

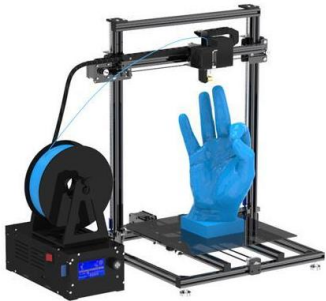
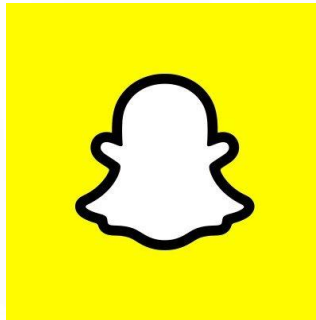
# Artificial Intelligence: Good or Evil?

Philippe Herve  
*VP of Oil and Gas Solutions*

**“Most people overestimate what they can do in one year and underestimate what they can do in ten years.” –Bill Gates**



Ten years ago, none of these things existed in the mainstream:

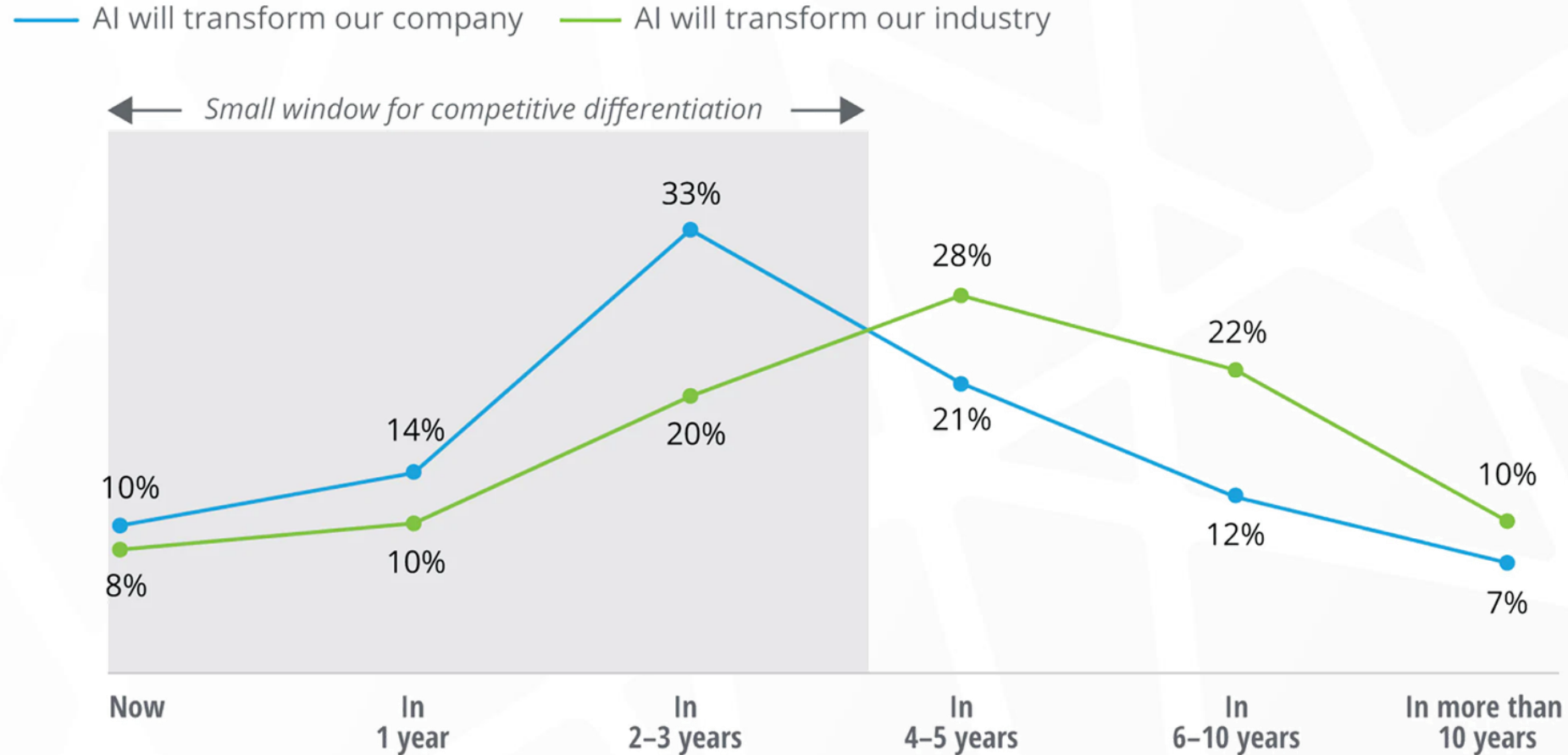


**Artificial  
intelligence**



**Technological disruption is coming faster than you think.**

# The Window for AI Differentiation is Closing



Source: Deloitte analysis based on Deloitte's "State of AI Enterprise, 2nd Edition" survey of 1,900 AI early adopters in seven countries. Note: Percentages may not add up to 100 due to a small number of respondents who answered, "Don't know." Copyright © 2019 Deloitte Development LLC. All rights reserved.

©SparkCognition, Inc. 2020. All rights reserved.



# Basics of Artificial Intelligence

# The History of Modern AI



## AI 1.0



**1950**

In 1950, Alan Turing publishes "Computing Machinery and Intelligence," a paper from which the Turing Test emerged.



**1956**

John McCarthy coins the term "Artificial Intelligence," defining it as "the science and engineering of making intelligent machines."

## AI 2.0



**1980s**

Rebirth of AI due to "expert systems," which are programs that answer questions or solve problems using logical rules.

## AI 3.0



**2000s**

Deep learning, a family of machine learning methods based on learning data representations, breaks through.

# Why is **AI 3.0** working?



Computing  
Power



Abundance of Data

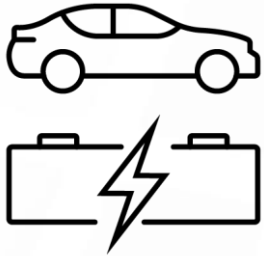


Algorithm & Research  
Breakthroughs





# Success of AI 3.0



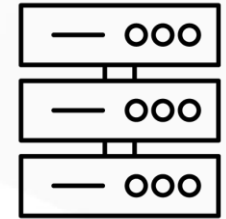
TESLA



IBM Watson

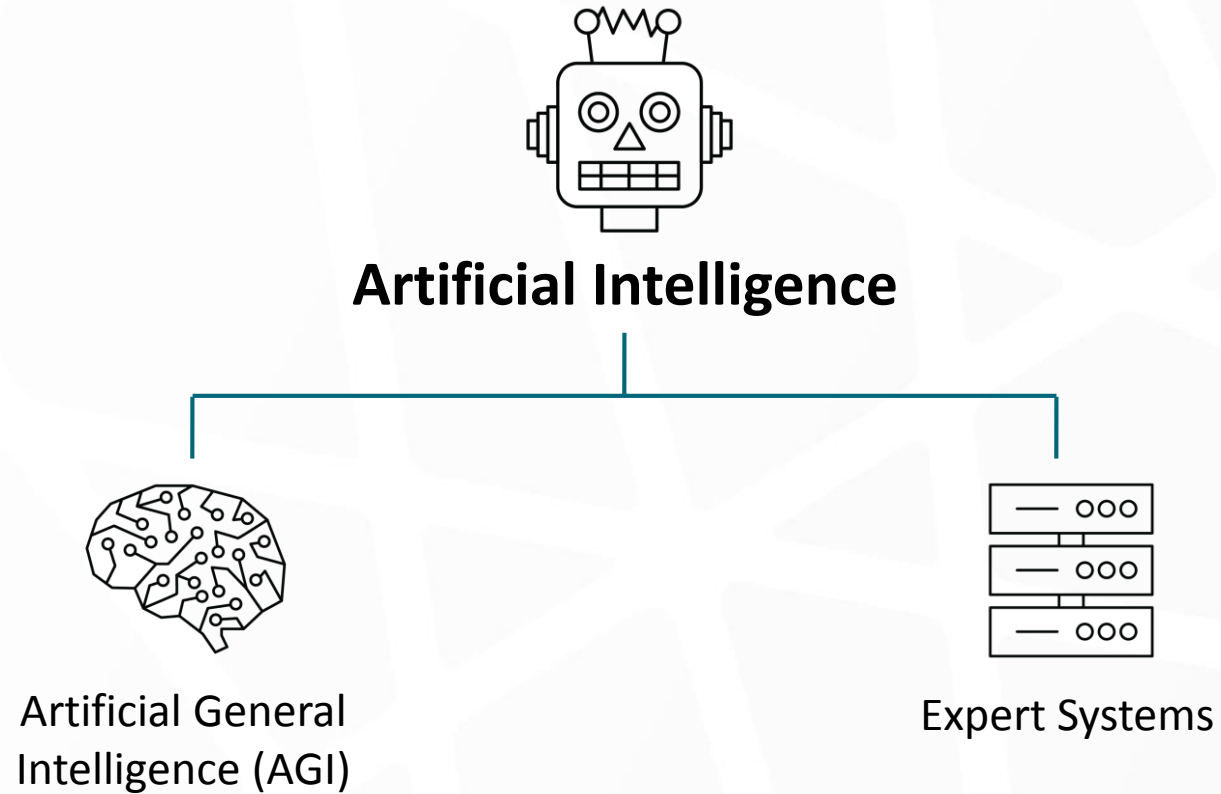


 sparkcognition™



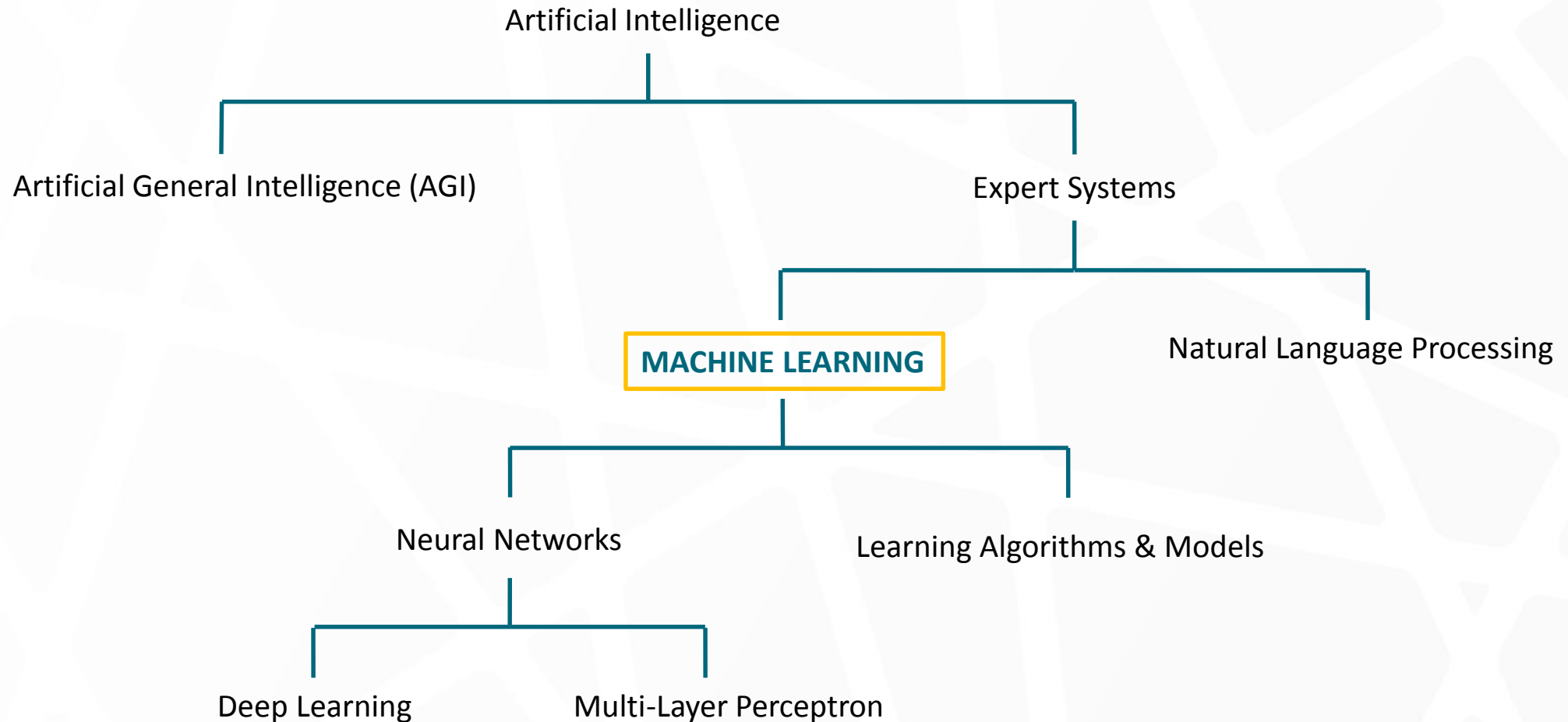
Google

# Artificial Intelligence





# AI vs. Machine Learning vs. Deep Learning



# How Would You Write the Code

...to tell the difference between a banana, an apple, and grapes?



**Bananas are Yellow**



**Apples are Red**



**Grapes are Green**



# Feed in Measurable Characteristics to an Algorithm



Height



**BANANA**



Width



**APPLE**



Color

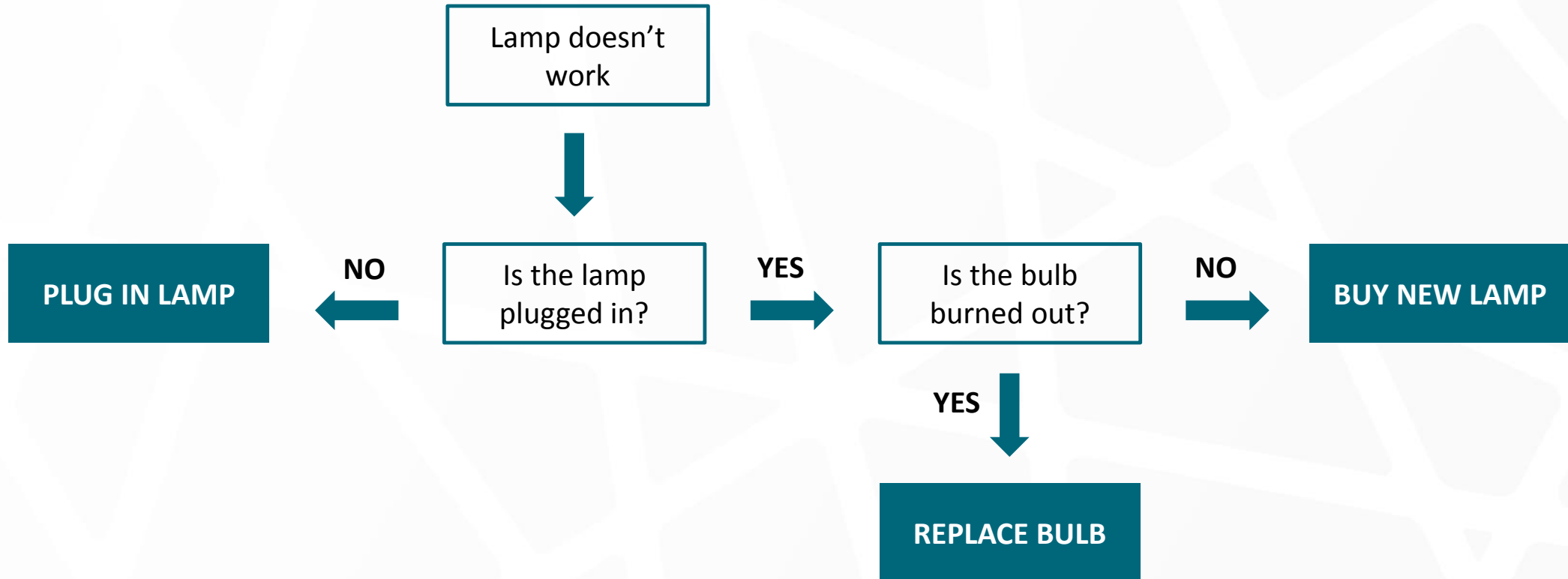


**GRAPES**

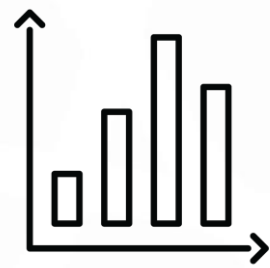
Shape

*Let the algorithm define the relationships between the measurable characteristics and the fruit they embody*

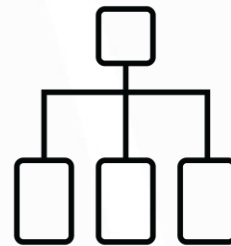
# What Is an Algorithm?



# What Is a Model?



**INPUT DATA**



**ALGORITHM**



**MODEL**

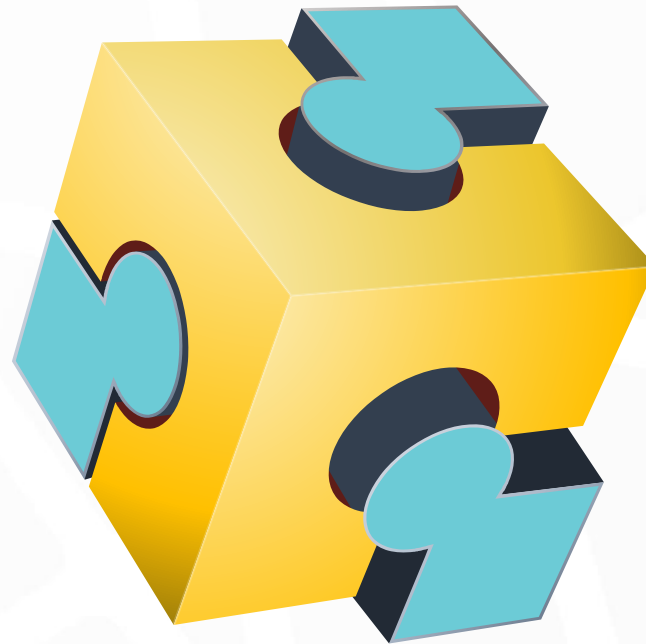
# Models Automate the Enterprise



What is a model? *A representation of a complex process or system*

**Identification of  
important trends  
and patterns**

**Ability to improve  
continuously over  
time**



**Ability to process  
big data and non-  
linearity**

**Ability to provide  
consistent predictive  
accuracy**

# Supervised vs. Unsupervised Machine Learning



## Unsupervised

Solving a given problem without a known desired outcome

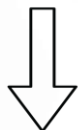
### Unsupervised Machine Learning Practices:



Clustering



Anomaly Detection

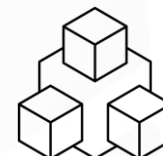


Dimensionality Reduction

## Supervised

Solving a given problem with a known outcome

### Supervised Machine Learning Practices:



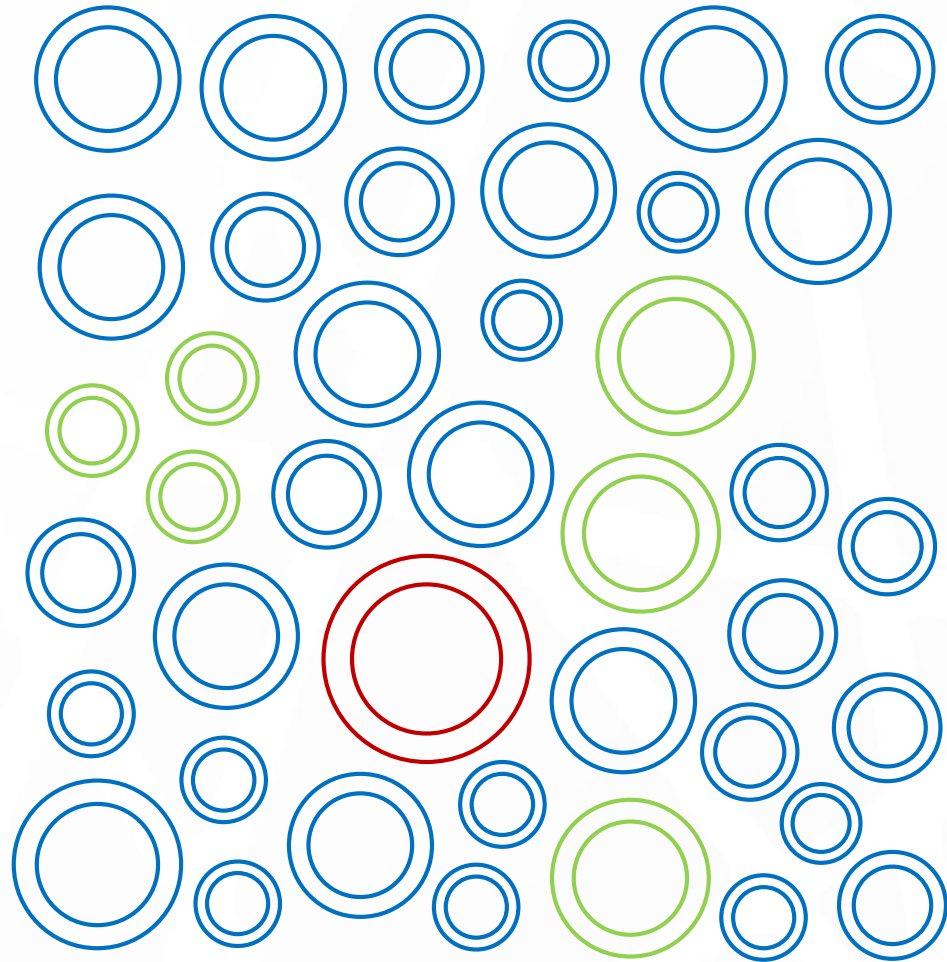
Classification



Regression



# Unsupervised Learning



Small



Medium



Large

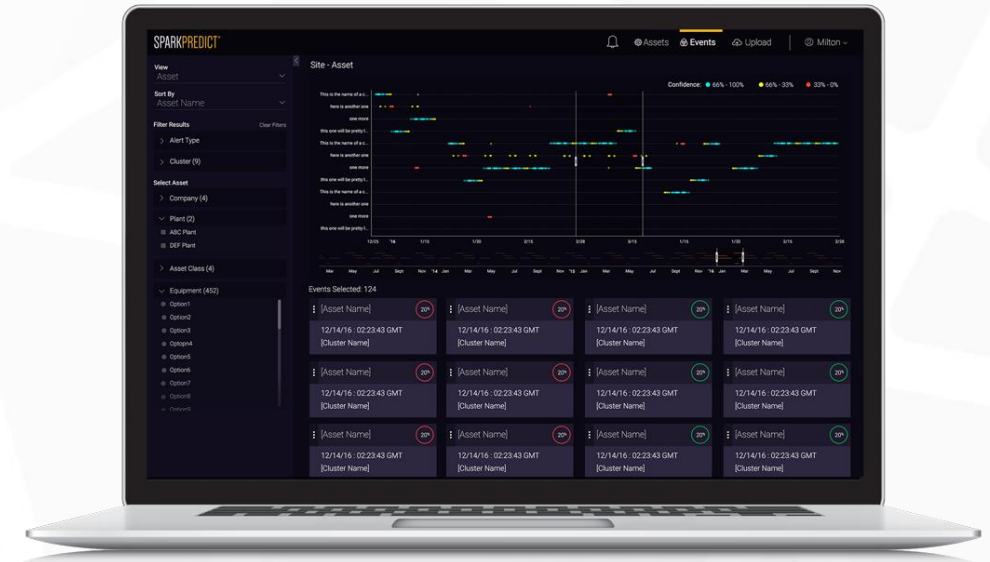
# Case study: Predicting turbine failures with machine learning



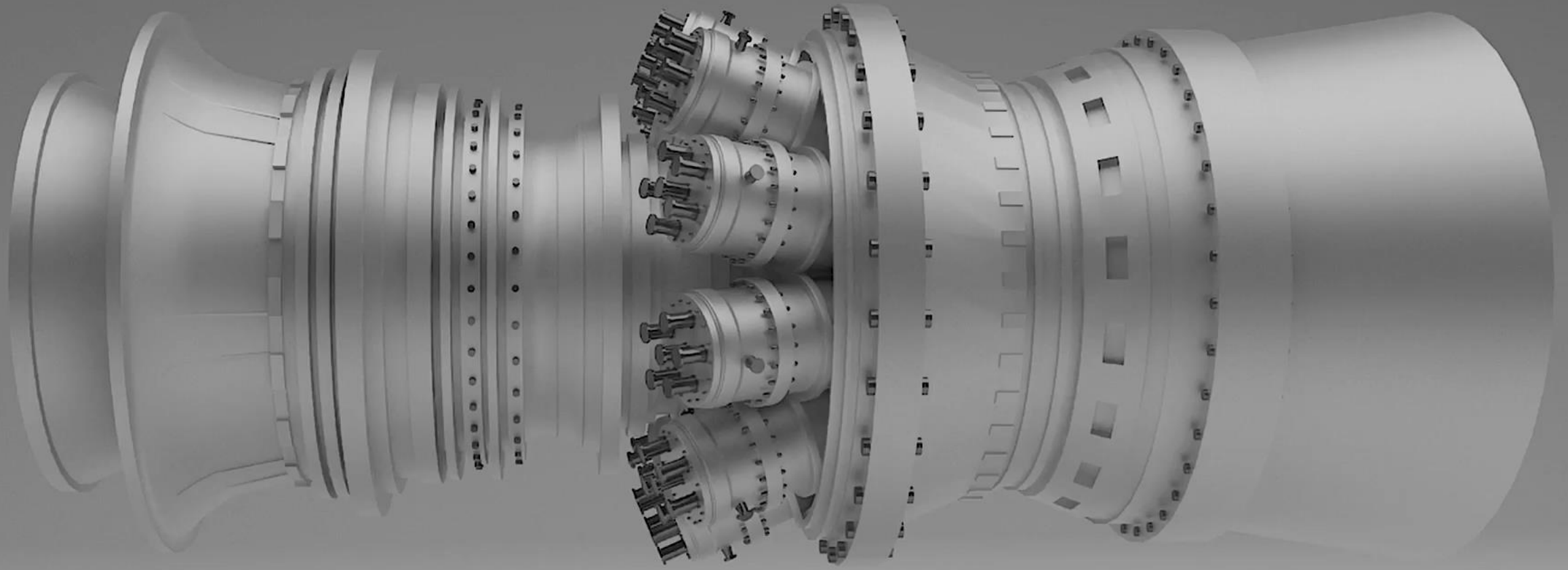
Why did this utility invest in AI and machine learning?

- Managing risk of the unknown
- Desire for new capabilities
- Desire to take productivity to new levels
- Limited SME experience for the utility and the OEM

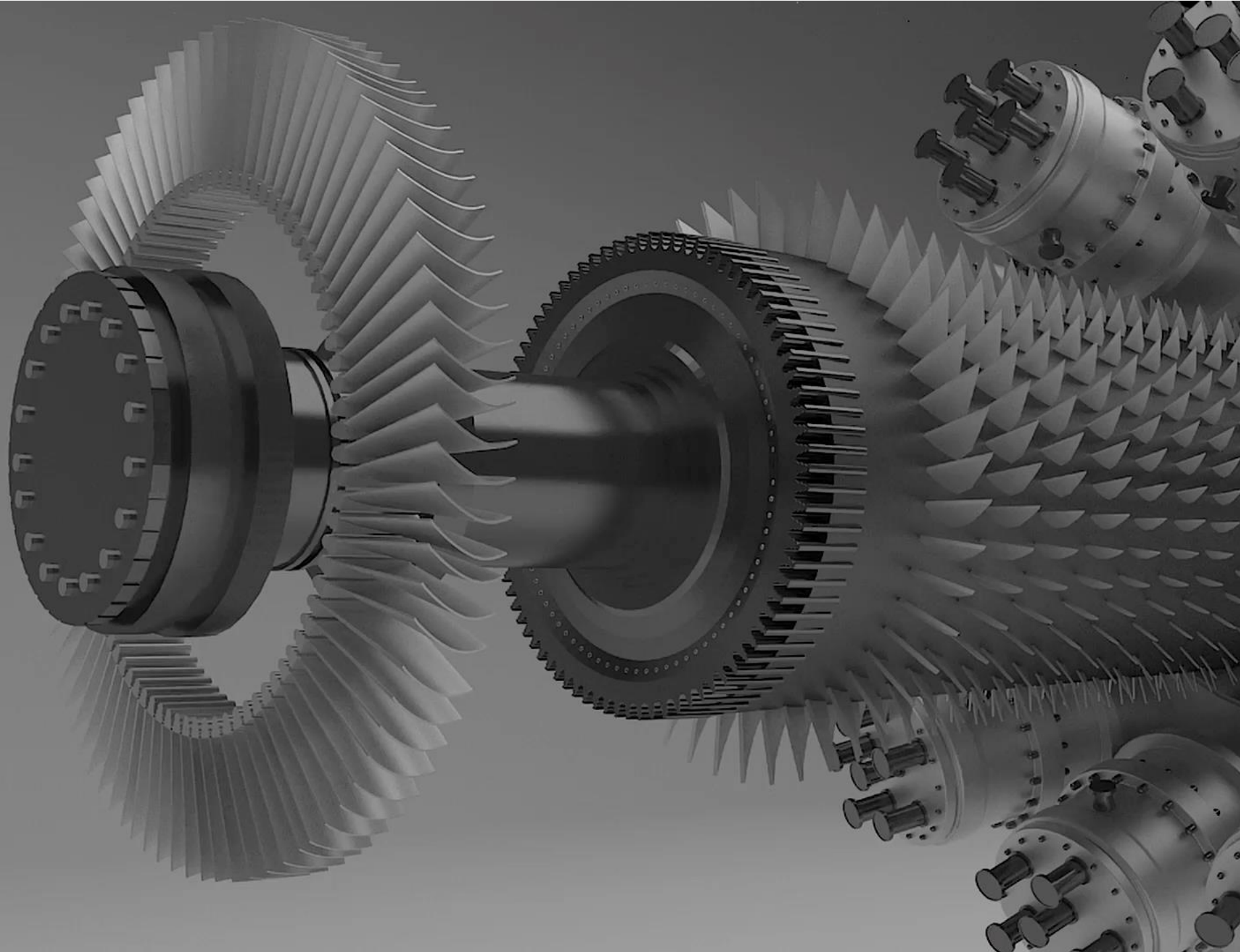
SPARKPREDICT®



# The failure

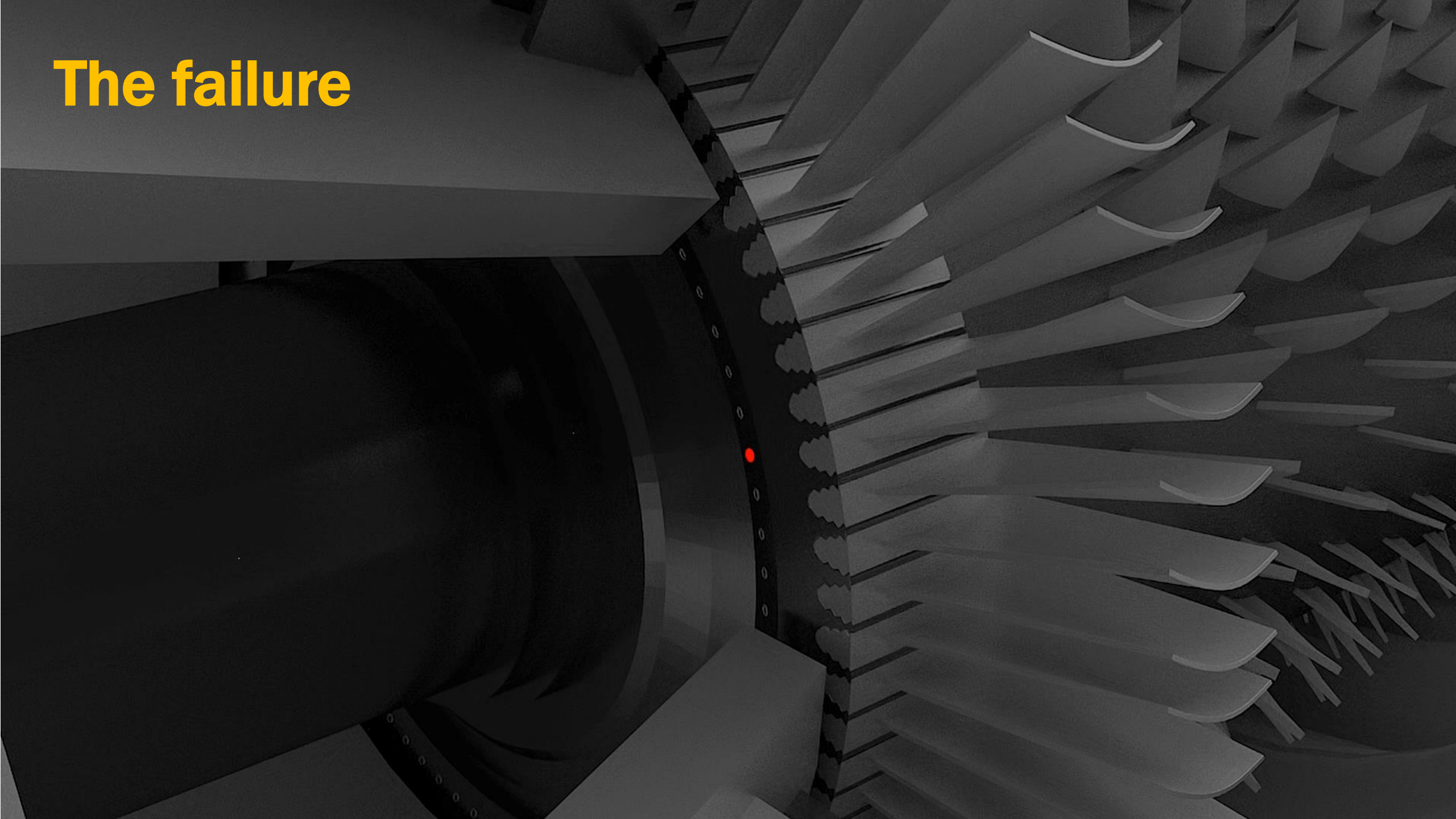


# The failure





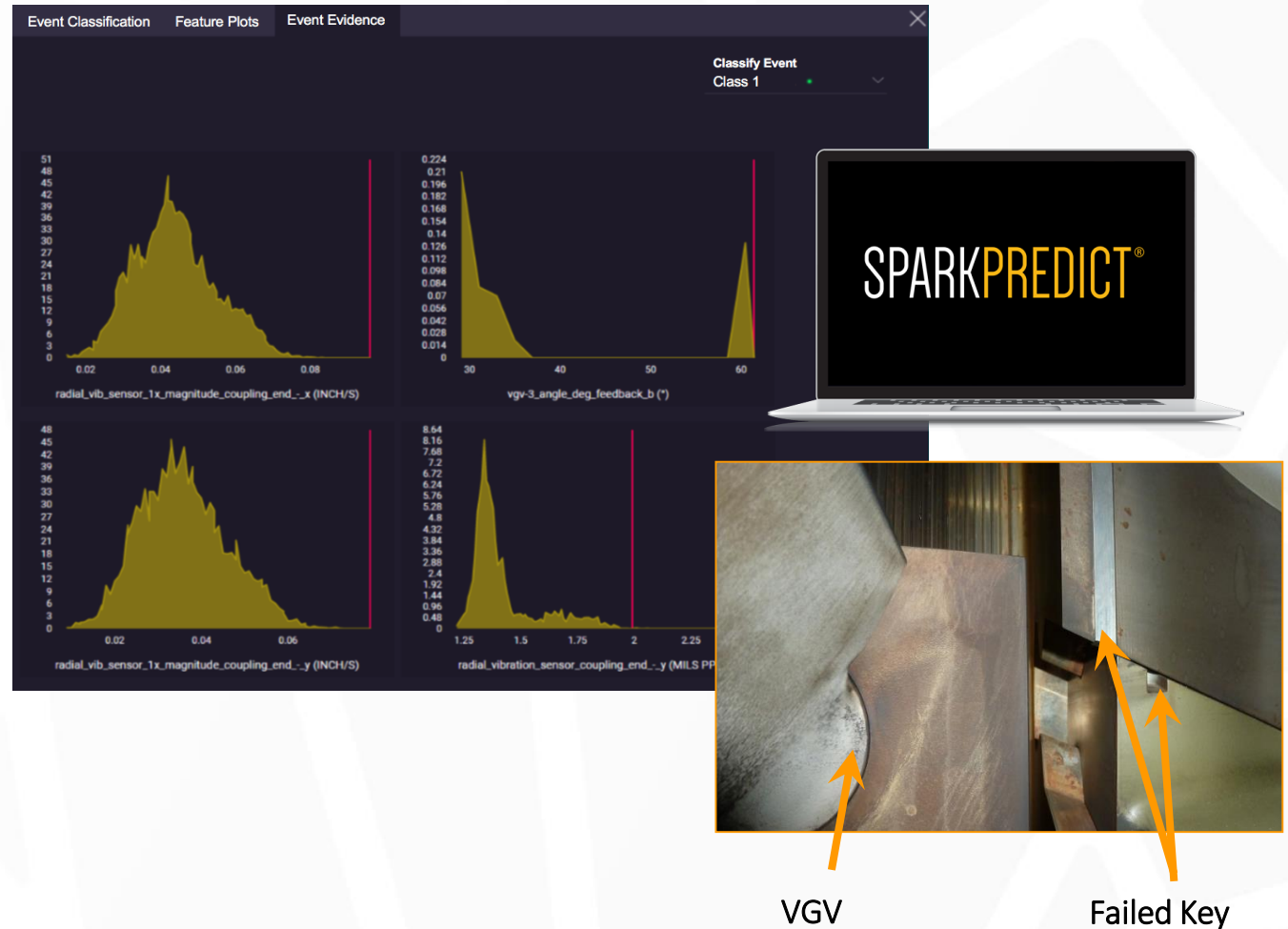
# The failure



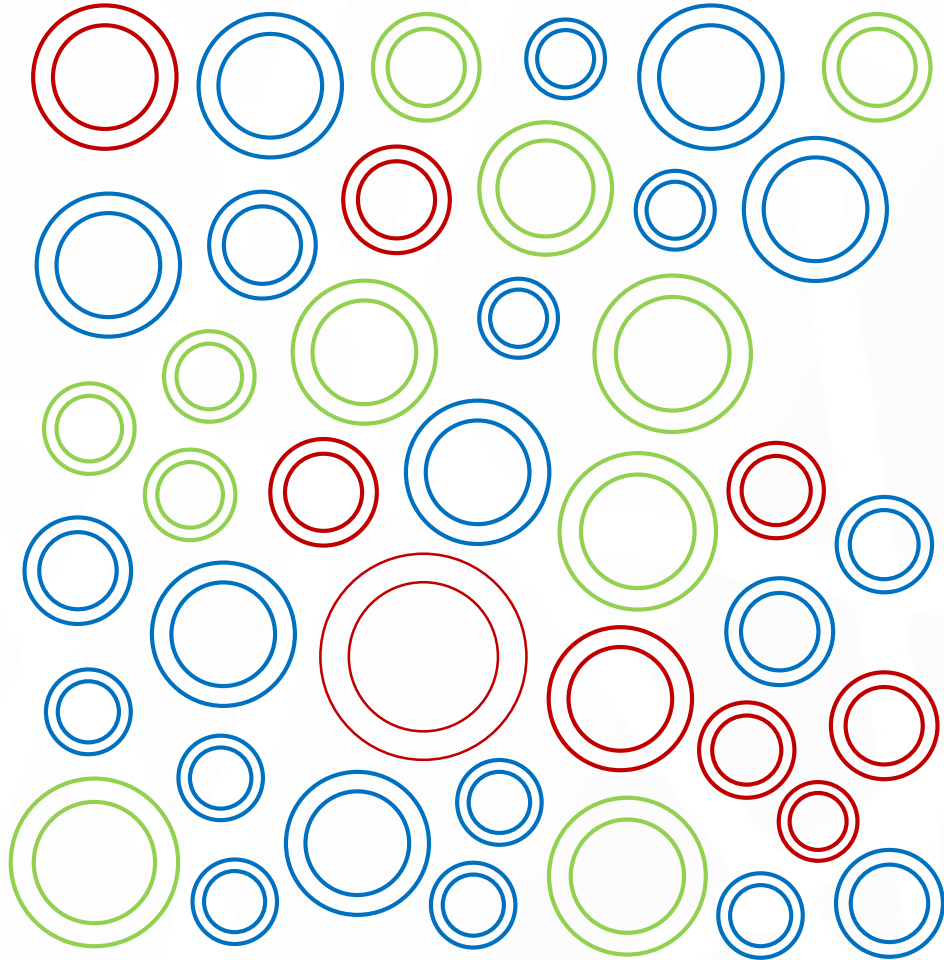
# How did machine learning deliver?



- Detected anomaly with one month lead time
- APR and **other monitoring methods did not detect it**
- Identified never-before-seen issue
- Correctly pointed to problem area of turbine (explainable AI)
- Failure was a manufacturing defect unlikely to occur again



# Supervised Learning



Red



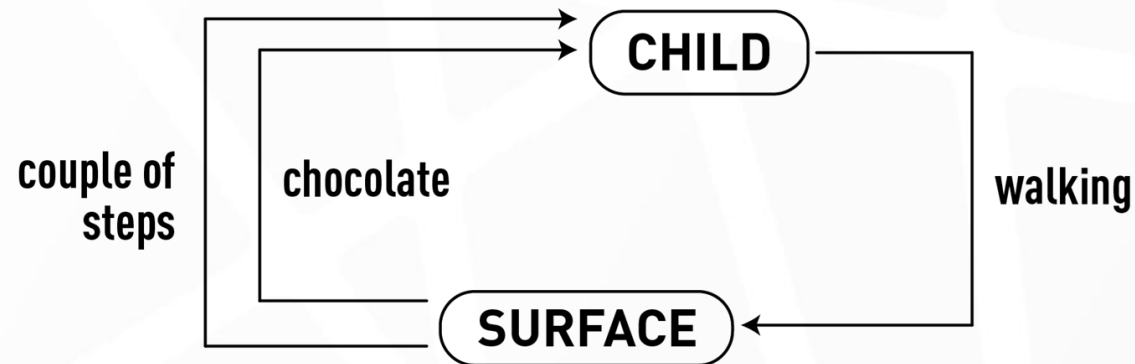
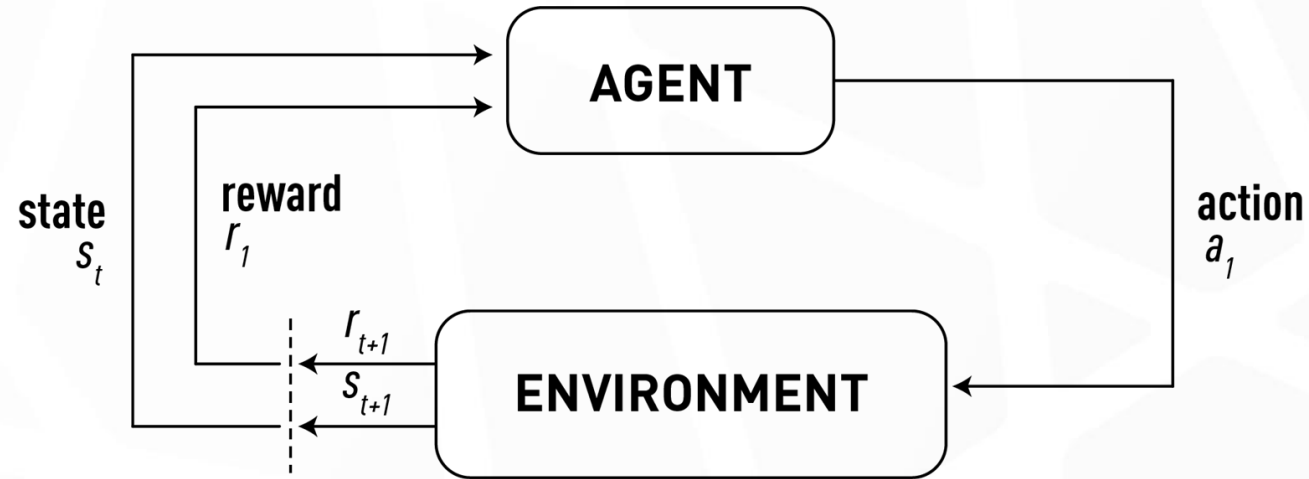
Green



Blue

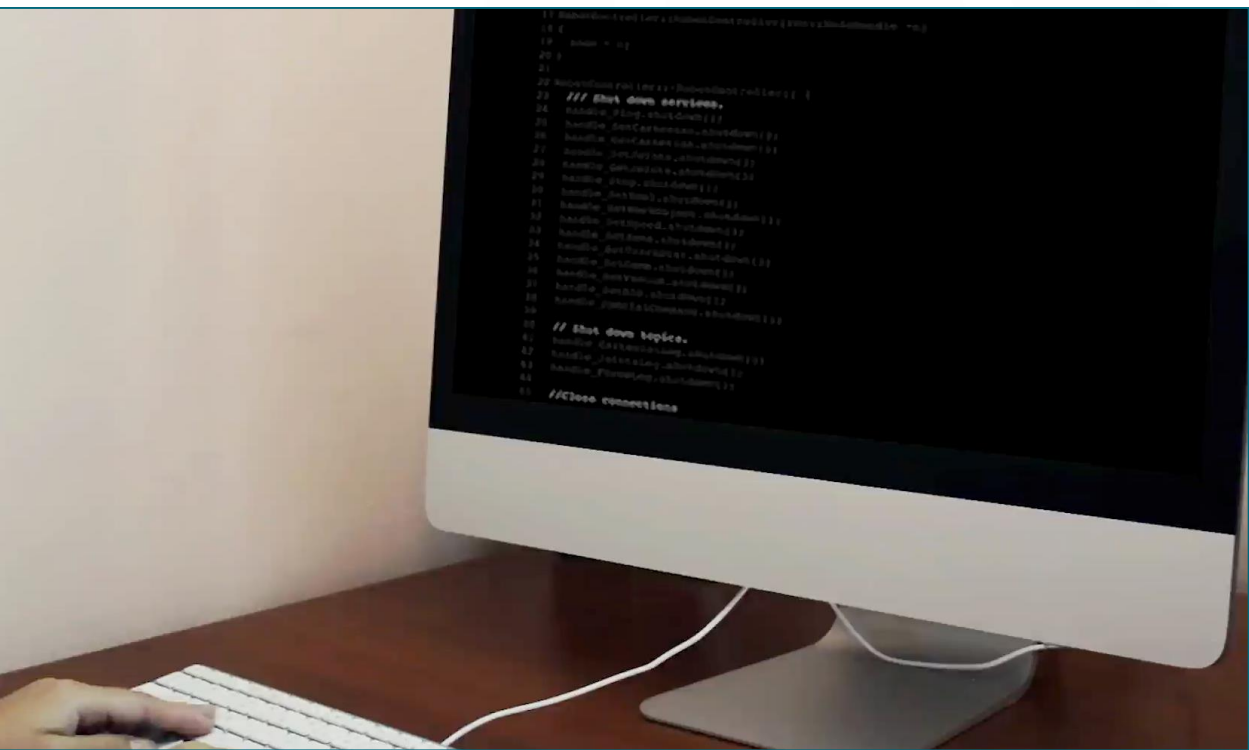


# What Is Reinforcement Learning?

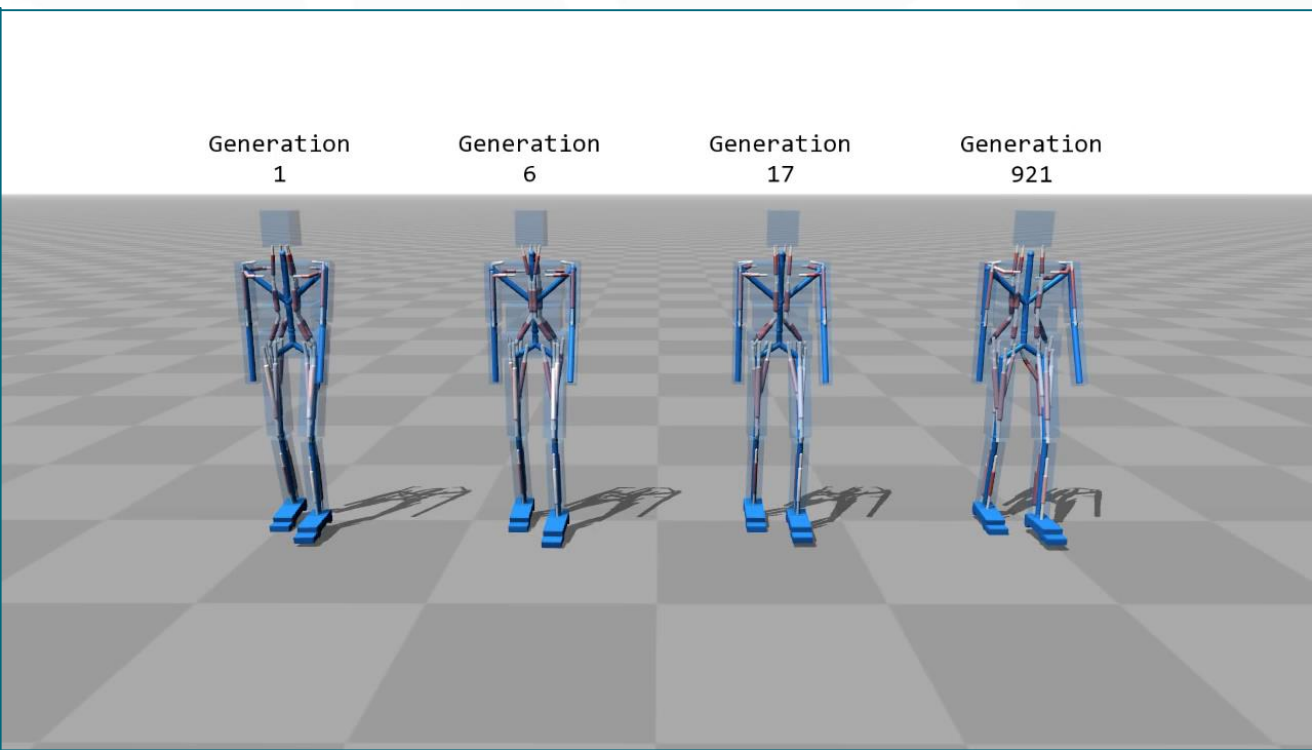




# Reinforcement Learning: The path to Automation



Explicit Instructions “If-Then-Else”



Genetic/Evolutionary Algorithms (AI  
Technique) Geijtenbeek et. al.

# Safety through automation



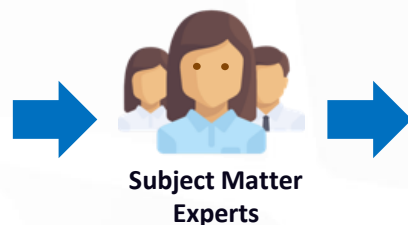
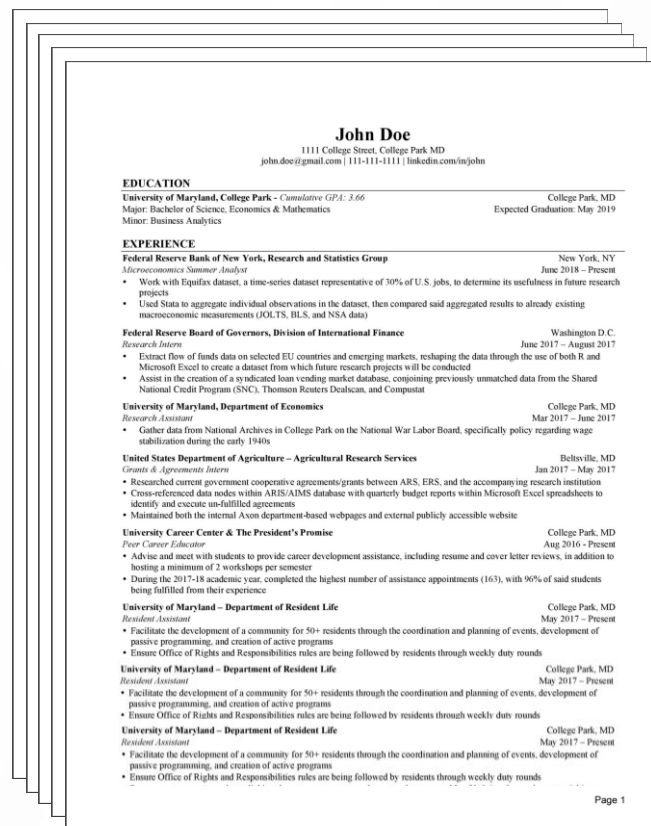
Engine startup sequence...





# Natural Language Processing

# How Do Humans Analyze Natural Language Content



Select  
Existing Information

Extract  
Information

Categorize  
Information

Content	Date	FileName	Candidate	Current Role	Org Fit
John Doe 1111 College Street, College Park ...	02.05.2019	jd_resume.pdf	John Doe	Microecon..	Sales
_____	_____	_____	_____	_____	_____
_____	_____	_____	_____	_____	_____
_____	_____	_____	_____	_____	_____
_____	_____	_____	_____	_____	_____
_____	_____	_____	_____	_____	_____
_____	_____	_____	_____	_____	_____

NLP Automates *Extraction* or *Categorization* of Information From Natural Language Content





# AI in Cybersecurity



**Evolving Landscape:** With the number of connected devices and new threats growing exponentially each year, machine scale is required to keep up with the evolving threat landscape

**50B**

Number of connected devices that will need to be secured in 2020, up from 9B in 2012



**50%**

Of connected devices will be IoT

**95%**

Cyber attacks originate from the endpoint and propagate through the network

**600M**

Number of new malicious threats created *last year* by hackers around the globe, up from 47M in 2010



**325M**

Ransomware payments were made from corporations to hackers in 2015

**4x**

Expected increase in ransomware payments in 2016

**<25%**

The initial detection rate of newly created malware by traditional (signature based) antivirus solutions

**4 weeks**

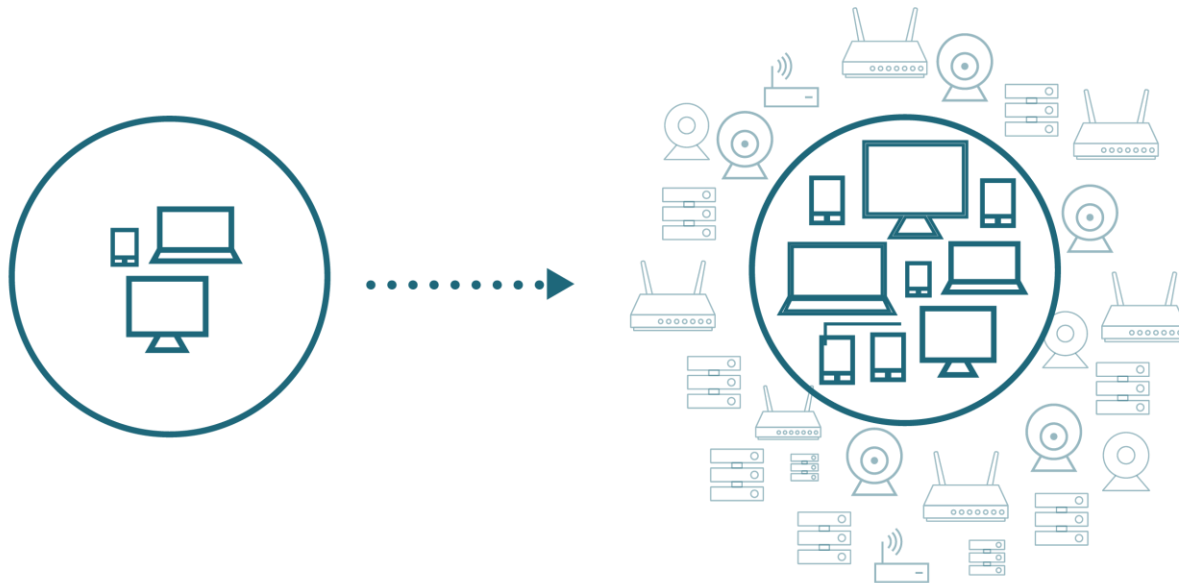
The time it takes most traditional antivirus vendors to detect a new virus

**OTC2017** \

OFFSHORE TECHNOLOGY CONFERENCE  
1-4 May 2017 \ Houston, Texas, USA \ NRG Park  
2017.otcnet.org

OTC-27895-MS • Using a Cognitive Analytic Approach to Enhance Cybersecurity on Oil and Gas OT Systems • Herve

The signature-based approach to endpoint protection is broken. A new approach is needed to keep up with the evolving threat landscape.



Traditional ***perimeter detection*** can no longer keep up with the **greater number of outside connections** in OT networks

Traditional ***endpoint detection*** can no longer keep up with the **proliferation of new sensors and endpoints** to guard

# AI Can Solve Major Cybersecurity Problems

*The average cost of a data breach has increased from \$5M to \$7.1M since 2017*



Cyber threats are increasing  
in scale and complexity

*Zero-day attacks are **4x more likely** to compromise organizations*



Cybersecurity teams are  
overworked and understaffed

***75%** of ~1400 survey respondents think that their IT security function is understaffed*



Companies are not getting  
value from their investments

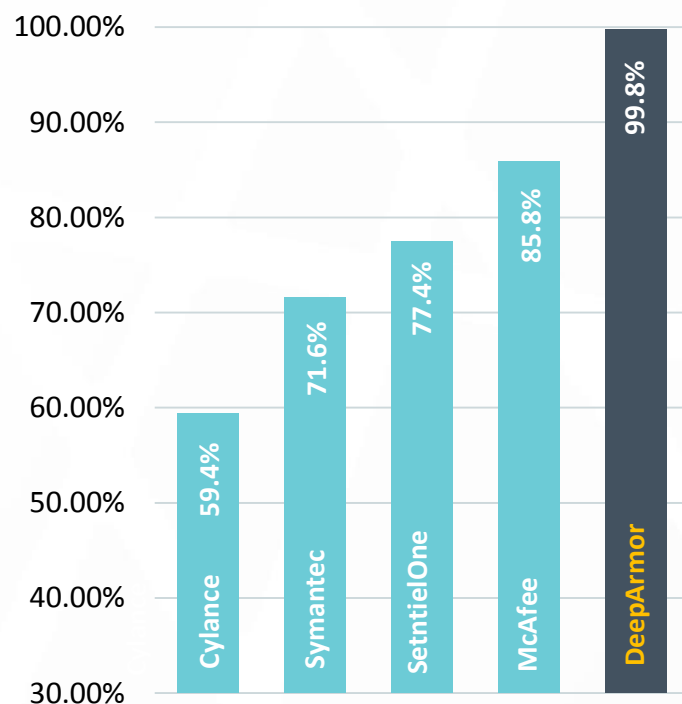
*Only **46%** of EDR tools are actively employed and used*

# Why Artificial Intelligence?

## Industry-Leading Zero-day Detection

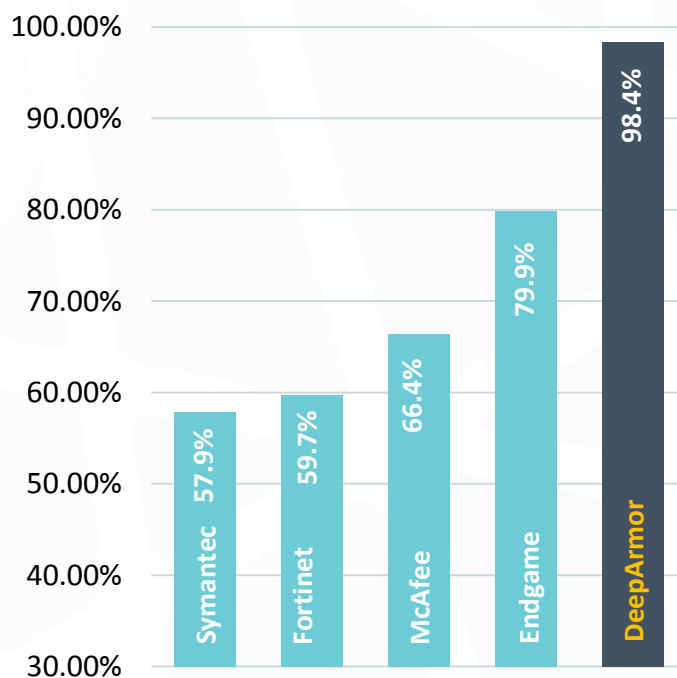


 Windows



Zero-Day Malware Protection Testing

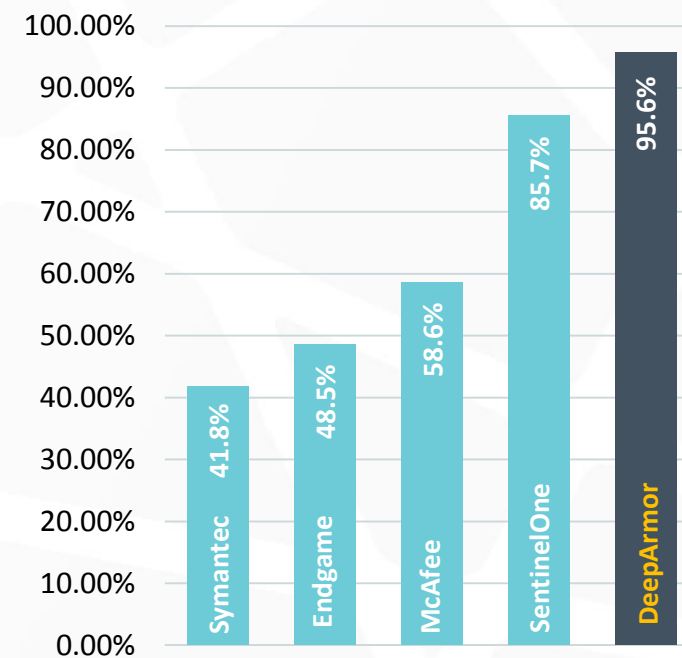
macOS



Zero-Day Malware Protection Testing

0.1% False Positive Rate on NEW Cleanware

 Office



Zero-Day Document Protection Testing

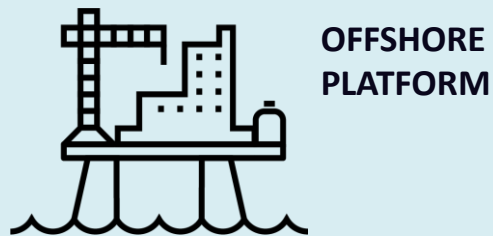


# Case Studies for AI in Oil and Gas

# Advanced Analytics Applications in the O&G Industry



## UPSTREAM



ONSHORE  
PUMP JACK

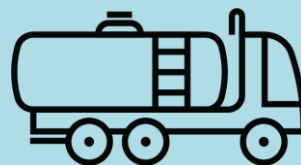
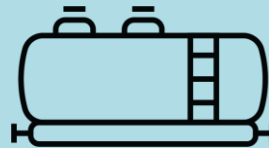


### APPLICATIONS

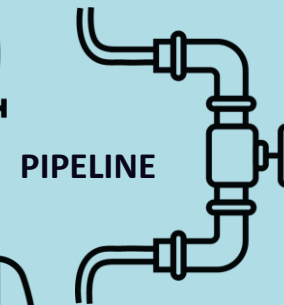
- Reliability and Predictive Maintenance
- Maintenance Workflow Automation
- Drilling and Production Optimization
- Asset Protection
- Supply Chain and Capital Decision Support

## MIDSTREAM

### PROCESS & STORAGE



TRANSPORT

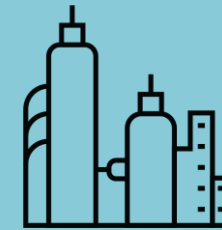


PIPELINE

### APPLICATIONS

- Transportation Optimization
- Pipeline Risk Assessment
- Storage Optimization

## DOWNSTREAM



DISTRIBUTION  
SALES | MARKETING | RETAIL



### APPLICATIONS

- Reliability and Predictive Maintenance
- Demand Optimization
- Trade Optimization
- Price Optimization



# Aker BP: Unmanned Production



Critical Multi-phase  
pump (MPP)



RELIABILITY

## TRANSFORMATIVE RESULTS

**75%** of Historical  
Failures Identified

**9-12** Days Advance  
Forewarning

**2** Weeks from Data  
to Deployment



# O&G Supermajor: Offshore Production



## RESULTS PER PLATFORM

**2** Months to deploy  
on 20 subsystems

**2X** Advance notice of  
failures (9 days)

**1-4%** Increase in  
annual production



## "Refinery of the Future" Partners



**Deloitte**

  
Hewlett Packard  
Enterprise

**FLOWSERVE**



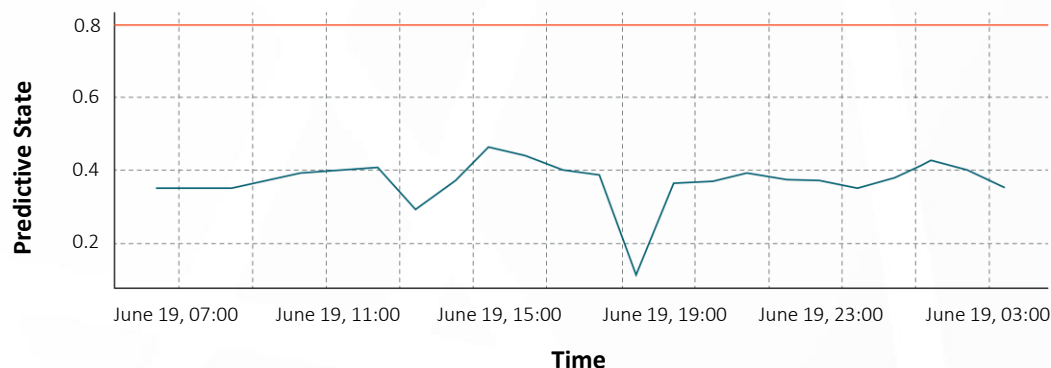
## TRANSFORMATIVE RESULTS

- Delivering predictability for critical pumps
- Improving refinery safety and product quality

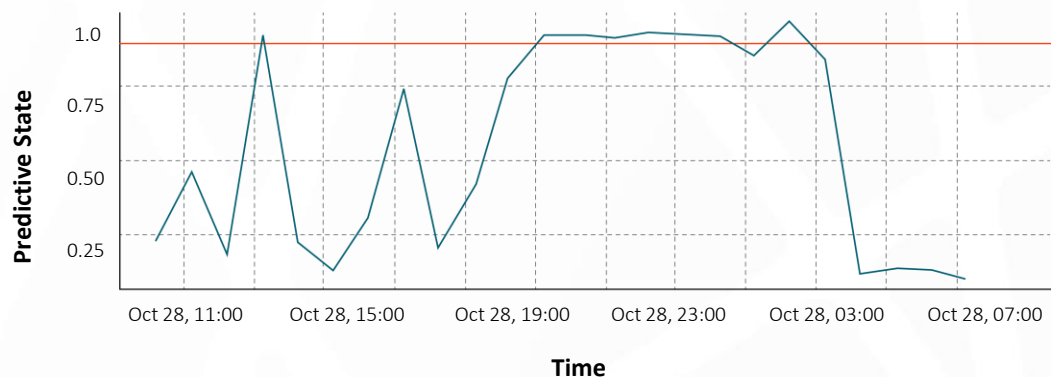
# Real-time drilling failure prediction



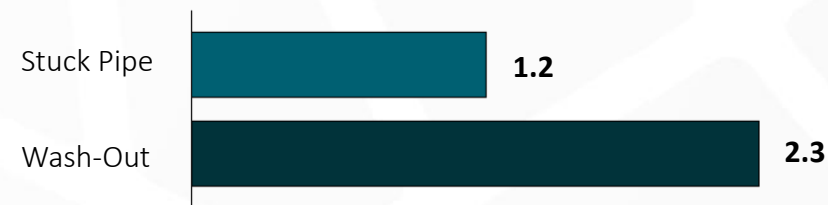
## Stuck Pipe Risk Index



## Wash-out Risk Index



## Avg. Failure Prediction Lead Time (Hours)



### Use Case:

Improve well efficiency by reducing stuck pipe and drill string washout failures

### Accuracy:

>90%, combined for misses and false positives

### Benefits:

Provides timely (>1 hour) notice for preventive action

Automatically accounts for state of drill rig

### Data:

Pressure, temperature, RPM, chemicals, flow rate, tubing material grade, wellbore temp, fluid density, composition etc.

Equipment make, supervisor experience etc.



# Meanwhile In Aerospace





Aerial Operating System  
Autonomous Aircrafts



# Artificial Intelligence: Good or Evil?



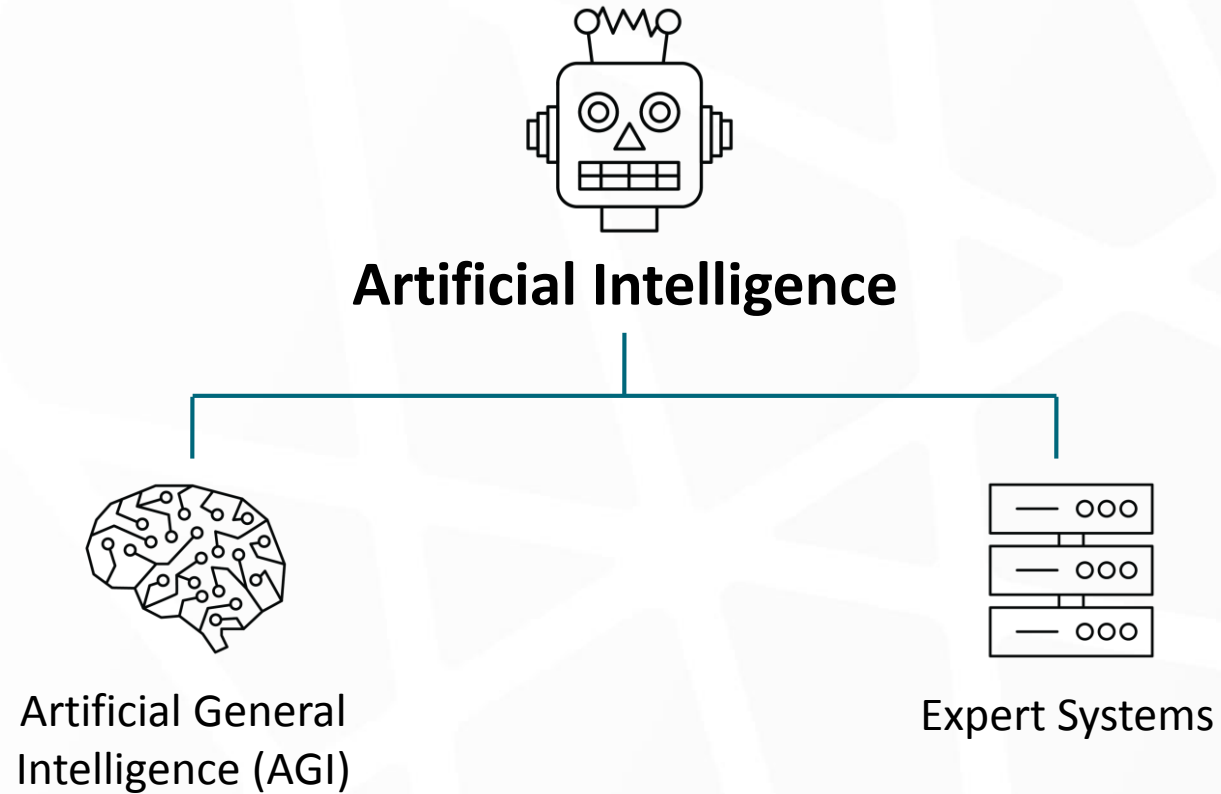


# The Car: Good or Evil?

---



# Artificial Intelligence



I trust Humanity

I am Optimistic.  
AI will positively  
transform our world.





# Questions?



Philippe Herve

*VP of Oil and Gas Solutions*

[pherve@sparkcognition.com](mailto:pherve@sparkcognition.com)