

Data & risks



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The first algorithm?

BC

1843

1956

1997

The first algorithm!

**300
BC**

- Euclid / Euclides
- Ca. 325 – 270 BC
- Alexandria, Greece
- Mathematician

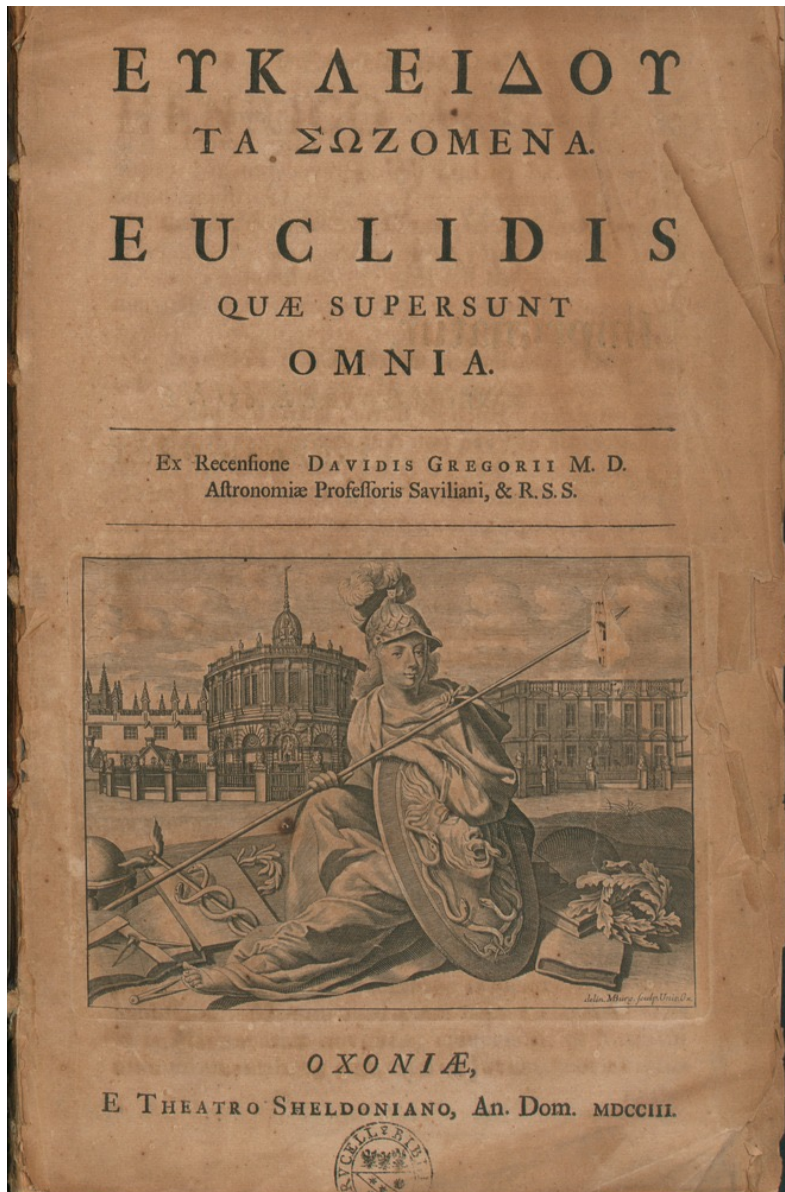


Algorithm still used today!

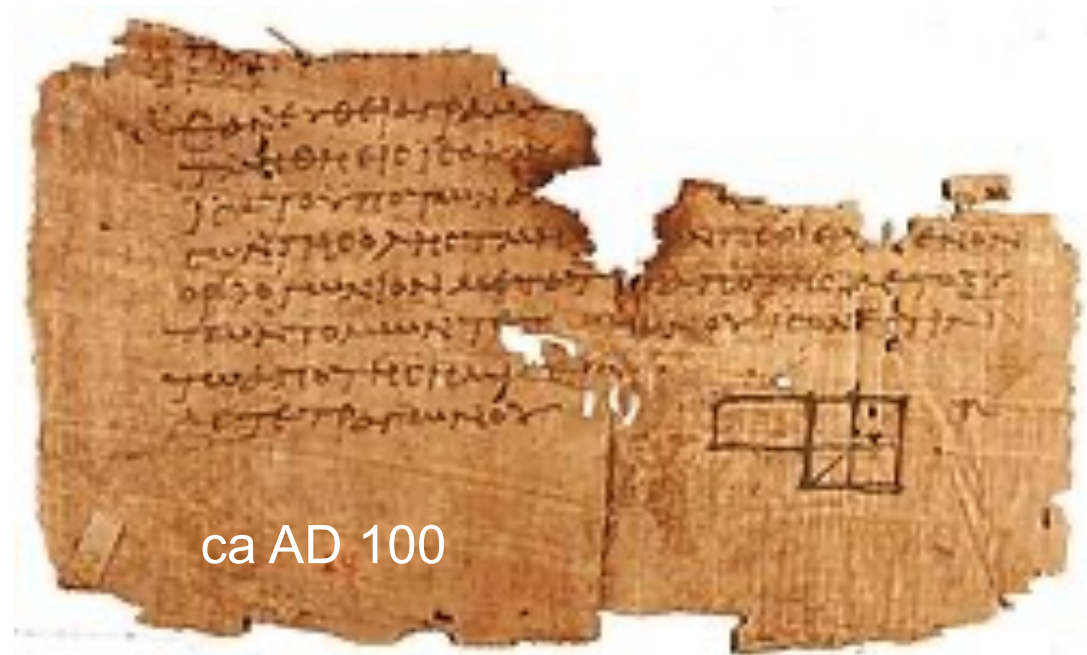
Raphael
School of Athens
ca 1500



Claim to fame: **The elements**

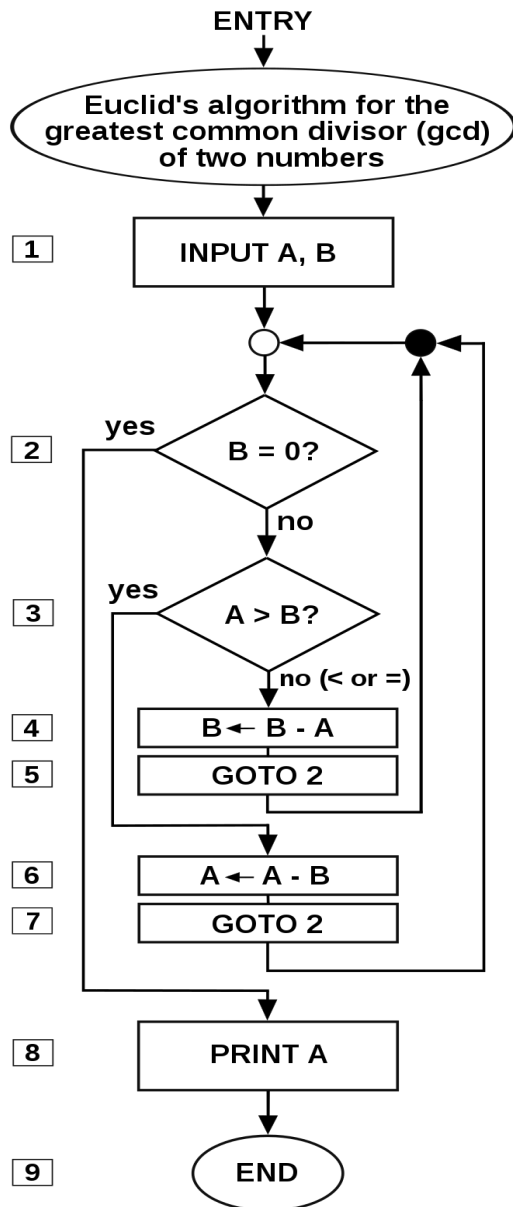


- Numbers & geometry (triangles, cones)
- Euclid: first come up with procedure
- Algorithm
 - step wise procedure =
 - list of instructions =
 - recipe



ca AD 100

For nerds only: Euclid's algorithm



- computes greatest common divisor
- $\text{gcd}(30, 18) = 6$

Idea / trick / magic

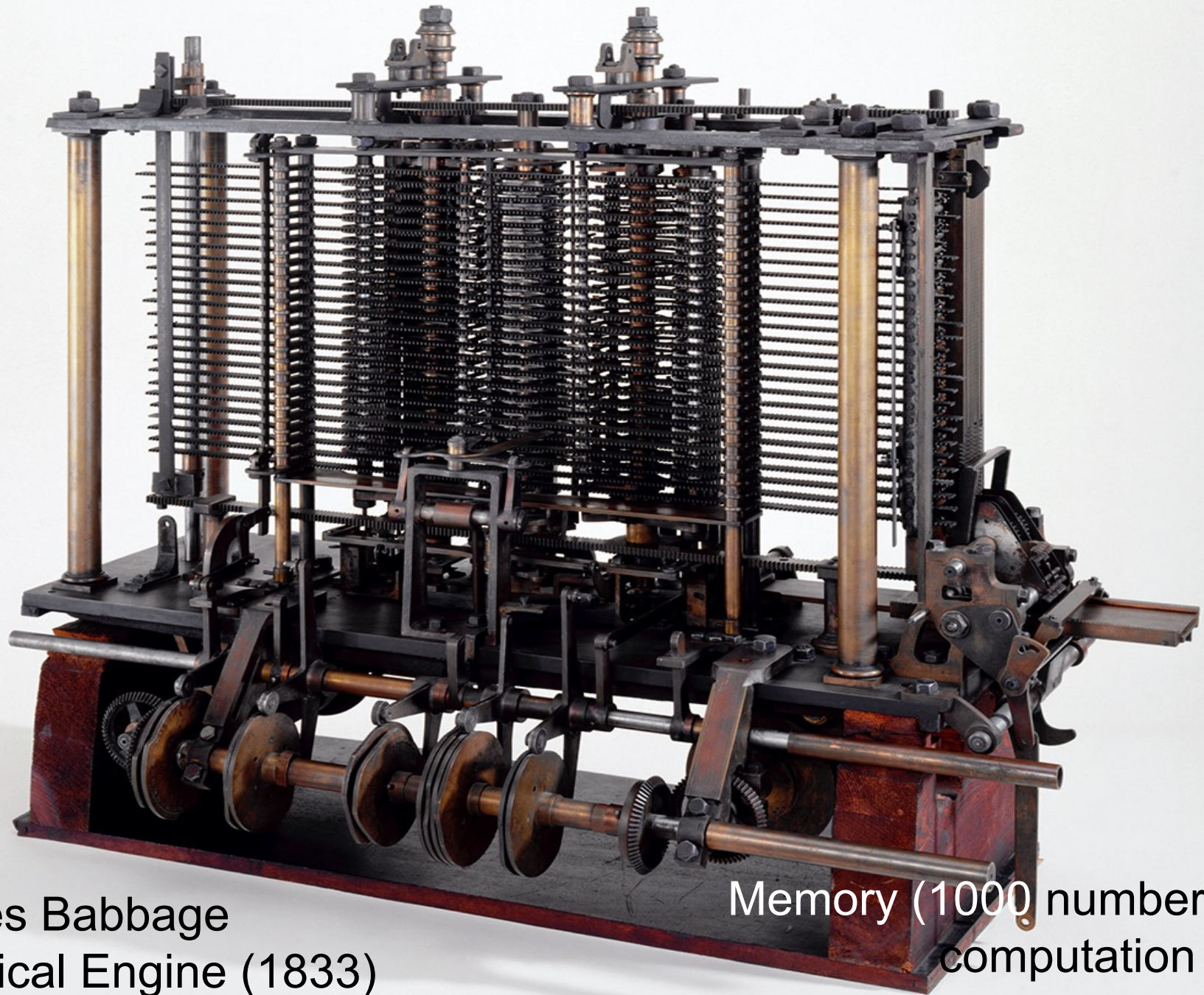
$$\begin{aligned}\text{gcd}(30, 18) &= \\ &= \text{gcd}(30 - 18, 18) = \text{gcd}(12, 18) \\ &= \text{gcd}(12, 18 - 12) = \text{gcd}(12, 6) \\ &= \text{gcd}(12 - 6, 6) = \text{gcd}(6, 6) = \\ &= \text{gcd}(6, 6 - 6) = \text{gcd}(6, 0) = 6\end{aligned}$$

Algorithms: no computer needed



The first algorithm for computers

1843



Charles Babbage
Analytical Engine (1833)

Memory (1000 numbers) +
computation unit

Ada Lovelace

- Lived 1815 –1852
- Daughter of Lord Byron

Analytical Engine

- Translated + extended description
- Invented calculation method for Bernoulli numbers
→ first computer algorithm

Vision

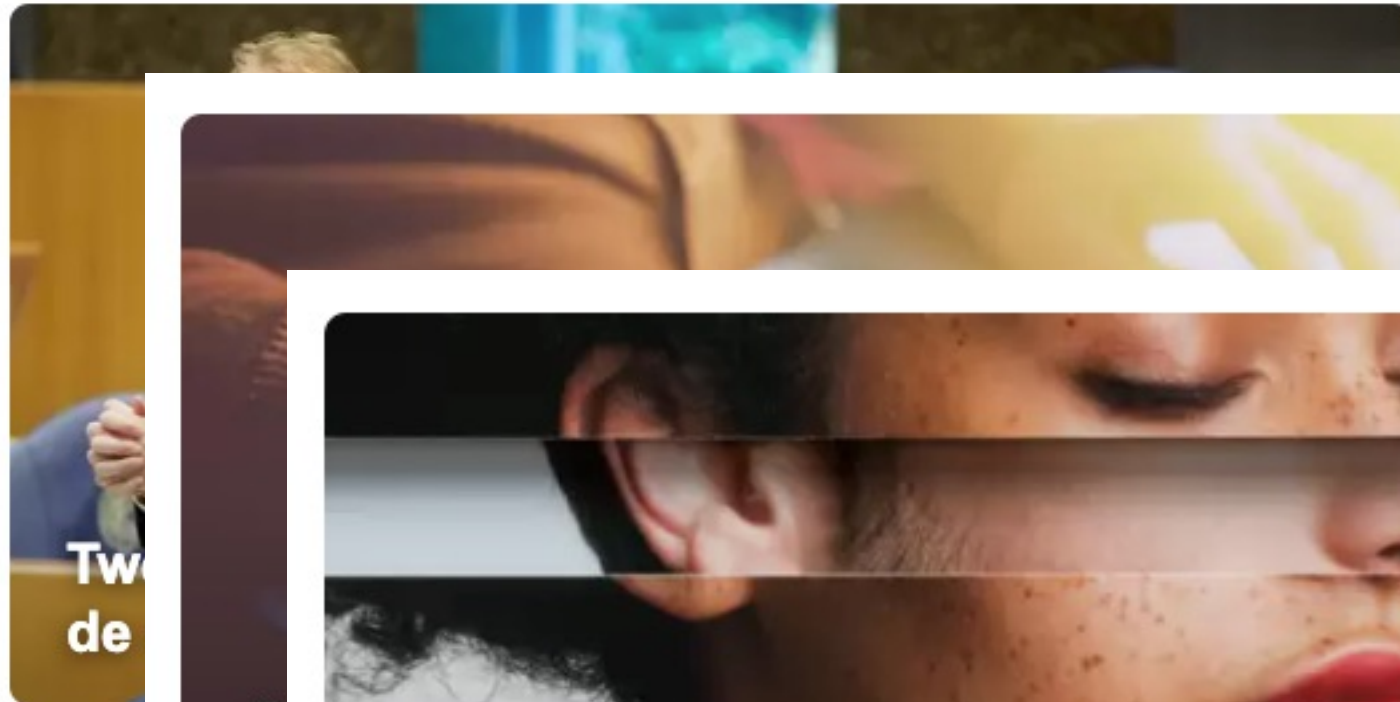
computers go beyond calculating + number crunching



From Ada's notes

Number of Operations Nature of Operations		Variables for Data						Working Variables									Variables for Results	
		1V_0	1V_1	1V_2	1V_3	1V_4	1V_5	0V_6	0V_7	0V_8	0V_9	${}^0V_{10}$	${}^0V_{11}$	${}^0V_{12}$	${}^0V_{13}$	${}^0V_{14}$	${}^0V_{15}$	${}^0V_{16}$
		+	+	+	+	+	+	+	+	+	+	+	+	+	+	+	+	+
		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
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		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
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		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
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		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
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		0	0	0	0	0	0	0	0	0	0	0	0	0				

Algorithms: **What about the fuss?**



Tw
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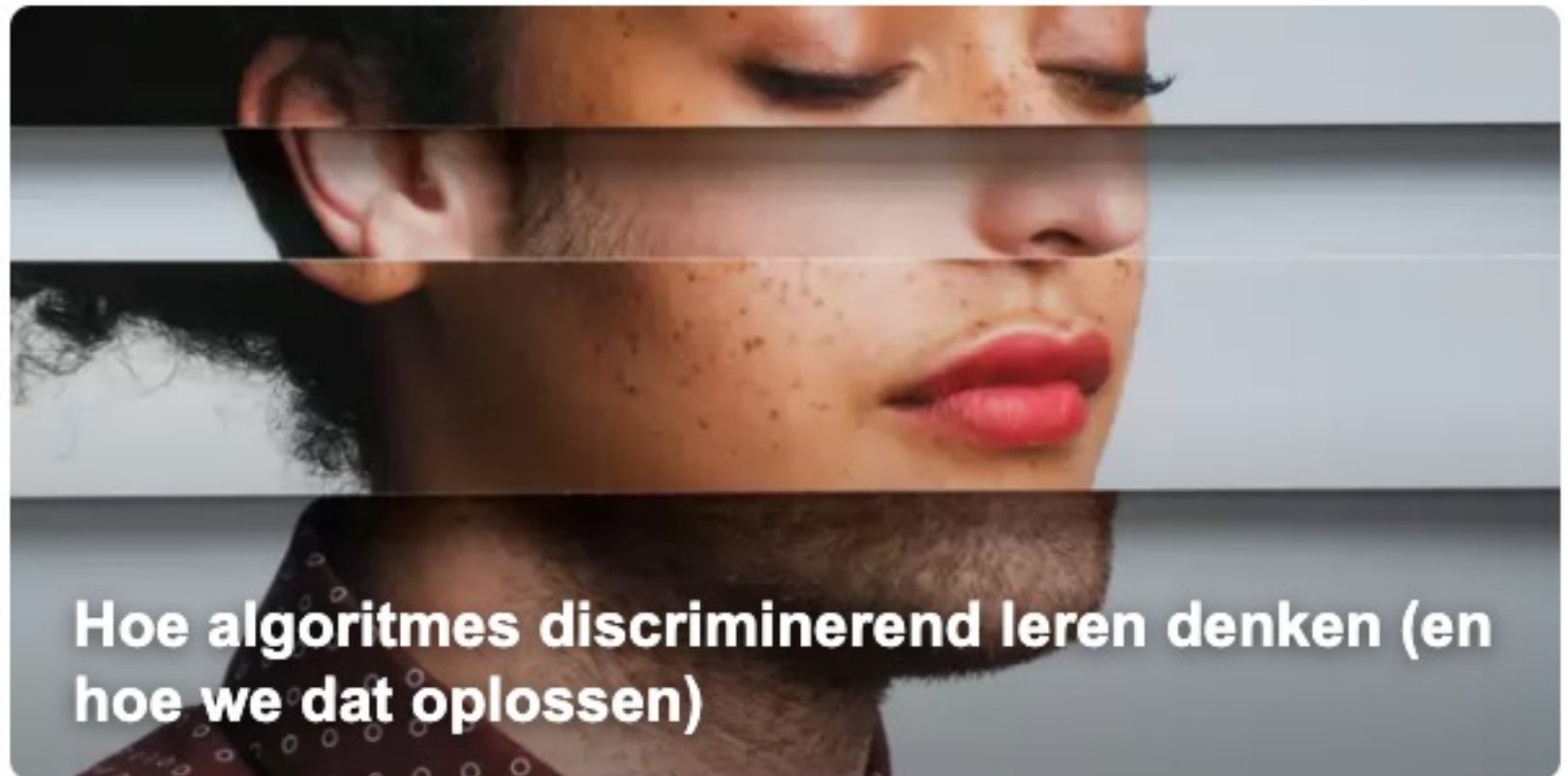
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Rotte



Hoe algoritmes discriminerend leren denken (en hoe we dat oplossen)

25 september 2020 16:38

Laatste update: 26 september 2020 17:15



247 NUjjj-reacties



The first self-learning algorithms



AI algorithms

- Not programmed by humans
- But: Self-learning
 - Learn automatically from data / examples
 - Eg: cats vs dogs

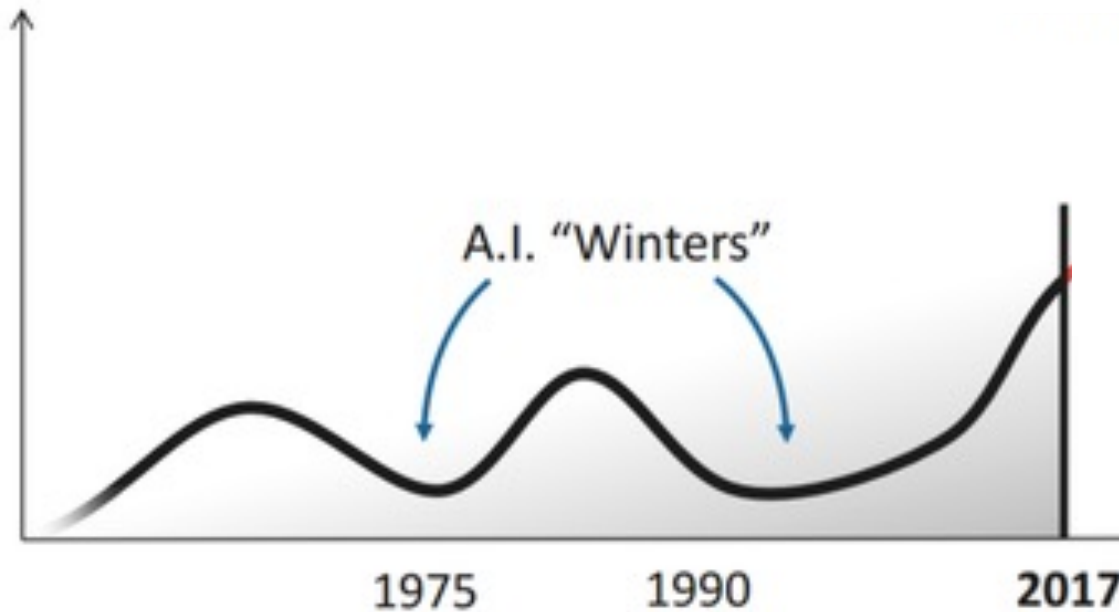
Buzz words

- Artificial Neural Networks (“deep learning”)
- Bayesian Networks
- Genetic algorithms
-
- 1956 dardmouth workshop
- The problem: *they did not work!*

1956

AI winters

Buzz, impact & pervasiveness



Winters

- AI solutions failed
- No investments

Why these winters?

- Problems are complex
- Computer power + data limited



AI Winter
1969 - 1990

The first **working AI** algorithms

1997

- DeepBlue beats Kasparov
- Classical algorithms

2013

- AlphaGO defeat2 Lee Sedol
- Neural networks



Why did it suddenly work?

Increased computer power

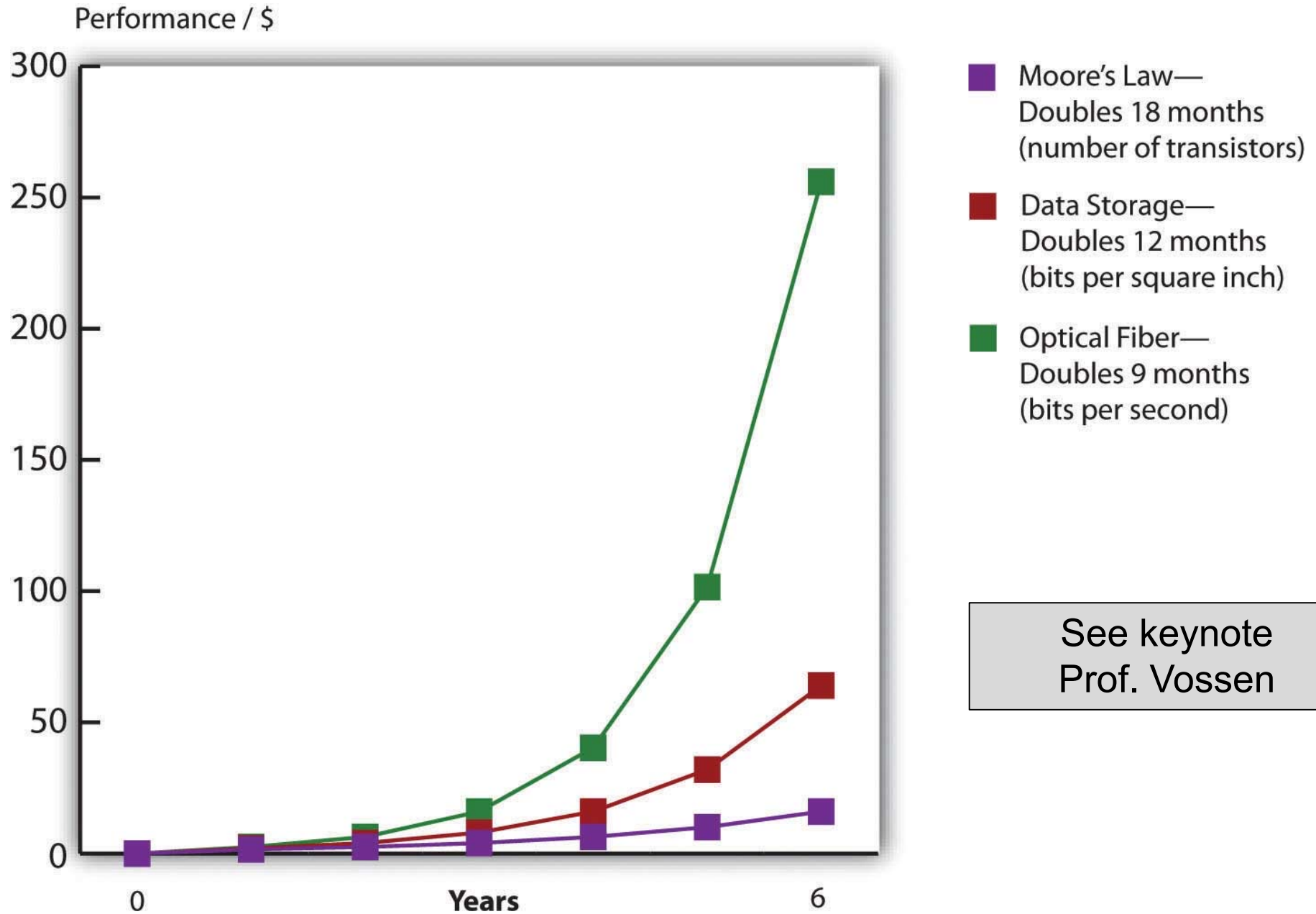
- Moore's law

Data storage cheaper

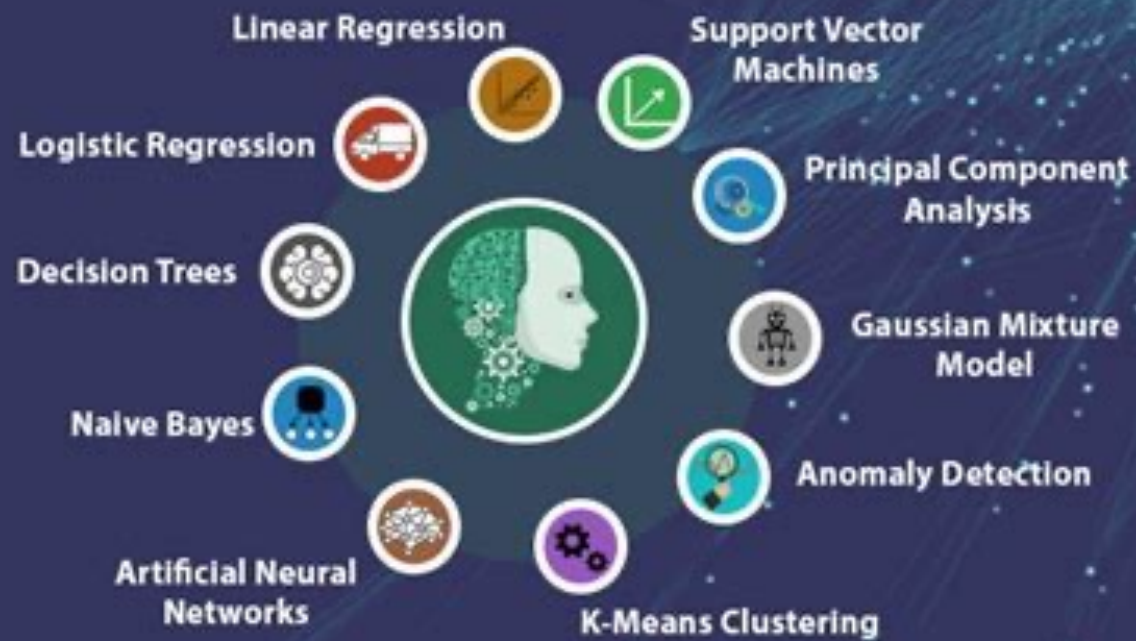
- More data available



Exponential growth: computing power & data storage



Top Machine Learning Algorithms



Summary

300 BC

- First classical algorithm: *stepwise procedure*

1843

- First computer algorithm: *on analytical machine*

1956

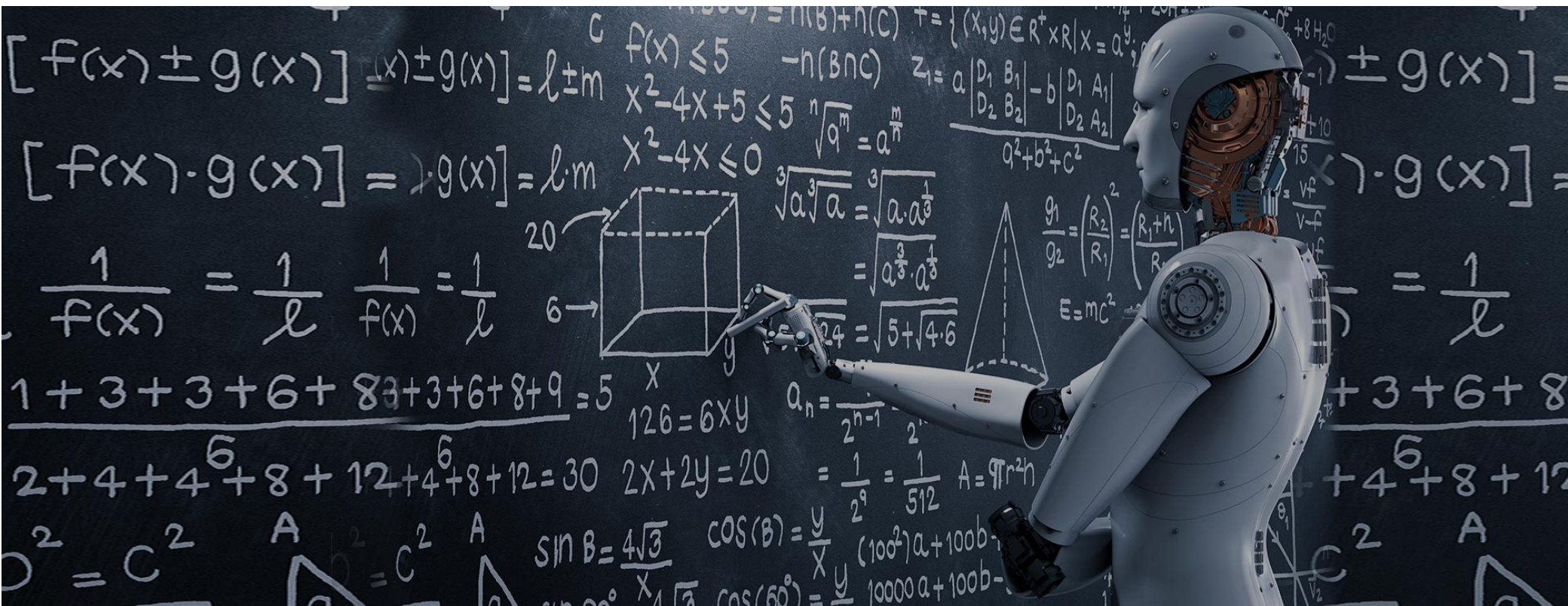
- First AI algorithms: *self-learning*

1997

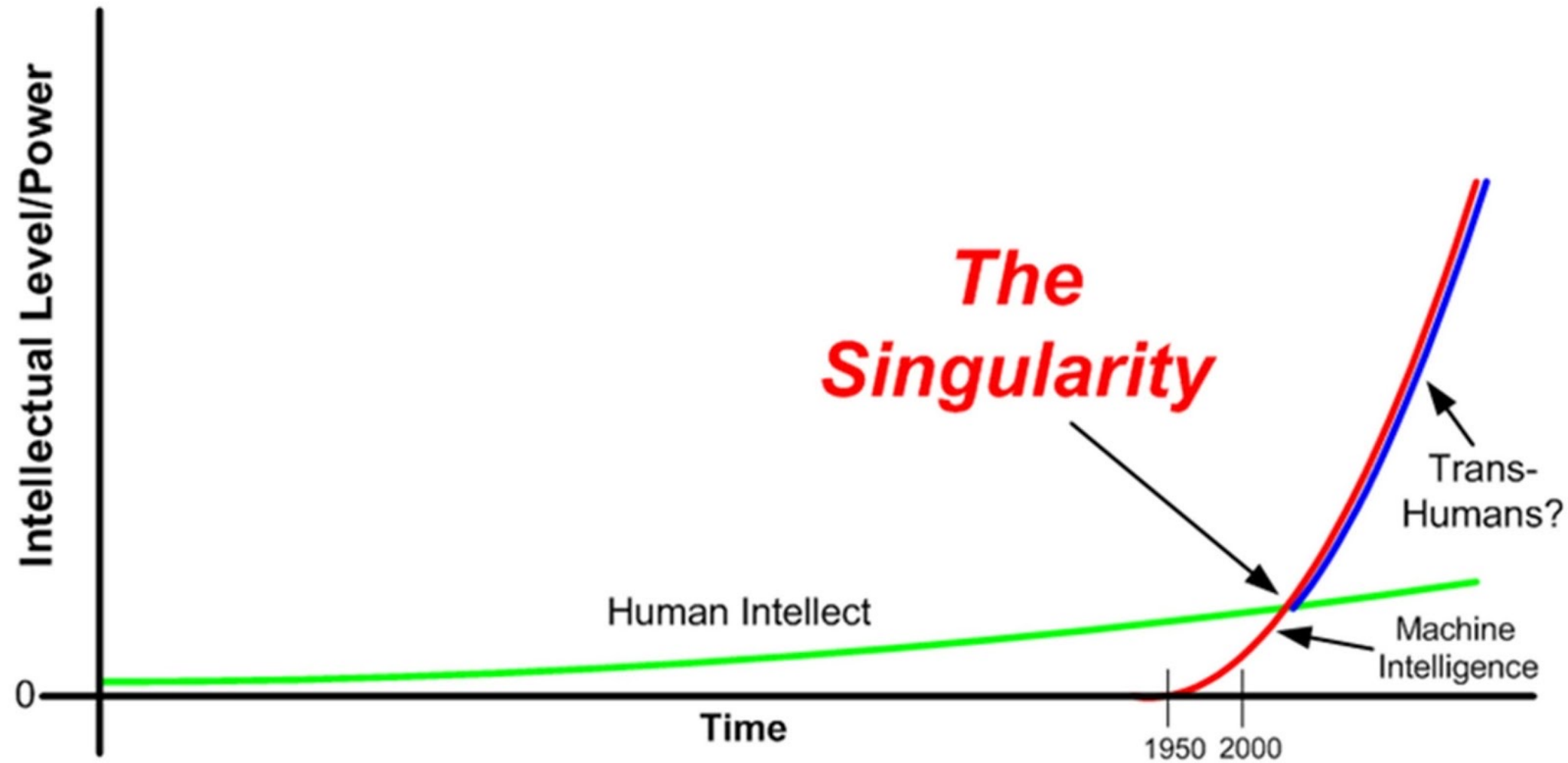
- First AI algorithms: *now effective*

If you hear the word “algorithm”, is it human-written or self-learning?

Where will we end up?



Where will we end?



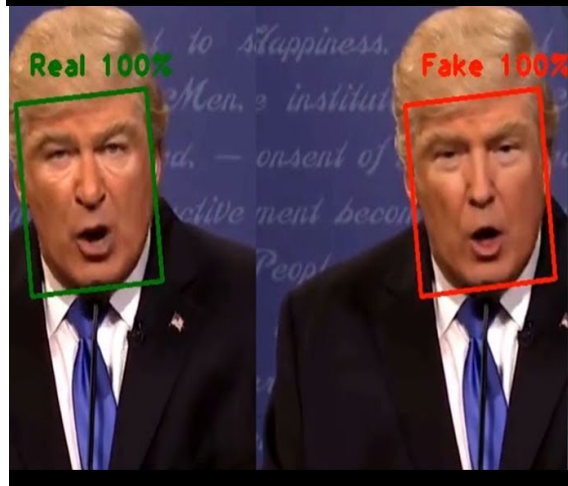
How does it all work?

Applications

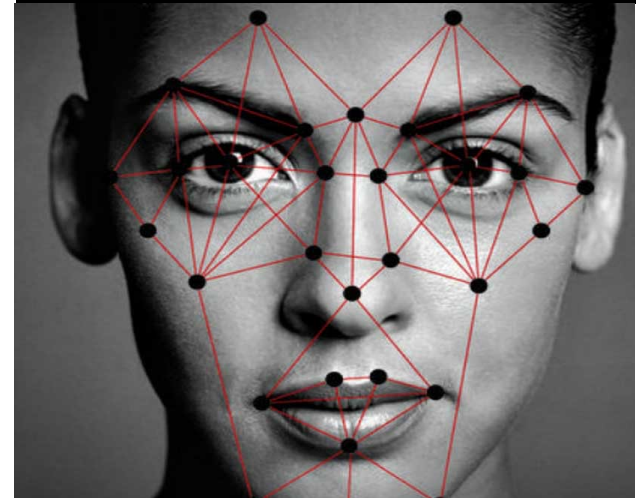
Self-driving cars



Deep fake



Face recognition



Sales forecasting



profiling



Applications in Risk Analysis

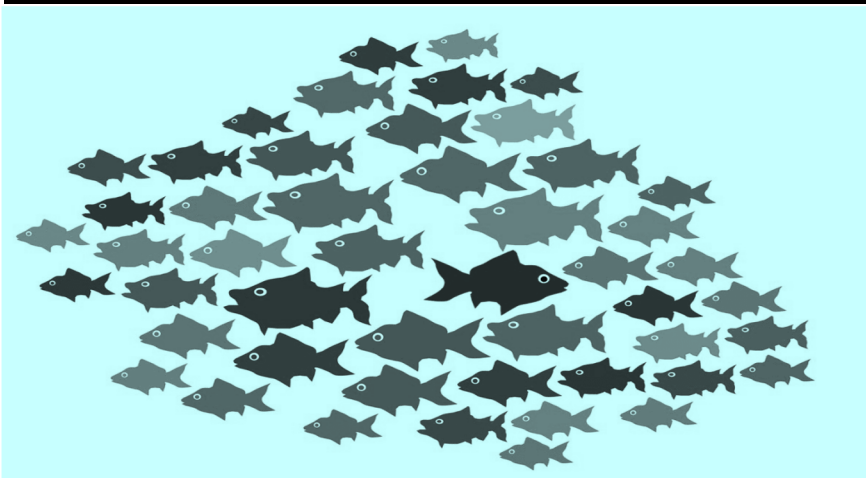
Spam filtering



“Spoorlopers”



Anomaly detection



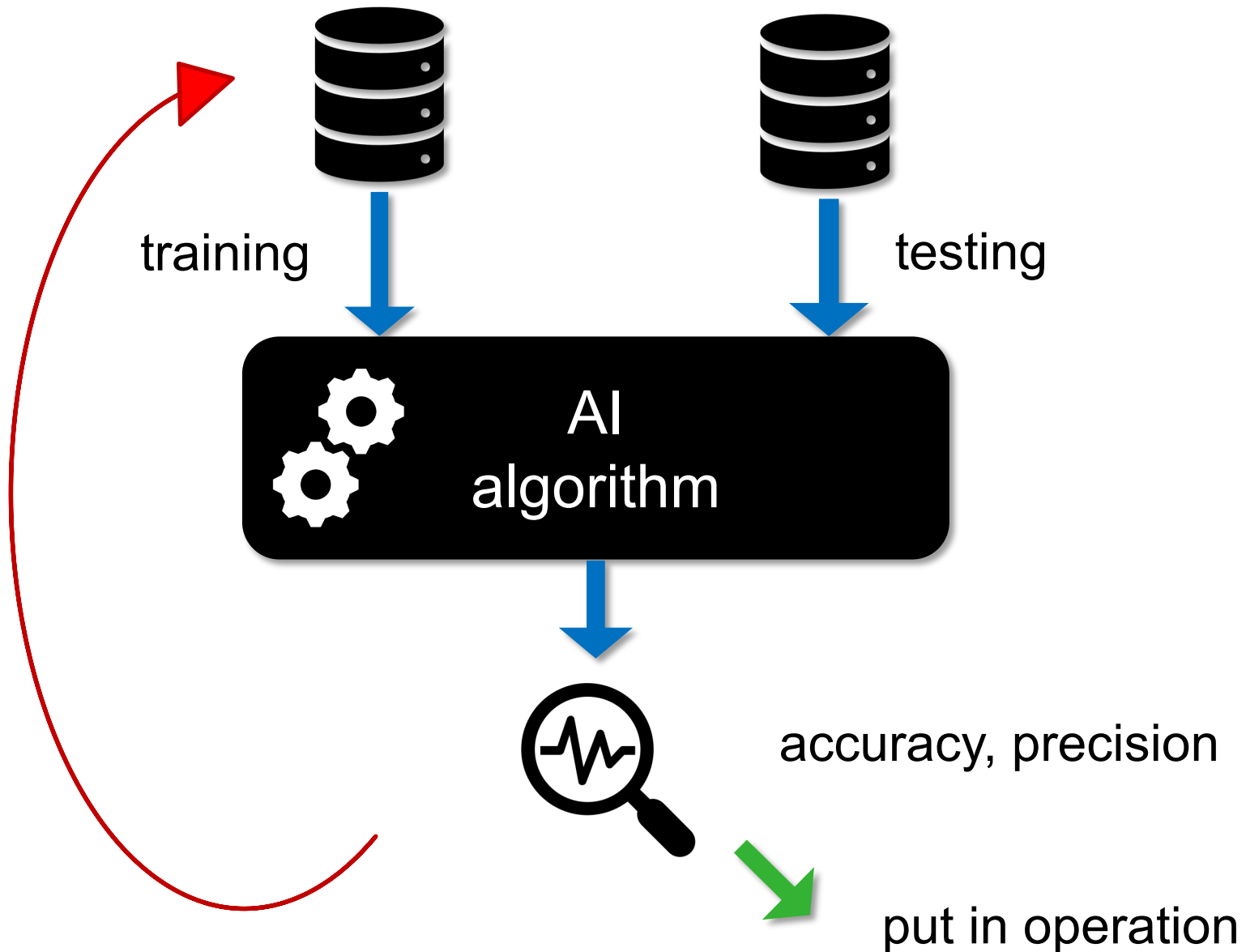
Predictive maintenance



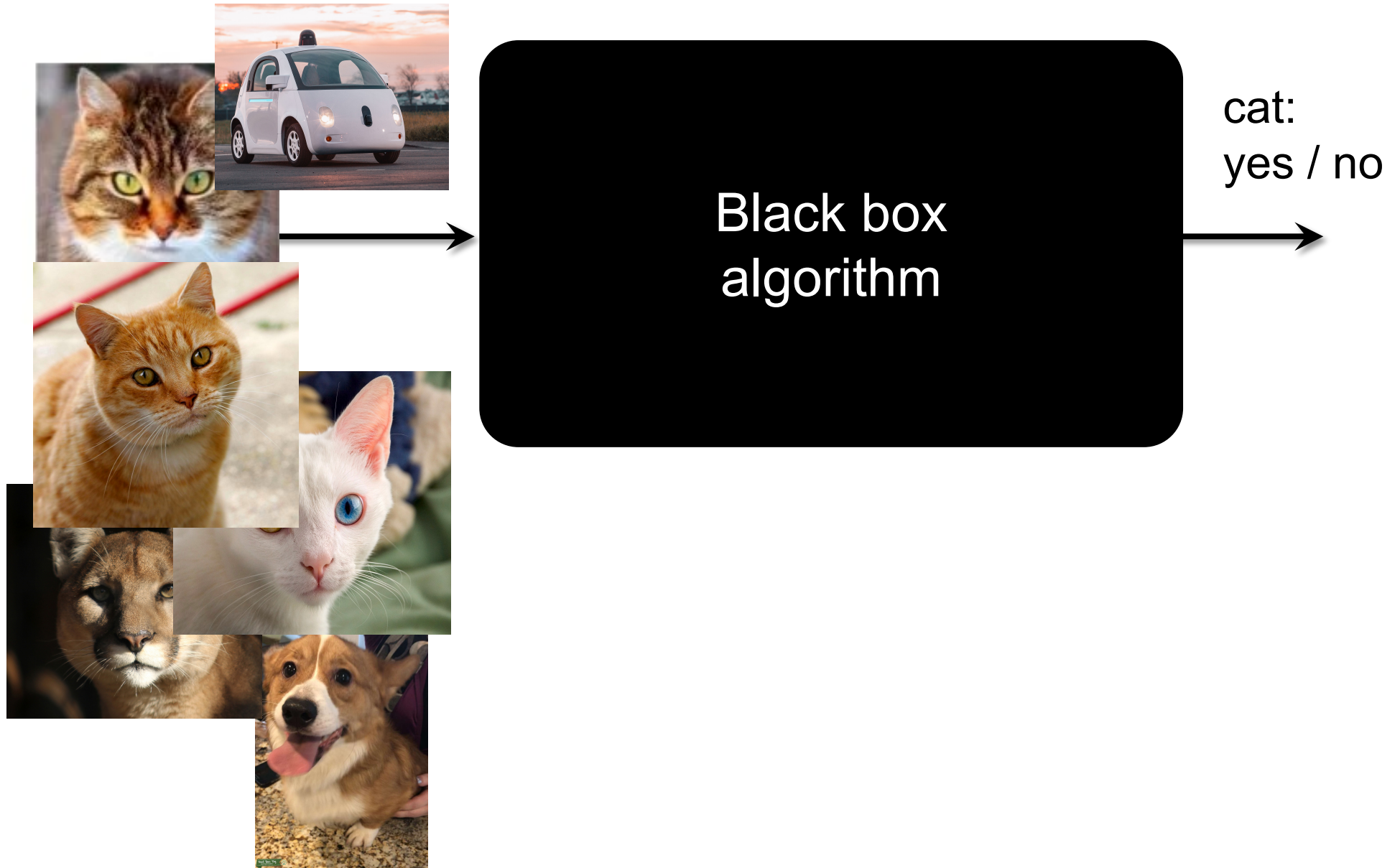
Top Machine Learning Algorithms



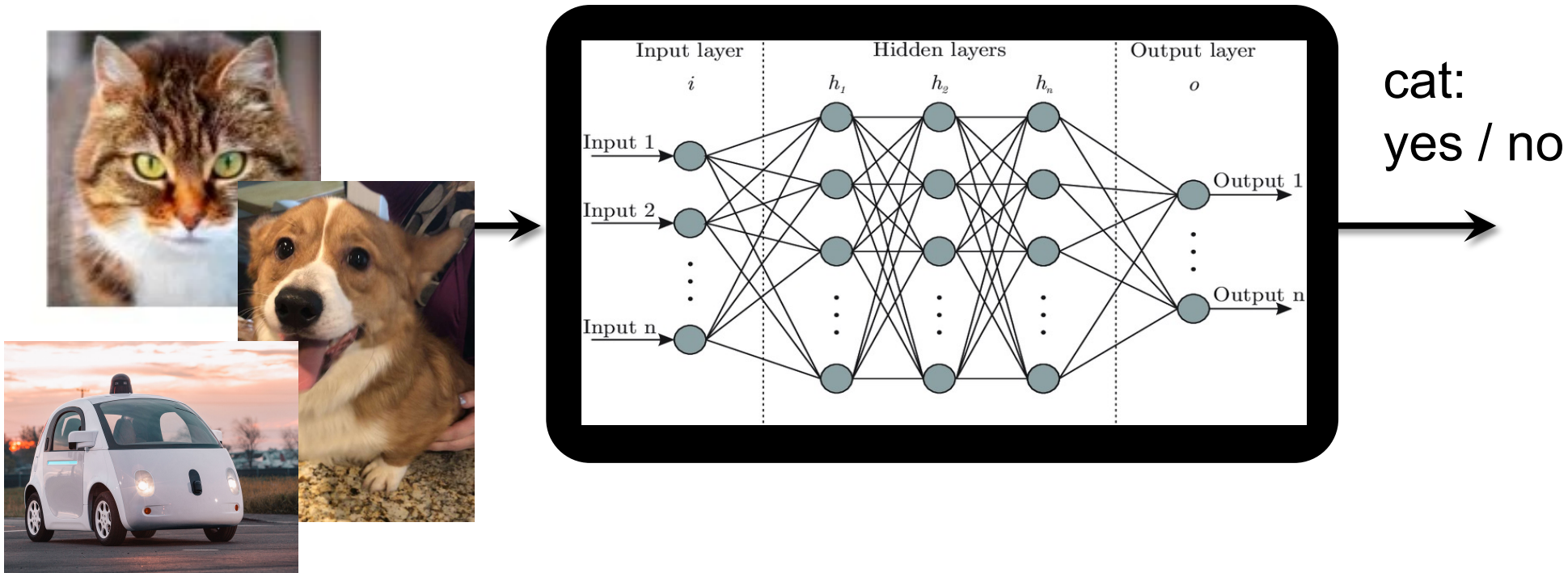
How it works: Self-learning algorithms



example: neural networks



neural networks: inside

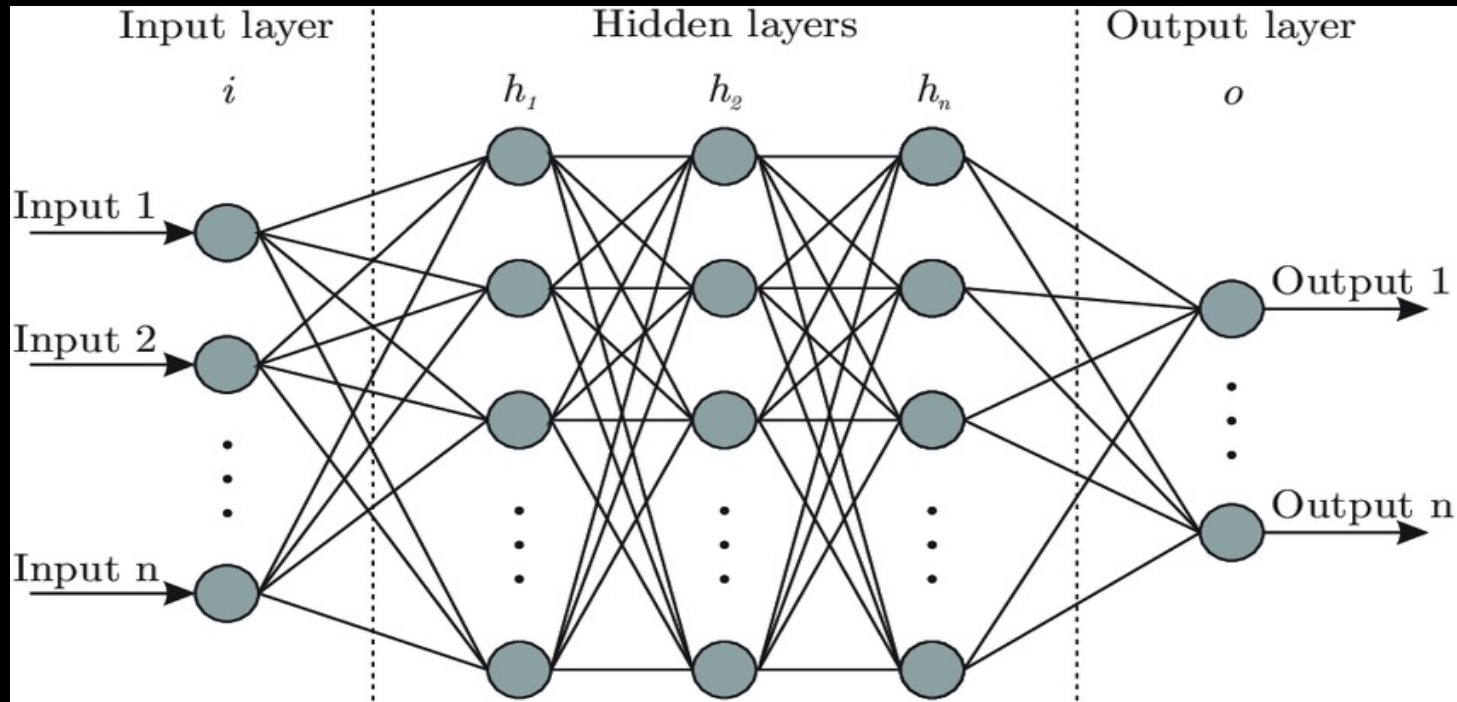


	algo: cat	algo: no cat
picture: cat	true positive	false negative
picture: no cat	false positive	true negative

Accuracy = % correctly classified pictures



neural networks: inside



inputs

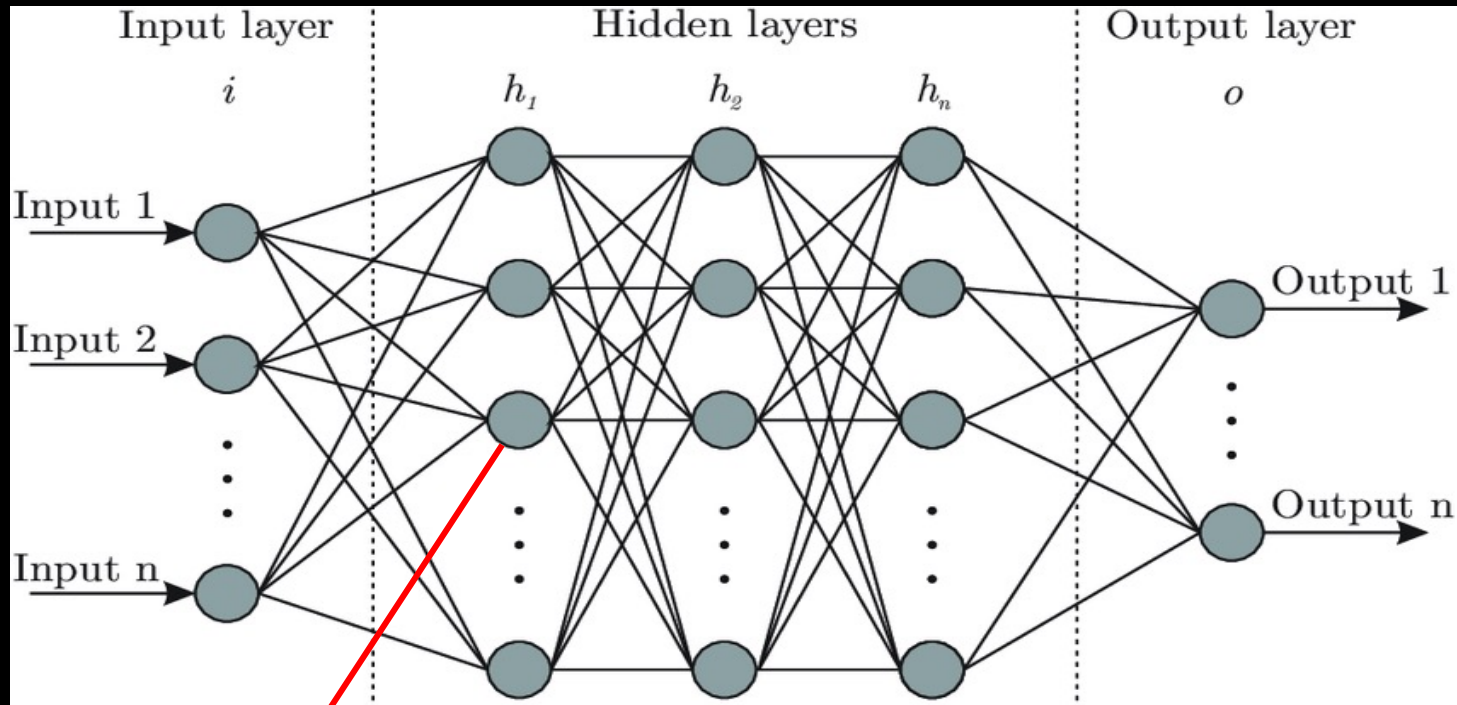
- (cat) picture: 64 x 64 pixels
- grey scale in $[0,1]$: 0=white, 1= black, 0.8 rather black

Output

- binary: 1 (yes, cat) or 0 (no cat)



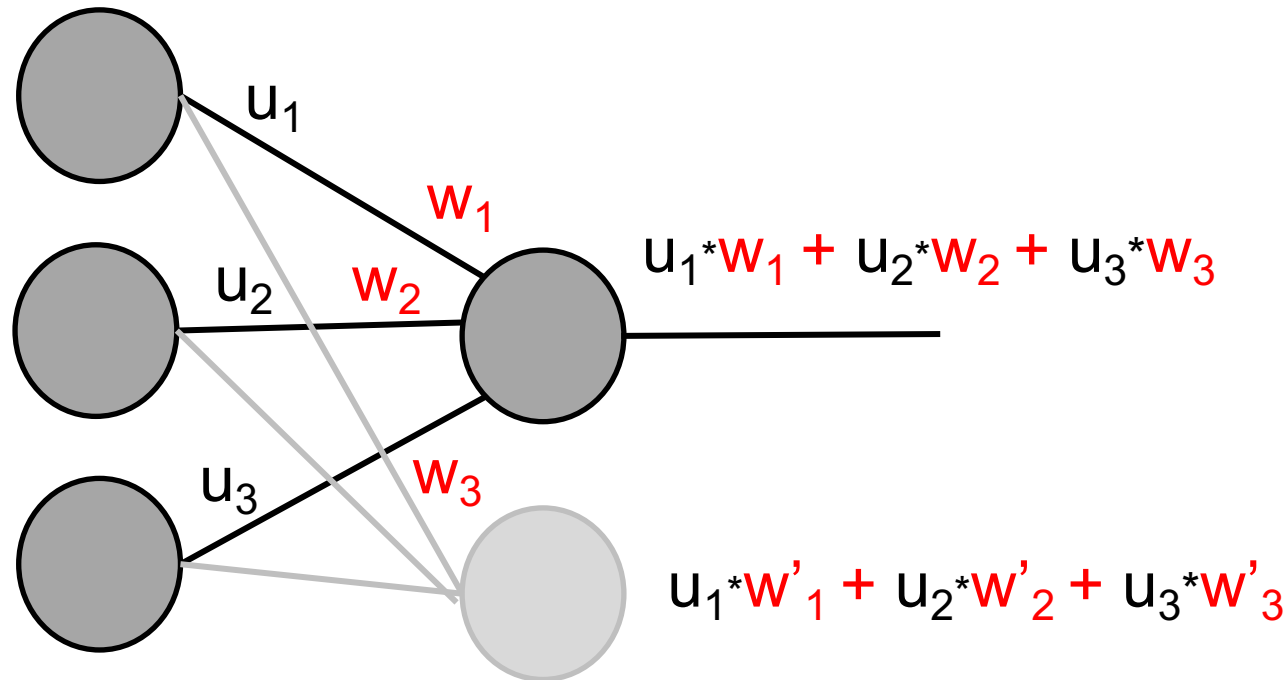
neural networks: inside



firing neurons

- a neuro fires iff their weighted sum of its inputs $> T$
- training = finding the weights
- many variants: recurrent, convolutional, long short term memory

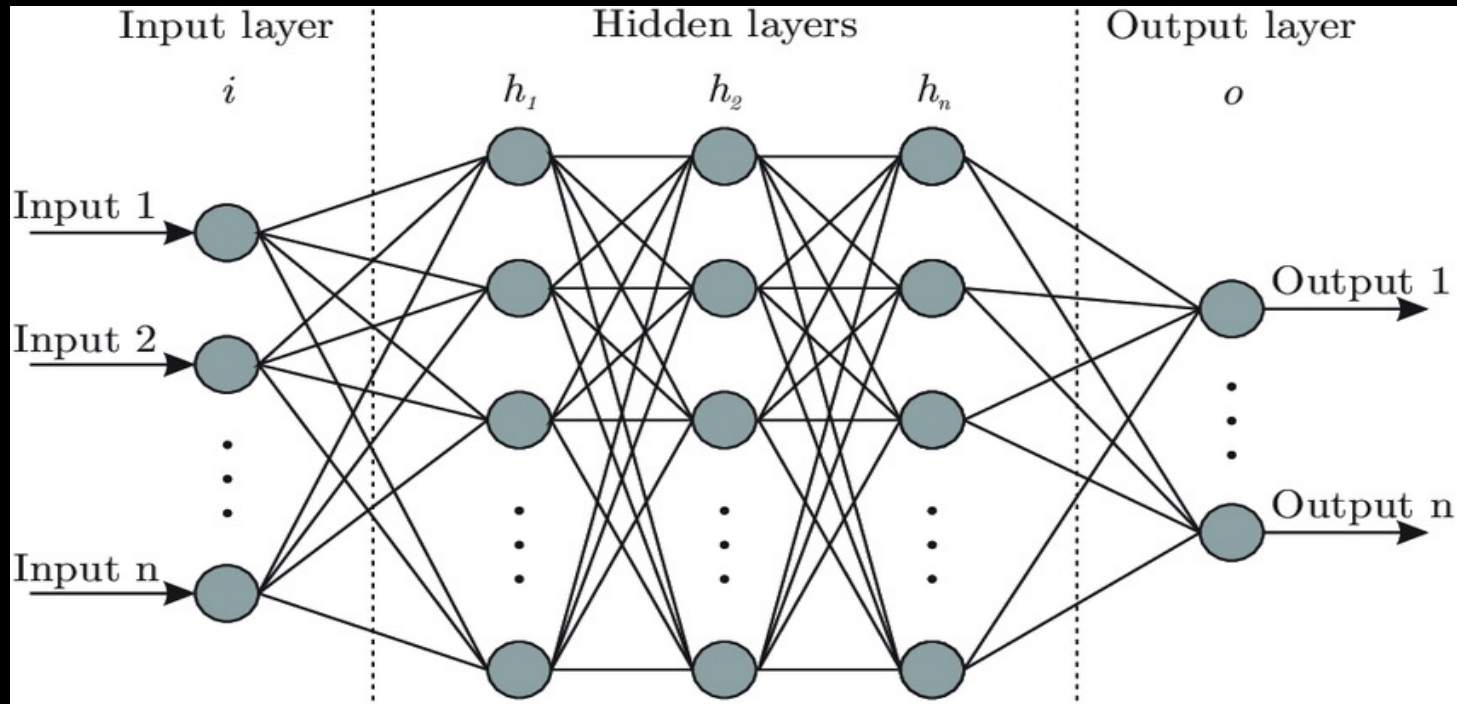
neural networks: inside



- a neuro fires iff their weighted sum of $> T$
- (renormalize to get numbers in $[0,1]$)
- Learning = finding good weights



neural networks: black box



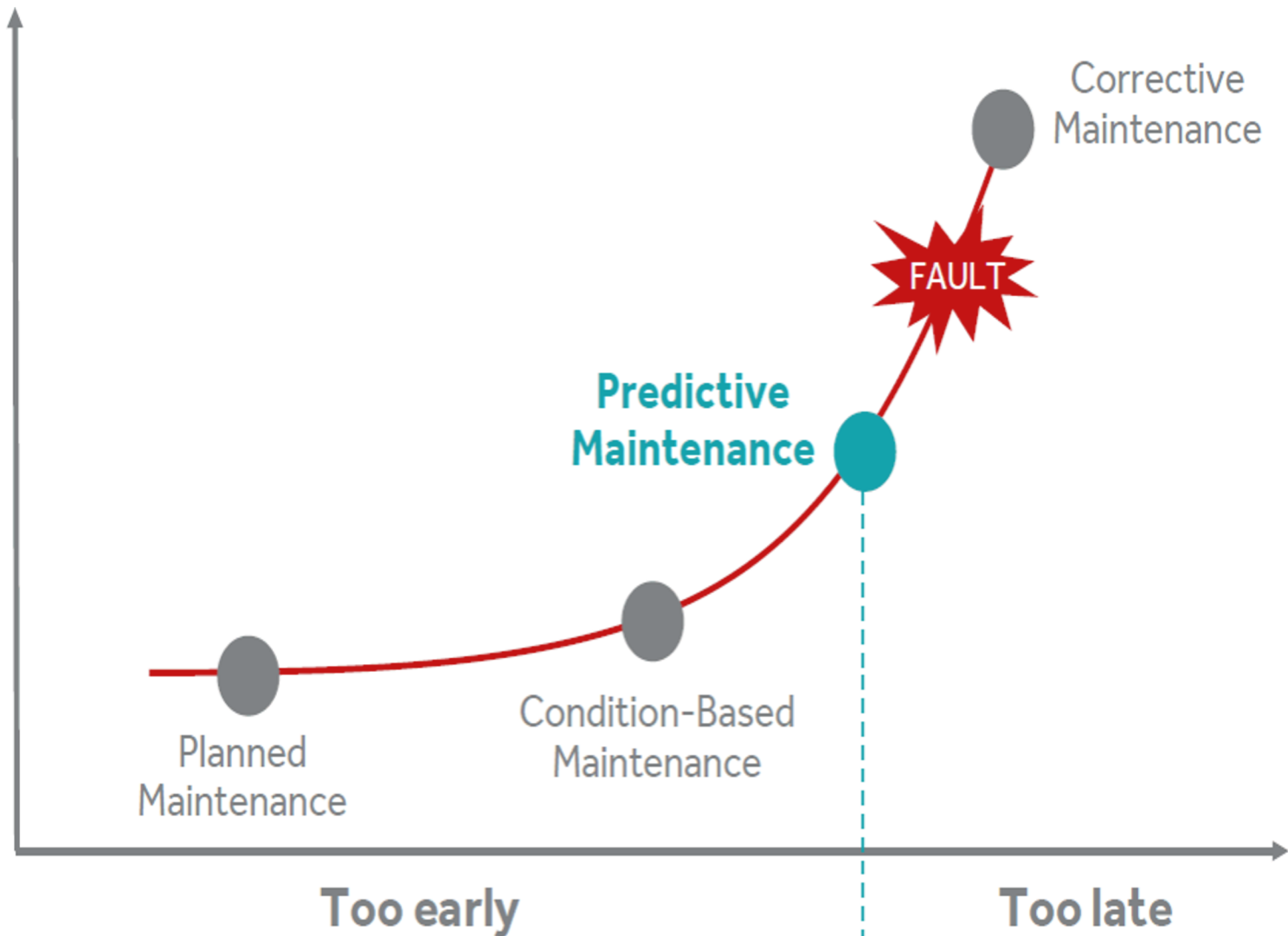
Problem: not understandable

- GIGO: garbage in, garbage out
- No patterns in NN (eg eyes, ears)
- Acceptable? Responsible? Liable?

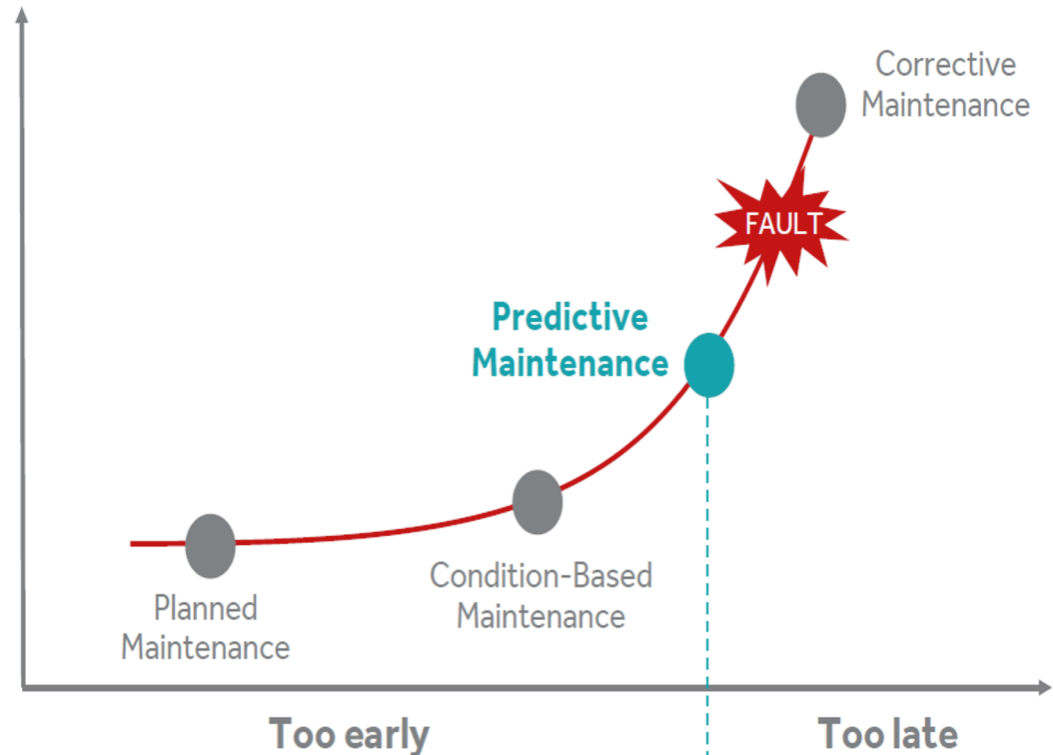
PART 2: Predictive maintenance



Predictive maintenance



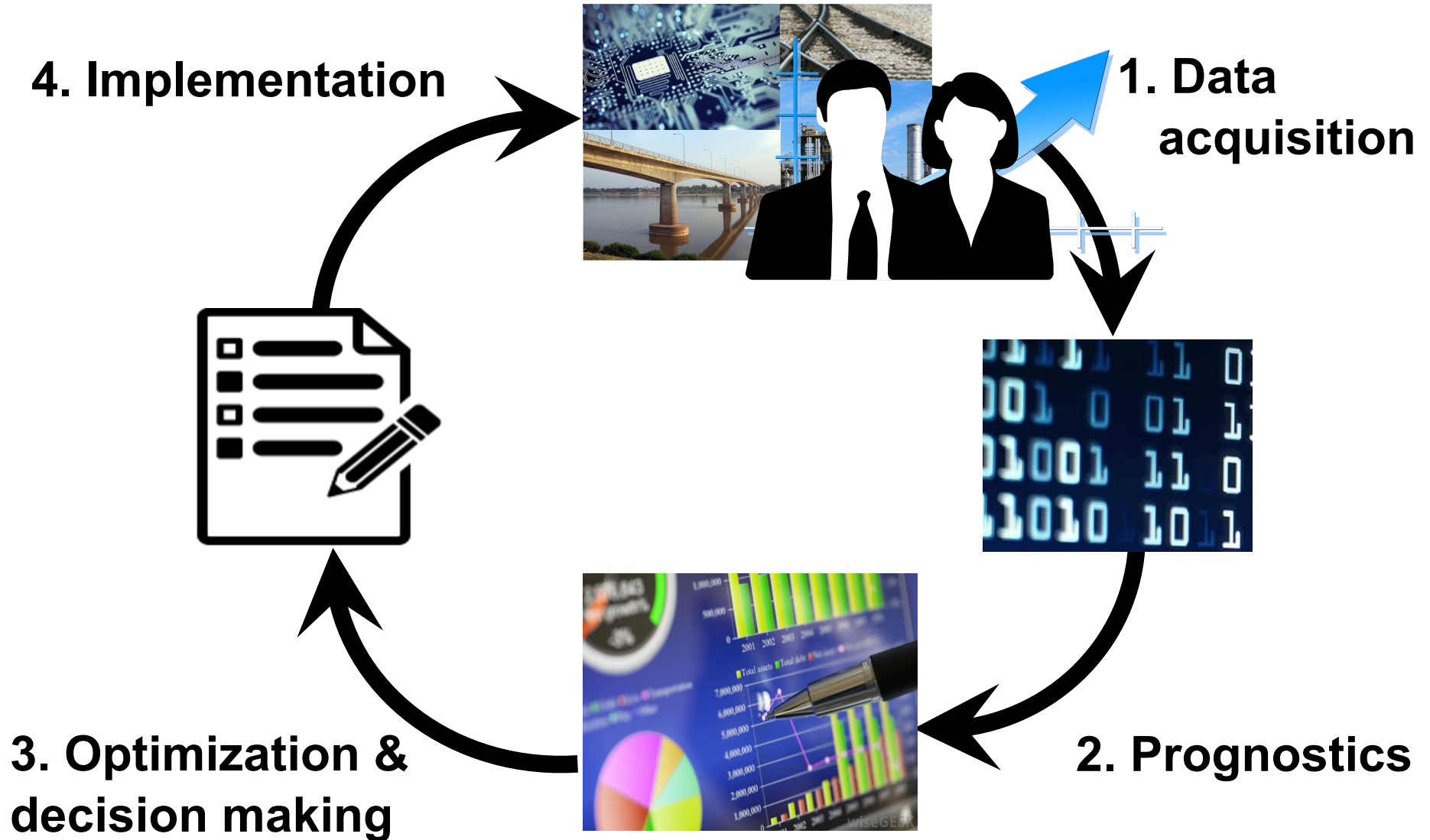
Predictive maintenance



McKenzie

- PdM is most promising IoT application
- 63 billion / year cost savings
- 50% downtime reductions

workflow: predictive maintenance



Discussion

I am curious to hear

- Which data could help your organization?
- What would you like to predict?

