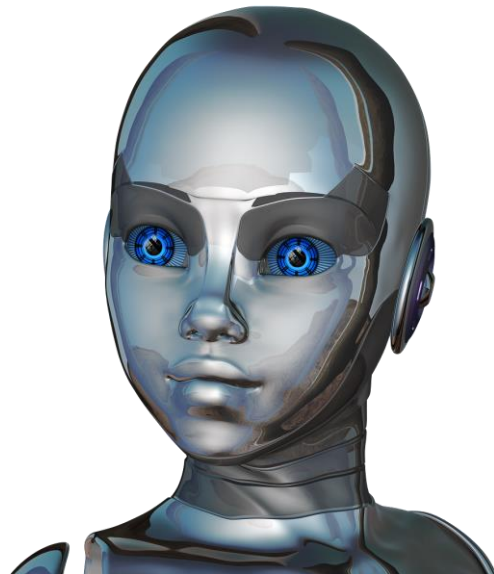


# Overview

:

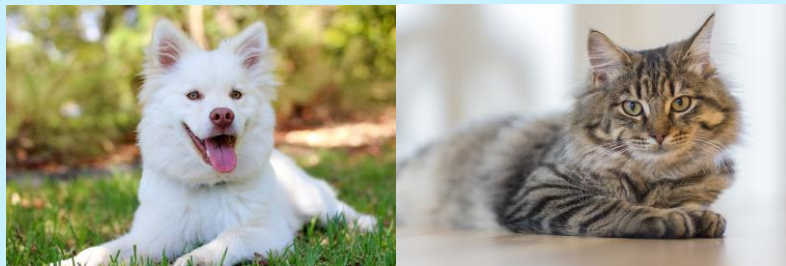
- Define “Artificial Intelligence” (AI),  
“Machine Learning” (ML), and “Deep Learning” (DL)
- Explain how DL helps solve classical ML limitations.
- Brief History of AI
- Differentiate modern AI from prior AI.
- Relate sample applications of AI.



*AI robot*

# AI Breakthroughs

## Image classification



*“Dog”*

*“Cat”*

As of 2015, computers can be trained to perform better on this task than humans.

## Machine translation

*“I am a student”*



*“Je suis étudiant”*

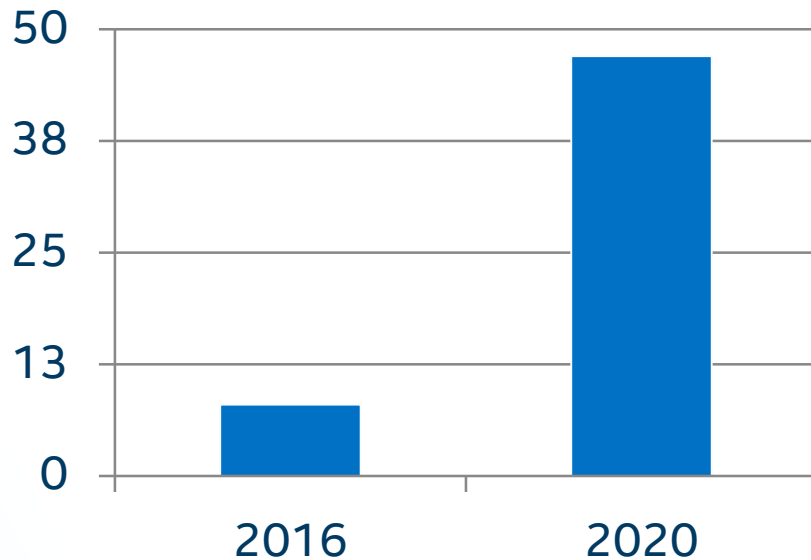
As of 2016, we have achieved near-human performance using the latest AI techniques.

# AI Is The New Electricity

*"About 100 years ago, electricity transformed every major industry. AI has advanced to the point where it has the power to transform...every major sector in coming years."*

*-Andrew Ng, Stanford University*

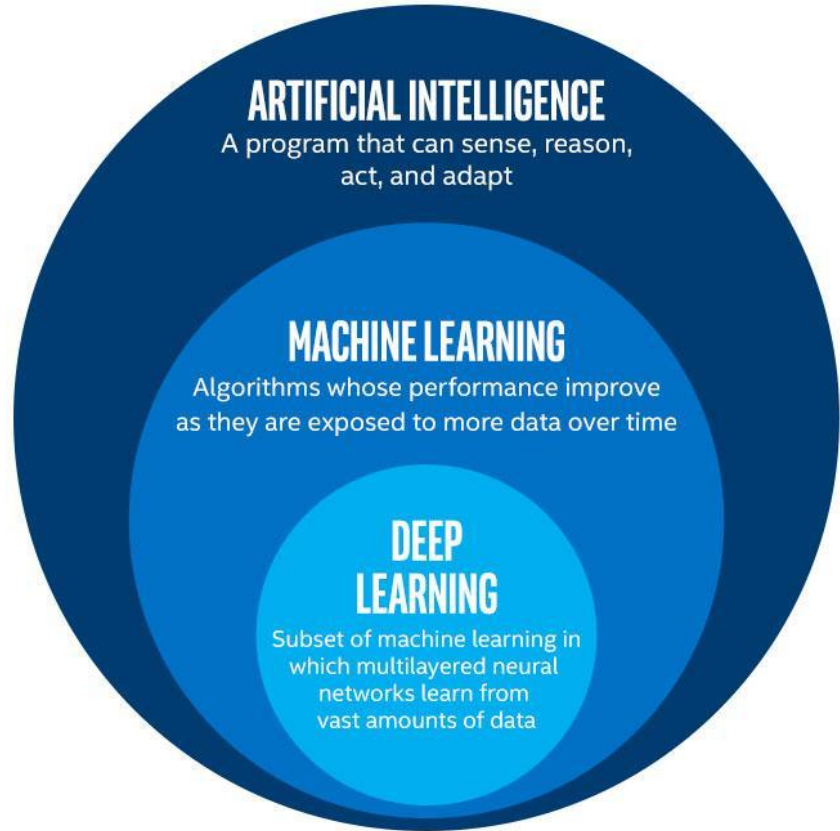
*Projected Revenue (in billions USD)  
Generated from AI, 2016-2020 (IDC)*



# DEFINITIONS

# Definitions

- Artificial Intelligence
- Machine Learning
- Deep Learning



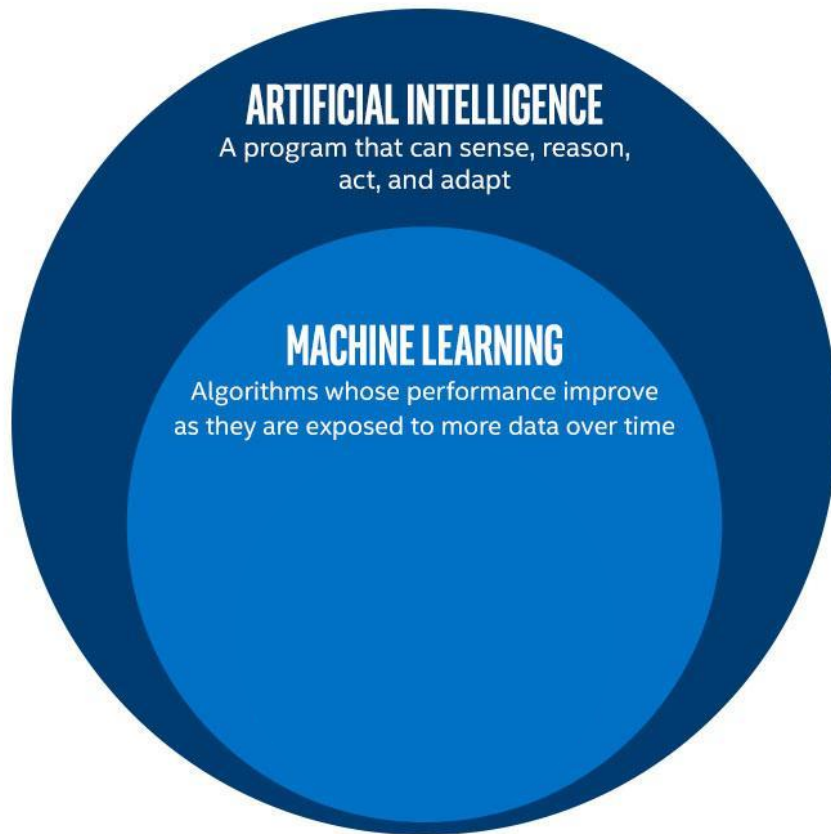
# Artificial Intelligence

“A branch of computer science dealing with the simulation of intelligent behavior in computers.” (Merriam-Webster)

“Colloquially, the term ‘artificial intelligence’ is applied when a machine mimics ‘cognitive’ functions that humans associate with other human minds, such as ‘learning’ and ‘problem solving’.” (Wikipedia)

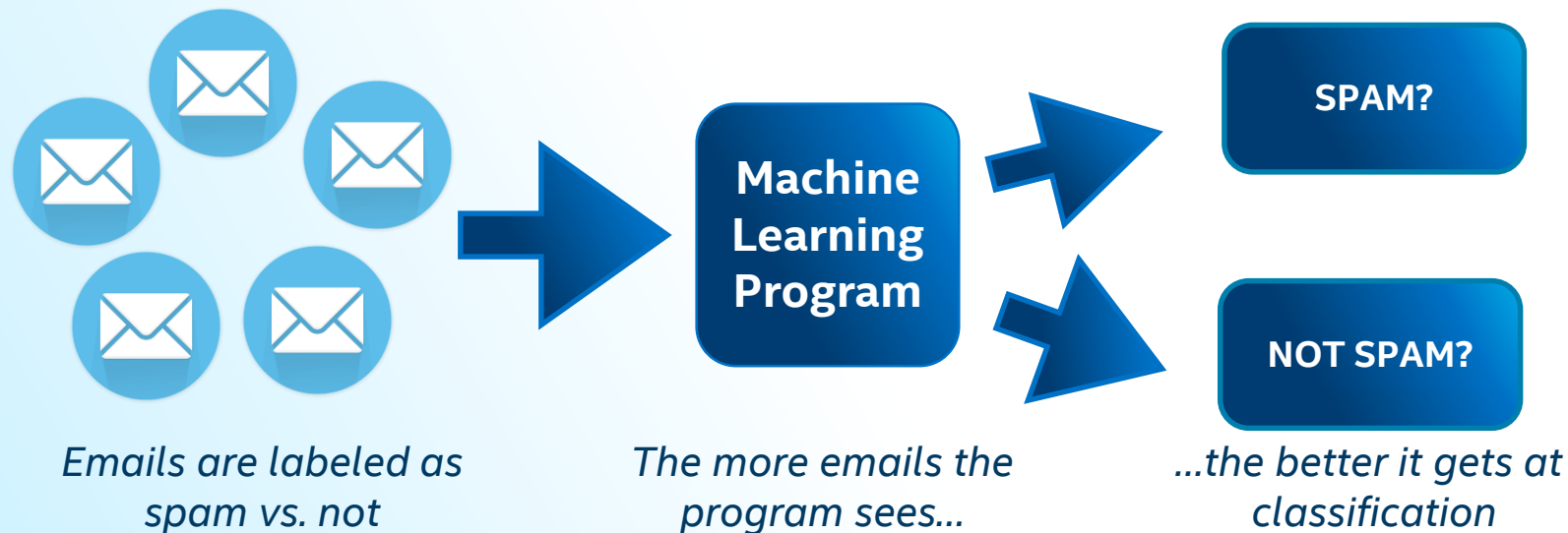
# Machine Learning

“The study and construction of programs that are *not explicitly programmed*, but learn patterns as they are exposed to more data over time.”



# Machine Learning

These programs learn from repeatedly seeing data, rather than being explicitly programmed by humans.





# Machine Learning Terminology

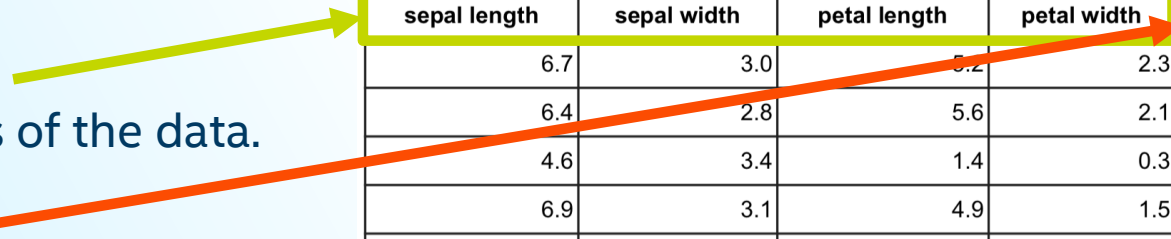
This example is learning to classify a species from a set of measurement features.

**Features:**

Attributes of the data.

**Target:**

Column to be predicted.



The diagram illustrates the relationship between features and the target variable. A yellow arrow points from the 'Features' label to the first four columns of the table (sepal length, sepal width, petal length, petal width). An orange arrow points from the 'Target' label to the 'species' column. The 'species' column is highlighted with an orange border, while the feature columns are highlighted with a yellow border.

sepal length	sepal width	petal length	petal width	species
6.7	3.0	5.2	2.3	virginica
6.4	2.8	5.6	2.1	virginica
4.6	3.4	1.4	0.3	setosa
6.9	3.1	4.9	1.5	versicolor
4.4	2.9	1.4	0.2	setosa
4.8	3.0	1.4	0.1	setosa
5.9	3.0	5.1	1.8	virginica
5.4	3.9	1.3	0.4	setosa
4.9	3.0	1.4	0.2	setosa
5.4	3.4	1.7	0.2	setosa

# Two Main Types of Machine Learning

	Dataset	Goal	Example
Supervised Learning	Has a target column	Make predictions	Fraud detection
Unsupervised Learning	Does not have a target column	Find structure in the data	Customer segmentation

# Machine Learning Example

- Suppose you wanted to identify fraudulent credit card transactions.
- You could define features to be:
  - Transaction time
  - Transaction amount
  - Transaction location
  - Category of purchase
- The algorithm could learn what feature combinations suggest unusual activity.



*Credit card transactions*

# Machine Learning Limitations

- Suppose you wanted to determine if an image is of a cat or a dog.
- What features would you use?
- This is where **Deep Learning** can come in.

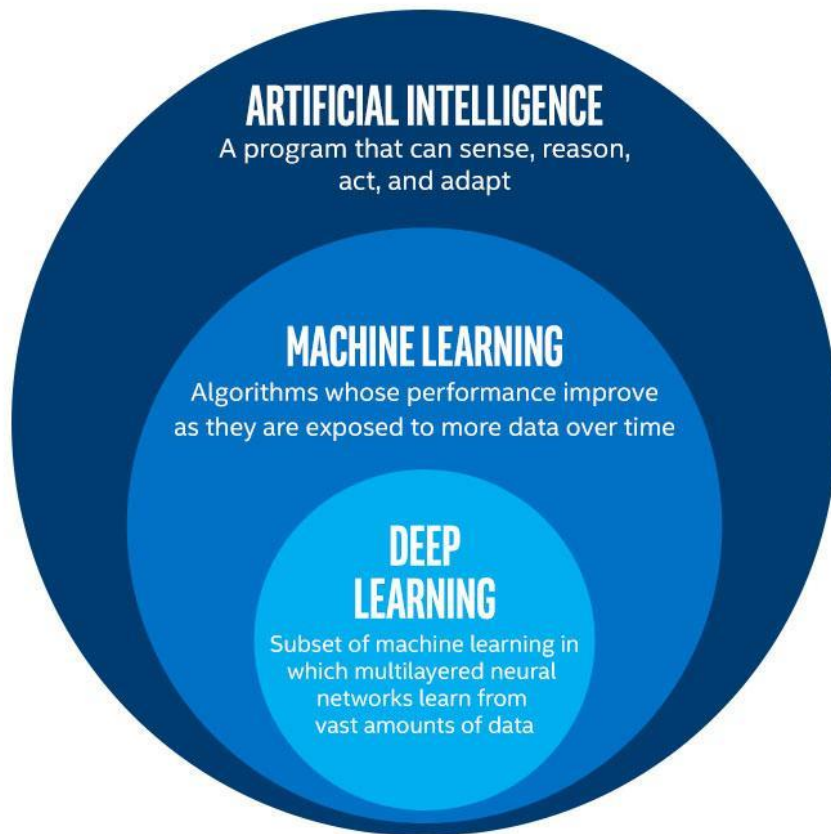


*Dog and cat recognition*

# Deep Learning

“Machine learning that involves using very complicated models called “deep neural networks”.

*Models* determine best representation of original data; in classic machine learning, humans must do this.



# Deep Learning Example

## Classic Machine Learning

Step 1: Determine features.  
Step 2: Feed them through model.



Feature Detection

Machine Learning Classifier Algorithm

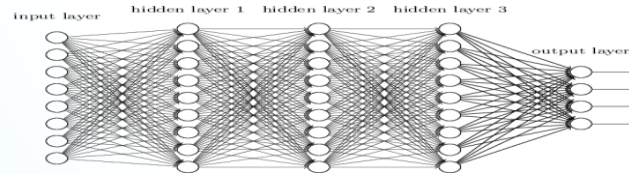
"G Hinton"

## Deep Learning

Steps 1 and 2 are combined into 1 step.



Neural Network

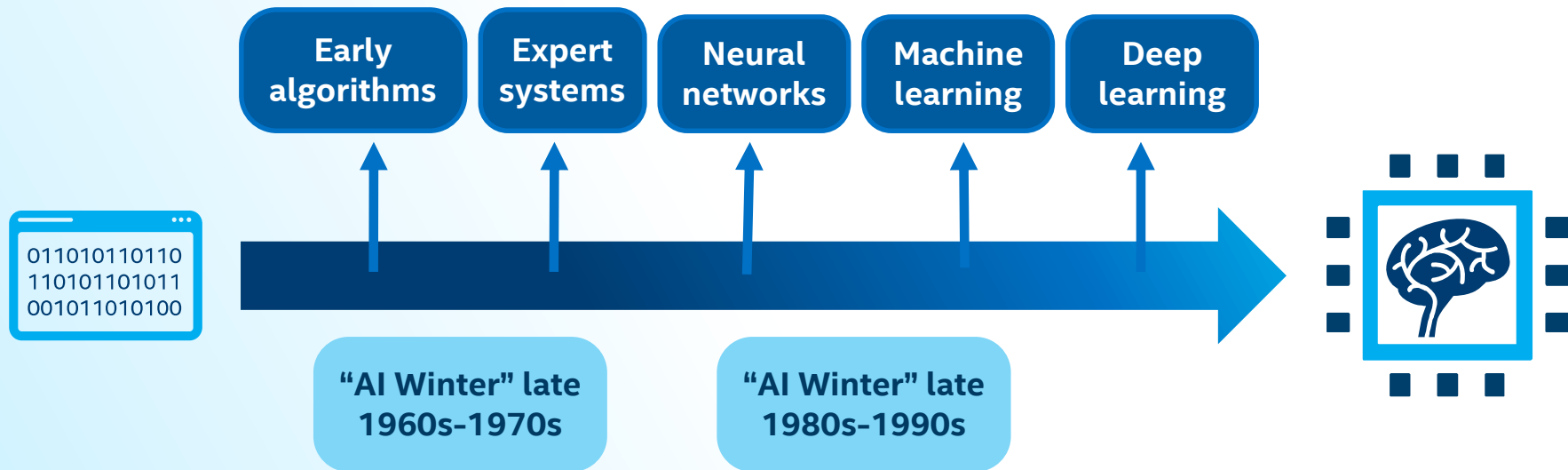


"G Hinton"

# HISTORY

# History of AI

AI has experienced several hype cycles, where it has oscillated between periods of excitement and disappointment.





**MODERN AI**

# Deep Learning Breakthroughs (2012 – Present)

- In 2012, deep learning beats previous benchmark on the ImageNet competition.
- In 2013, deep learning is used to understand “conceptual meaning” of words.
- In 2014, similar breakthroughs appeared in language translation.
- These have led to advancements in Web Search, Document Search, Document Summarization, and Machine Translation.



*Google Translate*

# Deep Learning Breakthroughs (2012 – Present)

- In 2014, computer vision algorithm can describe photos.
- In 2015, Deep learning platform TensorFlow is developed.
- In 2016, DeepMind's AlphaGo, developed by Aja Huang, beats Go master Lee Se-dol.



*Autonomous Mars rover*

# Modern AI (2012 – Present): Deep Learning Impact

## Computer vision



Self-driving cars:  
object detection



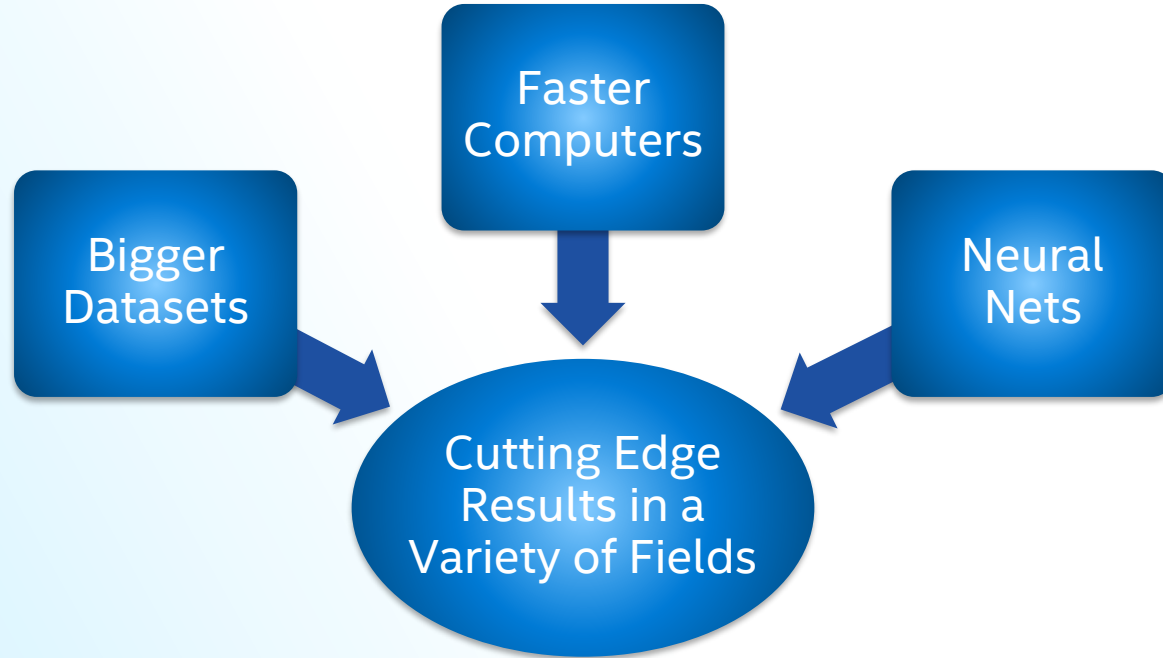
Healthcare:  
improved diagnosis

## Natural language



Communication:  
language translation

# How Is This Era of AI Different?



# Other Modern AI Factors

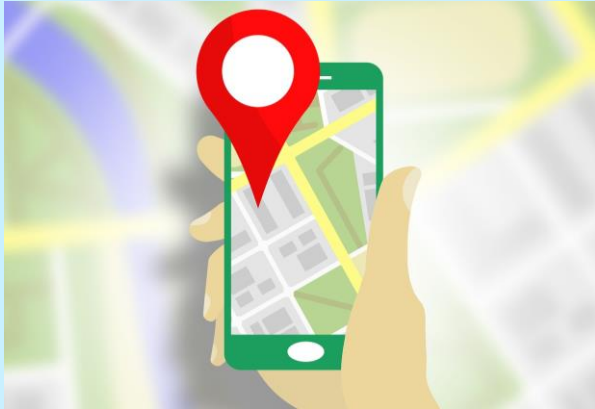
- Continued expansion of open source AI, especially in Python™, aiding machine learning and big data ecosystems.
- Leading deep learning libraries *open sourced*, allowing further adoption by industry.
- Open sourcing of large datasets of millions of labeled images, text datasets such as Wikipedia has also driven breakthroughs.



# APPLICATIONS

# AI Omnipresence In Transportation

## Navigation



Google™ & Waze™ find the fastest route, by processing traffic data.

## Ride sharing



Uber™ & Lyft™ predict real-time demand using AI techniques, machine learning, deep learning.



# AI Omnipresence In Social Media

## Audience



Facebook™ & Twitter™ use AI to decide what content to present in their feeds to different audiences.

## Content

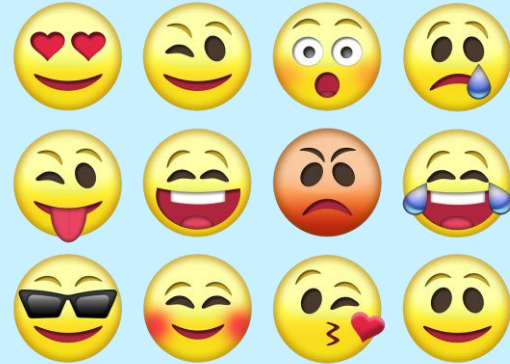


Image recognition and sentiment analysis to ensure that content of the appropriate “mood” is being served.

# AI Omnipresence In Daily Life

## Natural language



We carry around powerful natural language processing algorithms in our phones/computers.

## Object detection

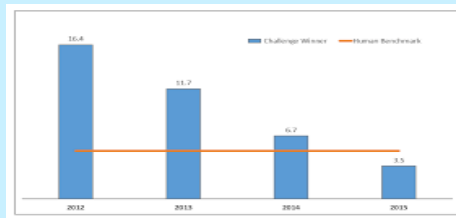


Cameras like Amazon's DeepLens™ or Google Clips™ use object detection to determine when to take a photo.

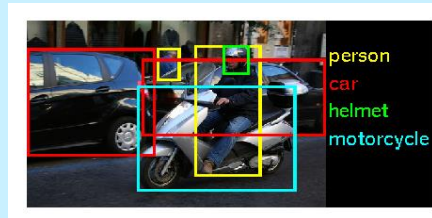
# Latest Developments: Computer Vision



Deep Learning  
“proven” to work for  
image classification.



Models outperform  
humans on image  
classification.



Object detection  
models beat previous  
benchmarks.

2012

2015

2016

# Application Area: Abandoned Baggage Detection

- We can automatically detect when baggage has been left unattended, potentially saving lives.
- This system relies on the breakthroughs we discussed:
  - Cutting edge object detection.
  - Fast hardware on which to train the model

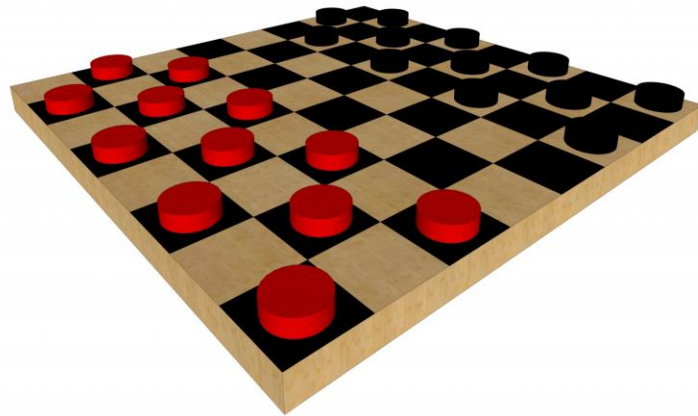


*Abandoned baggage*

# HISTORY

# 1950s: Early AI

- 1950: Alan Turing developed the Turing test to test a machine's ability to exhibit intelligent behavior.
- 1956: Artificial Intelligence was accepted as a field at the Dartmouth Conference.
- 1957: Frank Rosenblatt invented the perceptron algorithm. This was the precursor to modern neural networks.
- 1959: Arthur Samuel published an algorithm for a checkers program using machine learning.



*Checkerboard program*

# The First “AI Winter”

- 1966: ALPAC committee evaluated AI techniques for machine translation and determined there was little yield from the investment.
- 1969: Marvin Minsky published a book on the limitations of the Perceptron algorithm which slowed research in neural networks.
- 1973: The Lighthill report highlights AI's failure to live up to promises.
- The two reports led to cuts in government funding for AI research leading to the first “AI Winter”



*John R. Pierce, head of ALPAC*

# 1980's AI Boom

- Expert Systems - systems with programmed rules designed to mimic human experts.
- Ran on mainframe computers with specialized programming languages (e.g. LISP).
- Were the first widely-used AI technology, with two-thirds of "Fortune 500" companies using them at their peak.
- 1986: The "Backpropagation" algorithm is able to train multi-layer perceptrons leading to new successes and interest in neural network research.



*Early expert systems machine*



# Another AI Winter (late 1980's – early 1990s)

- Expert systems' progress on solving business problems slowed.
- Expert systems began to be melded into software suites of general business applications (e.g. SAP®, Oracle®) that could run on PCs instead of mainframes.
- Neural networks didn't scale to large problems.
- Interest in AI in business declined.

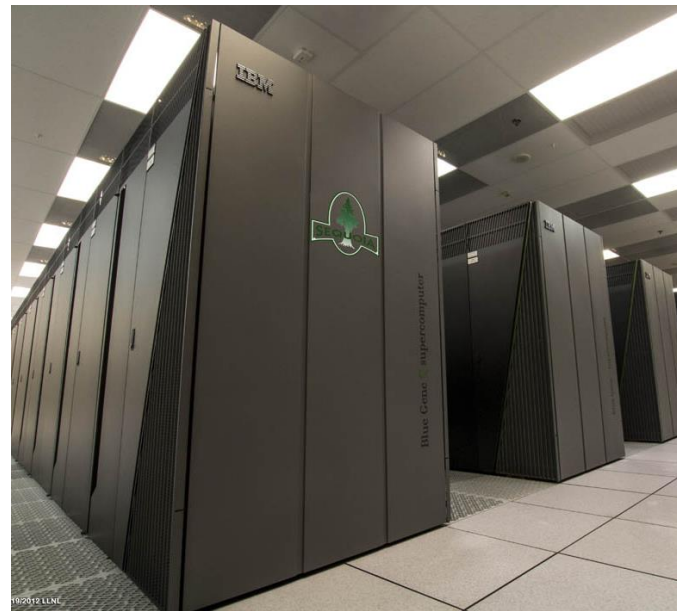


**ORACLE®**

*Software companies*

# Late 1990's to early 2000's: Classical Machine Learning

- Advancements in the SVM algorithm led to it becoming the machine learning method of choice.
- AI solutions had successes in speech recognition, medical diagnosis, robotics, and many other areas.
- AI algorithms were integrated into larger systems and became useful throughout industry.
- The Deep Blue chess system beat world chess champion Garry Kasparov.



*IBM supercomputer*

# 2006: Rise of Deep Learning

- 2006: Geoffrey Hinton publishes a paper on unsupervised pre-training that allowed deeper neural networks to be trained.
- Neural networks are rebranded to deep learning.
- 2009: The ImageNet database of human-tagged images is presented at the CVPR conference.
- 2010: Algorithms compete on several visual recognition tasks at the first ImageNet competition.



# Transformative Changes



## Health

Enhanced  
Diagnostics  
Drug Discovery  
Patient Care  
Research  
Sensory Aids



## Industrial

Factory  
Automation  
Predictive  
Maintenance  
Precision  
Agriculture  
Field  
Automation

Source: Intel forecast

# Transformative Changes



## Finance

- Algorithmic Trading
- Fraud Detection
- Research
- Personal Finance
- Risk Mitigation



## Energy

- Oil & Gas Exploration
- Smart Grid
- Operational Improvement
- Conservation

Source: Intel forecast

# Transformative Changes



## Government

Defense  
Data  
Insights  
Safety &  
Security  
Engagement  
Smarter  
Cities



## Transport

Autonomous  
Cars  
Automated  
Trucking  
Aerospace  
Shipping  
Search & Rescue

Source: Intel forecast



# Transformative Changes



## Other

Advertising  
Education  
Gaming  
Professional &  
IT Services  
Telco/Media  
Sports