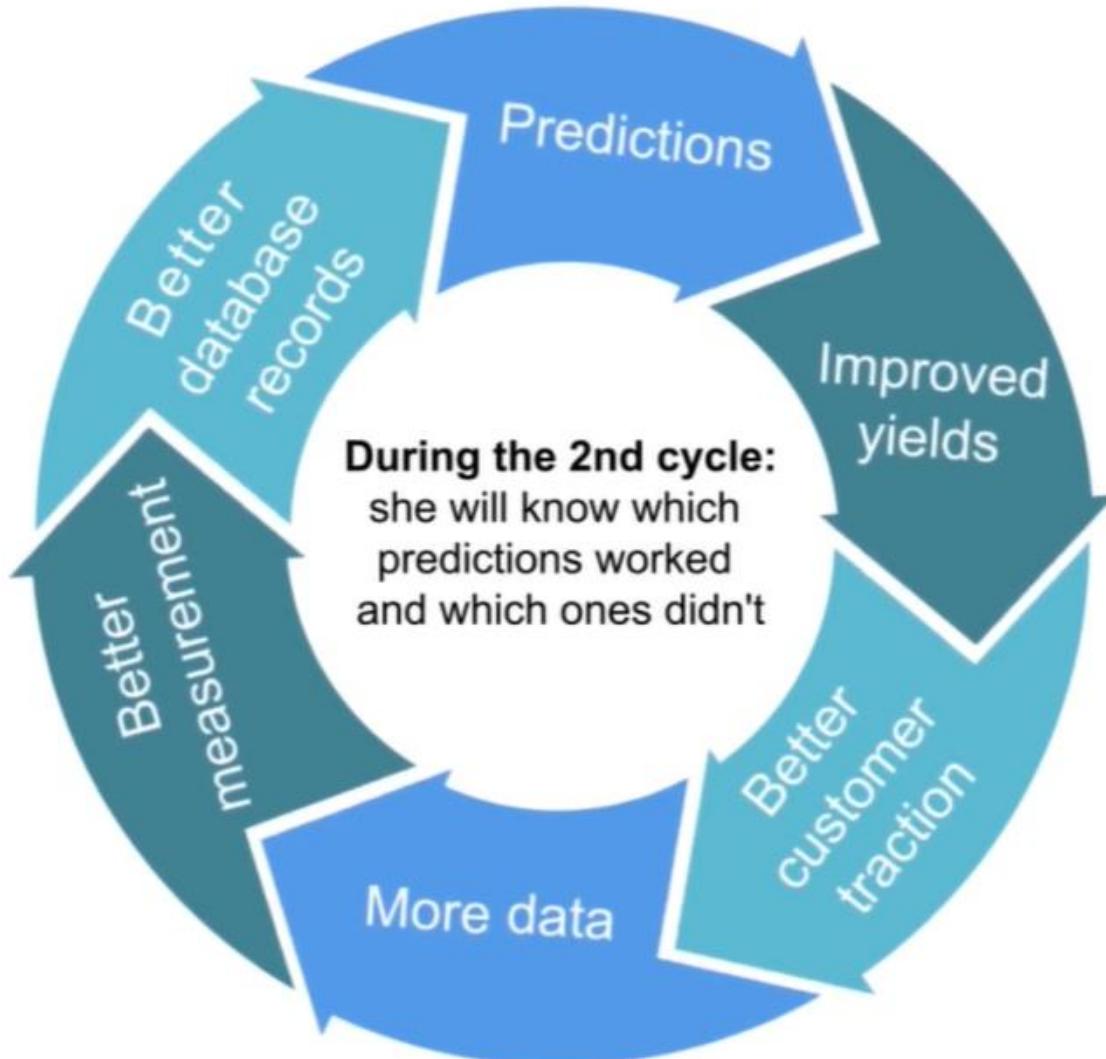
The iCow Story

- In the beginning, Kahumbu started with little or no data and relied primarily on theory and guesswork
- Later, when the data started flowing in, analytics kicked in

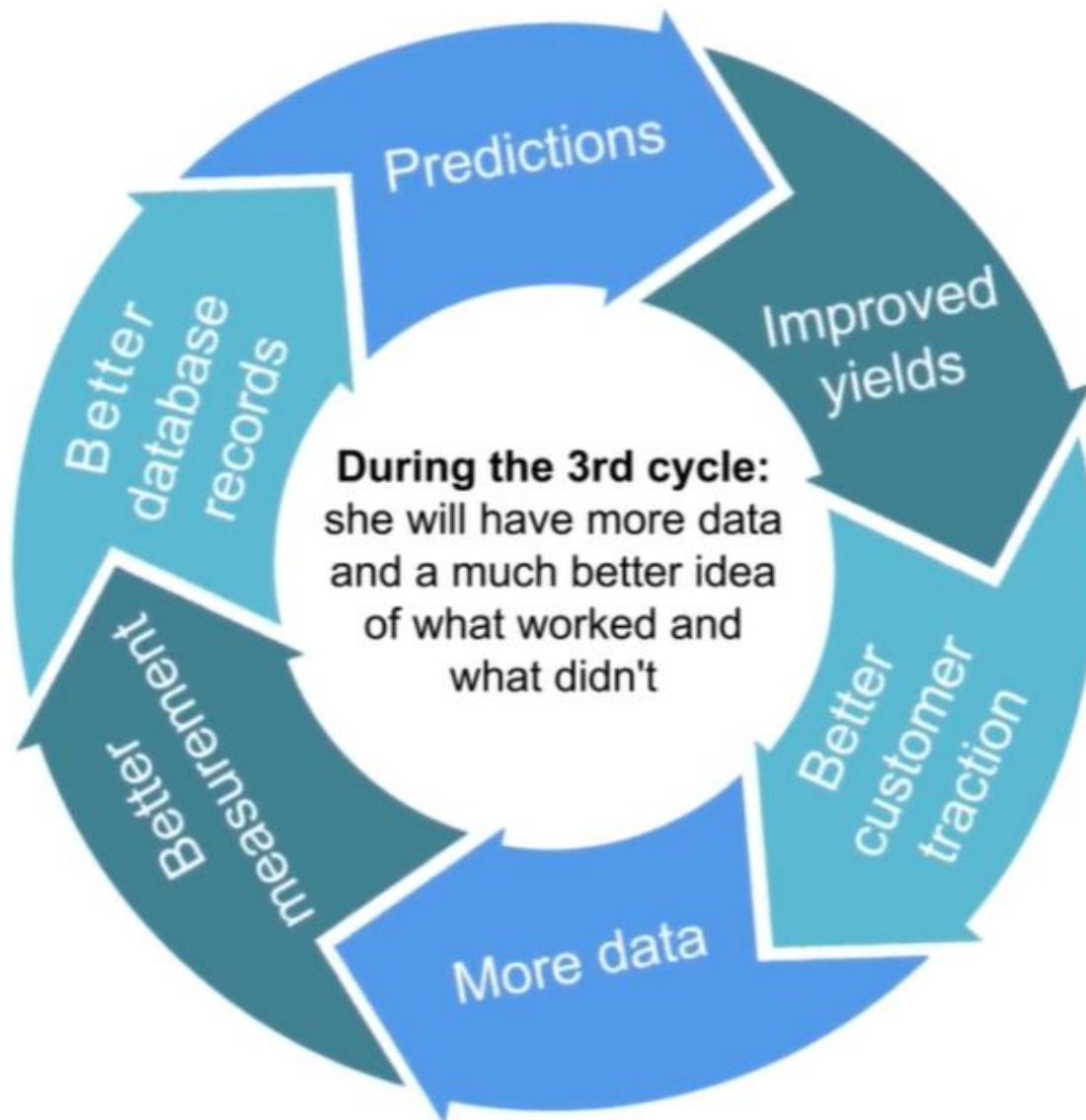
Role of Analytics in the iCow Story



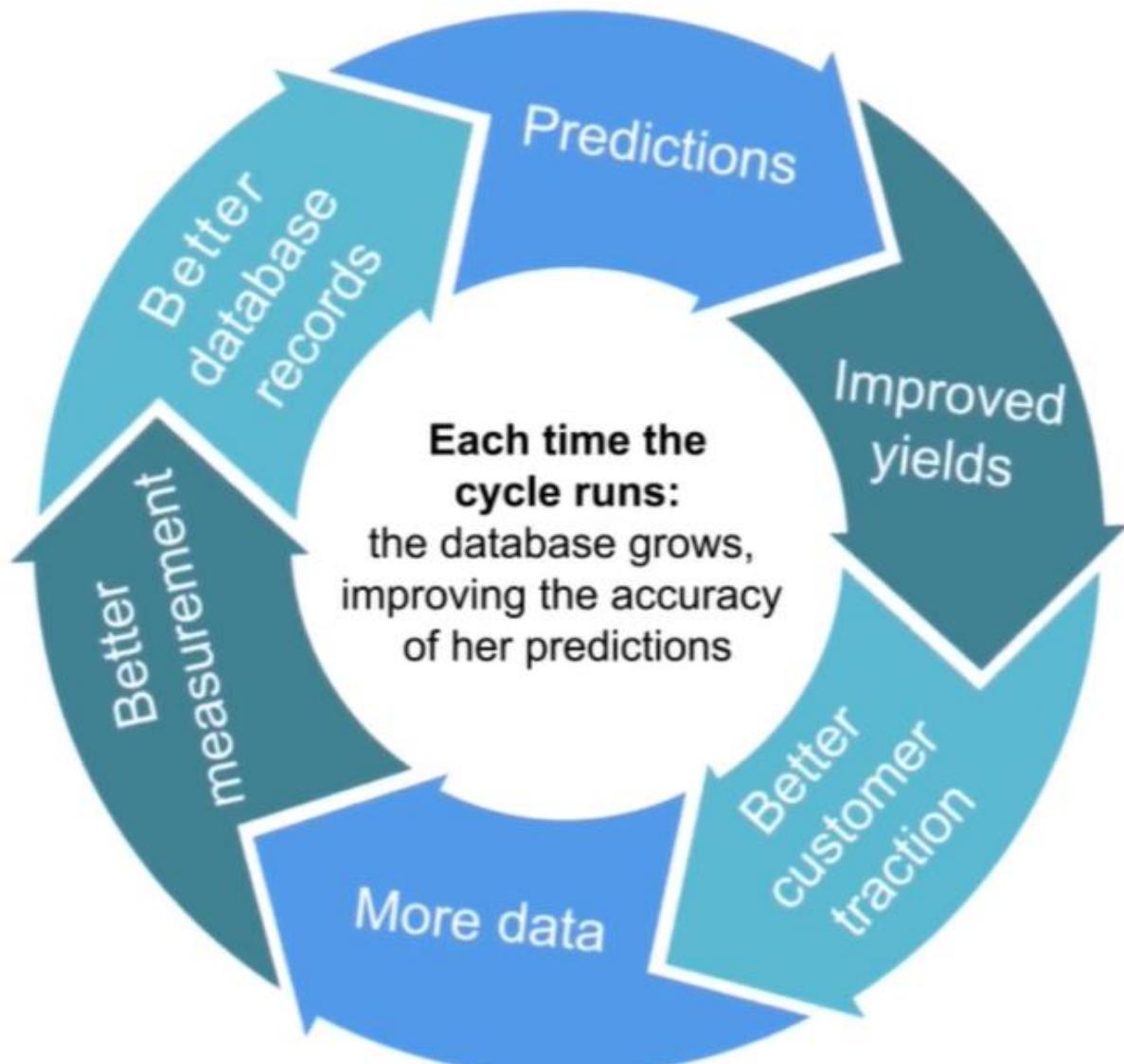
Role of Analytics in the iCow Story



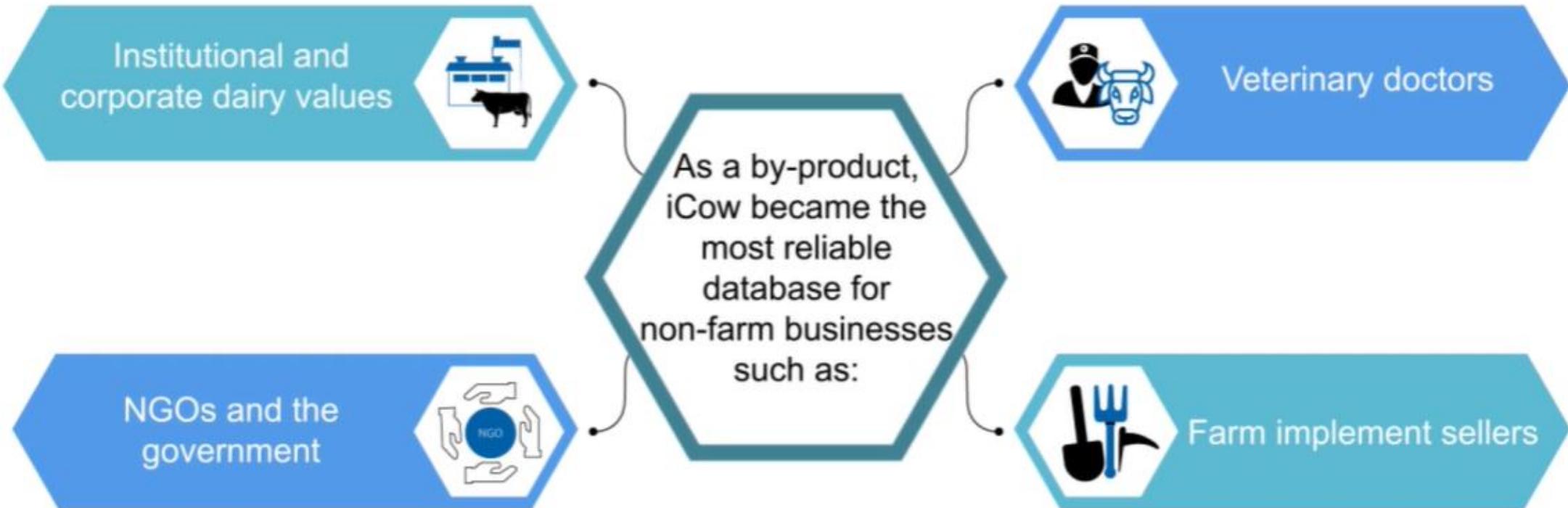
Role of Analytics in the iCow Story



Role of Analytics in the iCow Story



The iCow Story: Offering Reliable Information



The iCow Story

iCow could organically expand:

- its subscription business into value-added services for farmers
- its platform to offer B2B services to both small and large farmers

The iCow Story: Customer Valued Derived



- The average Kenyan farmer owned three cows
- Within seven months of using iCow, farmers reported a 33% jump in yield
- This was equivalent to owning a fourth cow
- Annually, the yields would increase by 50%

- Since iCow services were low-priced, for every \$1 farmers spent, they saw the returns of 7,600% (\$77)

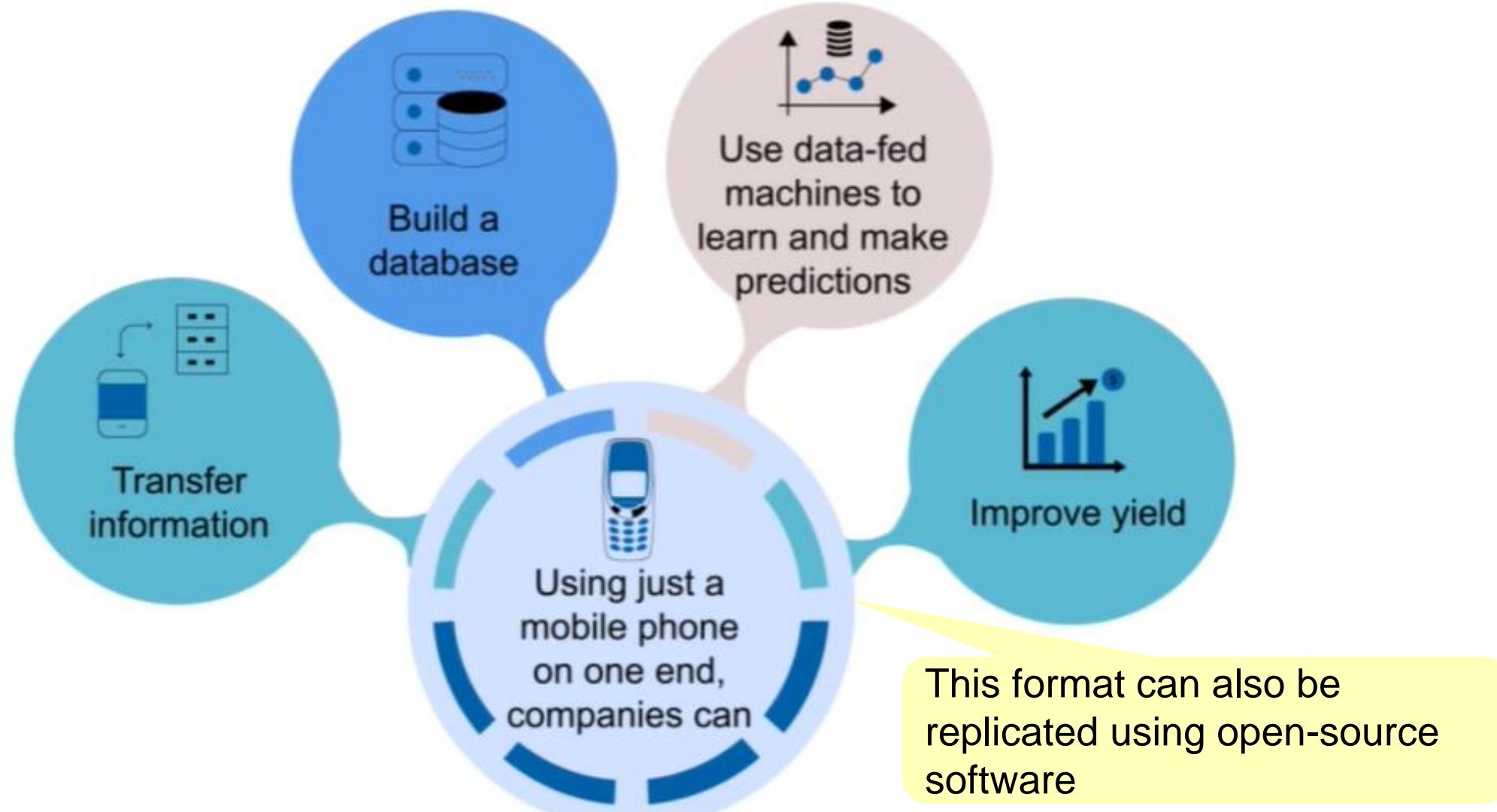
- Simplicity and convenience-farmers only needed to use a simple feature phone to avail of iCow's benefits

The iCow Story

The iCow success story:

Low end analytics not only helped yield huge returns but also created social benefits for farmers

How iCow's Template Can Be Used by Other Countries



iCow's Example: Conclusion

Kahumbu's problem statement:

To optimise dairy yield (Y) given a set of input variables (X)

Y variable (dependent)	X variable (input)
Dairy yield	Biological cycles of cattle
	Cattle feed
	Volatile market prices



Solution:

Use a model that predicts how inputs influence the outcome, a model that predicts what-ifs

This process of predicting input changes to outcome changes and mapping them is called modelling

ICow's Example: Conclusion

Problem statement:

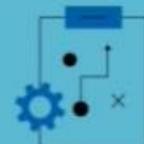
To optimise dairy yield (Y) given a set of input variables (X)



Solution:

Use a model that predicts how inputs influence the outcome, a model that:

Learns on its own



Makes predictions for individual cows



Mine data for patterns and match patterns



iCow's Example: Conclusion

- Data is the primary asset and the raw material from which we extract insight
- But data needs analytics

What is Business Analytics?

Business analytics is the process of connecting data, imposing structure on it, feeding data into a machine, knowing what modelling is possible by the machine and understanding how the machines learn

Business Analytics

Conceptual questions:

- What is a business?
- What are the objectives of a business?
- What is analytics? What does it do?
- What is data? Where does it come from?

The Objective of a Business

The two fundamental economic axioms

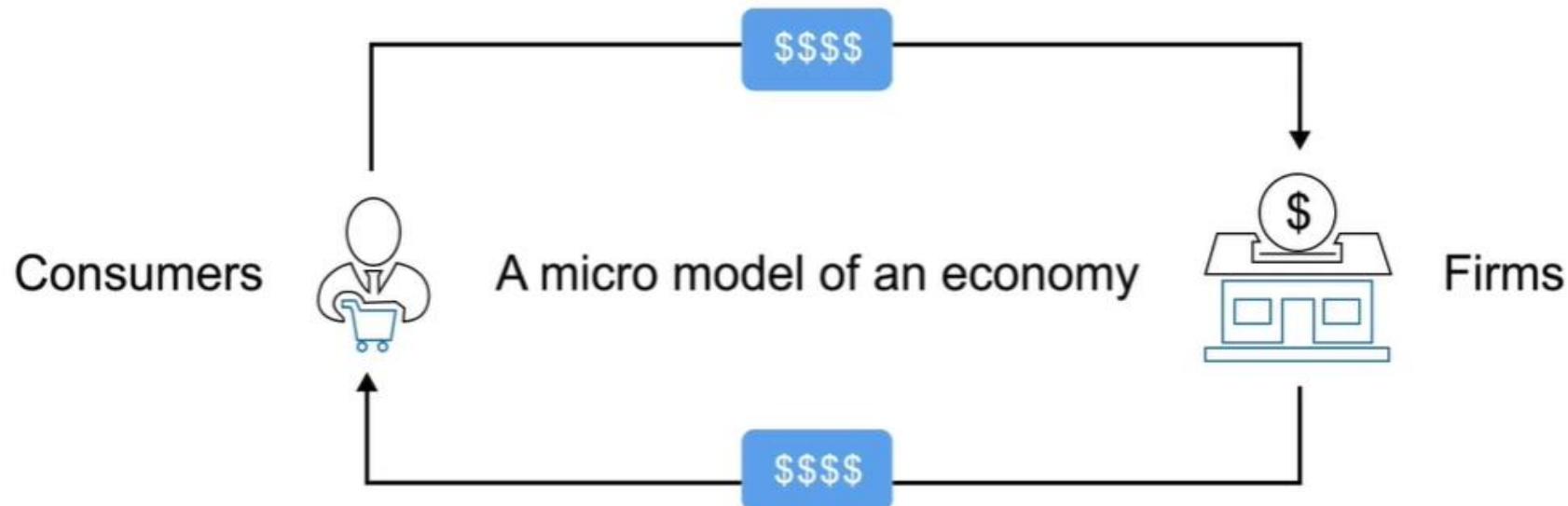
Axiom 1	Axiom 2
Firms exist to maximise economic profits	Consumers exist to maximise utility

An axiom is a statement that can neither be proved nor disproved

The Objective of a Business

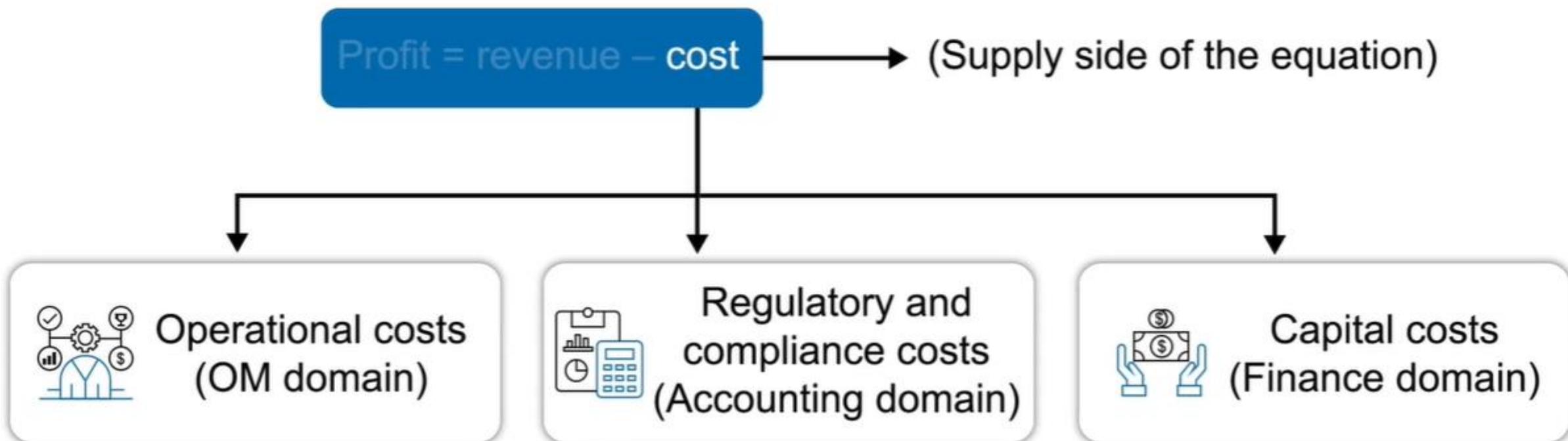
The two fundamental economic axioms

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The Objective of a Business

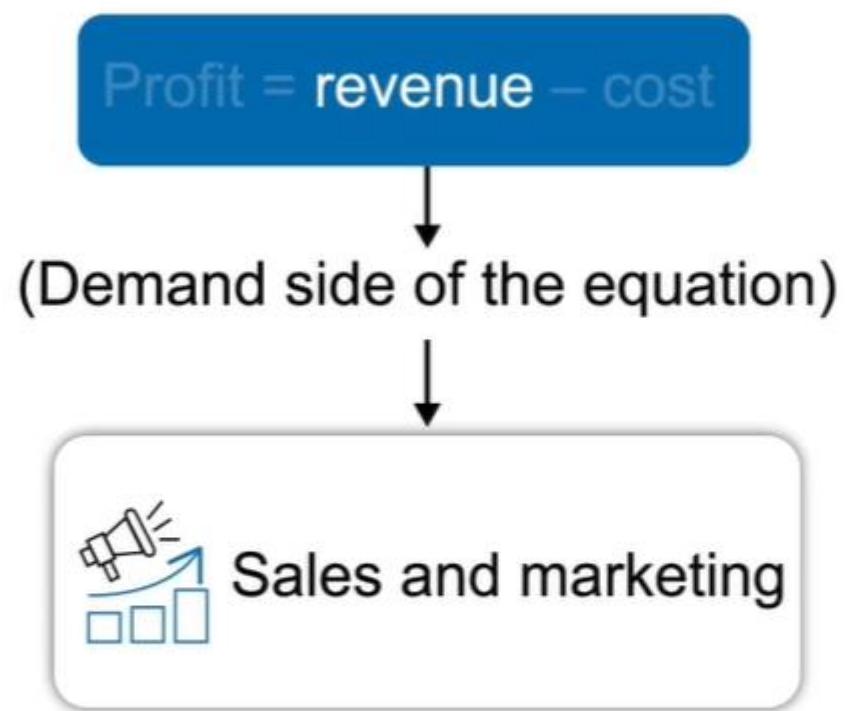
Maximising profits



Each domain tries to minimise its respective costs in order to meet the goal of maximising profits

The Objective of a Business

Maximising profits

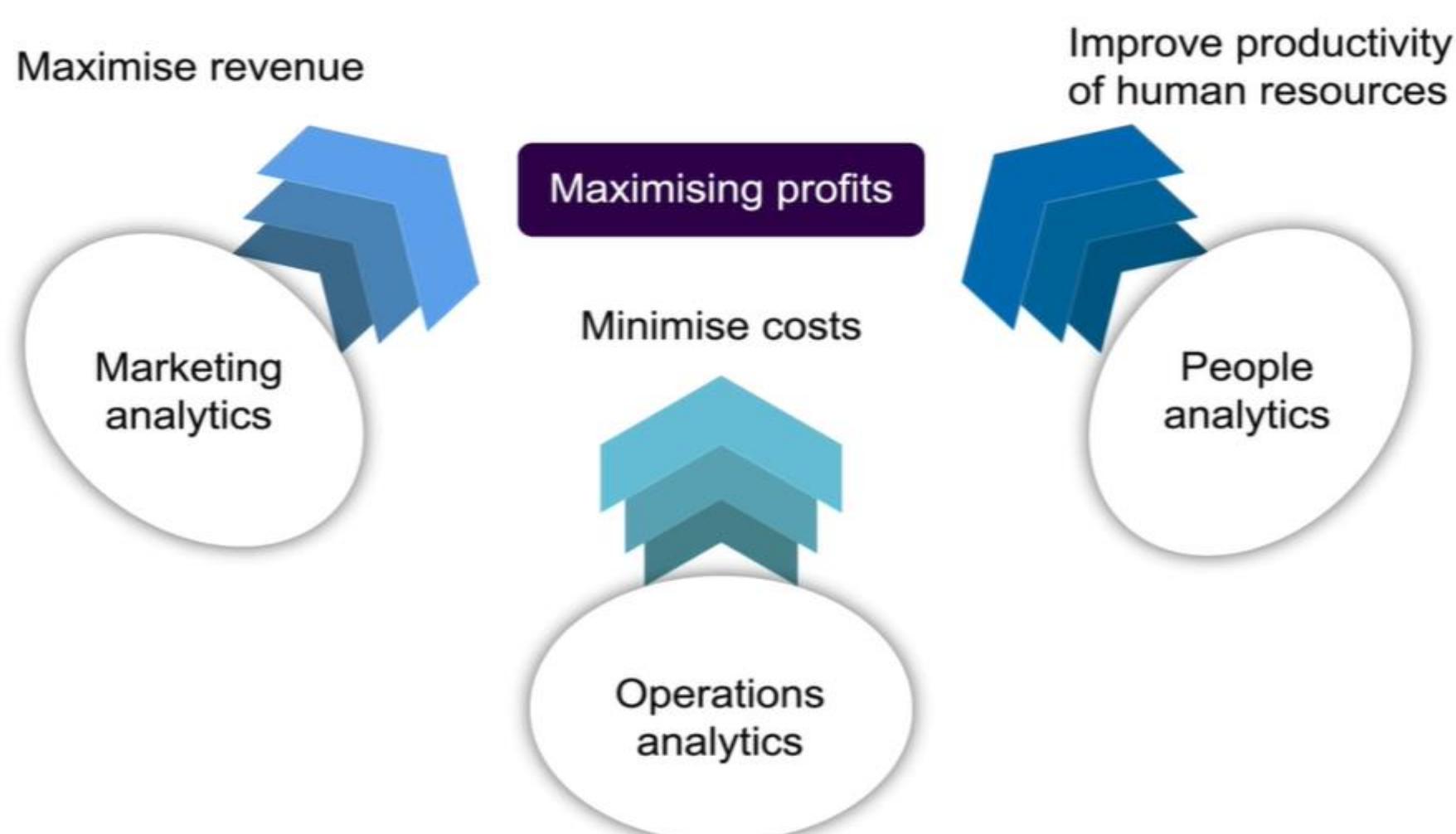


Sales and marketing contributes to maximising profit by maximising revenue

Different Functions of a Business

- Different business functions including sales and marketing, operations, accounting, corporate finance constitute an enterprise
- These are different functional parts of an enterprise
- Each of these functions yields its own analytics

How Different Functions Contribute to Profits

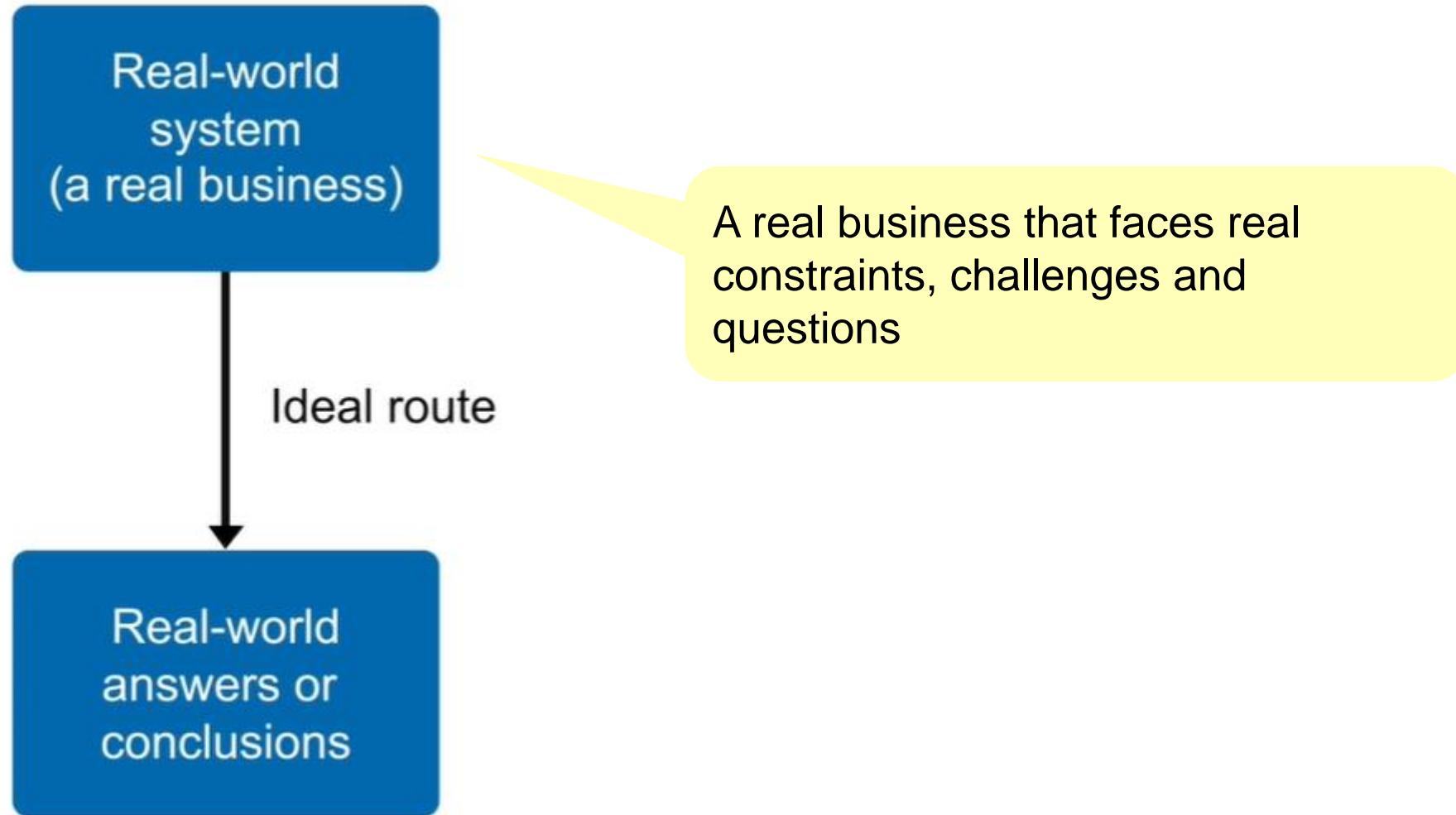


Market Power of a Firm

A firm will have market power if it does well on the demand side or the supply side

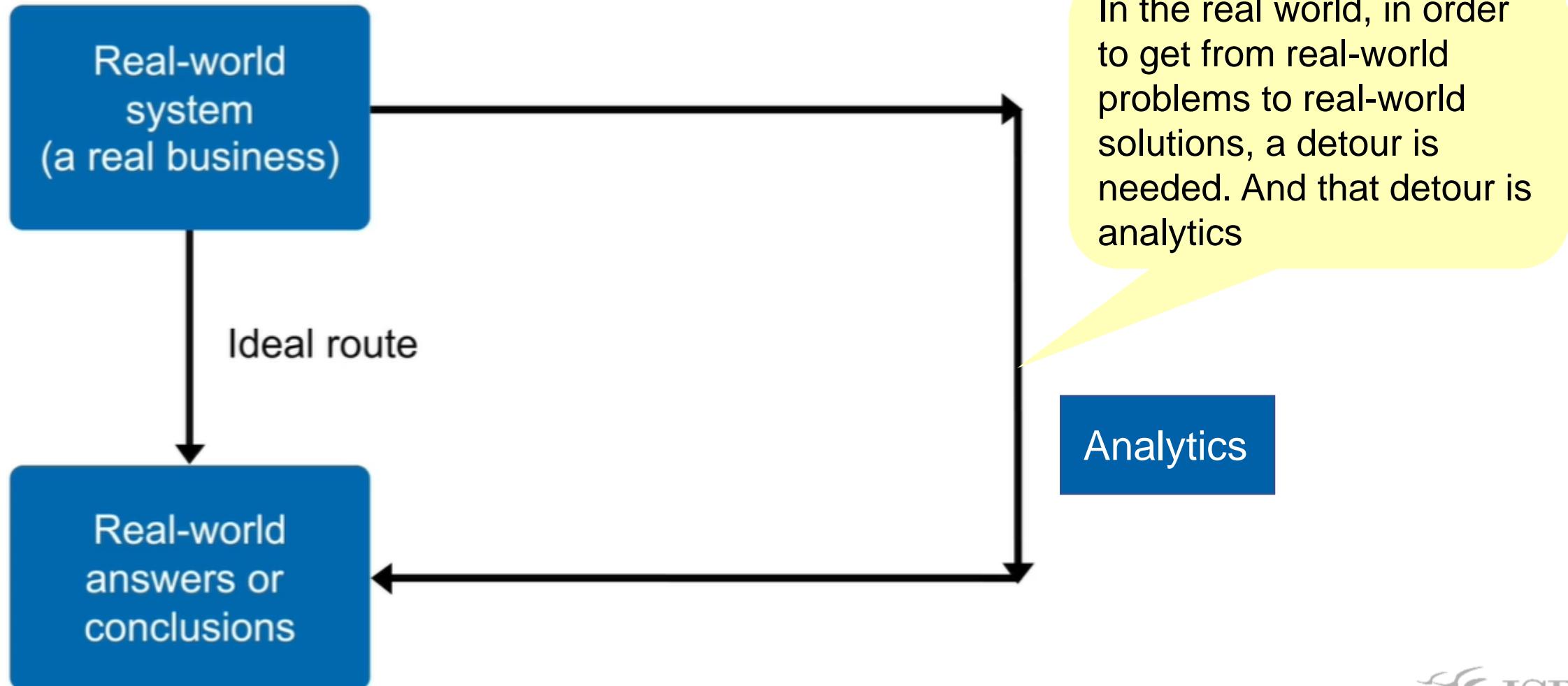
What Is Analytics

The anatomy of analytics



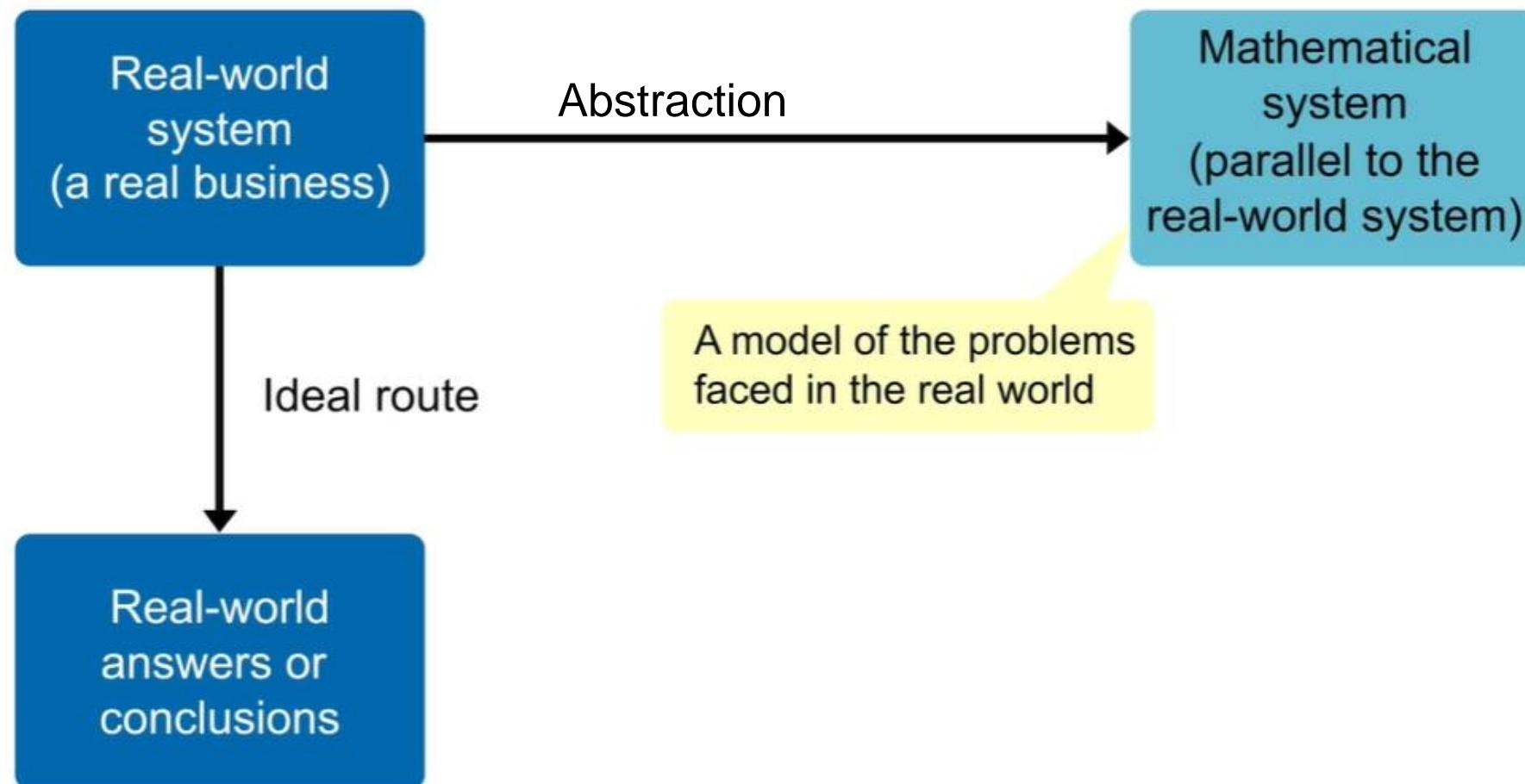
What Is Analytics

The anatomy of analytics



What Is Analytics

The anatomy of analytics

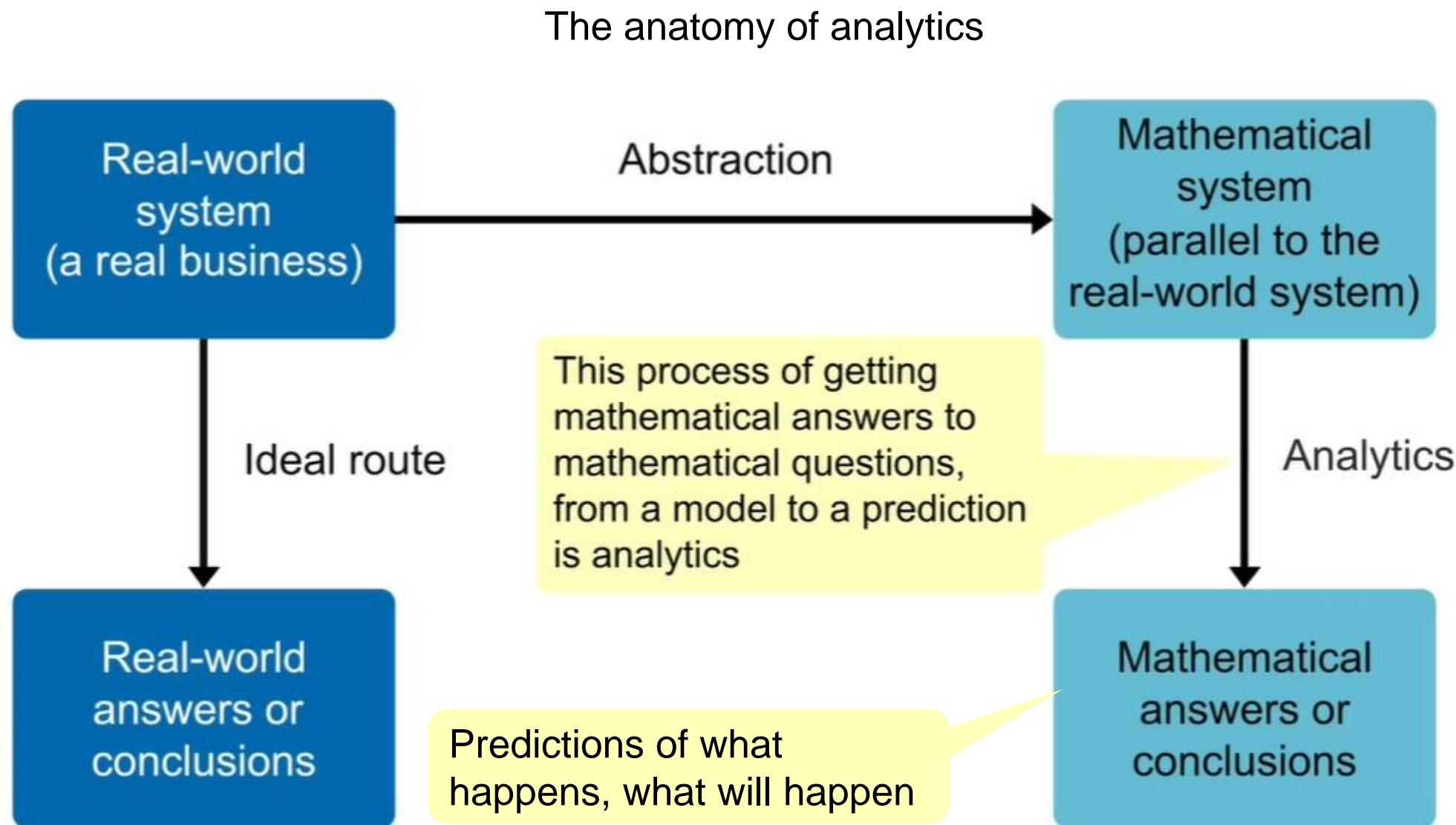


How the Detour of Analytics Works

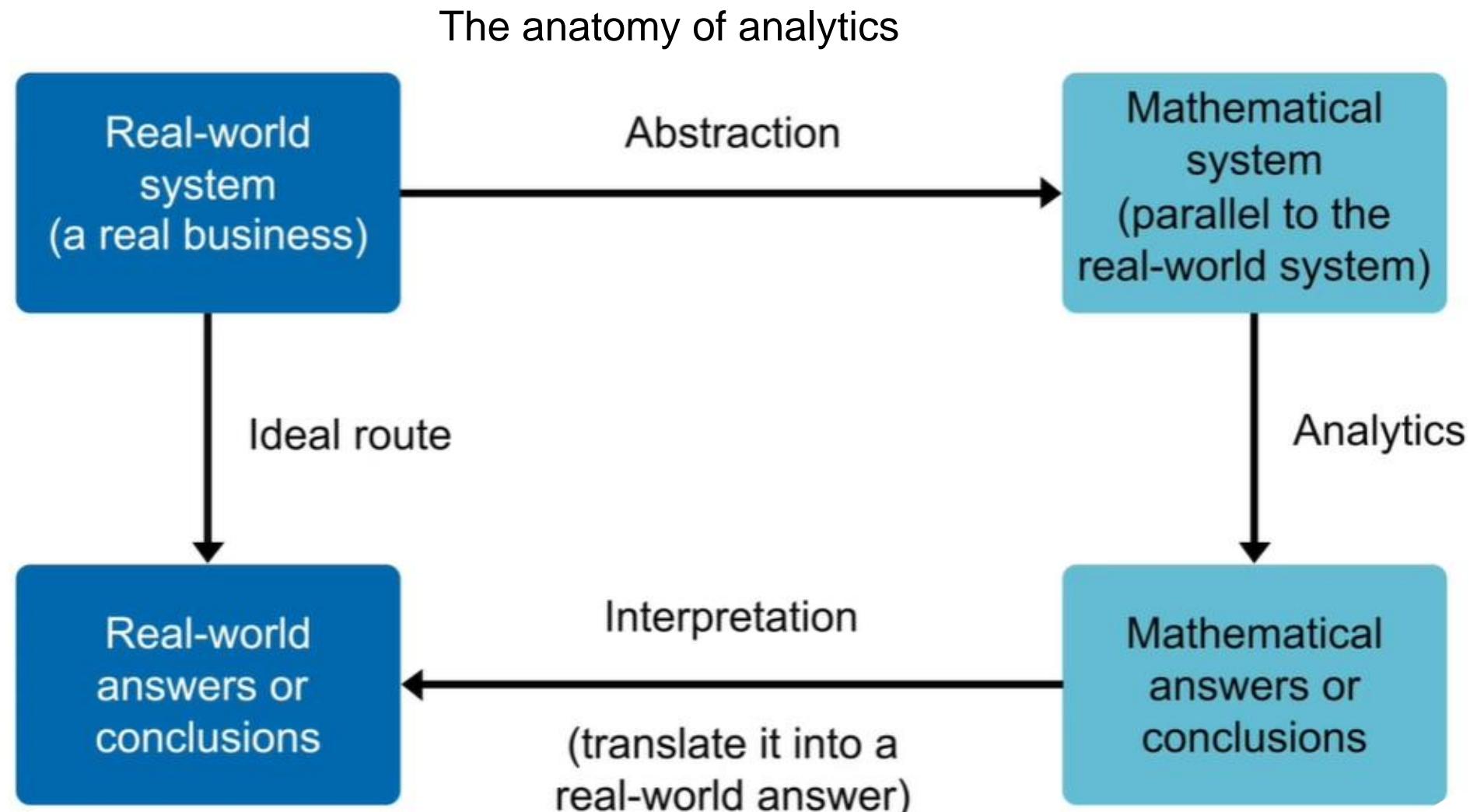
In order to get from real-world problems to real-world solutions, a mathematical model needs to be built

Building a model is necessary
because it helps explore
different possibilities and
optimise outputs

What Is Analytics



What Is Analytics

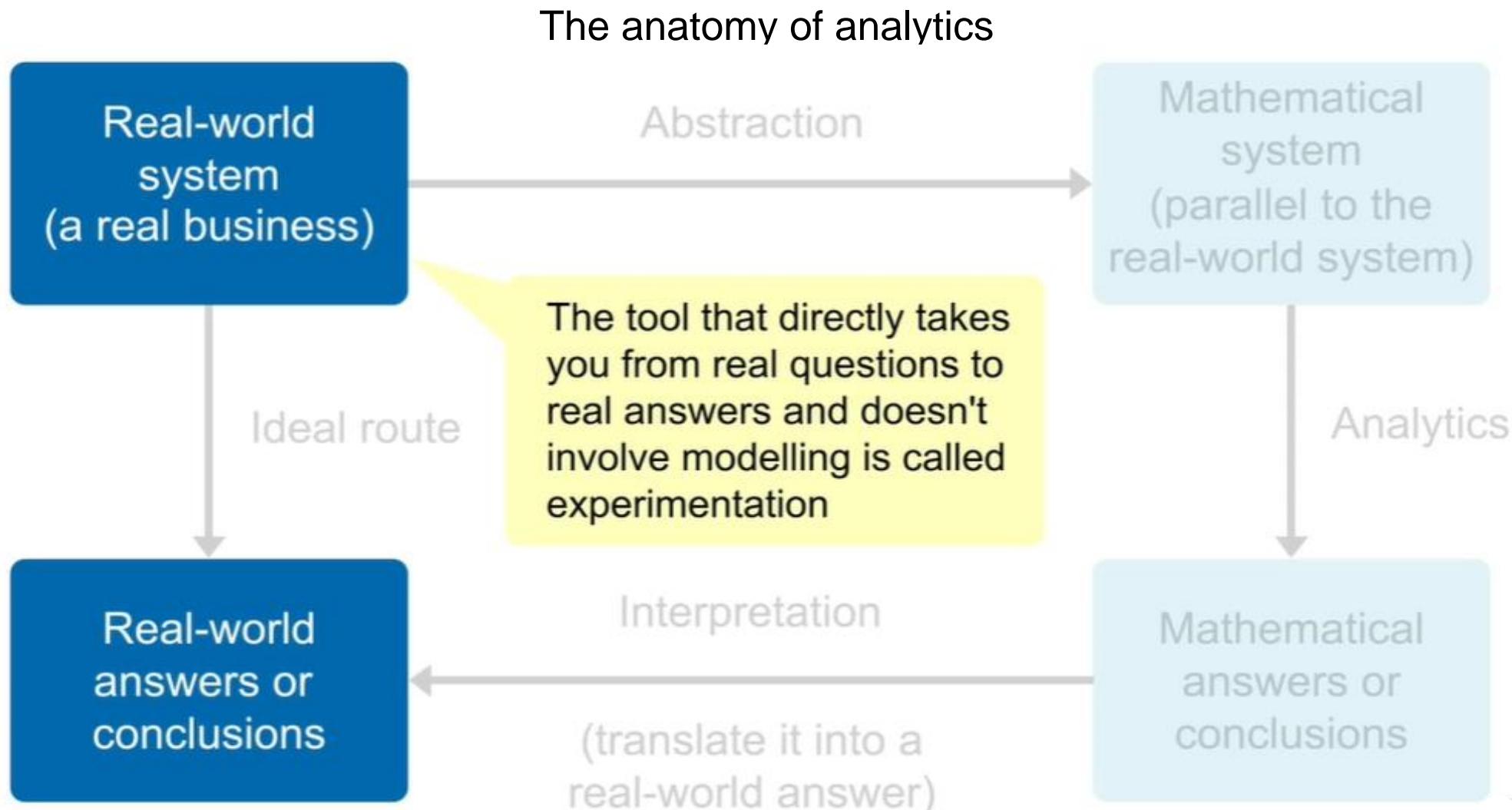


Why Analytics Is Useful

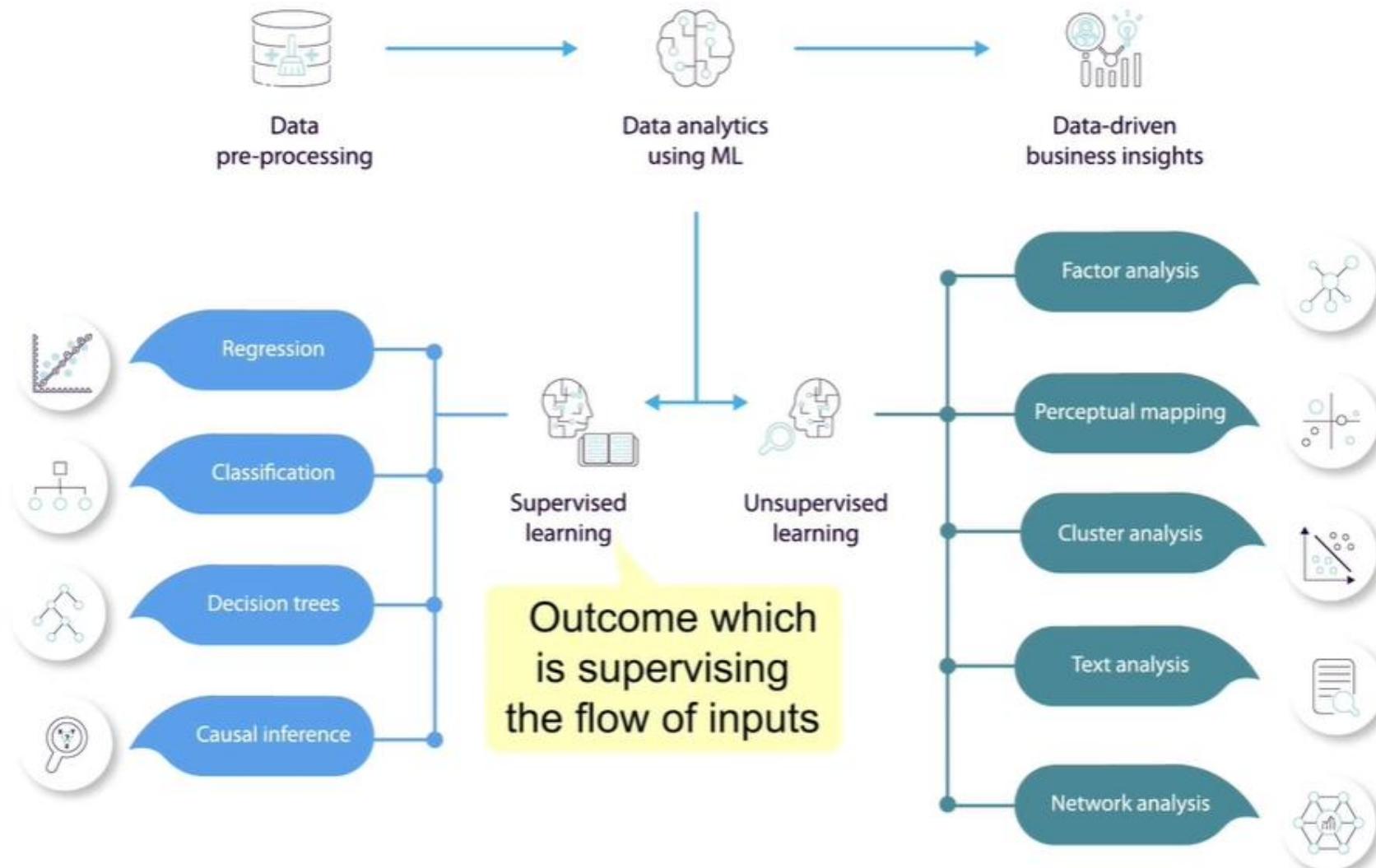
Analytics de-risks your job because:

It allows you to make mistakes in a mathematical model without any real-world cost implications

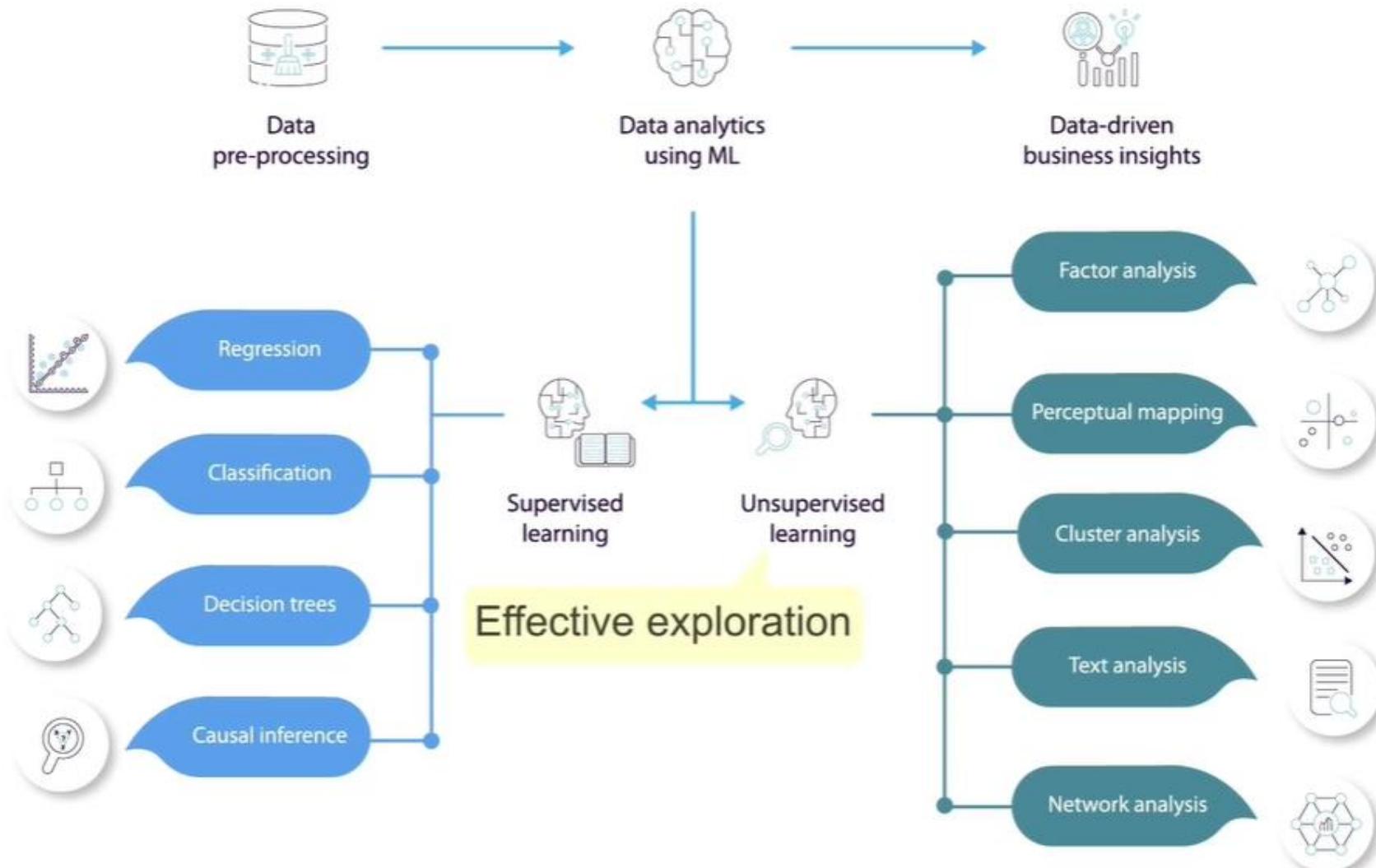
What Is Analytics



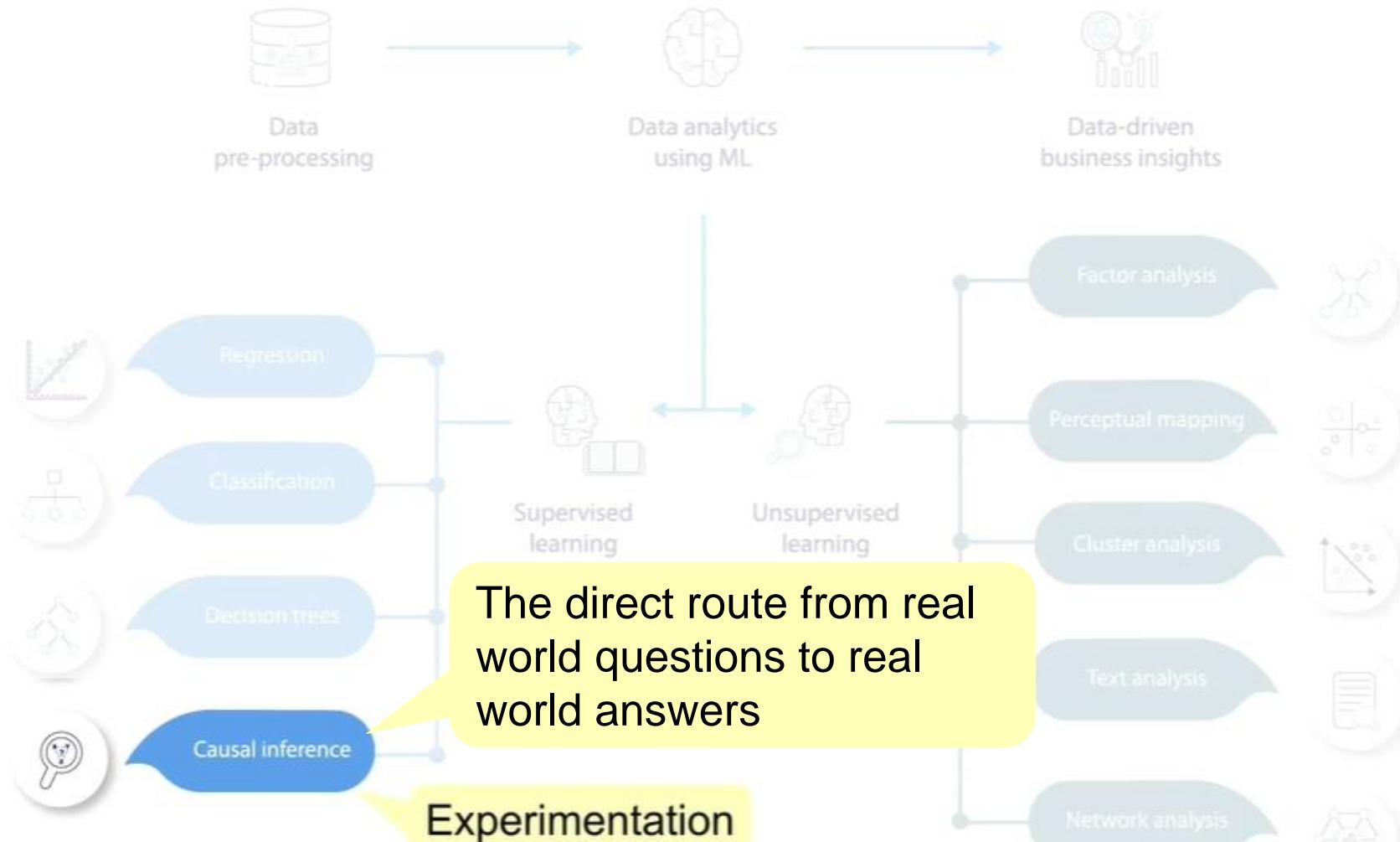
Business Analytics Programme Map



Business Analytics Programme Map



Business Analytics Programme Map



Problem Formulation

**Problem formulation is like building a box
within which we find solutions**

- A box that's too small will not provide optimal solutions
- A box that's too big will become time and effort-intensive to find the right solution

The Anatomy of a Business Problem

Musings of a decision-maker and country head for a corporation

“Sales fell short last year. But sales would've approached target except for six territories in two regions where results were poor. Of course, we implemented a price increase across-the-board last year, so our profit margin goals were just met, even though sales revenue fell short. Yet, two of our competitors saw above-trend sales increases last year. Still, another competitor seems to be struggling, and word on the street is that they have been slashing prices to close deals. Of course, the economy was pretty uneven across our geographies last year and the two regions in question, were weak anyway, particularly so last year. Then there was that mess with the new salesforce compensation policy coming into effect last year. One of the two weak regions saw much salesforce turnover last year.....”

Business Problems

Question on Business Problems

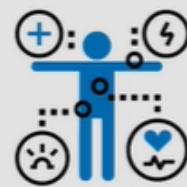
Is reality orderly?

- No. Reality is not orderly, but messy
- Life seldom simplifies issues (As in textbook, once you see the question, you know the solution is given in the question itself along with the book and the chapter it is from. So, you know which way to start thinking. But in life it is exactly opposite.

You don't even know where to start with

The Medical Analogy for Business Problems

Symptoms versus ailments



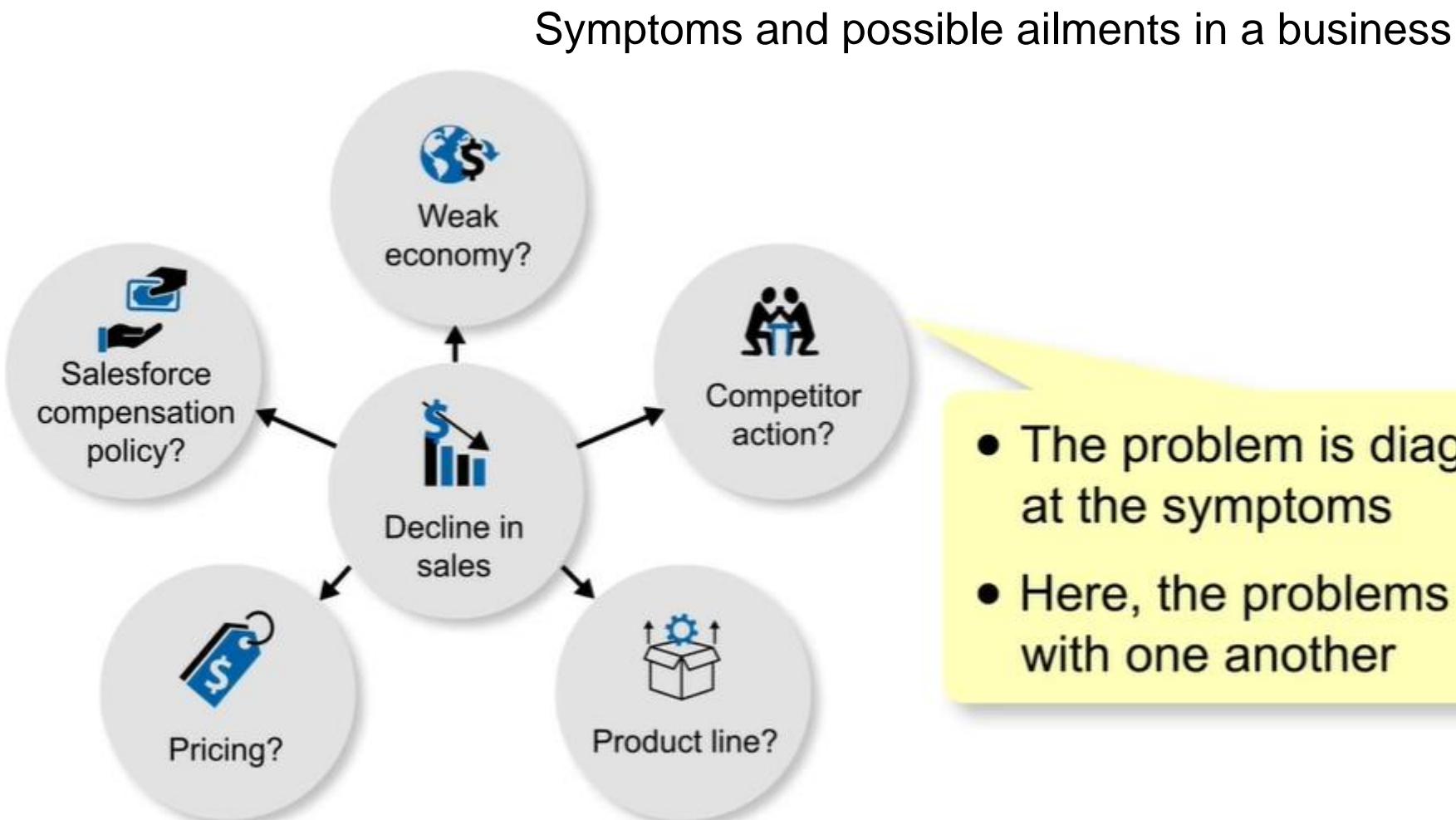
A symptom is the visible manifestation of an underlying problem (the ailment)



If a business is sick, fighting the symptoms may not be enough to cure the ailment

Decline in sales is the most common symptom of a sick business

The Medical Analogy for Business Problems



- The problem is diagnosed by looking at the symptoms
- Here, the problems could be interacting with one another

The Medical Analogy for Business Problems

- In businesses, using data can be a fast and cheap tool to narrow the field of ailments (problems)
- The data required to solve the problem depends on how the problem is formulated

Problem Formulation Framework



Messy reality

In an ideal world, one can go straight from messy reality to toolbox. However, things are not always ideal

Toolbox

Problem Formulation Framework

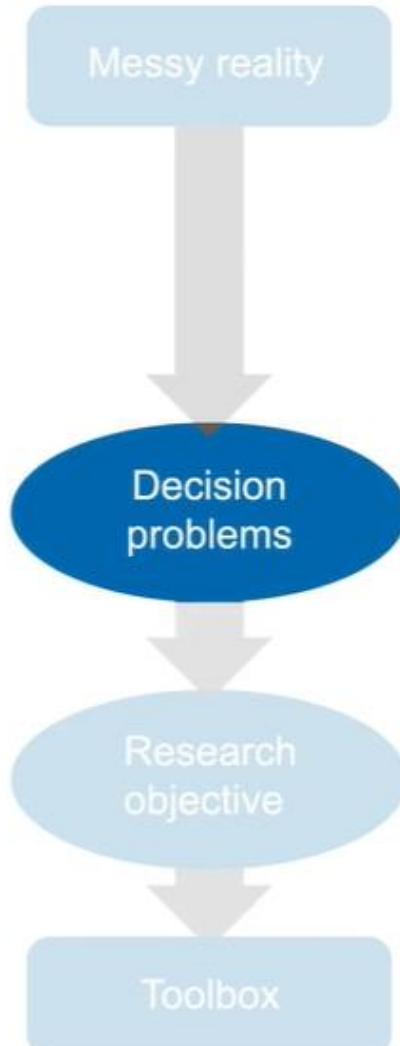
Probable causes:

- Product line is obsolete
- Customer connect is ineffective
- Product pricing is uncompetitive



These come from exploratory research which helps shortlist the major probable causes

Exploratory tools help us extract major problem heads from messy reality
Once we do this, we move from messy reality to a decision problem



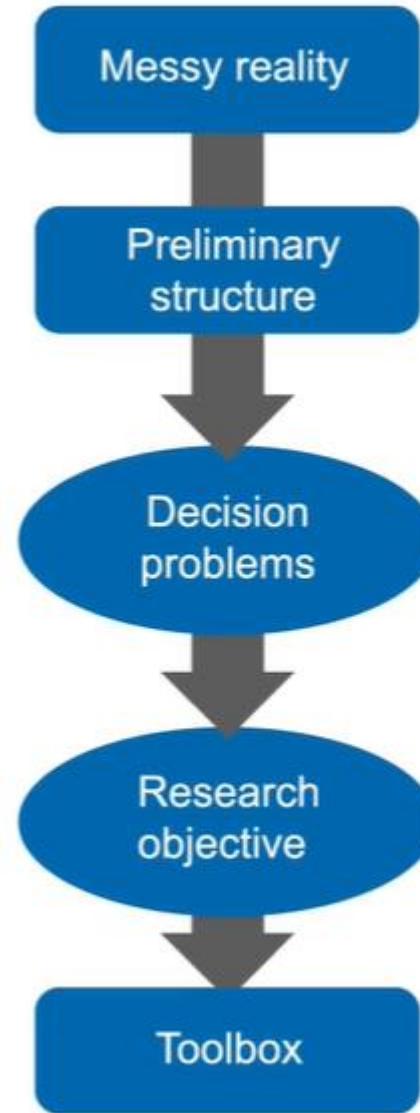
Problem Formulation Framework

Once we move from messy reality to a set of DPs, we impose preliminary structure on messy reality

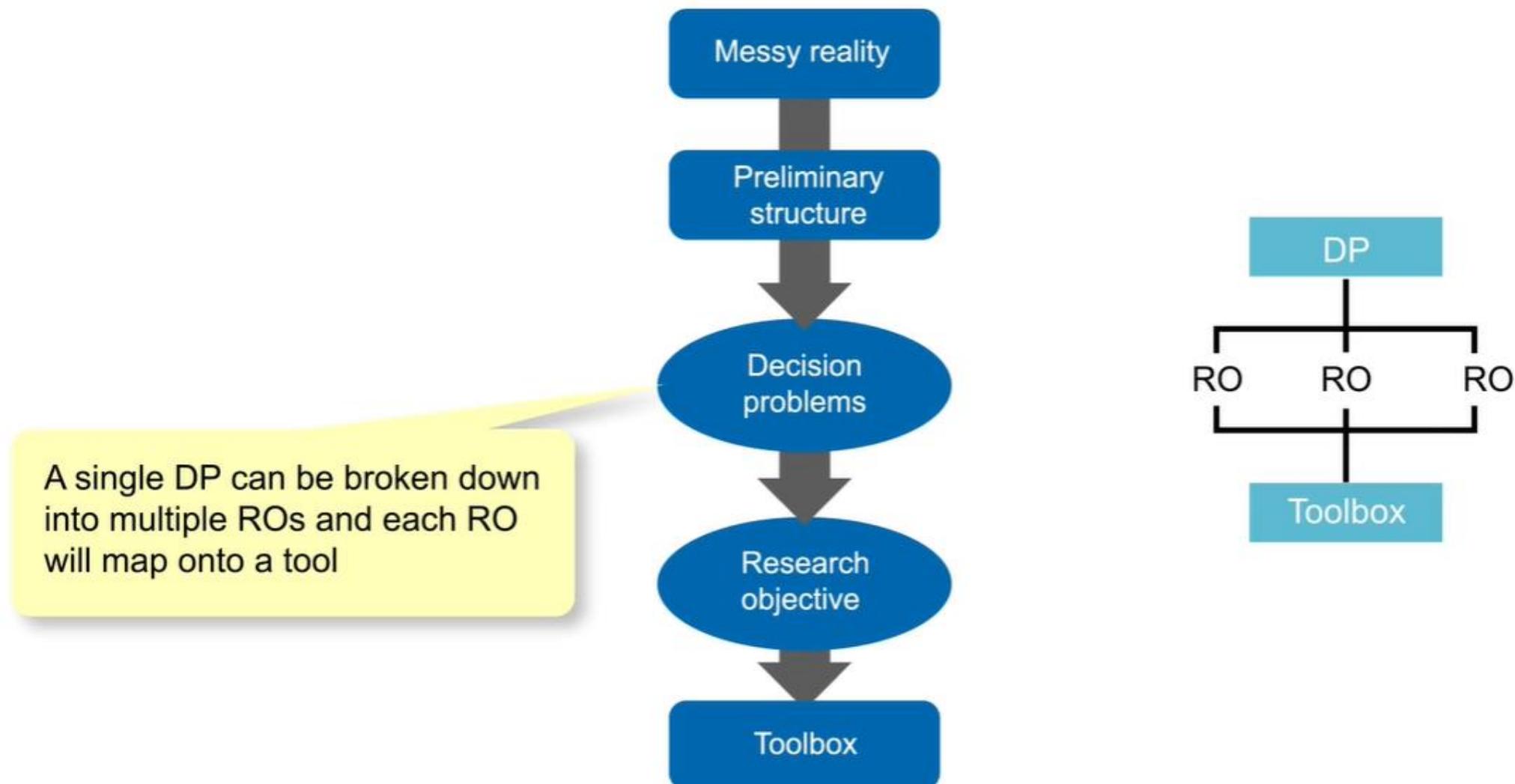


Problem Formulation Framework

- May not contain sufficient information to map directly onto tools
- So another level of refinement may sometimes be needed (RO)

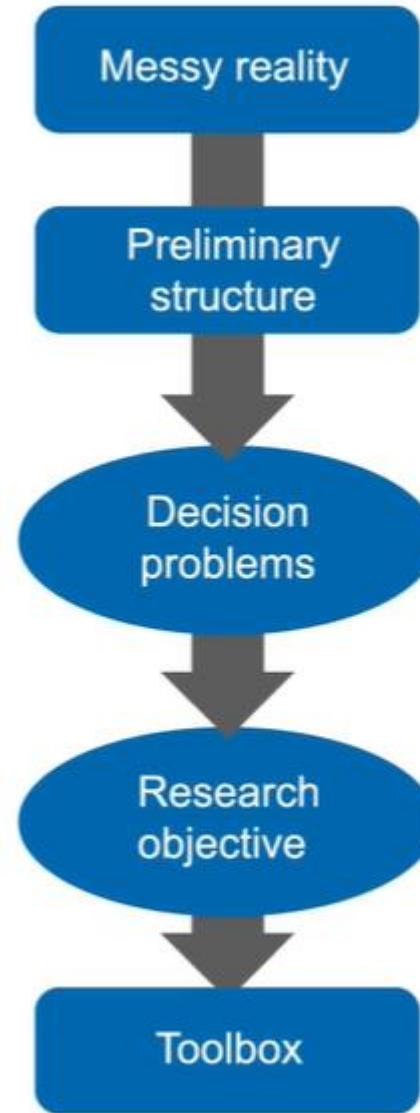


Problem Formulation Framework



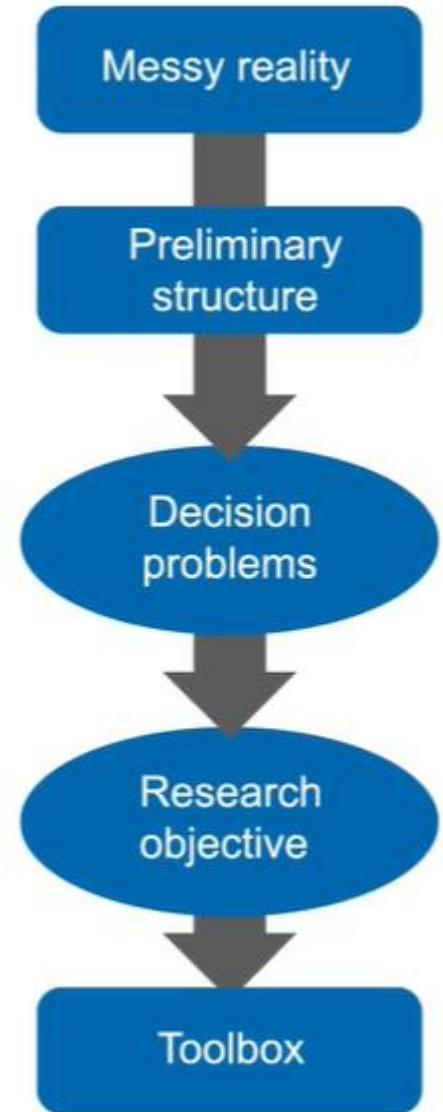
Problem Formulation Framework

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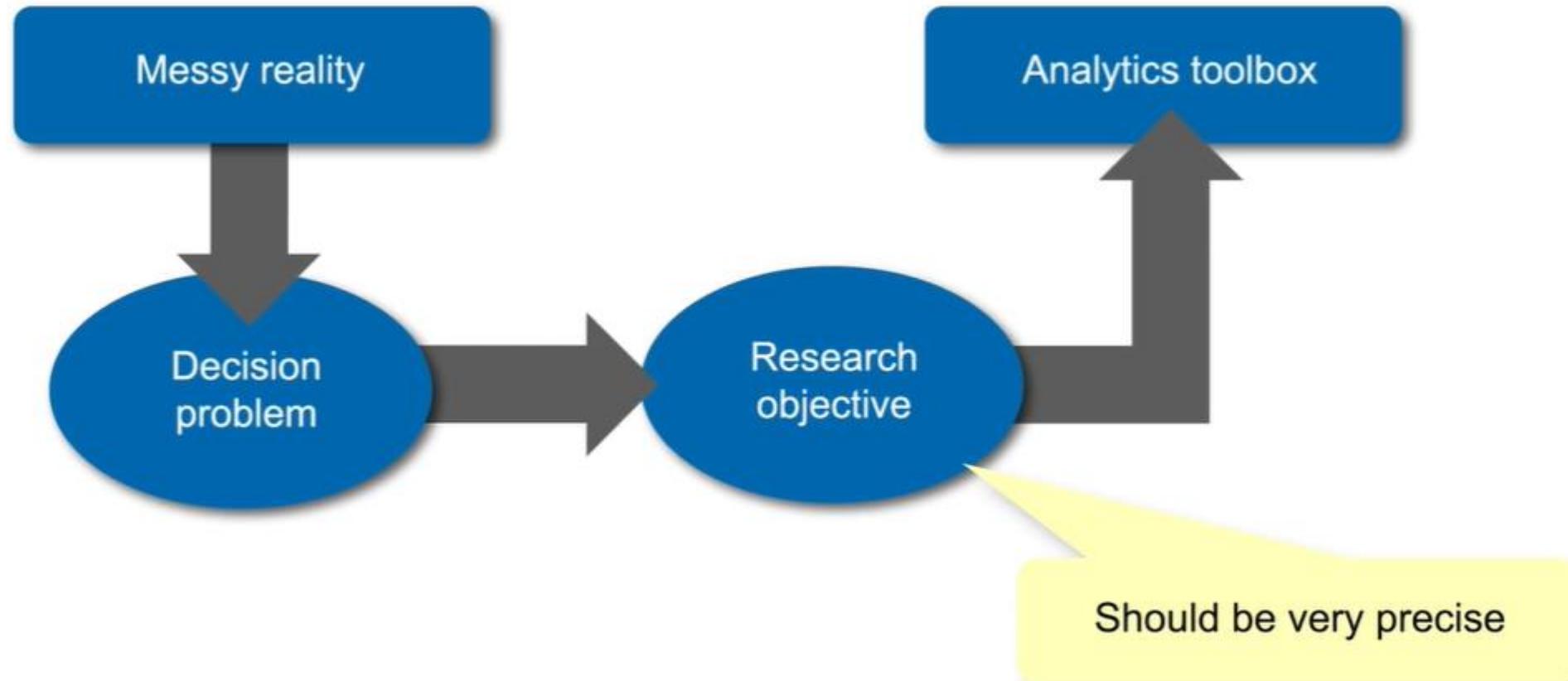


Problem Formulation Framework

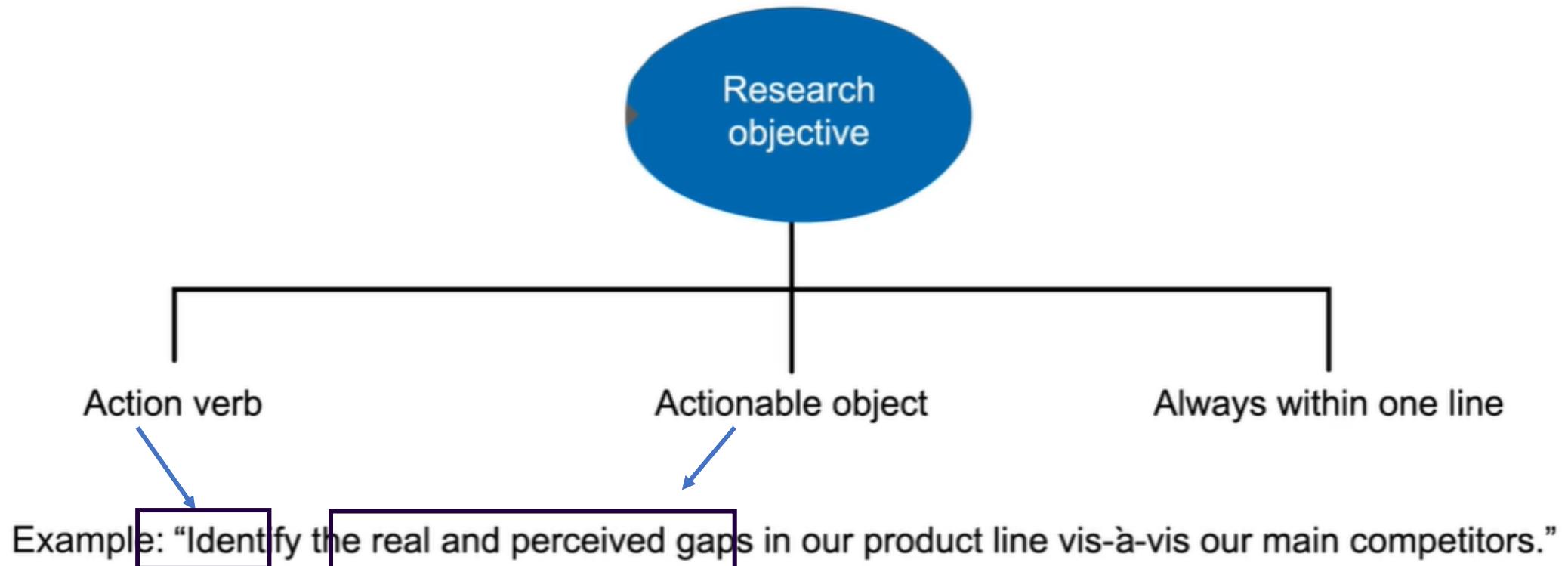
The two-way stations to move from messy reality to analytics tools are decision problems and research objectives



The Anatomy of a Research Objective



The Anatomy of a Research Objective



Problem Formulation: Examples of Research Objectives

	Verb	Actionable objects
1	Identify	Problems, opportunities, choice criteria...
2	Define	Concept, design, potential...
3	Describe	Design process, usage, work environment...
4	Explore	Perceptions, reactions, remedies...
5	Generate	Hypotheses, alternatives, explanations...
6	Evaluate	Feasibility, attractiveness, potential...
7	Select	Product, concept, ad execution...
8	Test	Preference, direction, ad execution...
9	Measure	Growth, size, frequency...
10	Prioritise	Segments, needs, opportunities...
11	Monitor	Trends, competition, events...
12	Track	Spending, satisfaction, awareness...

Once the RO is clearly defined, the toolbox is ready to open and deploy

What is Business Analytics: Summary



iCow's example demonstrated that analytics can be very powerful to solve real-world business problems even in a low-tech environment



Conceptual preliminaries to understand the anatomy of a business are based on questions such as “what are the objectives of a business?” and “what is data analytics?”



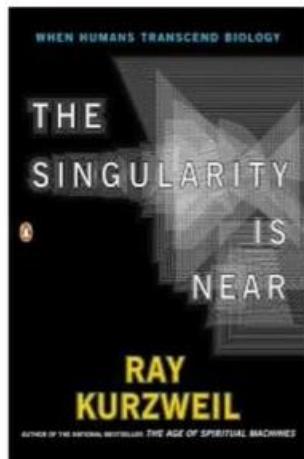
The problem formulation framework for solving business problems demonstrated that it is important to precisely formulate problems so that the tools that we have will have maximum ROI

Humans and Machines



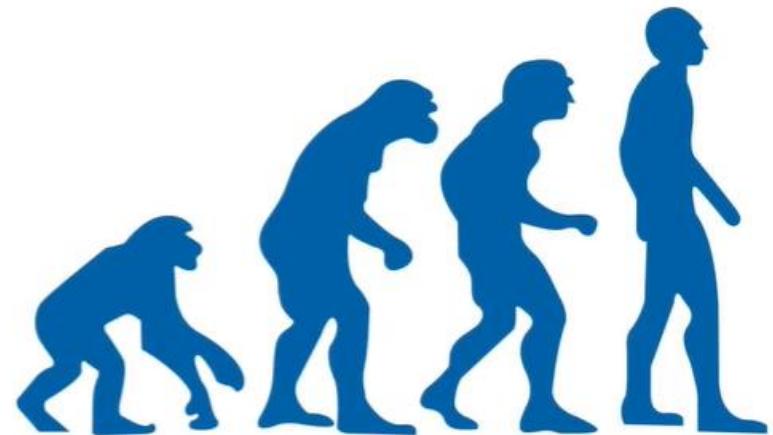
“Our technology, our machines,
are part of our humanity. We created
them to extend ourselves, and that is
what is unique about human beings!”.
– *Ray Kurzweil*

Humans and Machines



A book by Ray Kurzweil

The book talks about,



Evolution of AI

How Human Tools and Machines Have Evolved



Hand-held tools



Big and heavy machines



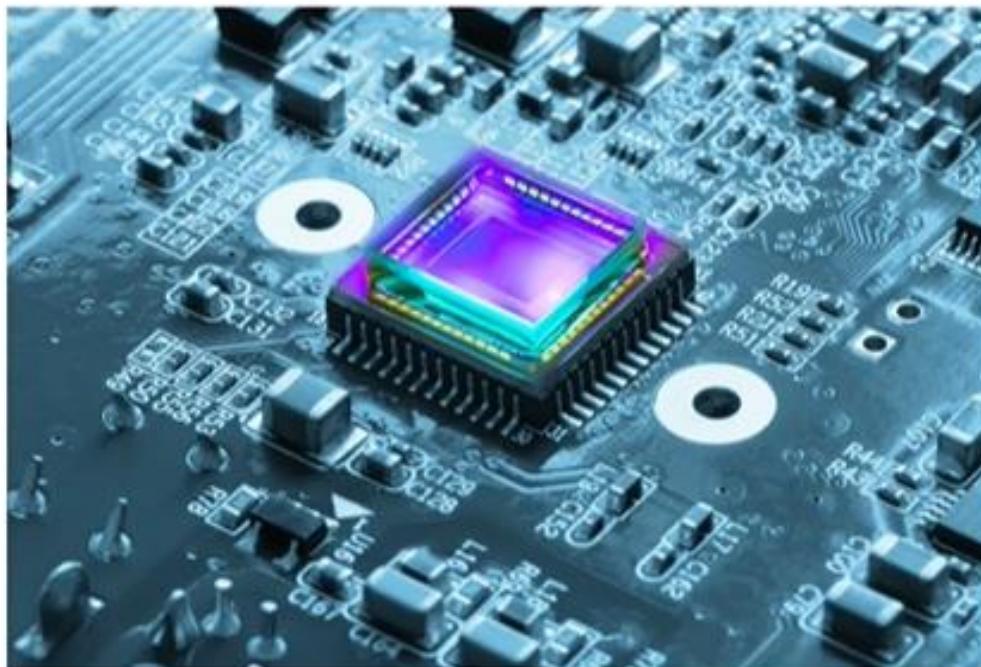
Lathe machines



Automated tools

How Human Tools and Machines Have Evolved

Sensors and communication devices have evolved over time



Better sensors



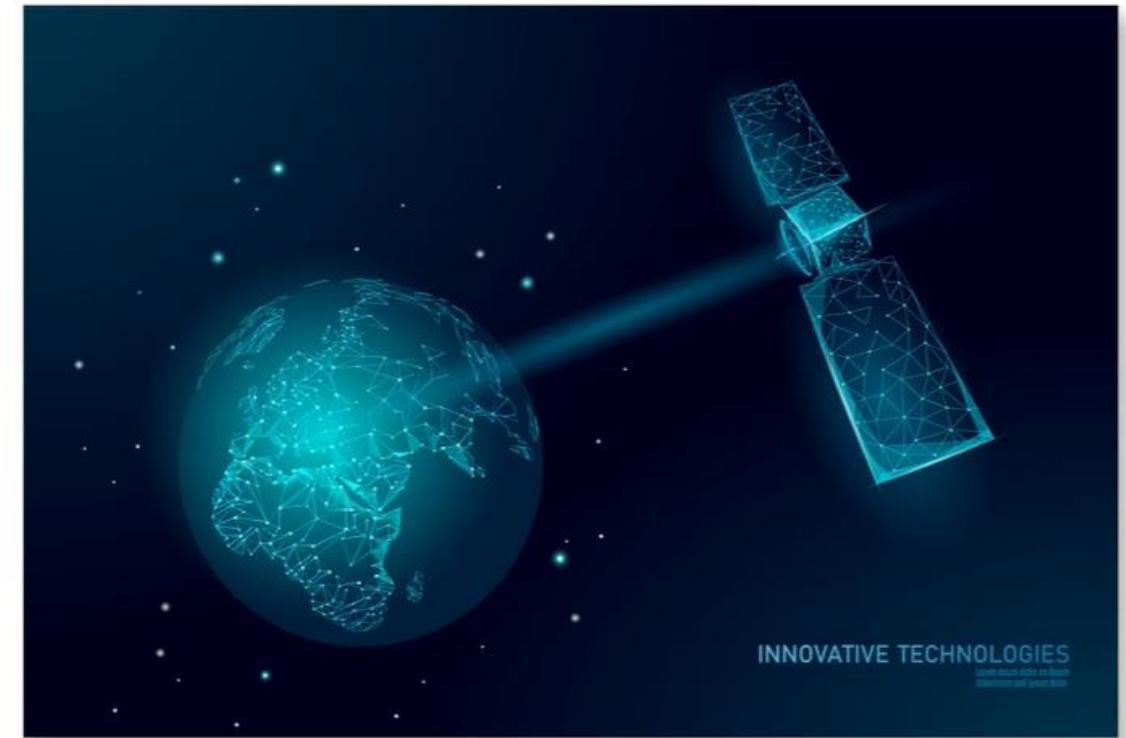
Smart communication devices

Technological Advancements

It is now possible to transfer terabytes of data and collect data from space



Data transfer through undersea cables



Data collection through satellites

Advancement in Computational Powers



From a room-sized computer



To a tiny microchip



Quantum computers to boost computation

Advancement in Computational Powers

It is now possible to store zabytes of data through cloud servers



Cloud servers to store data

Data Generation

Data is generated through:



Grocery stores
transactions



Credit card
transactions

638011	- 27640	%	▲ 28021	+ 11.02
470185	- 11822	%	▲ 14.086	+ 5.029
157900	- 61830	%	▼ 01.228	- 0.009
118.223	- 20.586	%	▲ 10.637	+ 0.821
327640	- 72.006	%	▲ 12.022	+ 0.536
228.310	- 46.183	%	▼ 09.715	- 0.009
720.586	- 61830	%	▲ 30.821	+ 9.715
972.006	- 37.900	%	▲ 18.223	+ 1.235
379.542	- 38.011	%	▲ 05.680	+ 1.222
100.089	- 74.480	%	▼ 14.060	- 0.004
461830	- 20.586	%	▲ 11.822	+ 4.086
505680	- 32.764	%	▼ 01.854	- 0.003
281.744	- 27.640	%	▲ 05.328	+ 9.704

Stock exchange



Satellites

Data Generation

Data is generated through:



Blogs and social media



The Internet (roughly 2.5 quintillion bytes of data every day)



Internet of things (iot)

Using Data

Data needs to be organised in
order to find patterns or derive
insights from it

What is Artificial intelligence?

Artificial intelligence refers to the simulation of human intelligence in machines that are programmed to think like humans and mimic their actions

How AI Mimics Human Actions

Intelligence is the ability to predict or assign a label to a “new” observation based on the model built from past observations

j k l m n o P Q R S
J L M N O P Q T S
J L m n o P q r s
J k / m n o P q r s
J K { M N O P Q r s
J K L m n o P Q r s
j K L m n o P Q r s
J K 1 M N O P Q R S
J K / m n o P Q r s
J R / m n O P e r s

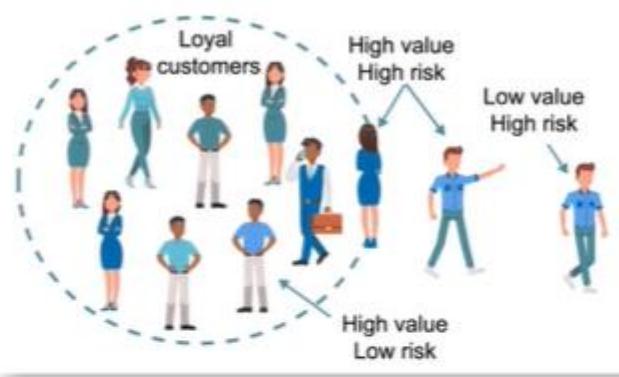
Can machines too identify letters?

Examples of AI

Can machines identify or predict



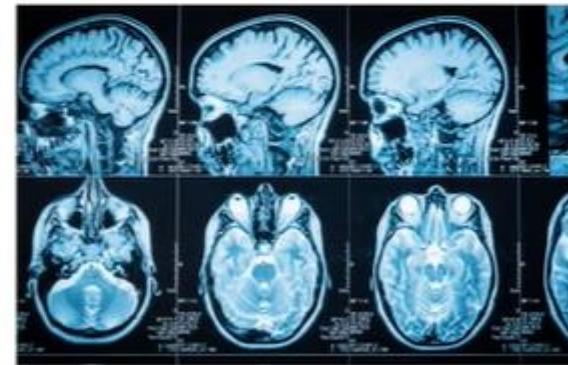
If this a fraudulent transaction?



Which customers are going to churn?



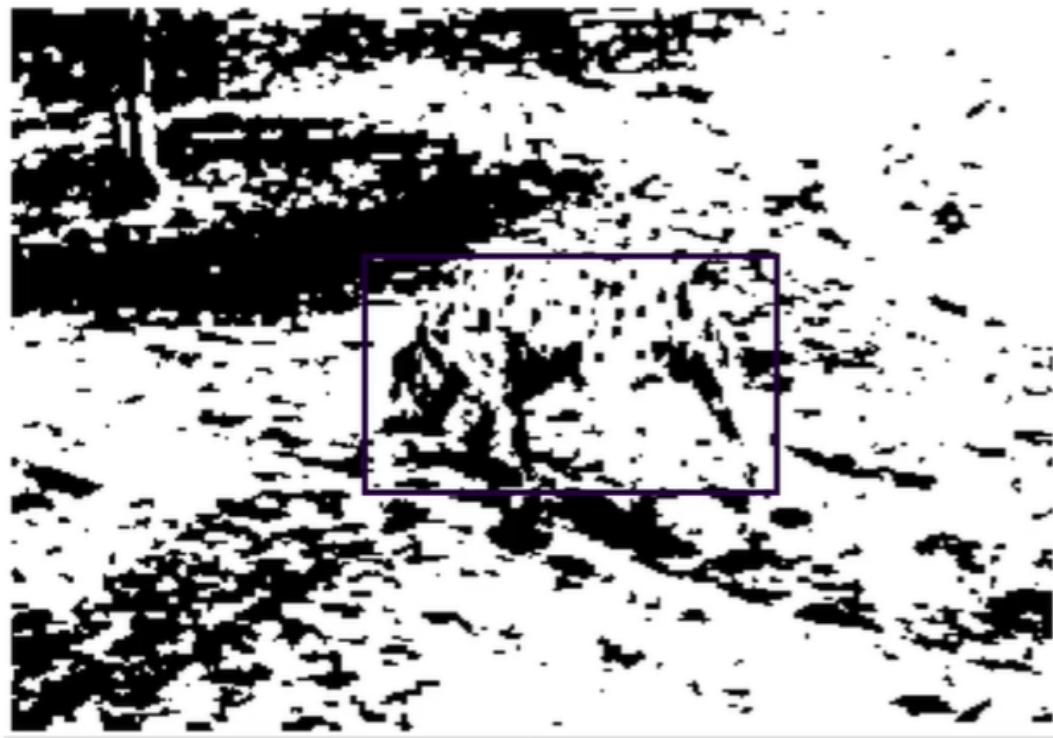
Who will respond to these coupons?



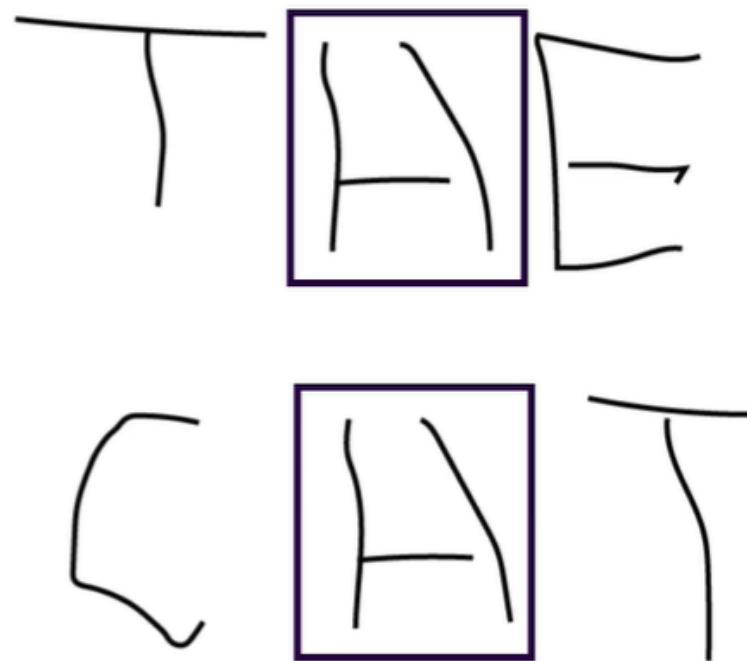
If there is a medical condition?

Examples of AI

Can machines identify or predict



An image of the pixels



What Is Intelligence

The ability to make sense of an individual component in the context is what intelligence is

Examples of AI

Can machines recognise?



Can machines (computer vision) identify different objects in a given image?

Examples of AI

Can machines recognise?

Named Entity Recognition

When we were in Spain, my mom taught me how to drive with a car.

Entity Mention Detection

When **we** **ORG** were in **Spain** **GPE**, **my** **PER** **mom** **PER** taught **me** **PRP** how to drive with **a car** **VBN**.

Relation Extraction

When **we** **SUBJ** were in **Spain** **LOC**, **my** **SUBJ** **mom** **OBJS** **SUBJ** taught **me** **PRP** how to drive with **a car** **LOC**.

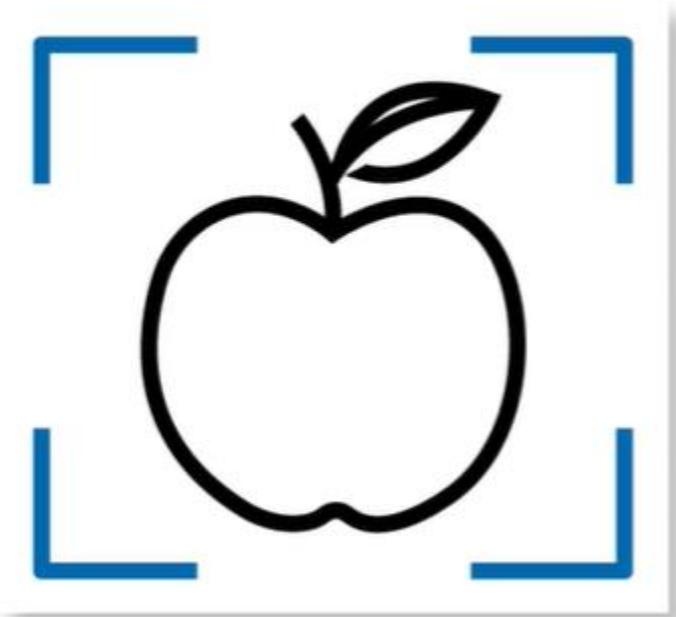
PHYS
PER-SOC
ART

Coreference Resolution

When we were in Spain, **my** mom taught **me** how to drive with a car.

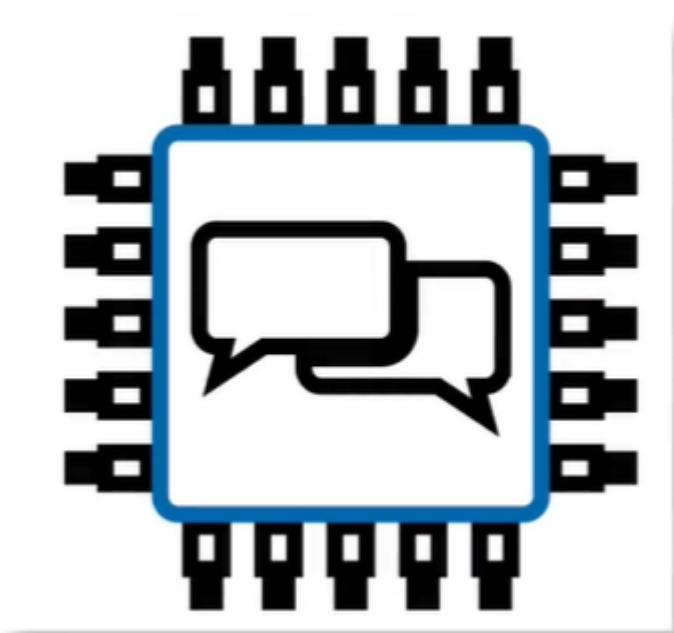
Can machines identify different words or parts of a text and their functions?

Advancements in AI



There are algorithms that help machines look at the pictures and identify objects in the picture

Advancements in AI

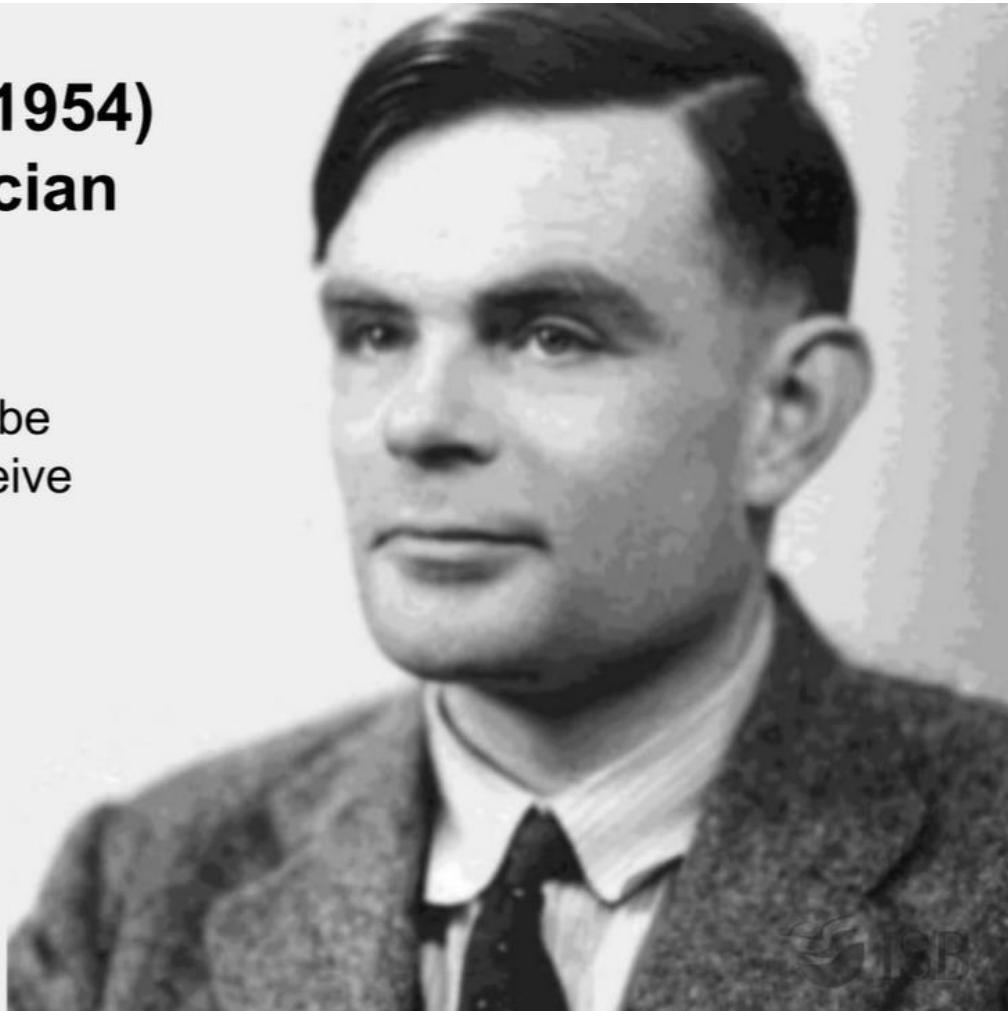


Natural language processing helps machines read and understand the meaning of a text and identify different parts of a text

Turing's Thoughts on an Intelligent Machine

Alan Turing, (1912 – 1954)
Educator, Mathematician

“A computer would deserve to be called intelligent if it could deceive a human into believing that it was human”

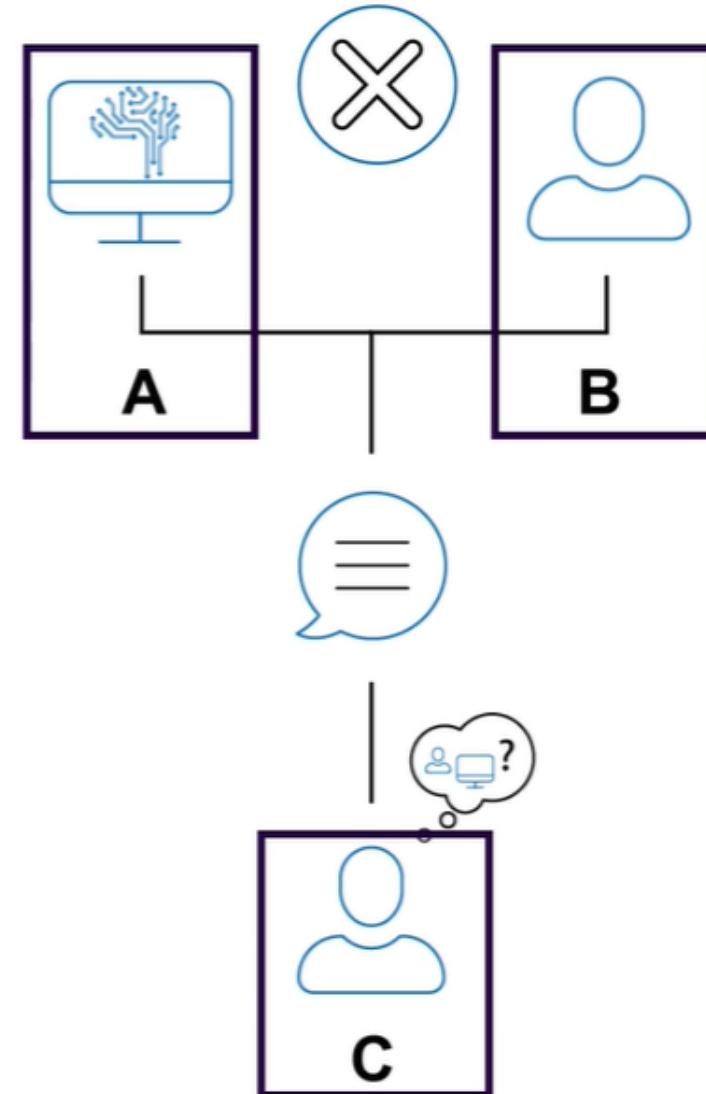


<https://www.biography.com/scientist/alan-turing>



The Turing Test for Intelligence

- C asks the question. Human and machine respond to the question. If person C identifies which is human's response and which is machine's, the test fails
- But if the machine answers and person C cannot identify or differentiate between the two answers, then the machine is considered intelligent



How Machines Have Outsmarted Humans

IBM's deep blue beat the world's best chess player



Algorithm created by Google beat humans in Go, a very complex game



Google's NLP was developed to translate various languages



1997

2011

2016

2018

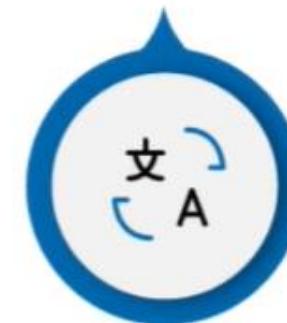
2019



IBM Watson beat humans in Jeopardy game



Portrait created by AI sold for \$432,000



AI-driven robo-taxi made 1000 trips

The Future of AI

“

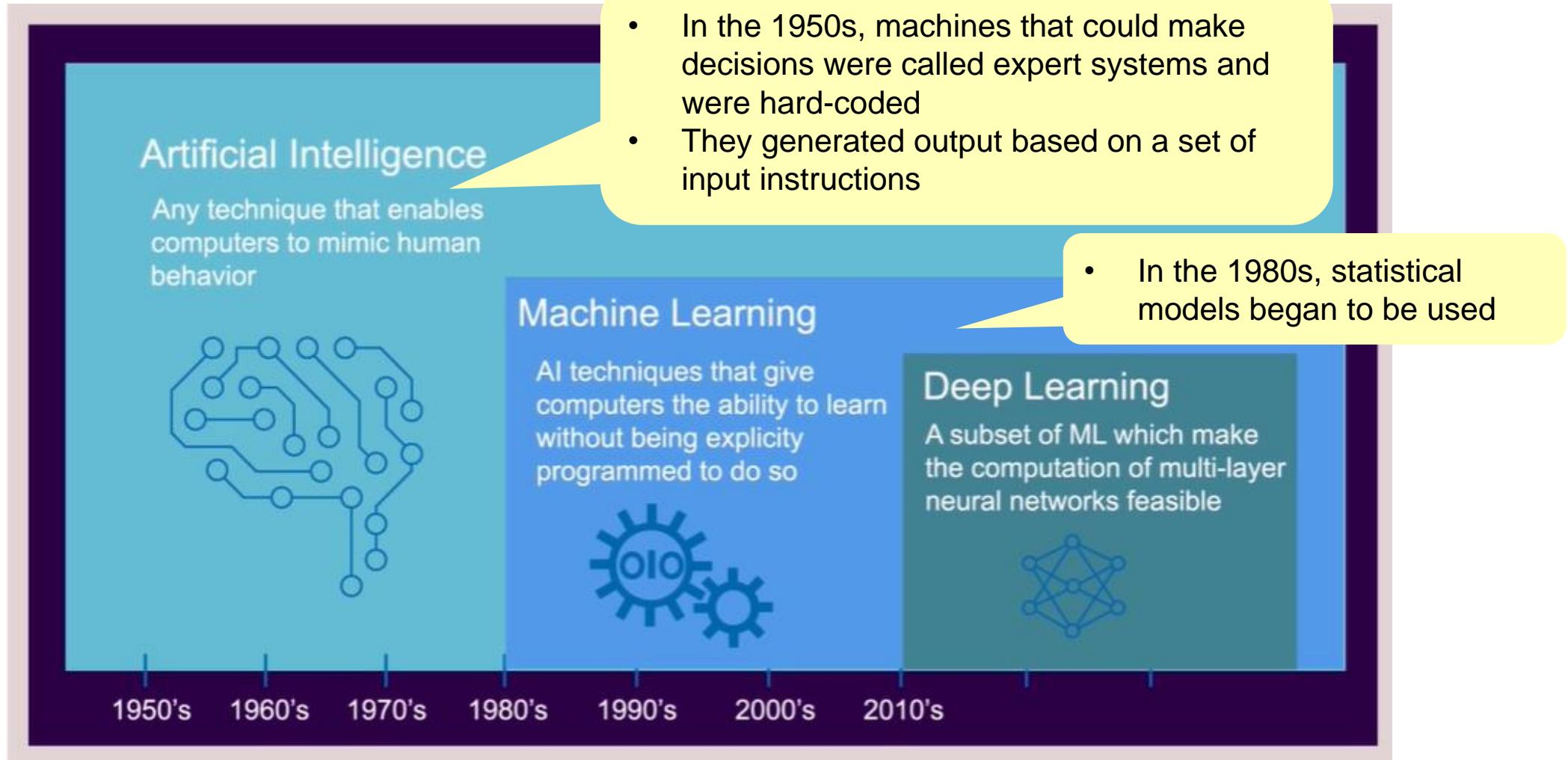
By 2050 robot "brains" based
on computers that execute 100
trillion instructions per second
will start rivaling human
intelligence – Hans Moravec,
March 2009, Scientific American

”

What is Artificial Intelligence

Artificial intelligence is a machine's
ability to mimic human intelligence

Artificial Intelligence Is Not New



Machine Learning vs Deep Learning

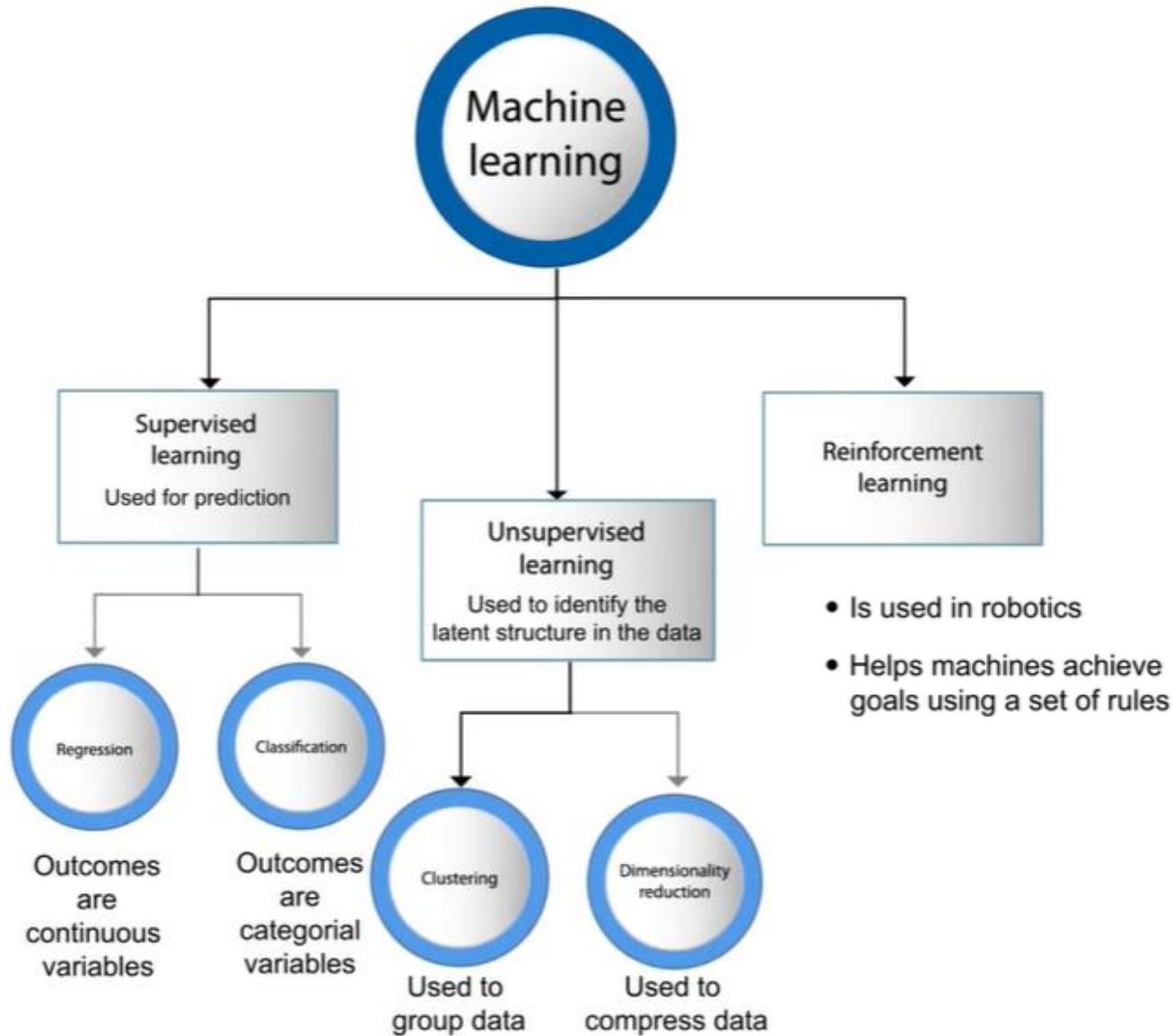
Machine learning

- Takes input in a matrix form—structured data with columns and rows
- Requires prepared structured data for processing

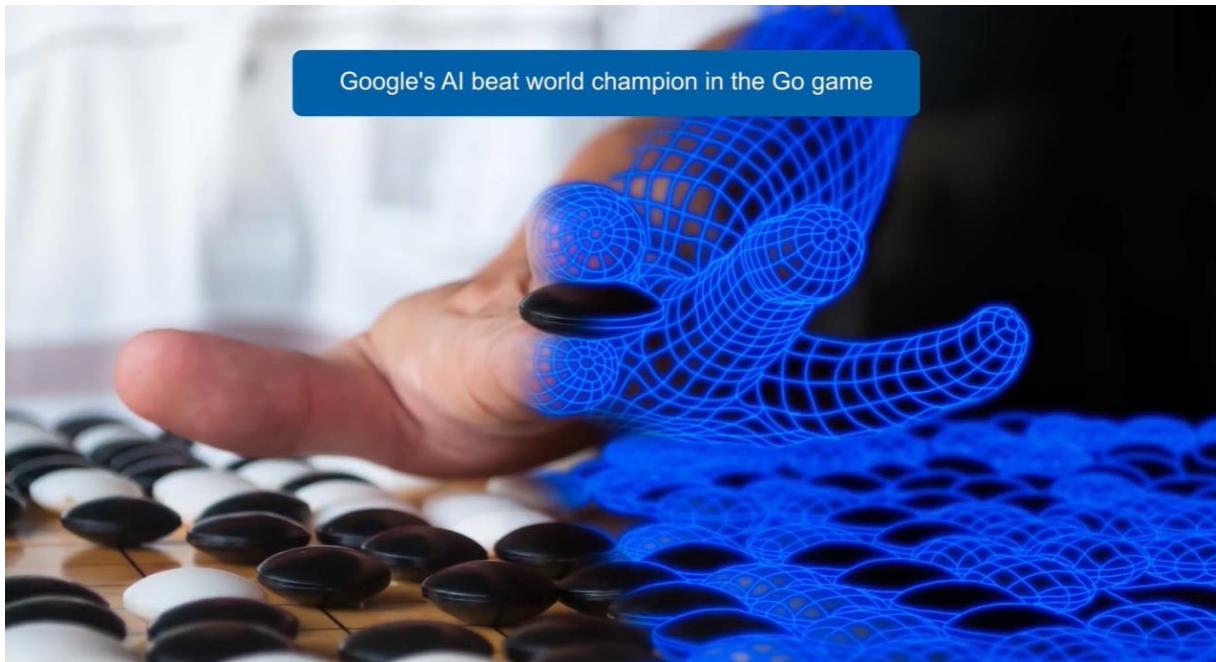
Deep learning

- Does not require feature data; instead, convert raw input into output
- Does not require guidance; follows self-learning
- Requires a large volume of data to learn

Machine Learning Techniques



Example of Reinforcement Learning



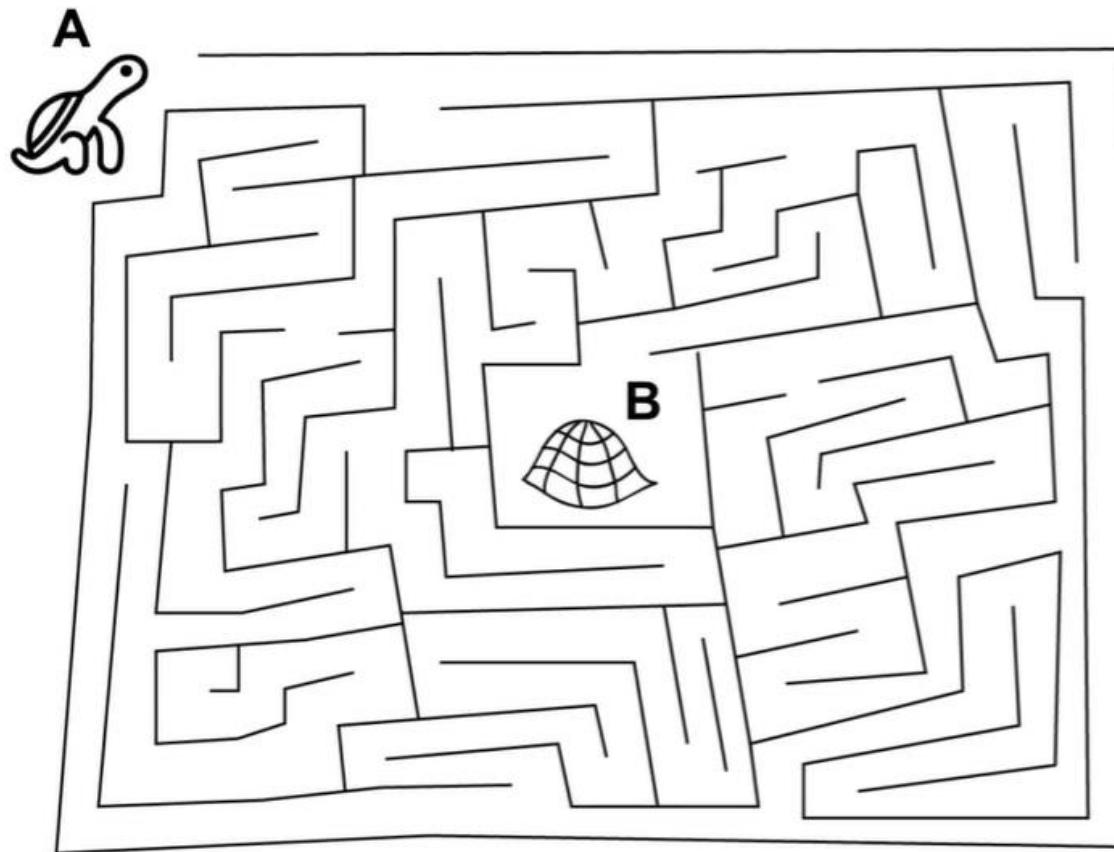
Google's AI beat world champion in the Go game



IBM's Deep Blue AI beat the world chess champion

Example of Reinforcement Learning

Help Tommy Tortoise find his shell



In reinforcement learning, machines are provided with:

- A goal
- A set of rules to achieve the goal

ML Approaches

Supervised learning

- Requires outcome variable (Y) of interest
- Goal:
To discover the relationship between input and outcome variables

Unsupervised learning

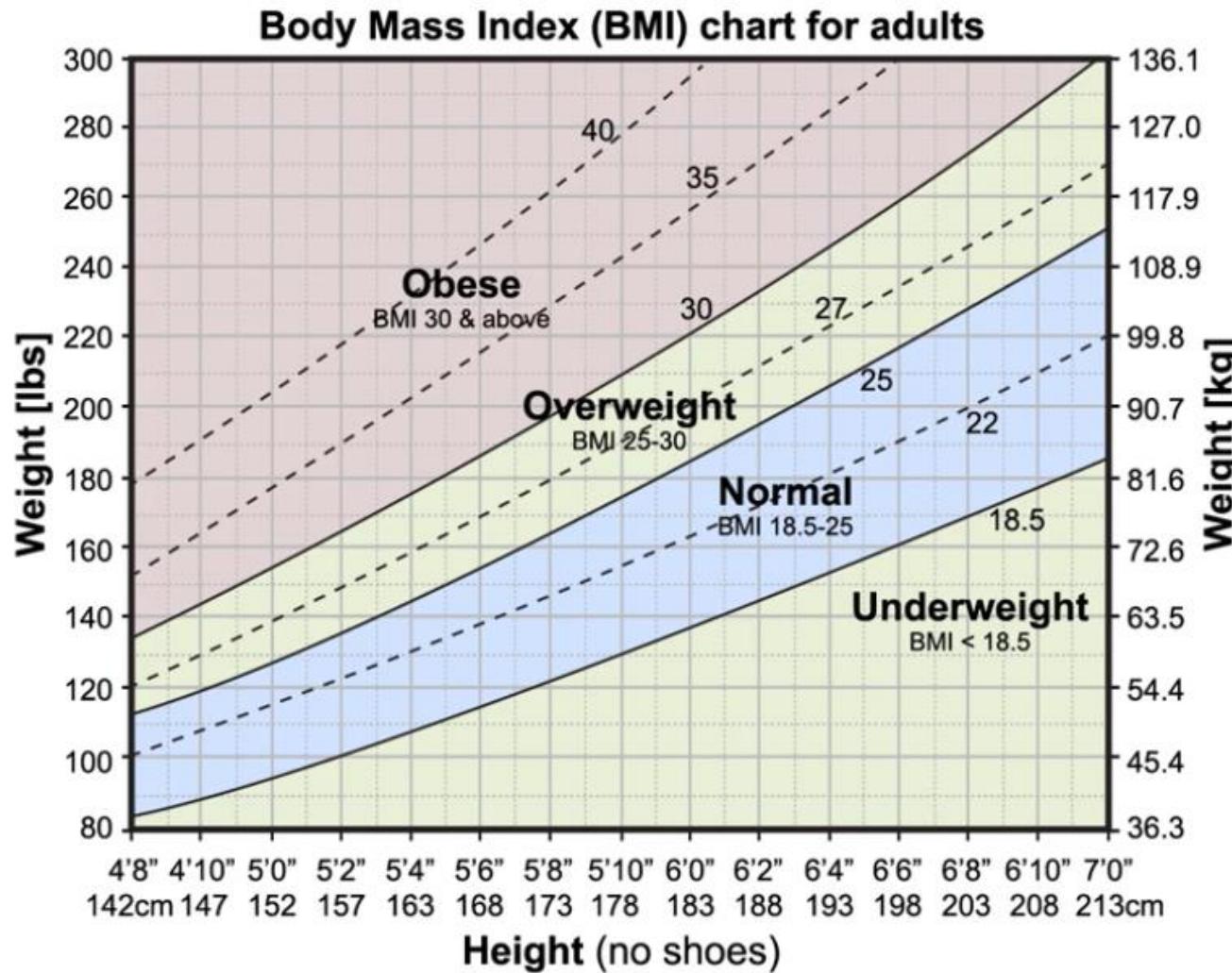
- No explicit outcome variable (Y)
- Goal:
To discover latent structure of the data

Reinforcement learning

- Requires deep learning models
- Goal:
To discover from start to end states

Classification Example

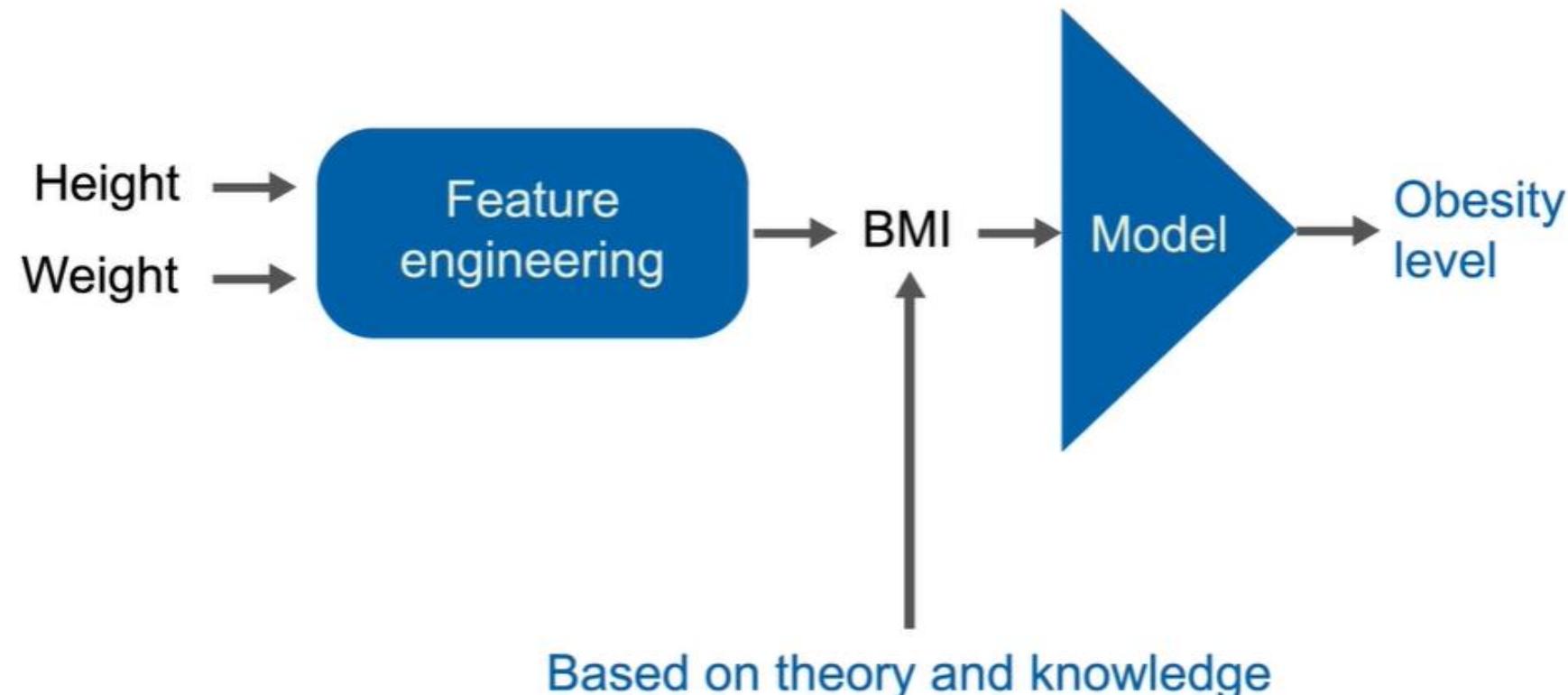
How doctors identify a patient's weight



Traditional Knowledge-Based Classification

Doctors measure height and weight and convert them into BMI index

In traditional knowledge-based classification, the input variables and the formula are converted into output variables using the theory



Machine Learning Context

In the machine learning context, the theory doesn't identify the rules. Rather, the machines discover the rules

Example of Machine Learning Classification

Given input and output,

Input	Output
Height	Obesity levels
Weight	

The machine will identify the formula for computing the input variables into the output

Machine Learning Classification

Example

Iris setosa



Iris versicolour



Iris virginica



It is possible for machines to identify these **using two approaches- machine learning and deep learning**

Approaches for Classification

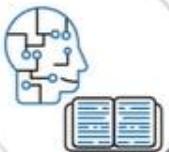
Iris setosa



Iris versicolour



Iris virginica



Machine learning

Guide the machine by feeding it information such as the:

- Length of the petal
 - Length of the sepal
 - Width of the petal
 - Width of the sepal
- to help the machine identify each flower correctly

Deep learning

- Feed pictures of the flowers into the machine
- The machine will analyse the pixels and make the associations

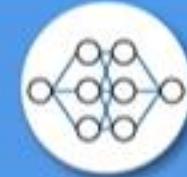


Machine Learning vs Deep Learning



Machine Learning

- Requires features to be fed into the machine
- Does not require the rule
- Discovers the rule using the features

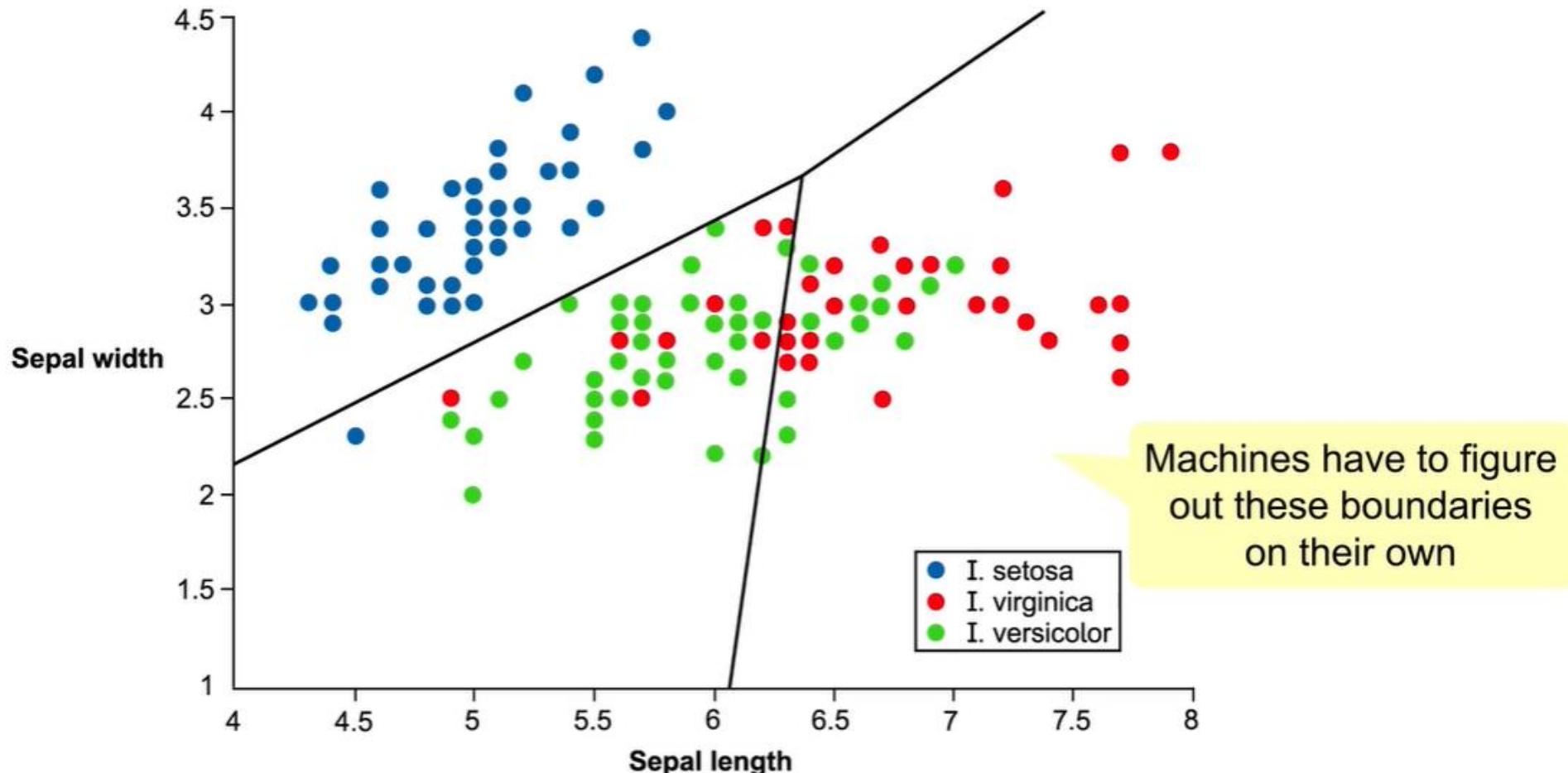


Deep Learning

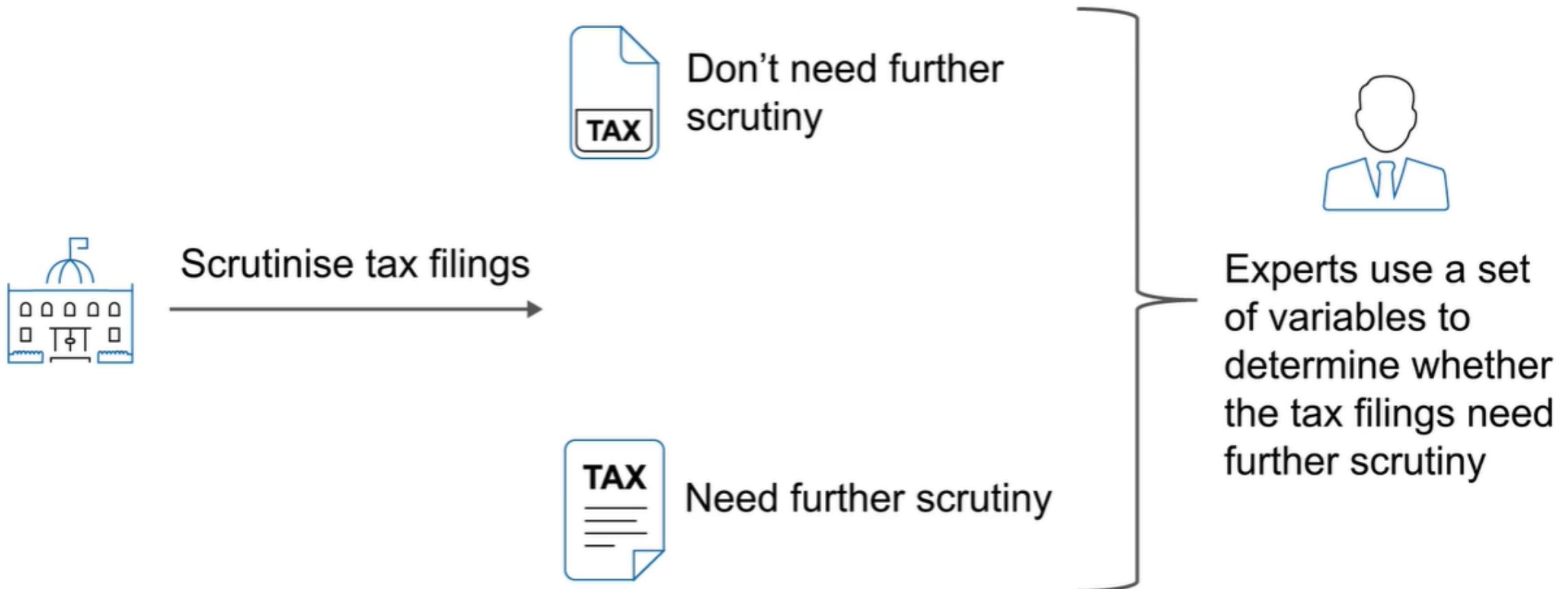
- Requires only raw data and not features
- Discovers how to convert the raw data into the outcome variable

Machine Learning-Based Classification

Learning rules from past decisions humans made



Example: How Tax Filings Are Assessed for Further Scrutiny



How Machines Scrutinise Tax Filings

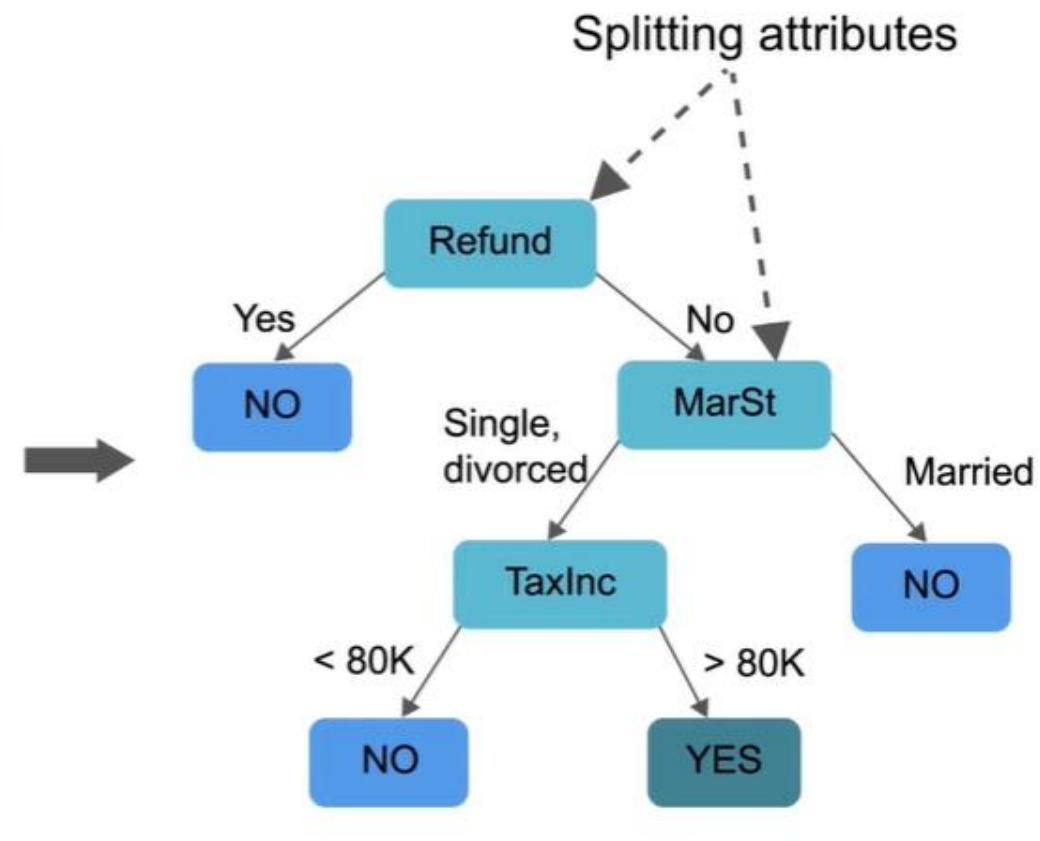
- Provide past tax filings data as input along with features such as:
 - Refund request status
 - Marital status
 - Taxable income
 - Previous classification status (further scrutiny or not)
- Have machines create a rule similar to human decision-making

Classification by Machine Learning

Machines create decision trees or if-else statements to determine how to classify each observation

Tid	Refund	Marital status	Taxable income	Cheat	class
1	Yes	Single	125K	No	categorical
2	No	Married	100K	No	categorical
3	No	Single	70K	No	continuous
4	Yes	Married	120K	No	continuous
5	No	Divorced	95K	Yes	continuous
6	No	Married	60K	No	continuous
7	Yes	Divorced	220K	No	continuous
8	No	Single	85K	Yes	continuous
9	No	Married	75K	No	continuous
10	No	Single	90K	Yes	continuous

Training data

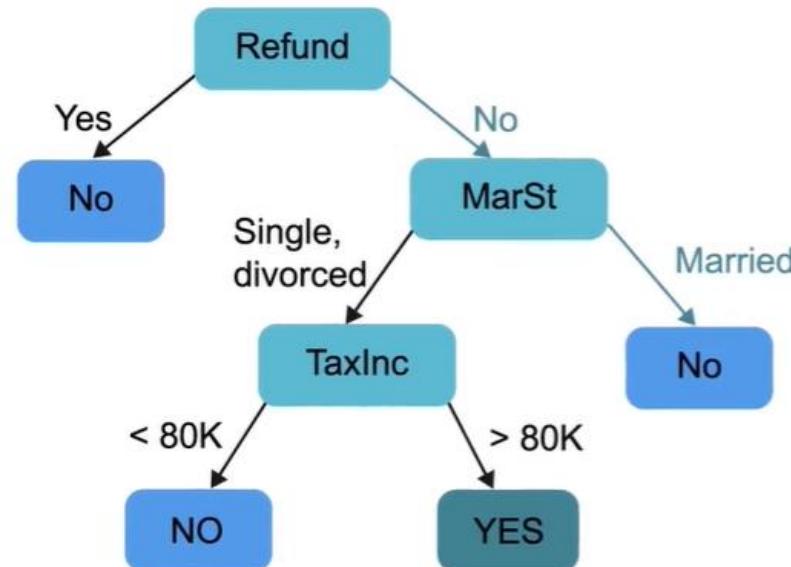


Classification by Machine Learning

- Once machines determine the rule, it can be used to classify every tax filing
- With the data as input, machines will predict whether a tax filing requires further scrutiny using the rule

Example Classification Tree

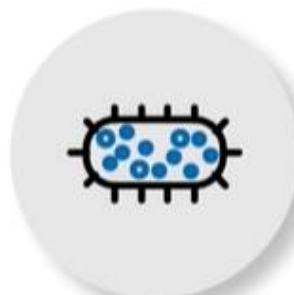
Instead of thousands of experts going through each and every tax, one single machine can identify all these tax filings and classify them for further scrutiny



Refund	Marital status	Taxable income	Cheat
No	Married	80K	?

Clustering Objects into Groups: Example 1

Having a structure helps classify new observations easily



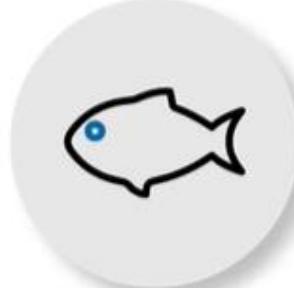
Bacteria



Plant



Human



Fish



Fungi

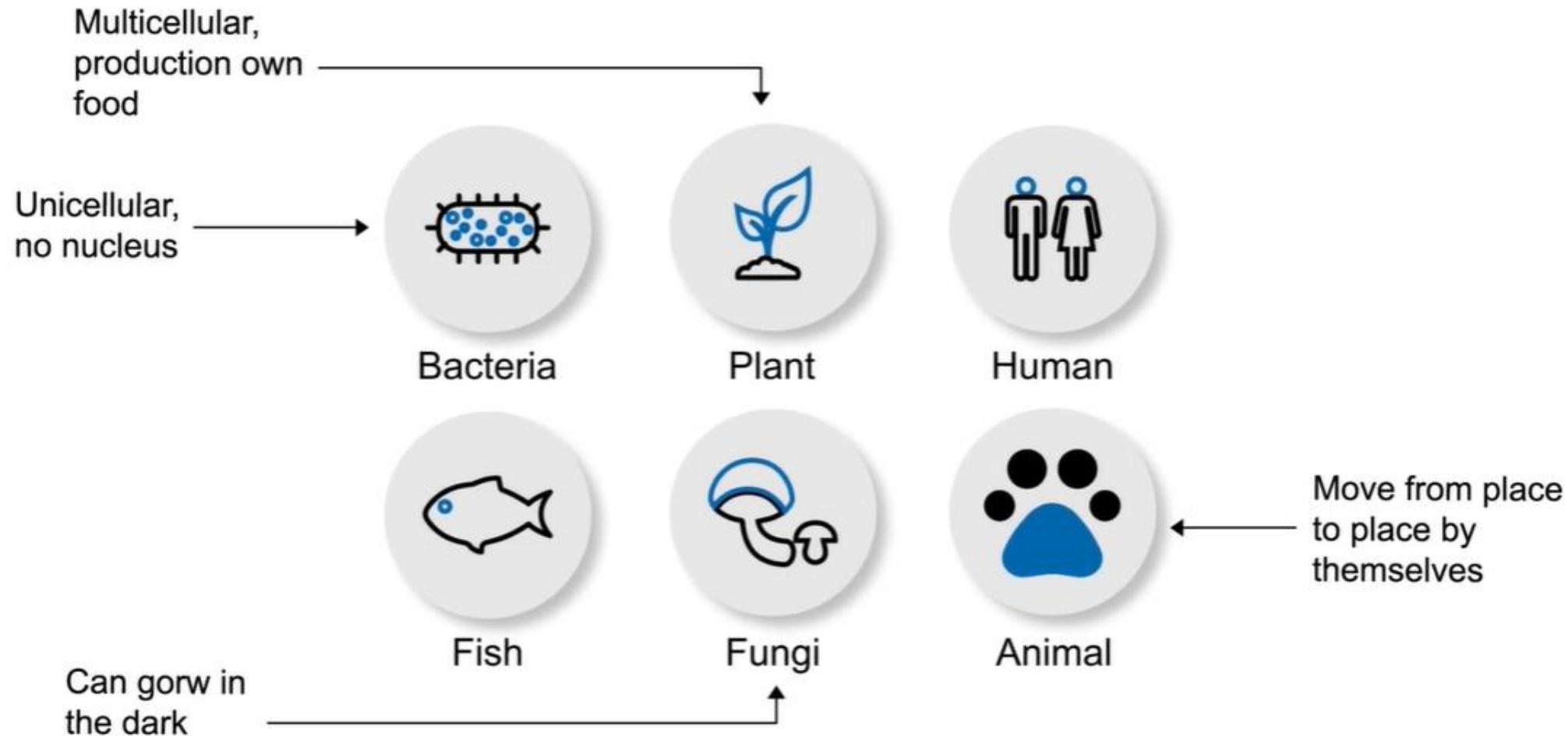


Animal



Clustering Objects into Groups: Example 1

Clustering observations into groups is based on domain knowledge and theory



Clustering Objects into Groups: Example 2

Periodic table of the elements

The periodic table is color-coded to categorize elements based on their properties:

- Non-metal** (Grey): Hydrogen (H), Nitrogen (N), Oxygen (O), Fluorine (F), Neon (Ne), Carbon (C), Nitrogen (N), Oxygen (O), Fluorine (F), Chlorine (Cl), Argon (Ar), Krypton (Kr), Bromine (Br), Iodine (I), Xenon (Xe), Radon (Rn).
- Alkali metal** (Orange): Lithium (Li), Potassium (K), Rubidium (Rb), Cesium (Cs), Francium (Fr).
- Alkaline earth metal** (Red): Beryllium (Be), Magnesium (Mg), Calcium (Ca), Strontium (Sr), Barium (Ba).
- Metal** (Dark Green): All transition metals (Scandium (Sc) through Zinc (Zn)), all lanthanides (Lanthanum (La) through Lutetium (Lu)), and all actinides (Actinium (Ac) through Lawrencium (Lr)).
- Metalloid** (Medium Green): Boron (B), Silicon (Si), Germanium (Ge), Arsenic (As), Sulfur (S), Phosphorus (P).
- Halogen** (Blue): Chlorine (Cl), Bromine (Br), Iodine (I), Xenon (Xe), Radon (Rn).
- Noble gas** (Light Blue): Helium (He), Neon (Ne), Argon (Ar), Krypton (Kr), Xenon (Xe), Radon (Rn).
- Lanthanide** (Orange-red): All lanthanides from Lanthanum (La) to Lutetium (Lu).
- Actinide** (Purple): All actinides from Actinium (Ac) to Lawrencium (Lr).
- Transition metal** (Purple): All transition metals from Scandium (Sc) to Zinc (Zn), plus Hafnium (Hf), Tantalum (Ta), Rhenium (Re), Osmium (Os), Iridium (Ir), Platinum (Pt), Gold (Au), Mercury (Hg), Thallium (Tl), Lead (Pb), Bismuth (Bi), Polonium (Po), Astatine (At), and Ununactinium (Uuo).

The table includes element symbols, atomic numbers, and atomic masses. The last two rows are marked with asterisks (*) and double asterisks (**).

1	H	HYDROGEN 1.0079	2	He	HELIUM 4.0026																									
3	Li	LITHIUM 6.941	4	Be	BERYLLIUM 9.0122																									
11	Na	SODIUM 22.989	12	Mg	MAGNESIUM 24.309																									
19	K	POTASSIUM 39.098	20	Ca	CALCIUM 40.078																									
37	Rb	RUBIDIUM 85.467	38	Sr	STRONTIUM 87.62																									
55	Cs	CAESIUM 132.906	56	Ba	BARIUM 137.327																									
87	Fr	FRANCIUM (223)	88	Ra	RADIUM (226)																									
*	57	La	58	Ce	59	Pr	60	Nd	61	Pm	62	Sm	63	Eu	64	Gd	65	Tb	66	Dy	67	Ho	68	Er	69	Tm	70	Yb	71	Lu
**	89	Ac	90	Th	90	Pa	92	U	93	Np	94	Pu	95	Am	96	Cm	97	Bk	98	Cf	99	Es	100	Fm	101	Md	102	No	103	Lr

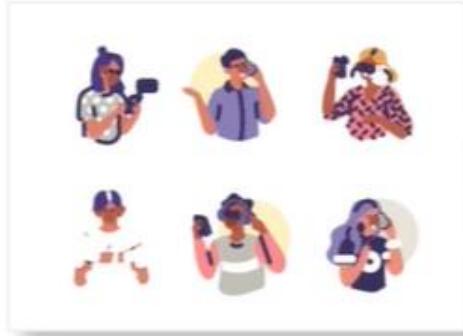
Benefit of Cluster Analysis

Cluster analysis helps organise objects into some order so that useful information could be retrieved

Other Examples of Clustering



Number and types of web pages



Types of consumers in a market



Types of people on social network



Types of genes in the human genome



Types of songs, podcasts or movies in a library

Clustering Objects into Groups: Example 3

Fruits can be clustered into groups based on their colours



Clustering Objects into Groups: Example 3

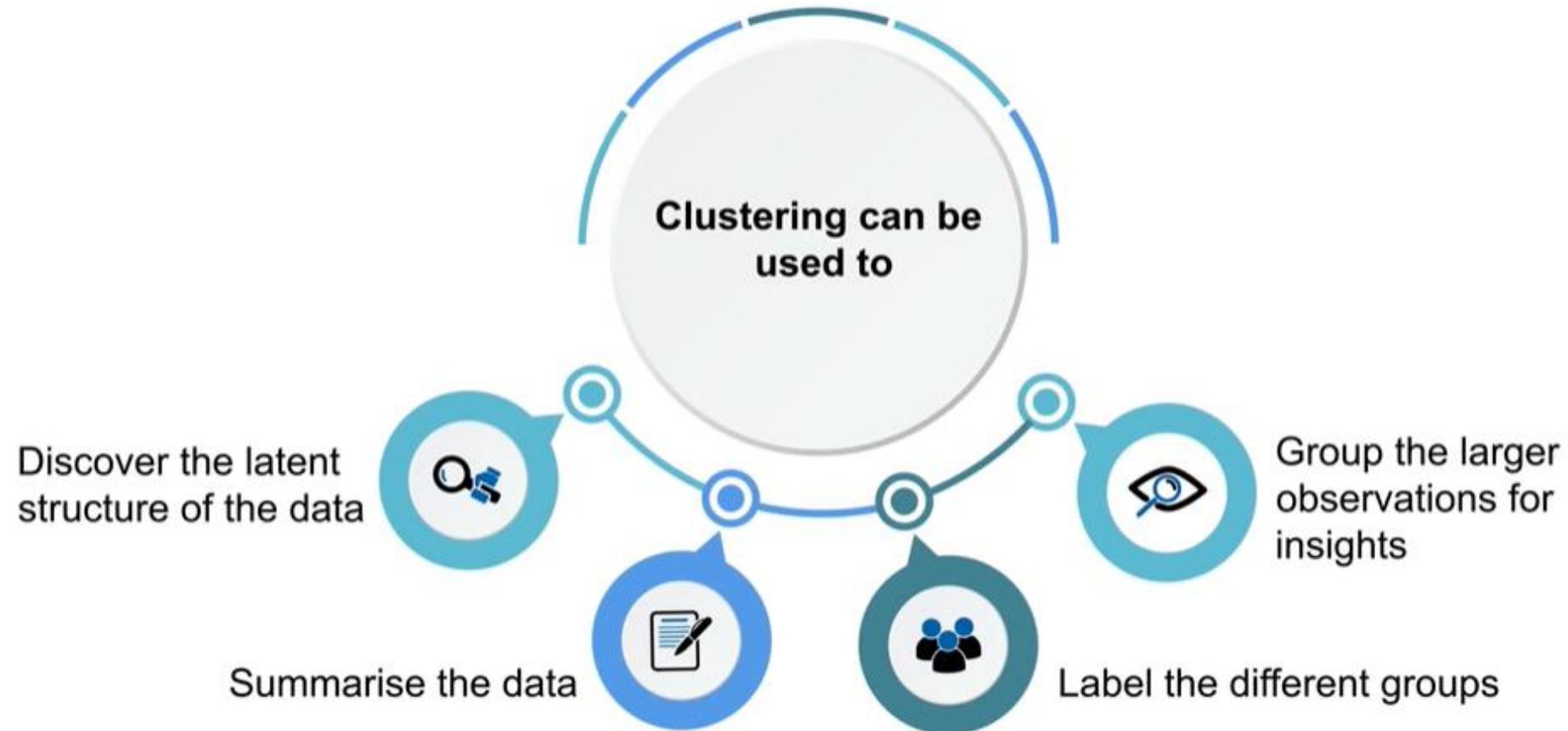
Fruits can be clustered into groups based on their shape



Cluster Analysis

- Clustering is exploratory in nature
- Based on a given set of features, machines can cluster objects into several groups
- Domain knowledge can then be applied to identify the useful clusters and extract insights

Why Do We Need Clustering?



Business Application of Clustering

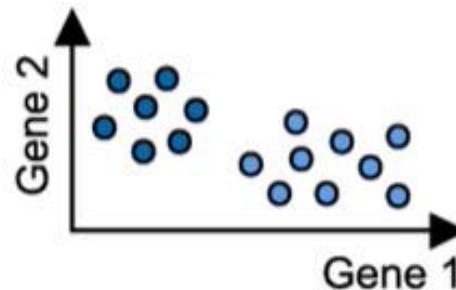
- Customers could be classified into groups based on features such as age, location and education
- Such customers groups could then be used to identify segments and better tailor products

Classification: Unsupervised vs Supervised learning

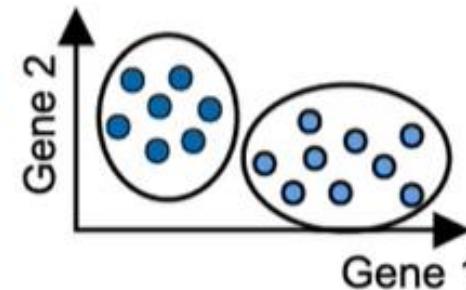
Unsupervised learning (Clustering)

Use features and factors describing a particular object to organise objects into different groups such that similar objects are in one group and two groups are very different

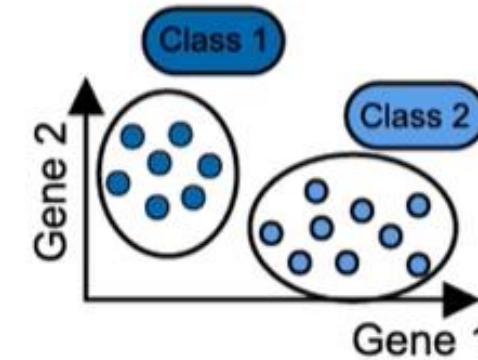
Unlabeled data set



Cluster samples



Assign labels

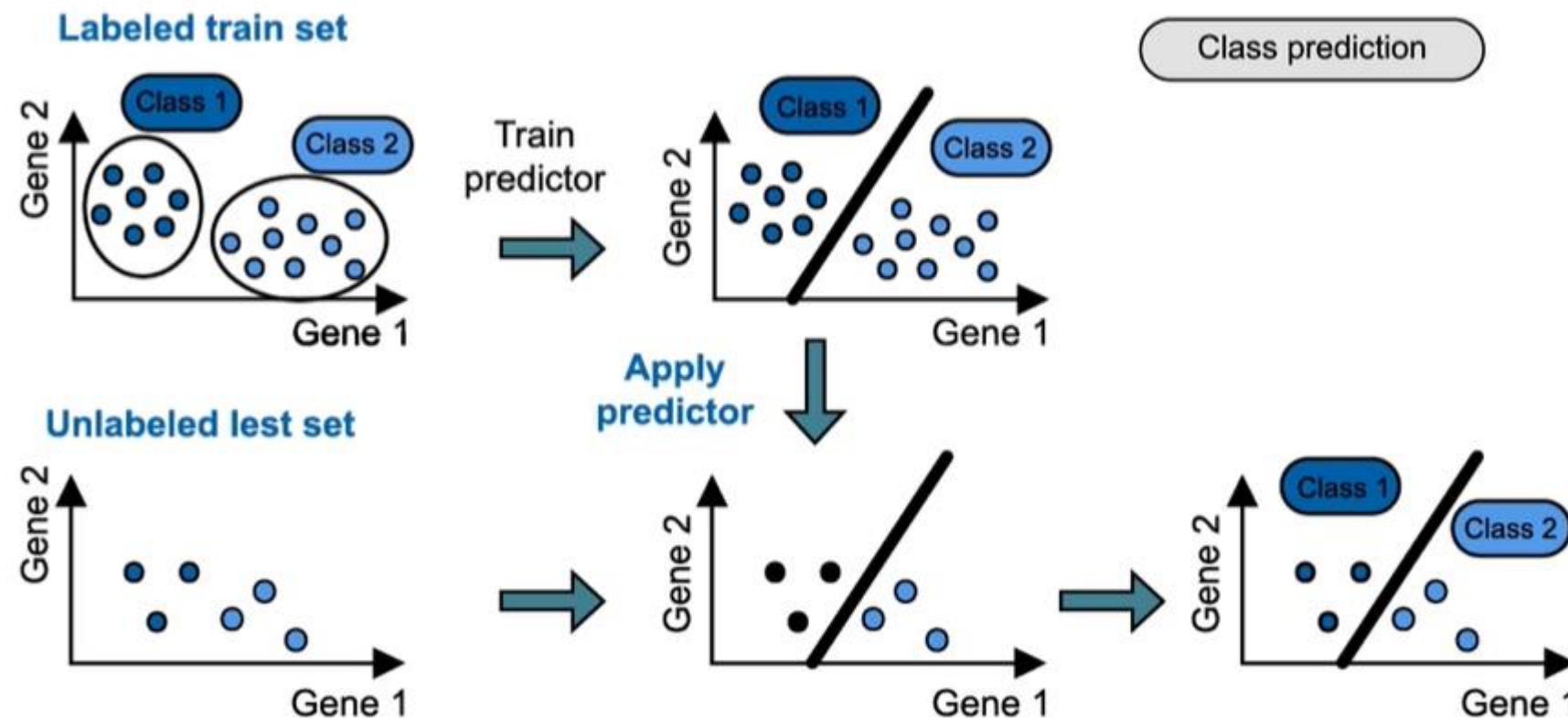


Class discovery

Classification: Unsupervised vs Supervised learning

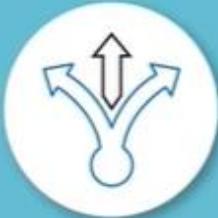
Supervised learning (Classification)

Based on a set of features with known classes, machine learning models predict the class of a new observation



ML Framework for Business Problems

Key analytics questions : business decisions



What decisions does
the business make?

Example:

Which offers should
be sent to which
customers?



On what basis are the
decisions made?

Example:

By observing the
past purchase
behaviour of the
customers



How do we measure the
success of our decisions?

Example:

- By measuring
whether a campaign
was successful
- By determining the
metrics of success

ML Framework for Business Problems

Key analytics questions: data analytics



**What kind of data
should be collected to
evaluate the decision?**

Example:

How many customers
redeemed the coupon?



**What data should be
collected to improve
the decision?**

Which variables or
features can help the
machine make better
decisions?



**What kind of
a model to use?**

- Supervised learning
 - Decision Tree
 - Logit Model
 - Regression
- Unsupervised learning



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