

# Artificial Intelligence Landscape

LECTURE 1



# ARTIFICIAL INTELLIGENCE PRACTITIONERS COURSE



< JOURNEY >

## AI Landscape – Lecture 1



## AI Industry Adoption Approaches – Lecture 2



## NLP and Virtual Assistants – Lecture 3



## Computer Vision – Lecture 4



## Machine Learning and Deep Learning – Lecture 5



## Future Trends for AI – Lecture 6



→ Lab 1 – Setting up your Cloud Account

→ Lab 2 – Gain Insights from AirBnB reviews

→ Lab 3 – Creating an AI Virtual Assistant

→ Lab 4 – Training AI to Host Customers

→ Lab 5 – Building your own Translator with AI

→ Lab 6 – Analyze, Classify, & Detect Objects

→ Lab 7 – Classifying Images Using Node-RED

→ Lab 8 – Fraud Prediction using AutoAI

# LECTURE 1

## ARTIFICIAL INTELLIGENCE LANDSCAPE

### OBJECTIVES

- Understand the effect of today's public opinion about AI in its mainstream adoption
- Discover applications of AI currently seamlessly embedded into popular digital products and services
- Understand the evolution of AI systems and methods
- Explore AI industry adoption scenarios



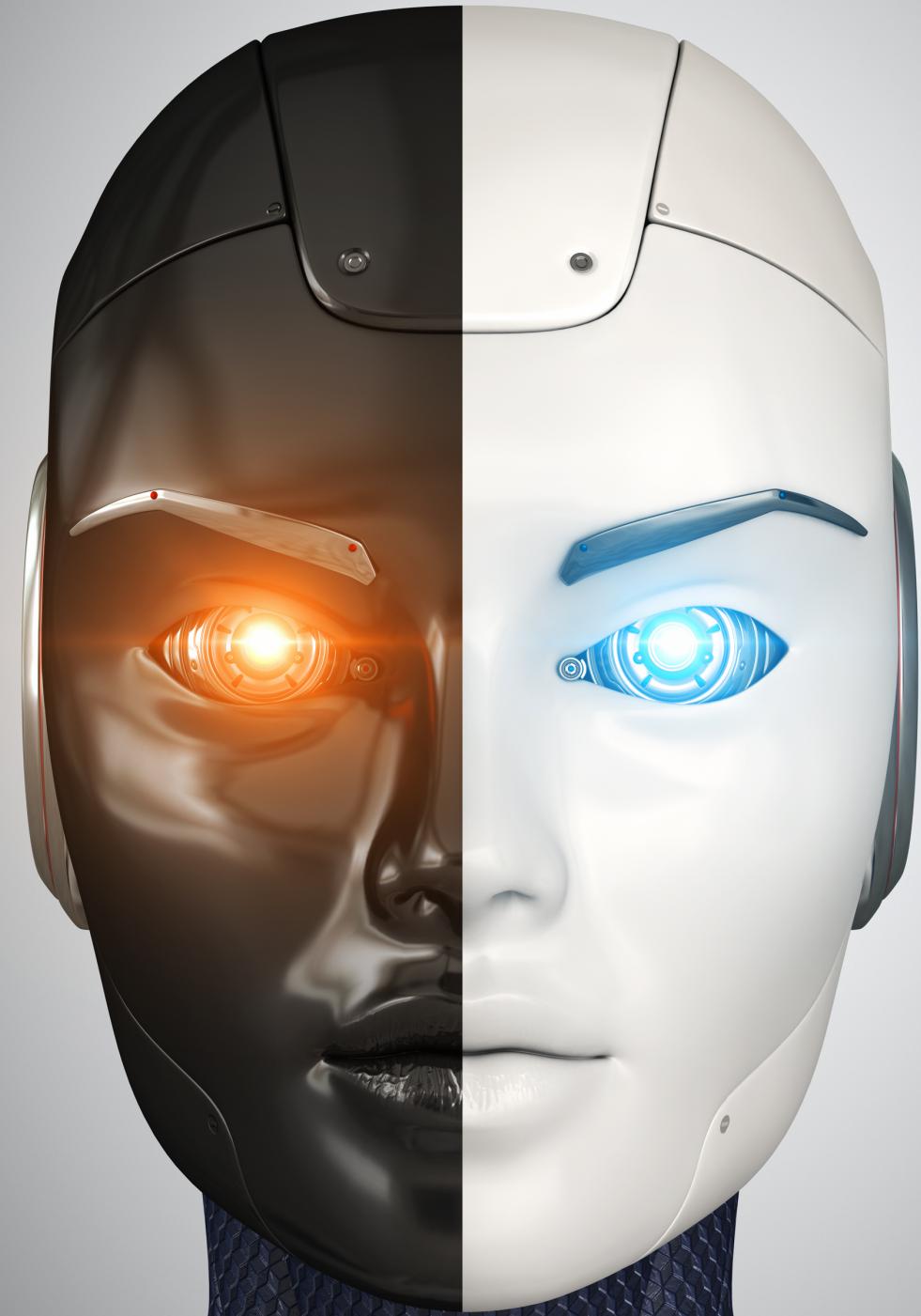
## LECTURE 1

### *Artificial Intelligence Landscape*

- ▶ 1. AI impact in the world today
- 2. History and Evolution of AI
- 3. AI Explained
- 4. AI Technologies
- 5. Summary & Resources

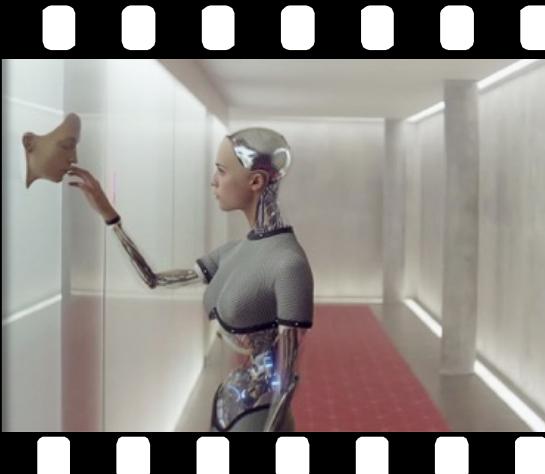
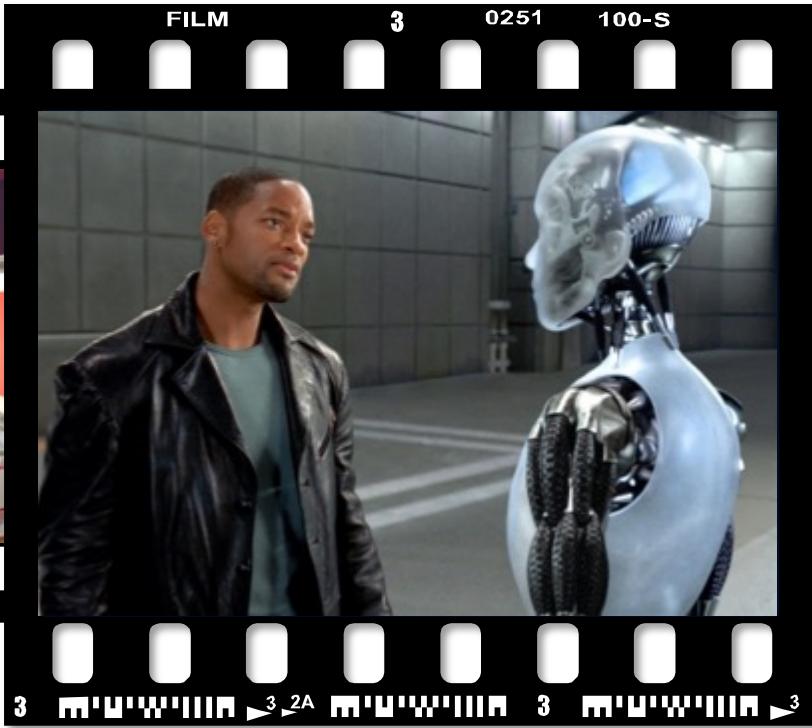


What is your  
definition of  
**Artificial  
Intelligence?**



## ARTIFICIAL INTELLIGENCE

# Perception *TODAY*



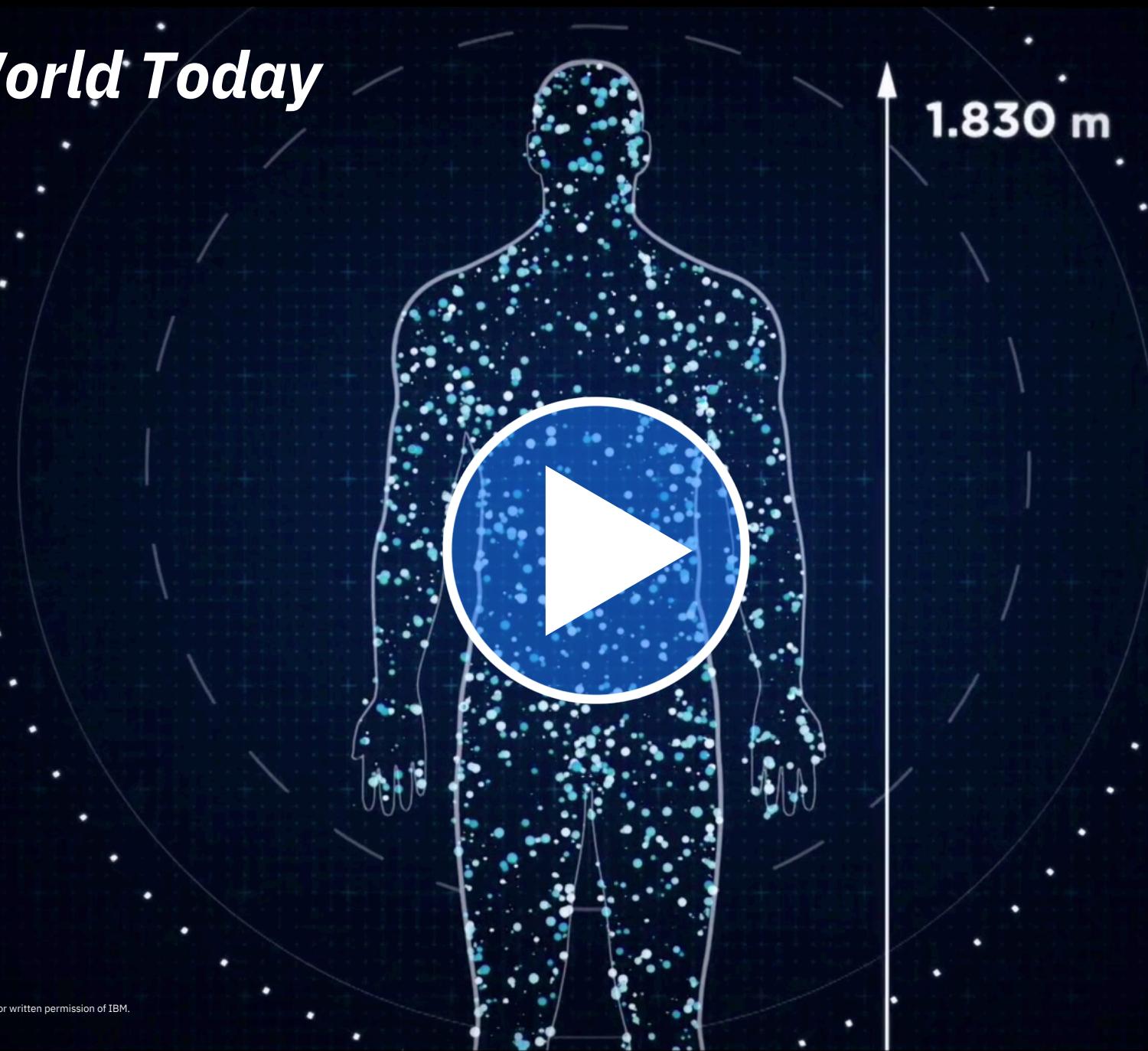
The Matrix

Her

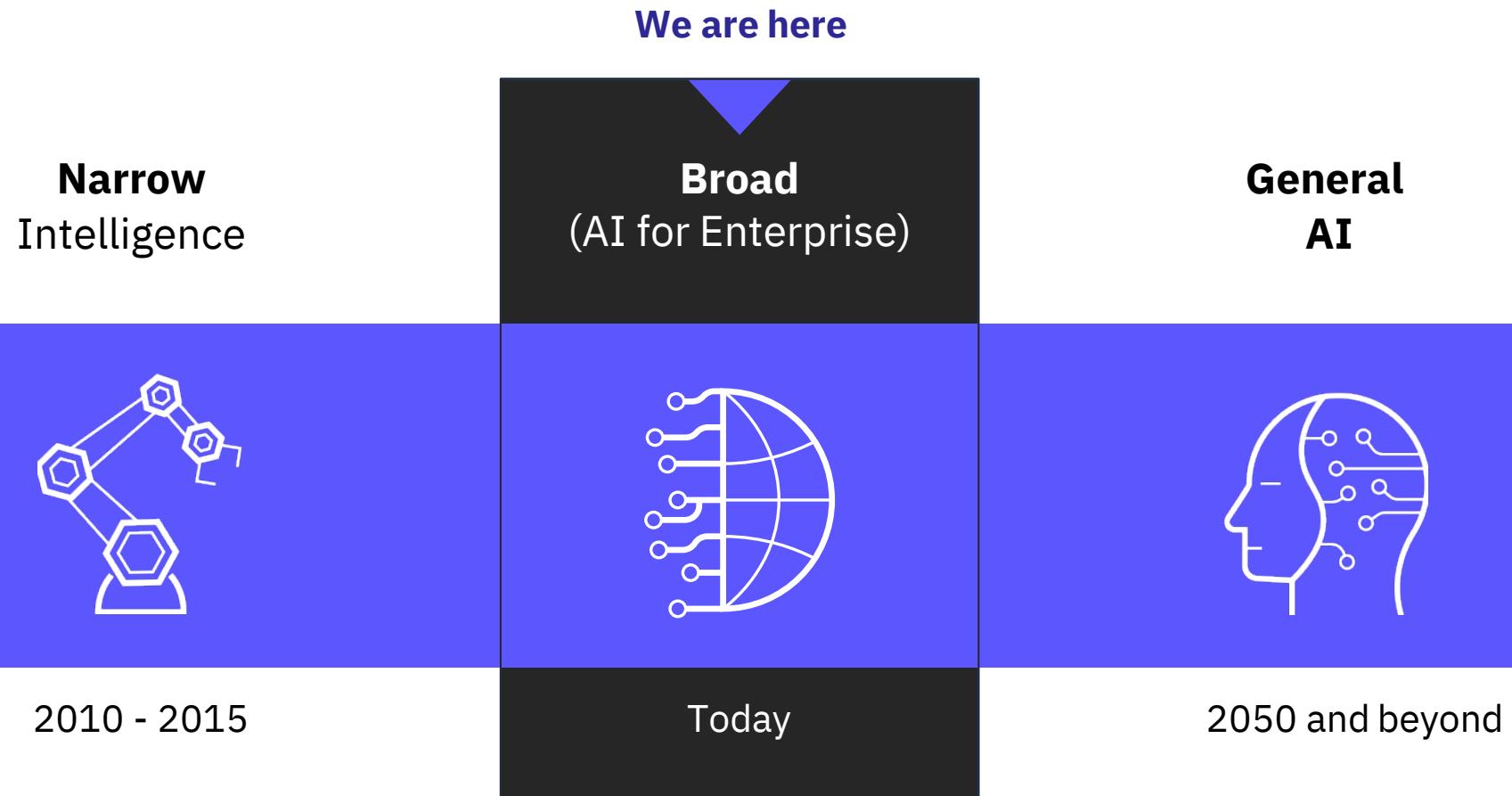
Highly influenced by media and entertainment industry

# AI in the *World Today*

Video 2:13 minutes



# Where are we today in AI?

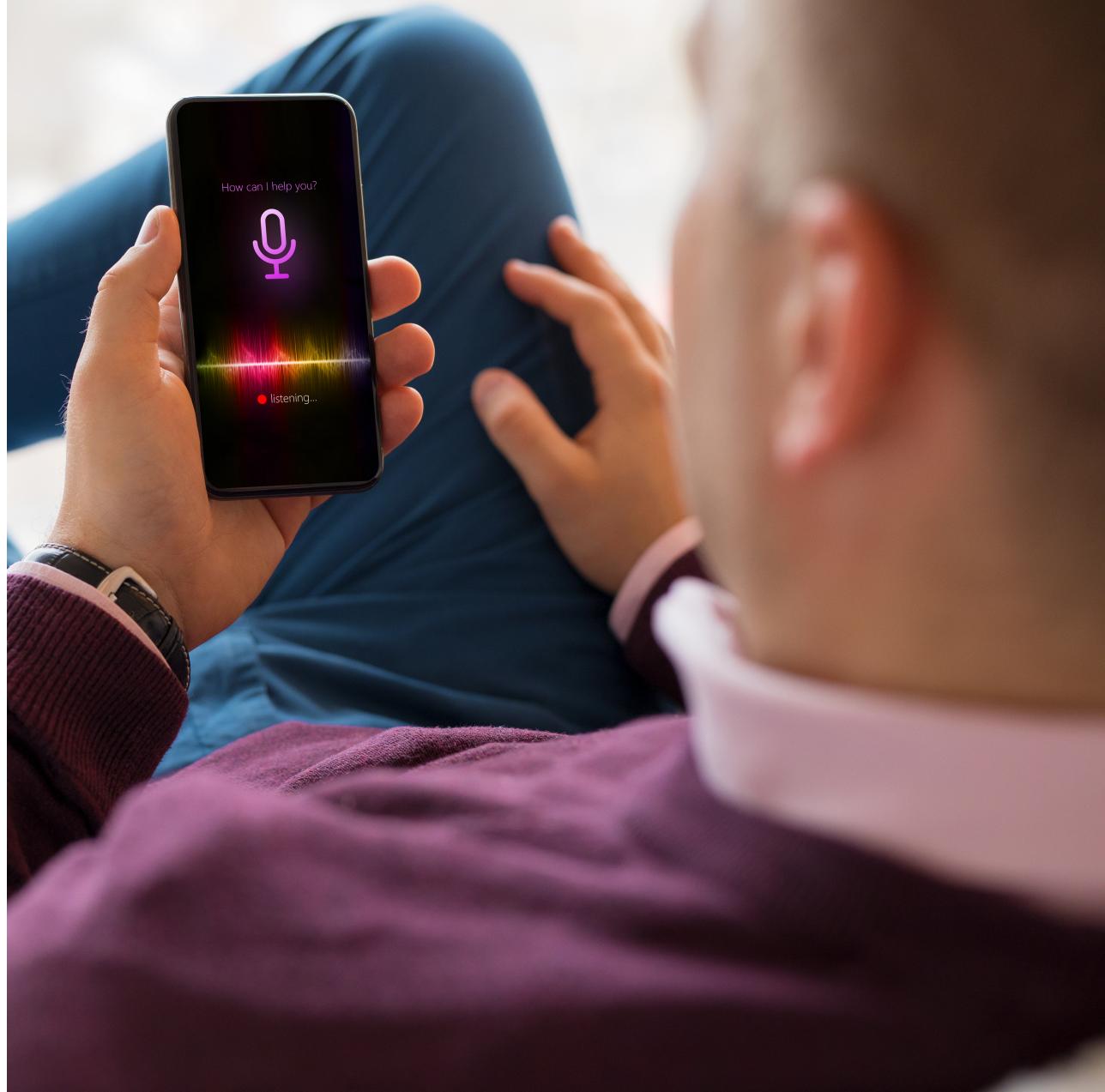


# Narrow Intelligence

Narrow AI is focused on addressing very focused tasks

## *Example*

Buying a book with a voice-based device based on “common knowledge”



# Broad AI

Broad AI is about integrating AI within a specific industry knowledge and data to train this type of system.

## *Example*

Self-driving cars are a collection of narrow AI systems that can make decisions



## BROAD AI - TODAY

# AI applications are transforming every industry



### Government

Campaign Content and Planning, Citizen Experience, Public Security, Policy Planning Support



### Finance

High Frequency Trading, Risk Modeling, Equity Research, Asset Mgmt, Underwriting, Investment Planning, Security



### Agriculture

UAV / Satellite Crop Field Analysis, Disease Recognition, Comprehensive Strategic Crop Planning



### Energy

Strategic Oil Drilling, Risk Minimization, Geological Analysis, Demand Prediction, Adjustment of Resource Generation



### Healthcare

Personalized Healthcare, Diagnostic Tools, Integrated Wellness and Health Systems, Behavior Tracking, Security



### Education

Personalized Education, Learning Content Indexing-to-Skill & Search, Custom Teaching Methods, Smart View Devices



### Science

Data Analysis, Experiments, Predictive Modeling, Theorem Proving, Deductive Reasoning, Experiment Planning



### Business Solutions

Interactive Chatbots that Learn from Experience with Customers, Regulatory Support, Prediction, Marketing

# General AI

General AI refers to machines that can perform any intellectual task a human can.

Currently AI does not have the ability to use previous experiences to come up with new creative ideas



“Our technology, our machines are **part of our humanity.**

We create them to extend ourselves and that is what is unique about human beings”

**Ray Kurzweil**  
Futurist



## LECTURE 1

### *Artificial Intelligence Landscape*

1. AI impact in the world today

## ► 2. History and Evolution of AI

3. AI Explained

4. AI Technologies

5. Summary & Resources



# History and Evolution of Artificial Intelligence

1906

*Warren McCulloch and Walter Pitts*  
researchers from University of Chicago  
propose the

**“nerve net”**  
– the first mathematical model of  
neurons and connections in the  
brain.

*Camilo Golgi and Santiago Ramon y Cajal*  
receive the Nobel Prize for developing the

**“neuron doctrine”** – the  
theory that the brain operates  
through interconnected  
individual cells.

• 1943

Alan Turing publishes the paper “Computing Machinery and Intelligence” introducing the

**“Turing Test”**  
– to determine machine’s human  
ability.

1950

# Artificial Intelligence Evolution

Dartmouth College

summer workshop with top scientists

**“Field of A.I. was born”**

– predicted that a machine as intelligent as a human being would exist in no more than a generation.

1957

MIT researchers Marvin Minsky and Seymour Papert publish

**“Perceptron's critique”**

– which drew very pessimistic conclusions about prospects for improving the Perceptron model of artificial intelligence.

• 1956

Frank Rosenblatt first demonstrated the Perceptron on an IBM 704 before building a custom machine.

**“Perceptron”** – the first trainable neural network algorithm, included photocells for image recognition.

• 1969

# Artificial Intelligence Evolution

## 1974-1980

Several researchers independently invent backpropagation training of neural nets. The algorithm successfully learned in many areas which Minsky and Papert had predicted would be impossible, helping launch a revival of interest in neural nets

### “AI Renaissance”

- The Snowbird neural networks workshop and the NIPS (Neural Information Processing Systems) conference.

## “FIRST AI Winter”

- the Lighthill report and governments’ funding restrictions.

## 1985

IBM researcher Gerald Tesauro creates a self-teaching neural net that learns to play backgammon

### “TD-Gammon”

- solely by playing against itself and learning from the game outcomes.

# Artificial Intelligence Evolution

## “SECOND A.I. Winter”

- Collapse of the Lisp machine market.

• **1987-1993**

Yann LeCun demonstrates a **Convolutional Neural Network**.

**1993**

• **“LeNet 1”** – capable of recognizing handwritten digits quickly and accurately.

IBM's Deep Blue, a computer capable of analyzing 200 million moves per second.

## “Deep Blue”

- defeats reigning world chess champion *Garry Kasparov*.

• **1997**

# Artificial Intelligence Evolution

2006 •

Researchers unveil a massive collection of human-annotated images

**“ImageNet” – the definitive data set for teaching deep learning algorithms that can see.**

Hinton and colleagues demonstrate that so-called “Deep Belief Nets” Their work launches today’s the movement

**“Deep Learning” – with many layers can be effectively trained.**

• 2009

IBM makes international headlines when their system appears on Jeopardy!—and wins

**“Watson” – Natural Language Processing**

# Artificial Intelligence Today

*For the first time, deep neural networks classify ImageNet images with greater accuracy than humans*

**“Open Source A.I.”**  
– Open Source Frameworks  
and data available to  
everyone.

• 2015

*Uber pilots self-driving cars program in Pittsburgh, PA*

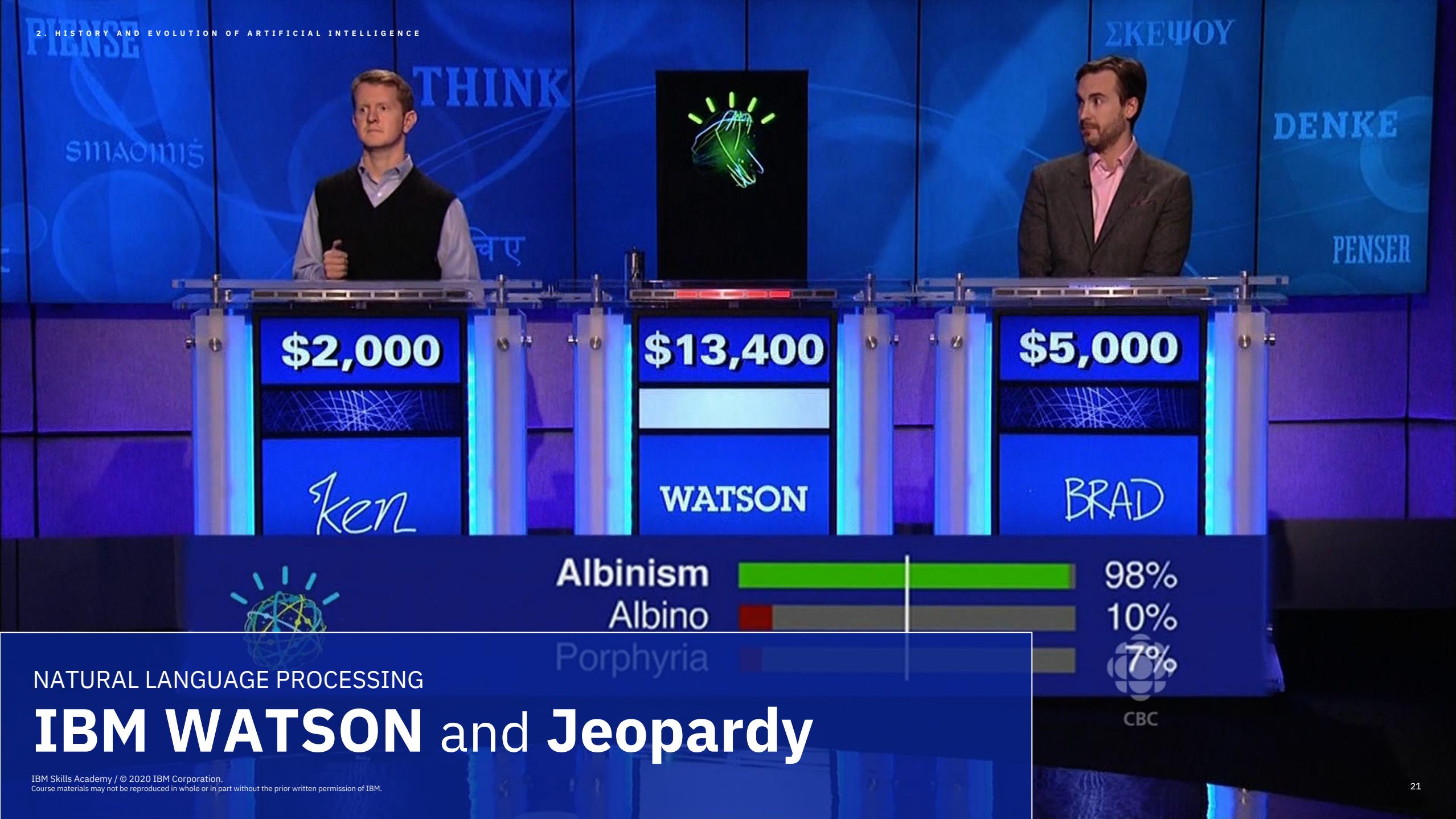
**“Self-driving cars”** –  
capable of transit through city  
traffic without a driver.

2016 •

IBM unveils Deep Learning as a Service, which allows anyone with an internet connection to take advantage of sophisticated AI algorithms, rich data platforms, and immense computing power.

**“Deep Learning as a Service”**

• 2018





# Watson beyond Jeopardy!

// IBM WATSON

**Watson is augmented intelligence:** helping people and machine work together to create knowledge from data that enhances human expertise

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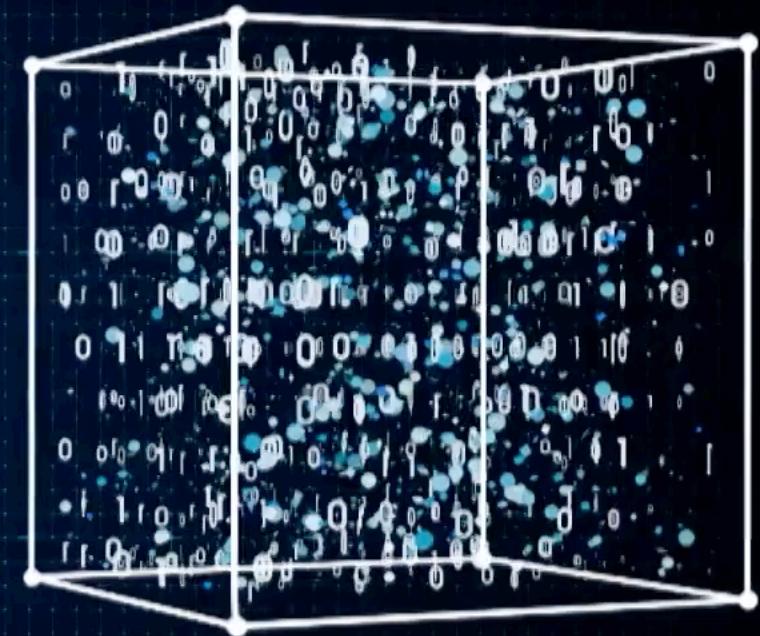
# How Machines Learn?



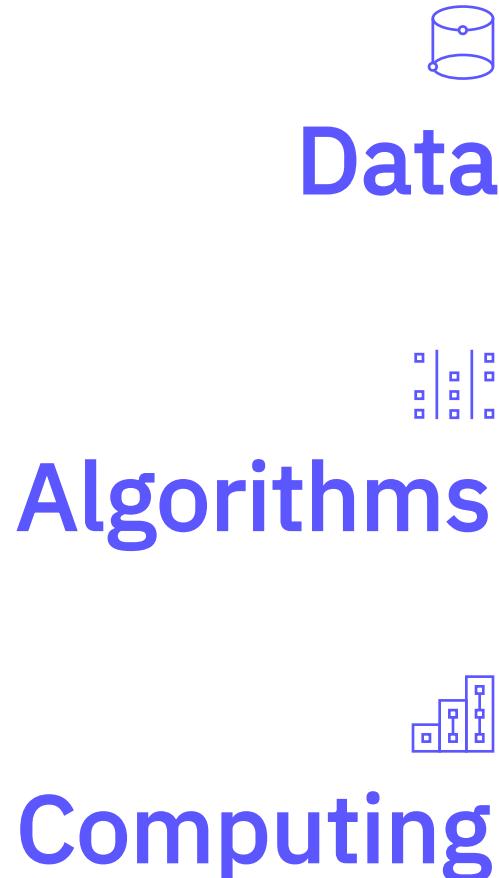
Video 6:30 minutes



$$\phi : \rho \left( \frac{\partial u_\phi}{\partial t} + u_r \frac{\partial u_\phi}{\partial r} + \frac{u_\phi}{r \sin(\theta)} \frac{\partial u_\phi}{\partial \phi} + \frac{u_\theta}{r} \frac{\partial u_\phi}{\partial \theta} + \frac{u_r u_\phi + u_\phi u_\theta \cot(\theta)}{r} \right)$$
$$\frac{u_\phi}{r \sin(\theta)} \frac{\partial u_\theta}{\partial \phi} + \frac{u_\theta}{r} \frac{\partial u_\phi}{\partial \theta}$$
$$\mu \left[ \frac{1}{r^2} \frac{\partial^2 u_\phi}{\partial r^2} \right]$$
$$\mu \left[ \frac{1}{r^2} \frac{\partial^2 u_\theta}{\partial r^2} \right]$$
$$\phi : \rho \left( \frac{\partial u_\phi}{\partial t} + u_r \frac{\partial u_\phi}{\partial r} + \frac{u_\phi}{r \sin(\theta)} \frac{\partial u_\phi}{\partial \phi} + \frac{u_\theta}{r} \frac{\partial u_\phi}{\partial \theta} + \frac{u_r u_\phi + u_\phi u_\theta \cot(\theta)}{r} \right)$$



# AI Evolution Factors



Exponential growth of available data, with the introduction of the Internet, social media, proliferation of sensors and smart devices, and the fact that data storage became cheaper

The development of more advanced algorithms has helped AI become more powerful and efficient.

Back when AI was just beginning to be developed, the computing power was minimal. Computers nowadays can take much more data and heavier algorithms than in the 1950s.

## AI TIMELINE

- 1950 • Turing publishes the Turing-test. “The point at which a machine has answers like a human”
- 1955 • AI first named by John McCarthy
- 1956 • “First” AI algorithm Logic Theorist by Simon and Newell
- 1957 • Rosenblatt invents the first self learning algorithm with the perceptron
- 1958 • IBM 305, the first hard drive, 5 MB
- 1969 • Backpropagation, one of the most important areas of a neural network, is proposed
- 1970 • IBM 1330, 100MB
- 1974 • Intel produces second generation general purpose chips
- 1974-1980 • First AI winter, the belief in machine learning and AI had dropped after multiple unsuccessful experiments combined with insufficient computing power, network capabilities and database capacity
- 1985 • IBM 0665 hard drive, 40 MB. But much smaller than the 1330
- 1989 • First convolutional neural network developed (used a lot in image recognition)
- 1991 • The internet is open for the public
- 1992 • First versions of natural language solutions set up.
- 1997 • IBM’s deep blue defeats Kasparov in Chess
- 1998 • Google’s Page rank is published
- 2000 • The adoption of Internet by the masses
- 2002 • Amazon brings cloud computing to the masses
- 2004 • Google develops an algorithm that can handle large amounts of data faster.
- 2005 • Stanford Robot drives automatically
- 2006 • IBM introduces Watson, A question answering machine that later wins from a Jeopardy champion
- 2010 • Worldwide IP traffic exceeds 20 exabytes (20 billion gigabytes) per month
- 2012 • Facebook gets a billion users
- 2014 • There are more mobile devices than humans in the world
- 2018 • Project debater of IBM shows ability to process very large data sets, including millions of news articles across dozens of subjects, and then turn snippets of arguments into full flowing prose—a challenging task for a computer

44 zettabytes

# Data evolution – growing at an exponential rate



4 billion pieces of content shared daily



30 million smart meters in Italy



200mb of data per cow per year



13 billion ad impressions per day



2.5 billion monthly page views



9 Million payments daily

◆ **80% of the world's data is dark data**

Ripe for discovery and exploration

We are here

UNSTRUCTURED DATA

**Structured data**

- Databases
- Formatted files

**Semi-structured data**

- XML/JSON
- E-mail
- Web Pages

**Unstructured data**

- Audio
- Video
- Image Data
- Natural Language
- Documents

2010

2025

STRUCTURED DATA

# How to provide insights on complex and volatile data?

Data is growing with time, but only a fraction of it is usable through traditional analytics

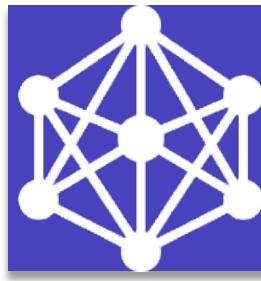
## DATA



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How to keep up with the mountains of contextual data available to you, even when most of it is unstructured in format

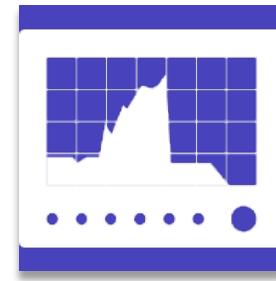
## COMPLEXITY



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How to overcome and solve for great complexity by giving the skill and knowledge of the informed few to the empowered many

## VOLATILITY

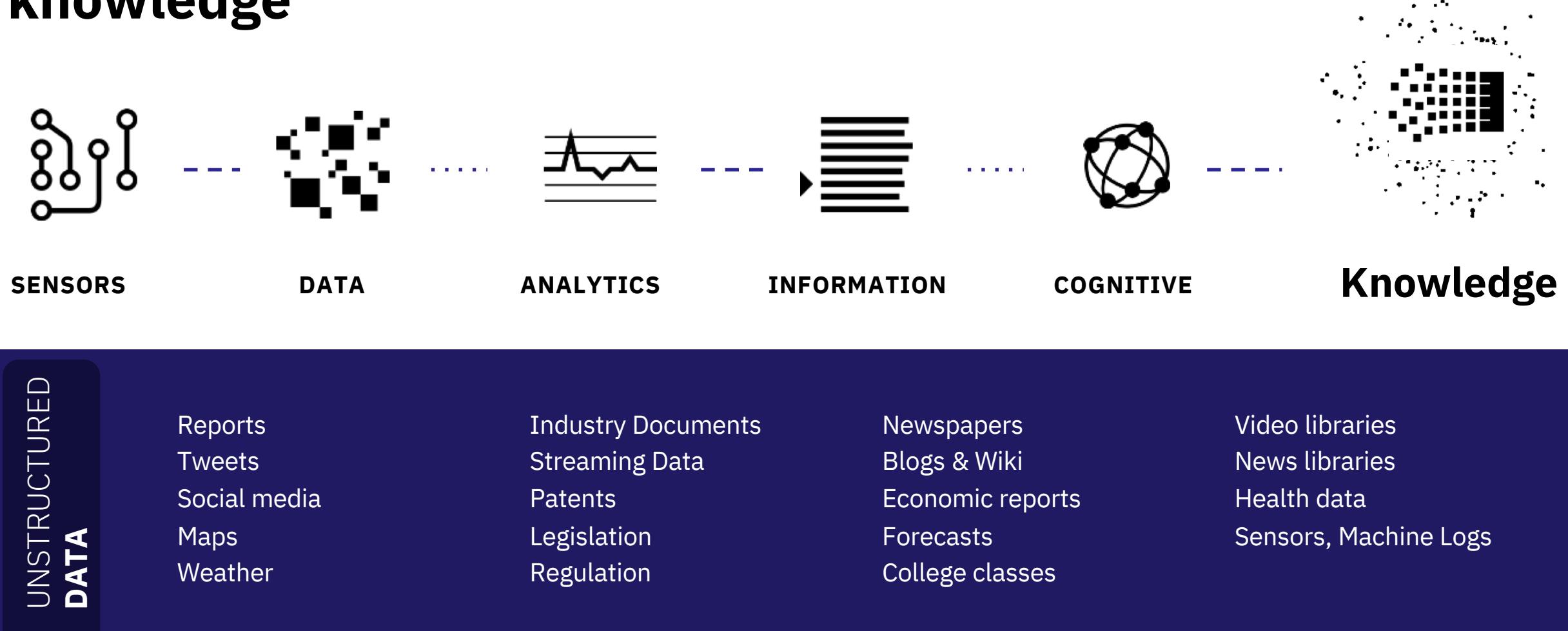


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How to stay ahead of the ever-changing expectations customers have for what's possible, leading your market segment in new ways

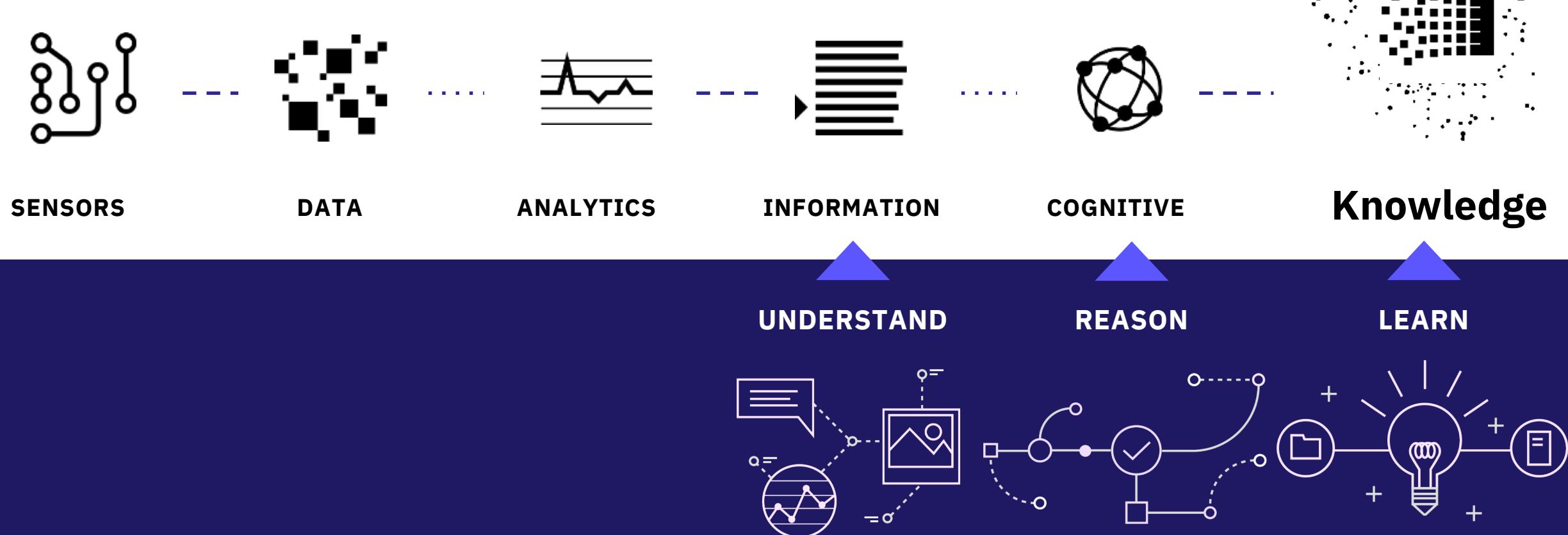
## WHAT IS THE SOLUTION?

**AI systems can transform unstructured data into knowledge**

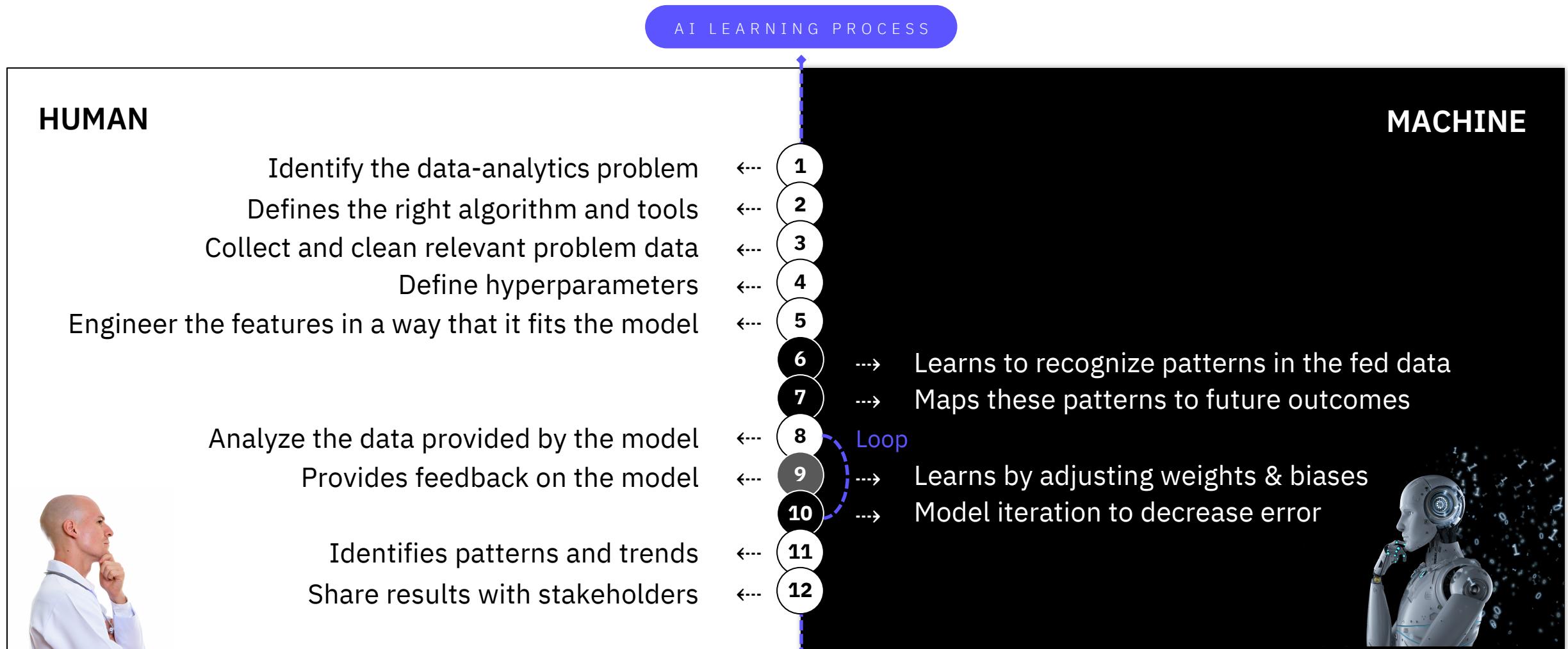


## WHAT IS THE SOLUTION?

**AI systems can transform unstructured data into knowledge**



# How does an AI system learn?

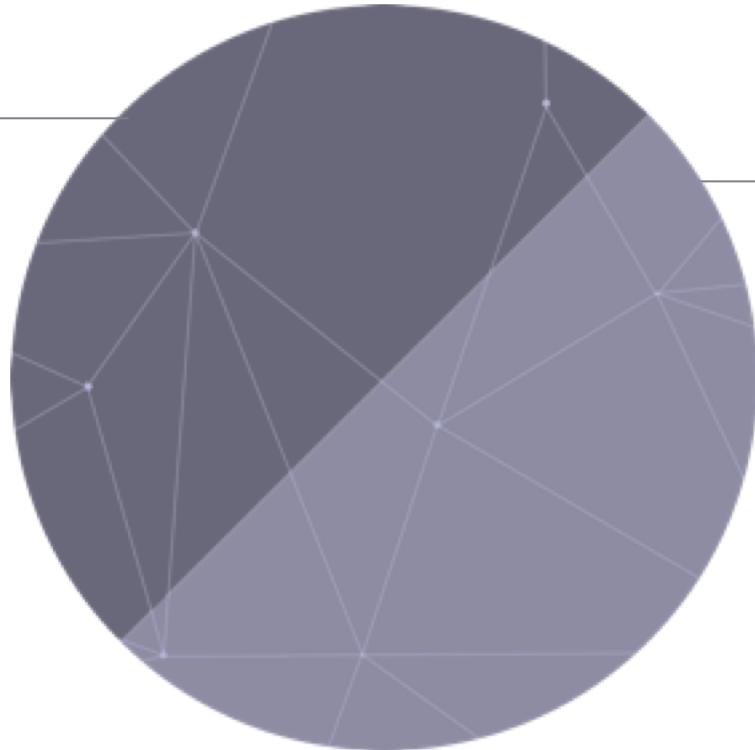


# AI enables a partnership between humans and technology

## Humans

excel at

- Morals
- Imagination
- Compassion
- Dilemmas
- Abstraction
- Dreaming
- Generalization
- Common sense (**but with many biases**)



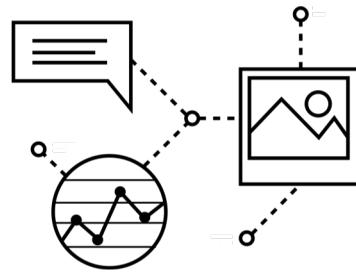
## AI Systems

excel at

- Locating knowledge
- Natural language
- Pattern identification
- Machine learning
- Providing endless capacity
- Eliminating Biases**

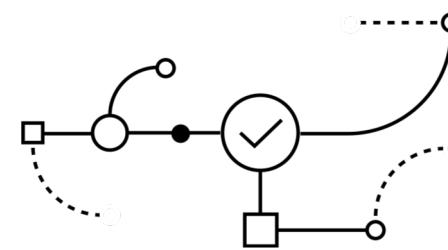
# AI systems can go beyond knowledge, they **understand, reason, learn, and interact**

## UNDERSTAND



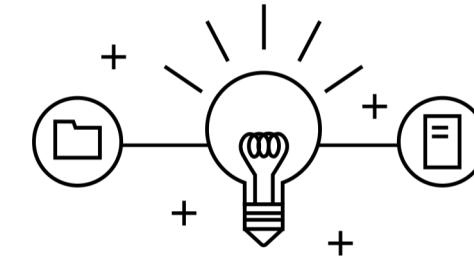
AI systems can understand unstructured information the same way humans do

## REASON



They can reason, grasp underlying concepts, form hypotheses, and infer to extract ideas

## LEARN



Each data point, interaction and outcome helps to continuously sharpen expertise

## INTERACT



With abilities to see, talk and hear, AI systems interact with humans in a natural way

# AI systems features

**We can take the best experts and make them better by introducing man and machine.**

- **Learn** at scale
- **Reason** with purpose
- **Interact** with humans naturally



# How AI can improve our world?

Video 30:57 minutes



Source: <https://www.youtube.com/watch?v=q7qElhGv7uY>

## LECTURE 1

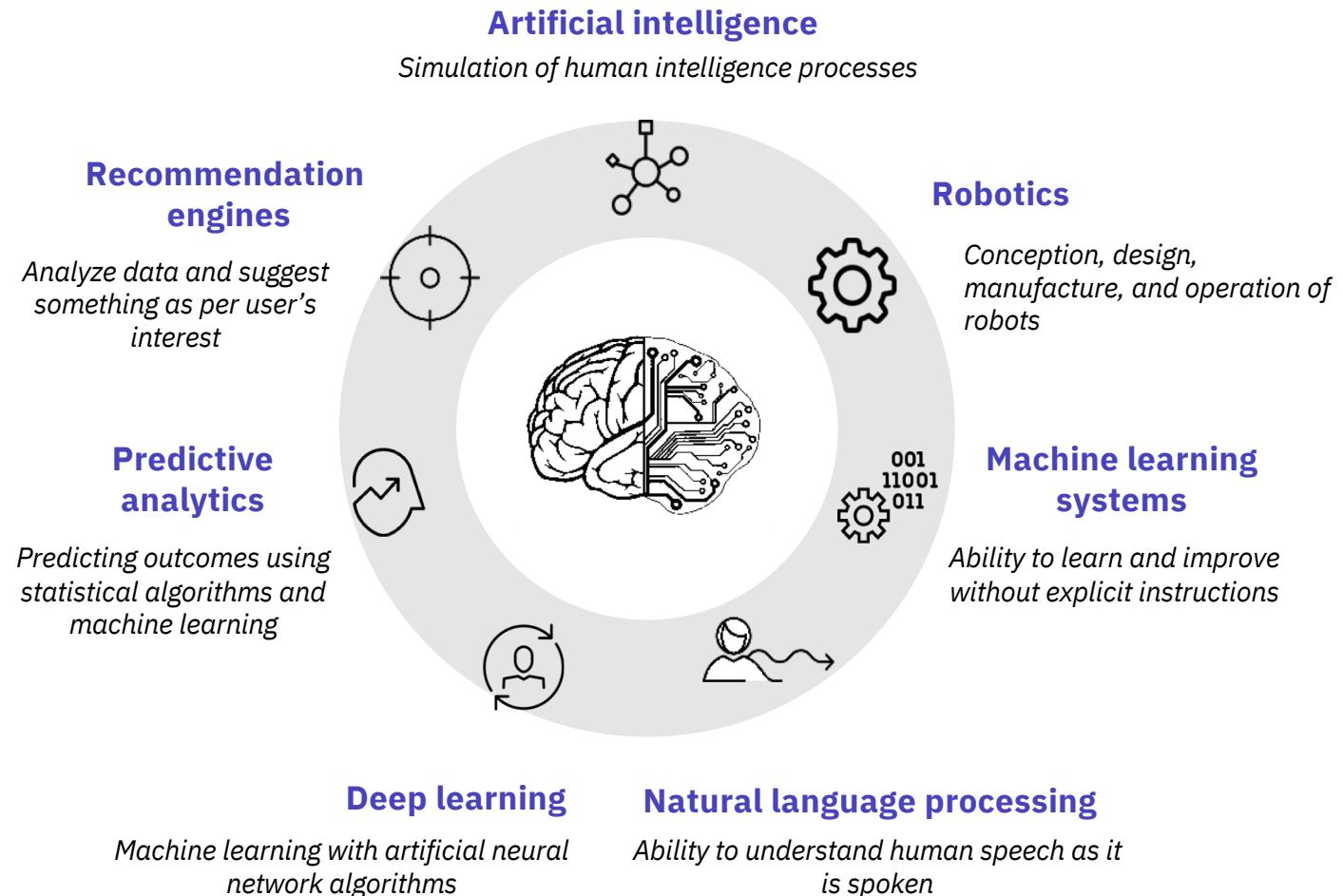
### *Artificial Intelligence Landscape*

1. AI impact in the world today
2. History and Evolution of AI
3. AI Explained
4. **AI Technologies**
5. Summary & Resources

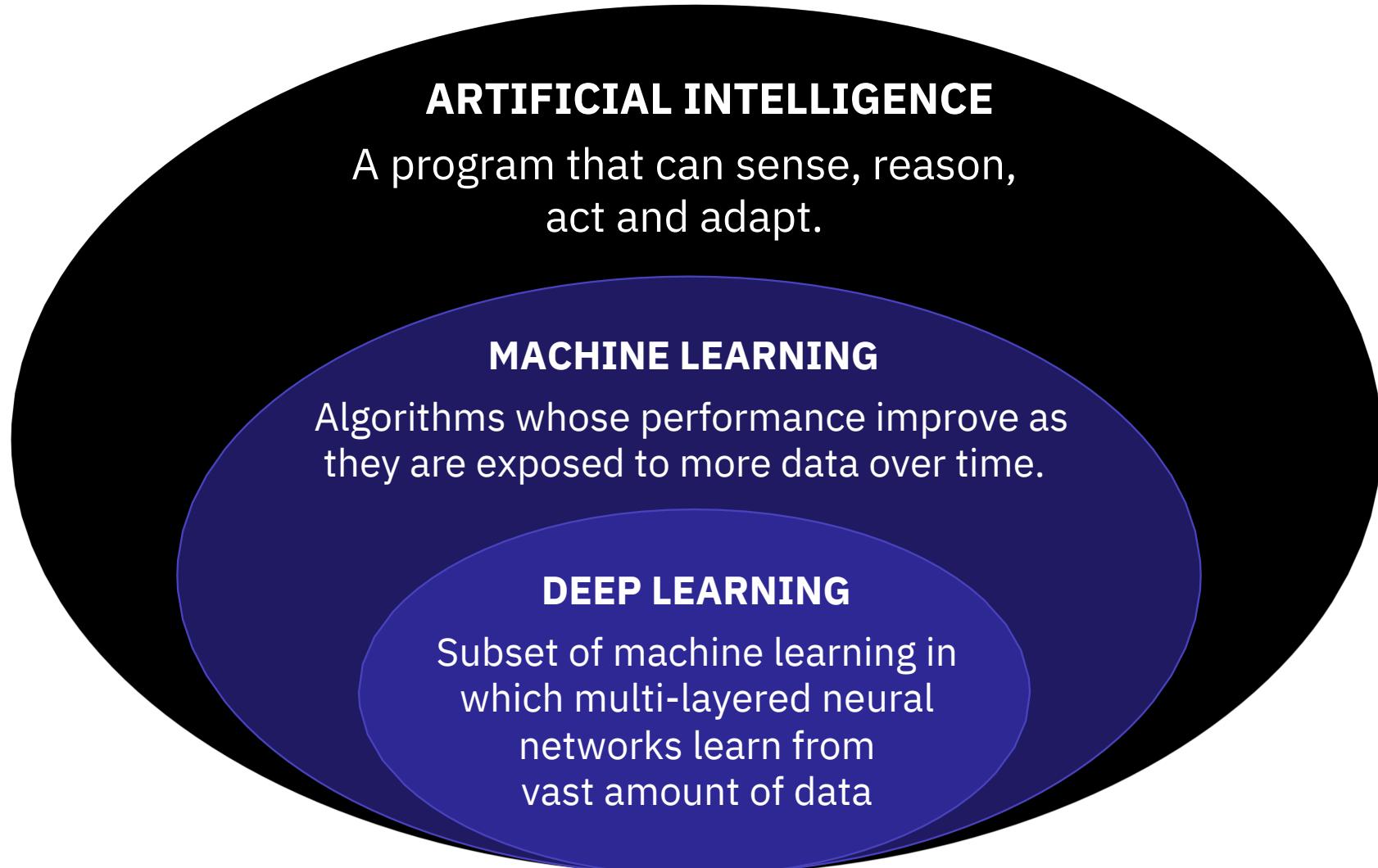


# AI systems are underpinned by advanced technologies

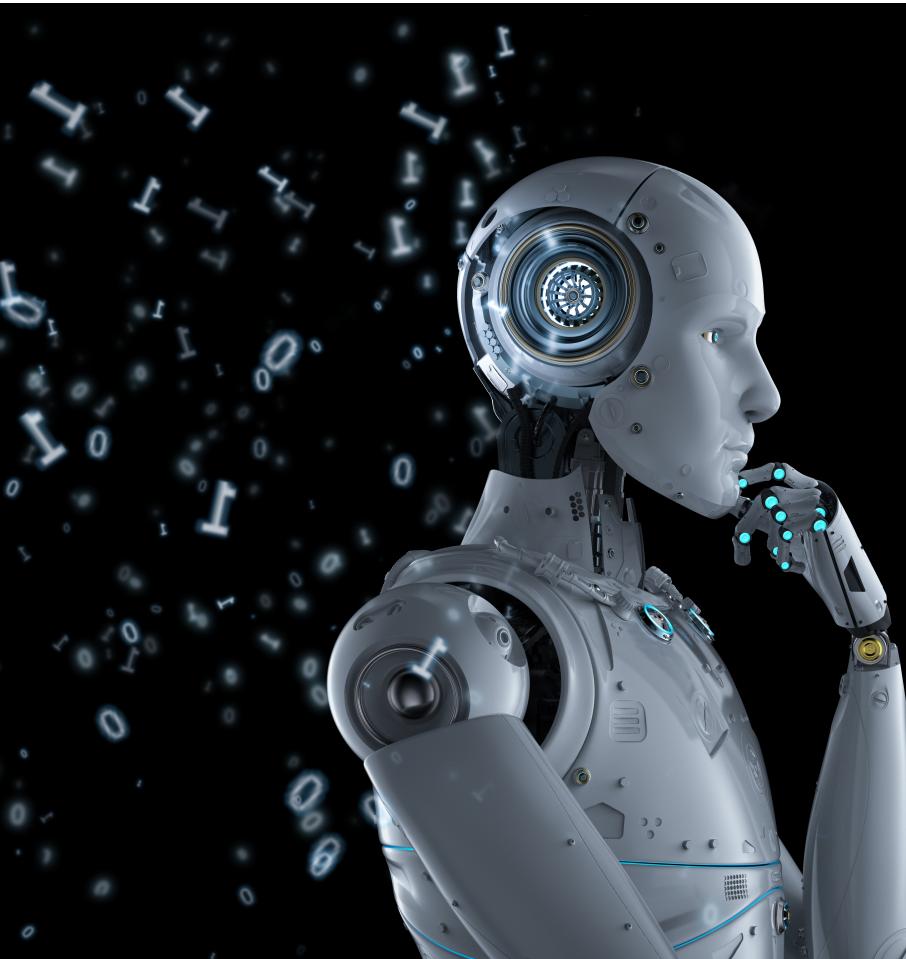
## Cognitive/AI Technologies



# Difference between AI, Machine Learning and Deep Learning



# Artificial Intelligence: Definition

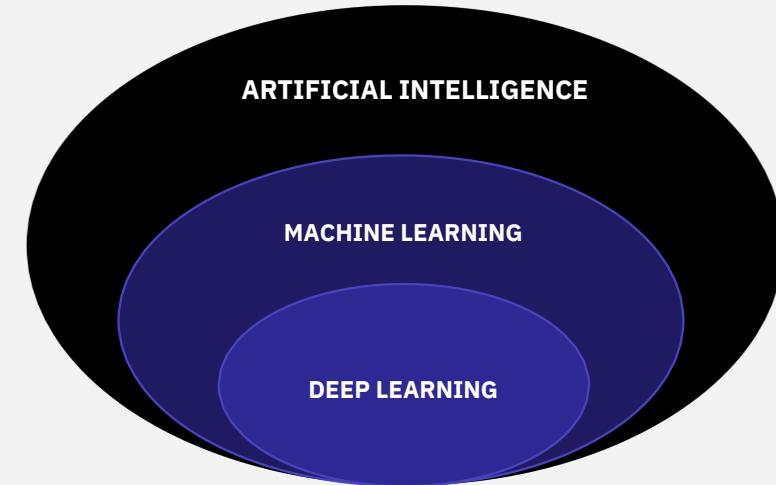


“Artificial Intelligence (AI) is a science and a **set of computational technologies** that are *inspired* by – but typically operate quite differently from – the ways people use their nervous systems and bodies to **sense, learn, reason and take action**.”

# Machine Learning

Machine Learning algorithms iteratively learn from data, thus allowing computers to find hidden insights without being explicitly programmed where to look.

Machine Learning is essentially teaching the computer to solve problems



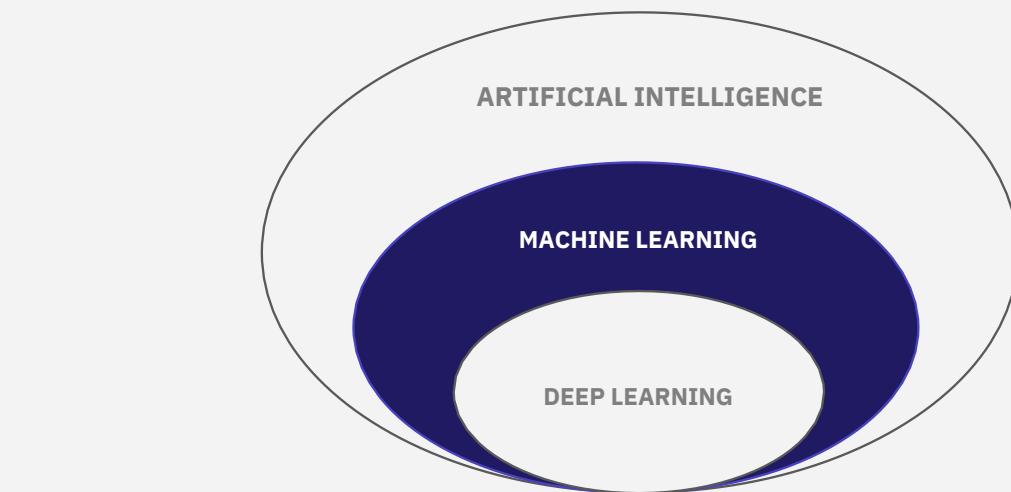
Three broad categories:

- 1. Supervised learning**
- 2. Unsupervised learning**
- 3. Reinforcement learning**

# 1. Supervised Learning

Supervised learning trains on large volumes of historical data and then builds general rules to be applied to future problems.

Example: Voter details labeled with their votes (label) in the previous years

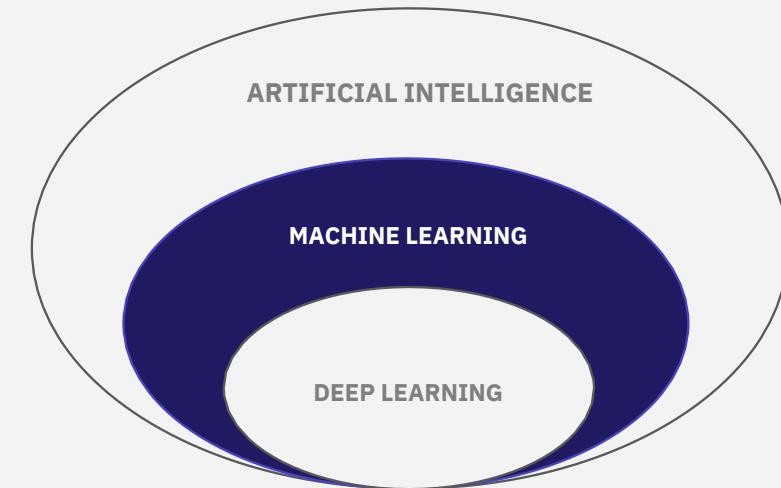


The most widely used supervised learning algorithms are [Support Vector Machines](#), [Linear Regression](#), [Logistic Regression](#), [Naive Bayes](#), and [Neural Networks \(multilayer perceptron\)](#).

## 2. Unsupervised Learning

While supervised learning relies on labeled or structured data (think rows in a database), unsupervised learning trains on unlabeled or unstructured data (the text of a book).

These algorithms explore the data and try to find structure.

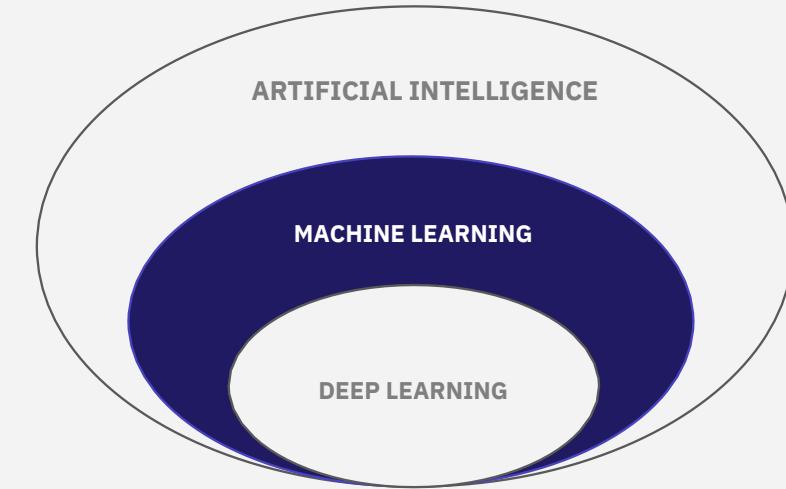


Here, widely used unsupervised learning algorithms are cluster analysis and market basket analysis.

### 3. Reinforcement Learning

It is not given a specific goal, but rather learns from trial and error.

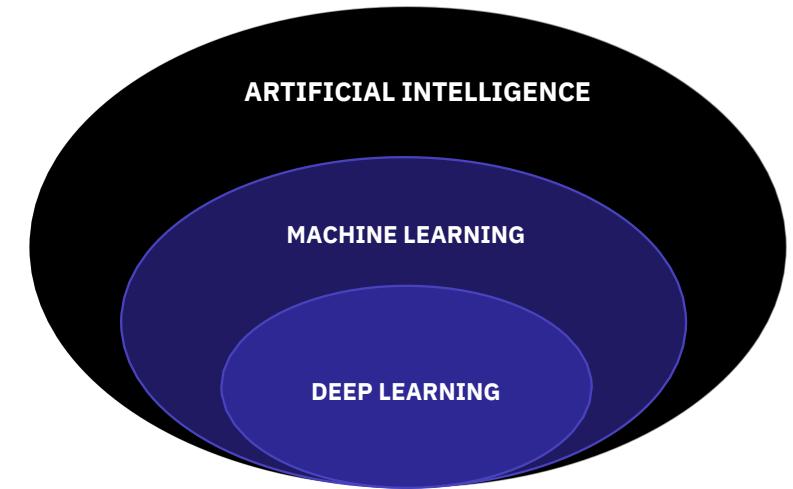
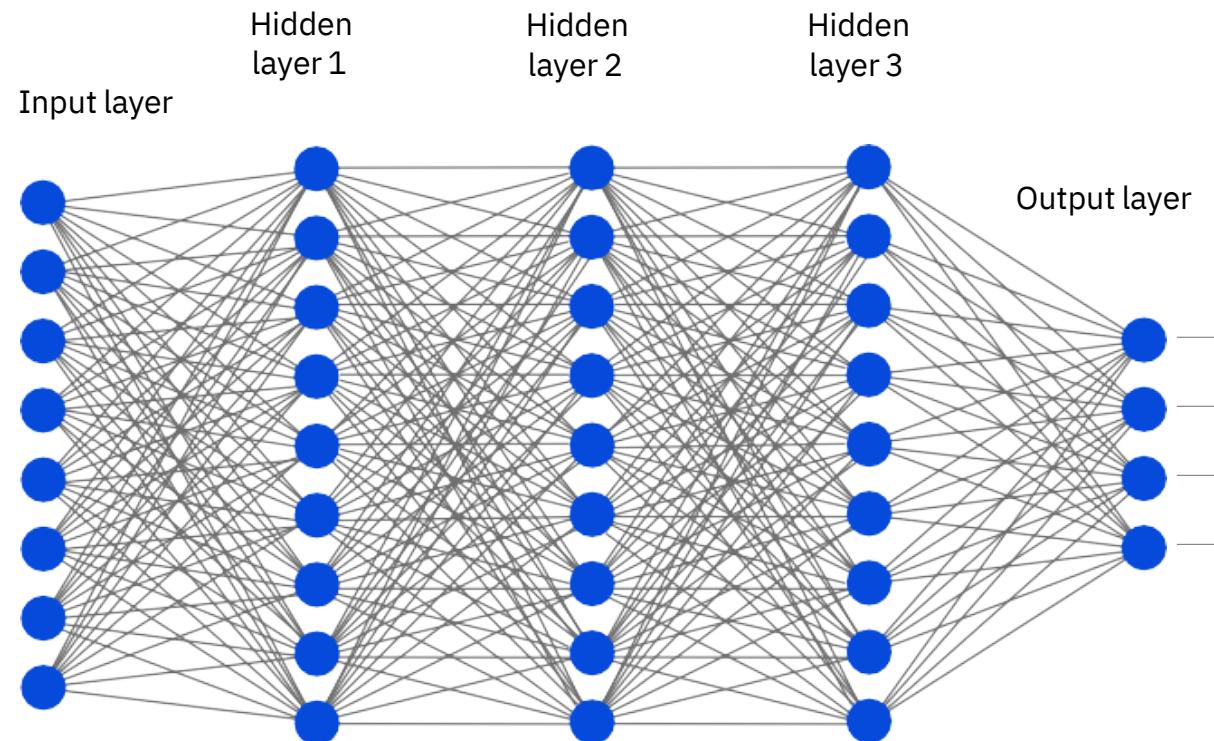
Reinforcement learning rewards the algorithm when it performs the correct action (behavior), and assigns a penalty when incorrect.



Reinforcement learning is most widely used in self-driven cars, drones, and other robotics applications.

# Deep Learning

**Is a family of algorithms that implements deep networks (many layers)**



DL almost always outperforms the other types of algorithms when it comes to:

- Image classification
- Natural language processing
- Speech recognition.

*Example:*

Recognizing melanoma or conducting machine translation, which was not possible using previous techniques.

# Transfer Learning

Transfer learning is the method of starting with a pre-trained model and training it for a new – related – problem domain.

The pre-trained network serves as transferred knowledge to be applied in another domain.

## Available pre-trained models:

[MobileNet](#) is model-trained on the ImageNet database (covering millions of images with 20,000 classifications). It can perform object detection, landmark recognition

[Object Detection](#) is capable of localizing and identifying multiple objects in a single image.

[Sentiment Discovery](#) This NLP model can identify sentiment of natural language, but also indicate through a heat map the positive and negative elements of text.

[YOLO for TensorFlow++](#) is a real-time object detection on mobile devices, can detect people and other objects in its field of view.

[Car Classification](#), using the Core ML framework (Apple Devices), takes images and can output a prediction of the vehicle present (up to 431 vehicle models)

[Lip Reading](#) is a model that can correlate an audio track to a video to properly orient the audio to the video based upon lip reading.

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# Summary

- AI is not a future statement, it is real today, and it fuels the 4<sup>th</sup> Industrial Revolution.
- Properly implementing AI requires careful evaluation and planning.
- There is no one magic algorithm that can solve all your problems.
  - Supervised learning is currently the most applied form of learning and provides the most value for a wide variety of enterprise applications.

# Resources page 1 of 2

1. This is A.I. - Discovery Channel and IBM TV Show  
<https://www.discovery.com/tv-shows/this-is-a-i/>
2. Beyond the hype: A guide to understanding and successfully implementing artificial intelligence within your business  
<https://www.ibm.com/downloads/cas/8ZDXNKQ4>
3. AI and Quantum: What's Next  
<https://www.ibm.com/blogs/systems/changing-lives-with-ai-and-quantum-computing-whats-next/>
4. A Practical Guide to Building Enterprise Applications: by Tom Markiewicz and Josh Zheng – Feb 2018 O'Reilly  
<https://tmarkiewicz.com/getting-started-with-artificial-intelligence/>
5. Beginner Guide to Machine Learning  
<https://developer.ibm.com/articles/cc-beginner-guide-machine-learning-ai-cognitive/>
6. Machine Learning, Deep Learning 101  
[https://developer.ibm.com/articles/l-machine-learning-deep-learning-trs/?mhq=deep%20learning%20101&mhsrc=ibmsearch\\_a](https://developer.ibm.com/articles/l-machine-learning-deep-learning-trs/?mhq=deep%20learning%20101&mhsrc=ibmsearch_a)
7. This is A.I. Discovery Channel and IBM TV Show  
<https://www.discovery.com/tv-shows/this-is-a-i/>

# Resources page 2 of 2

8. Dr Patrick Soon Shiong, NantHealth @ NHS Confederation Conference June 4, 2015  
<https://www.youtube.com/watch?v=TmXzi7AZxX8>
9. Global Mobile Data Traffic Growth  
<http://www.telecomtv.com/articles/mobile/cisco-predicts-292-exabytes-of-mobile-data-by-2019-12153/>
10. What Do Data Scientists Do?  
<https://datasciencedegree.wisconsin.edu/data-science/what-do-data-%20%20scientists-do/>
11. The future of Cognitive computing  
<https://www.ibm.com/blogs/bluemix/2015/11/future-of-cognitive-computing/>
12. The future of Cognitive computing – John Kelly Presentation  
<https://www.youtube.com/watch?v=q7qElhGv7uY>
13. Why are deep neural networks hard to train?  
<http://www.neuralnetworksanddeeplearning.com/chap5.html>
14. Transfer learning for deep learning  
<https://developer.ibm.com/articles/transfer-learning-for-deep-learning/>



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