

How Shape-Based Intelligence Transforms Data Insights

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AGENDA

01	Data Analytics Trends
02	Data Monetization
03	Shape-Based Intelligence
04	Hype, Magic, and Myths
05	Case Studies
06	Summary

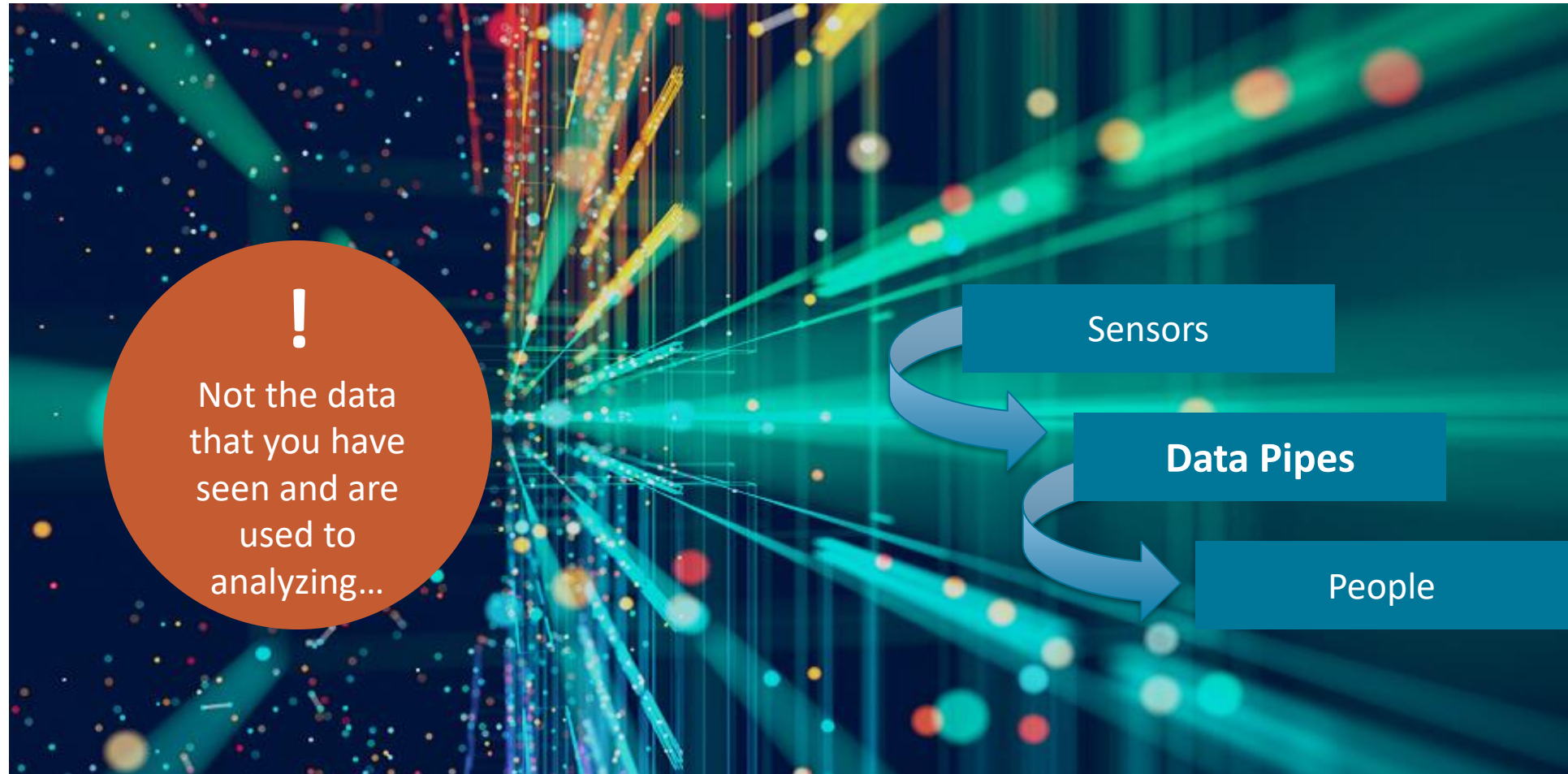
DATA ANALYTICS TRENDS



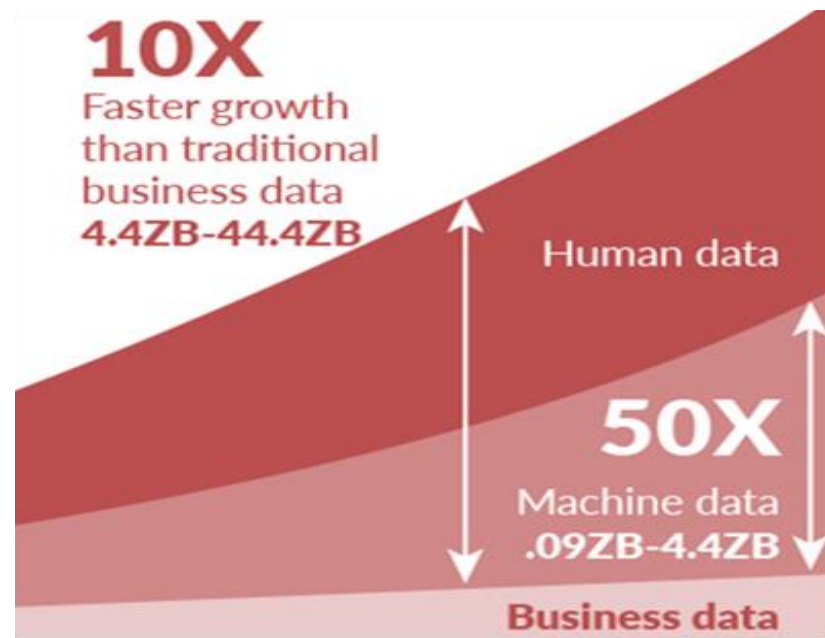
DIGITAL TRANSFORMATION ACCELERATED BY THE PANDEMIC



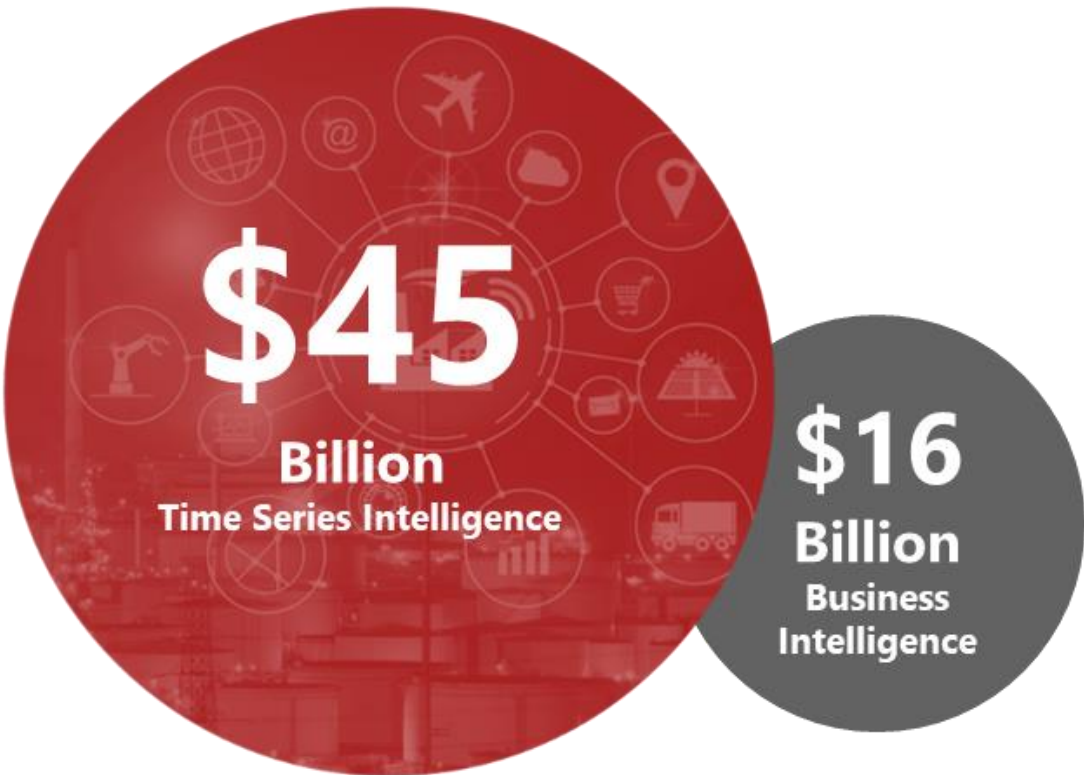
PATTERNS IN THE TIME SERIES DATA ARE THE NEW GOLD NUGGETS



TIME SERIES DATA AND ANALYTICS ARE GROWING FAST



Source: Aureus Analytics Report



Source: Maximize Market Research

WHAT IS GRANULAR TIME SERIES DATA?



- Granularity – This is the sampling rate, i.e., the rate at which measurements are taken. Typically expressed as minutes, seconds, milliseconds, nanoseconds.
- Dimensionality – The descriptors and metadata for the individual units of analysis, i.e., the SKU within a particular retail store, or sensor on a particular car engine, etc.

HOW IS THIS DIFFERENT FROM THE TIME SERIES DATA WE KNOW?

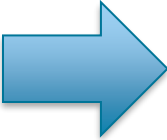
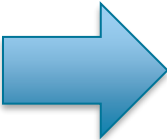
Traditional time series data for market performance

Stock market performance in 2019

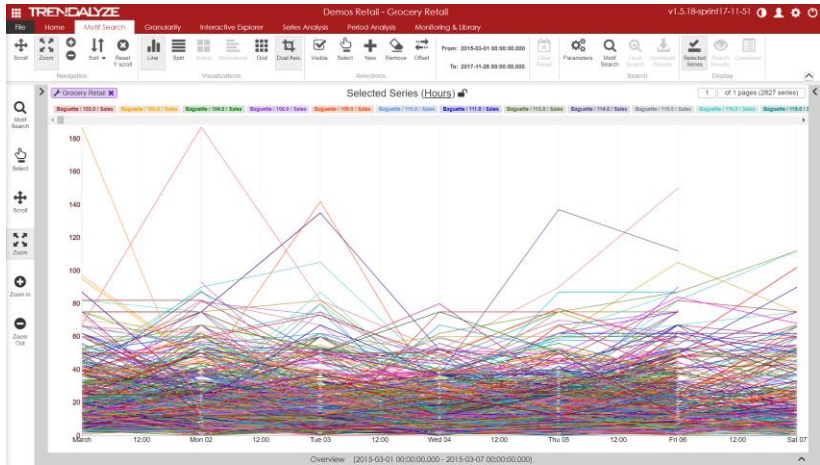
Annual return for the S&P 500, Dow Jones, and NASDAQ



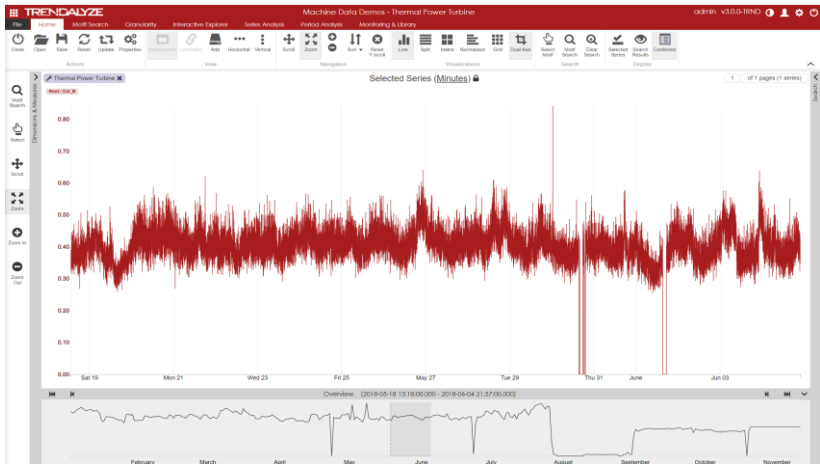
SOURCE: FactSet. Data as of market close on 12/31/2019.



Highly dimensional time series data



High frequency time series data



TIME SERIES DATA IN THE WORLD AROUND US

Vibrations



Predictive failure, maintenance

Trades



Buy / sell opportunities

Vehicle Operations




Driving patterns, maintenance

Vitals



Remote patient monitoring

Transactions



Promotion optimization

Data Traffic

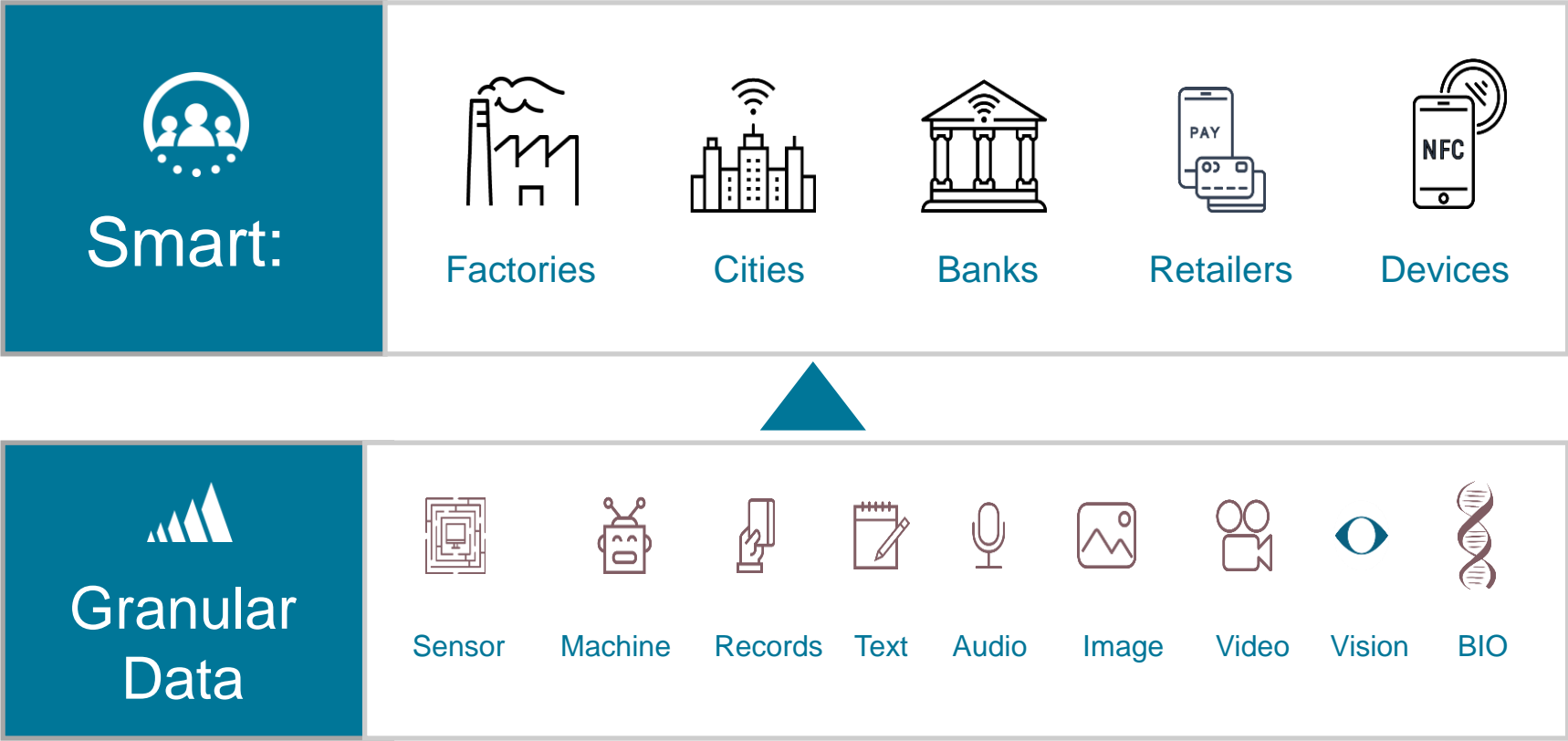


Capacity and utilization

DATA MONETIZATION



CONTEMPORARY ECONOMICS FAVORS DATA MONETIZATION



WHAT IS DATA MONETIZATION?

Using data and analytics
to create new data products
and business models



WHAT DIGITAL ASSETS ARE USED TODAY?



WHAT DATA PRODUCTS ARE USED TODAY?

Target the right leads at your accounts

Recommended Leads

Jenny Strand - 2nd IT Manager at Fresh Solar San Francisco Bay Area

Walter Quinn - 2nd IT Manager at Fresh Solar San Francisco Bay Area

2 common connections

Save in Sales Navigator

Within Salesforce, view recommended leads and find the entire buying committee

ed SALES NAVIGATOR

Requires Salesforce Professional edition or above. Requires Sales Navigator Team edition or above.

Sleep Stages

Monday 30 Day Avg Benchmark

Learn More

Ranges for men your age

Monday Typical Range

Awake 15-33%

NETFLIX

Just for Kids Genres Suggestions For You

Picks for You

DEXTER

Switch

PEARL HARBOR

TORCHWOOD

FINDING NEVERLAND

KEVIN-JAM PAUL REUBEN MAIL-GO

Popular on Facebook

THE EXPENDABLES

PODGE "152" BOX

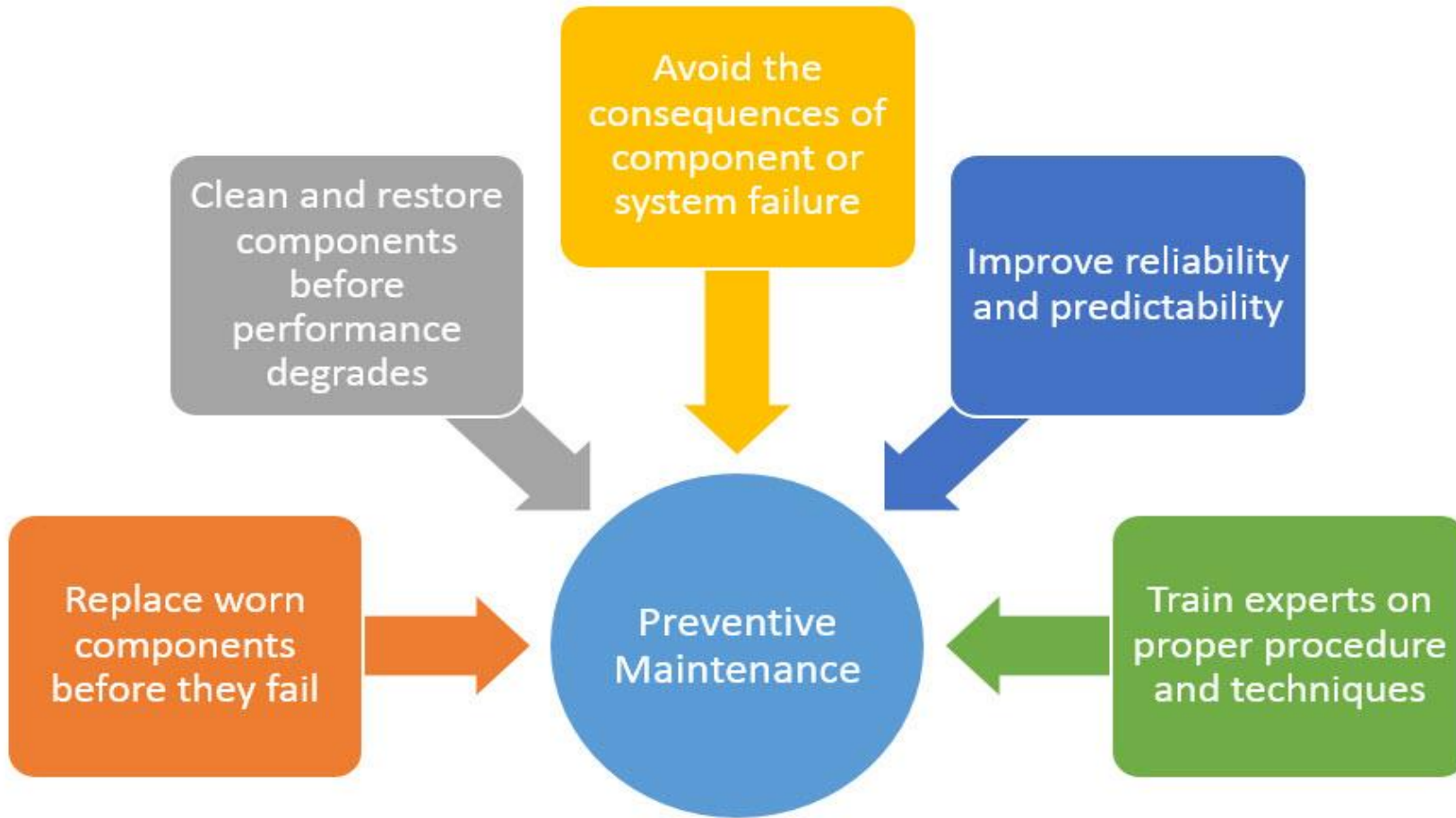
24

Bob the Builder


Go AdWo

Facebook

HOW IS DATA USED TODAY TO OPTIMIZE COSTS?




WHAT TYPE OF DATA SUPPORTS MONETIZATION?



Raw Data

Data capture,
data streams


Asset with potential,
actionable value is TBD



Processed Data

BI data,
analytic summaries

Descriptive, diagnostic,
predictive value



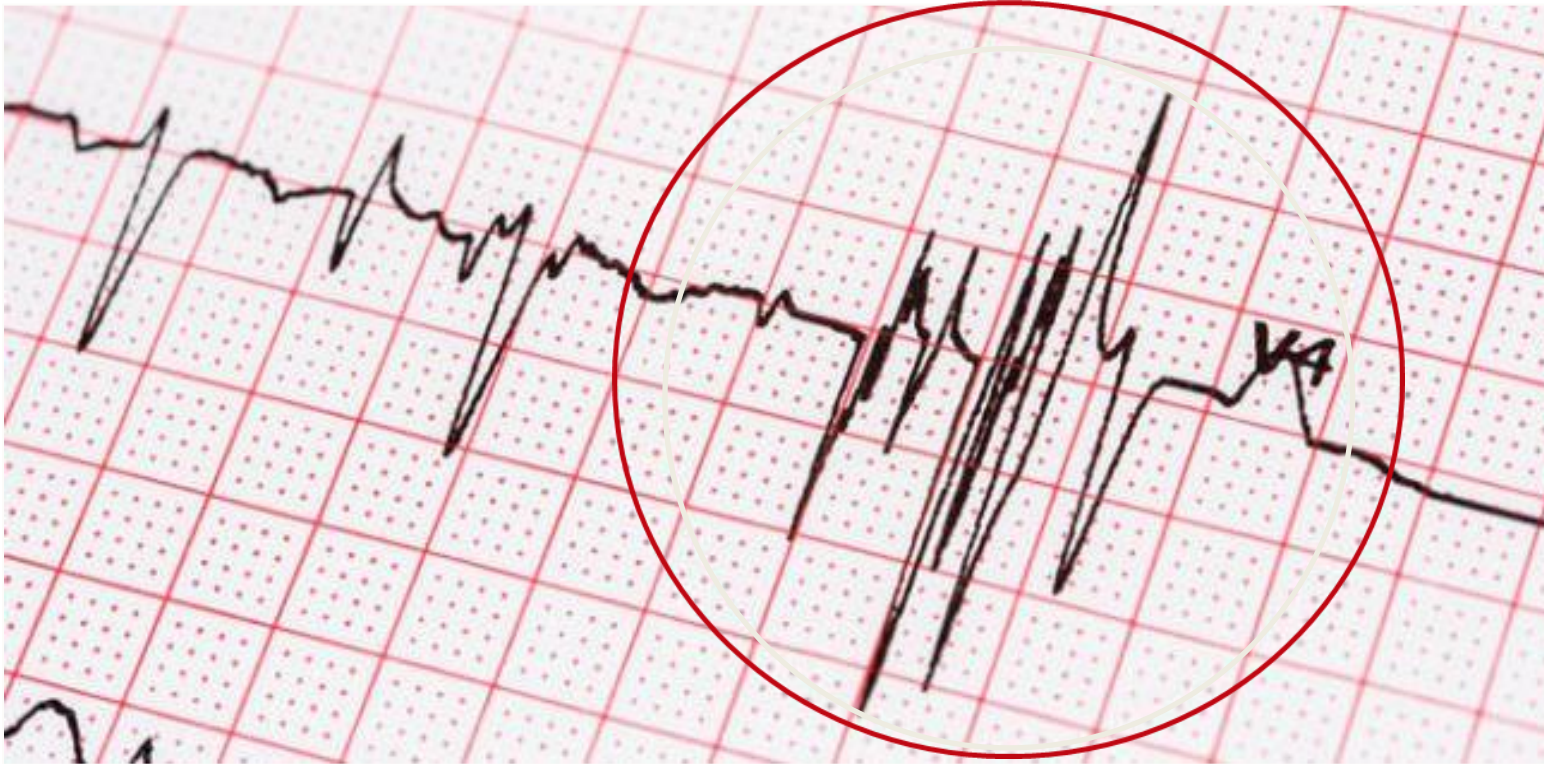
Granular Data

Detailed
temporal data

Trend and pattern-
based actionable value

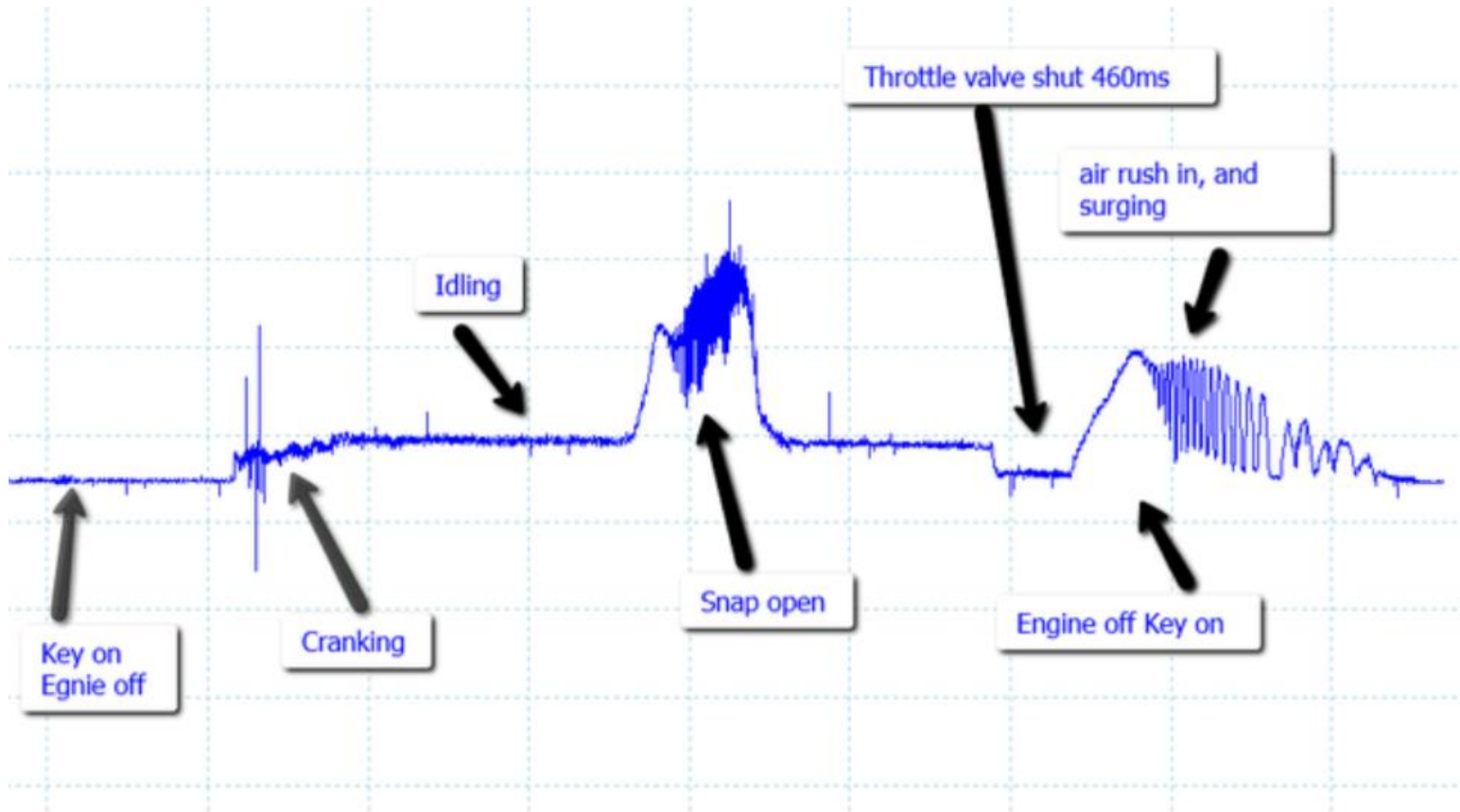


WHAT ARE THESE OPPORTUNITIES FOR MONETIZATION?



- Sensors and devices collect the “heartbeats” of everything
- Knowing the good and the bad “heartbeats” is precious
- Monitoring “heartbeats” is the key to saving or making money

THE OPPORTUNITIES FOR MONETIZATION EXIST IN EVERY INDUSTRY



- Engine diagnostics in fleet management sector
- New car mechanics are like doctors knowing the “heartbeats” of the engine
- Each of these cycles produce distinct “heartbeats” used to detect faults

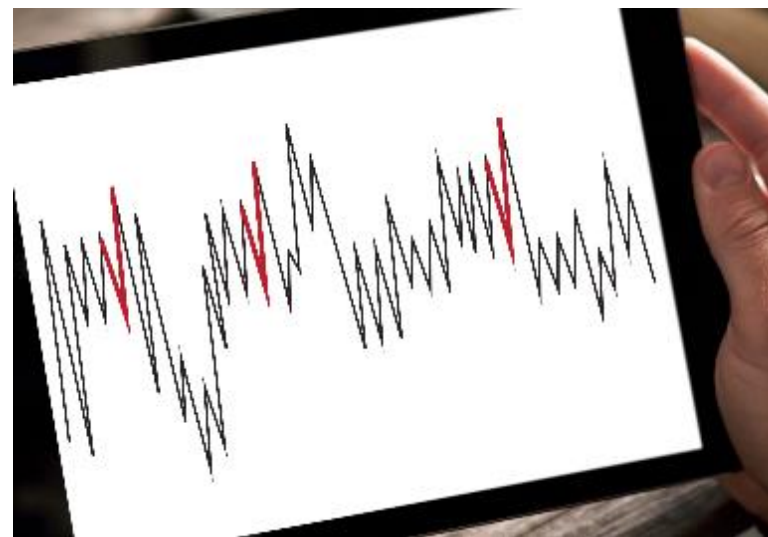
SHAPE-BASED INTELLIGENCE



WHAT IS SHAPE-BASED (MOTIF) INTELLIGENCE?

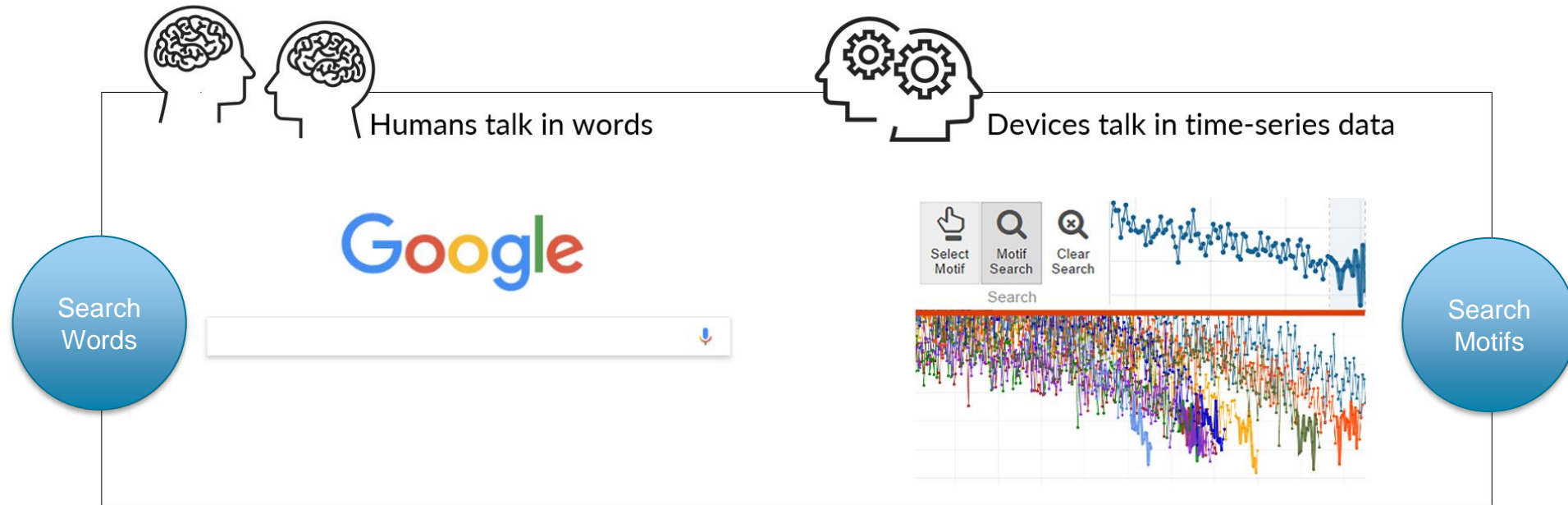


A single pattern/anomaly or first discovery is not a motif



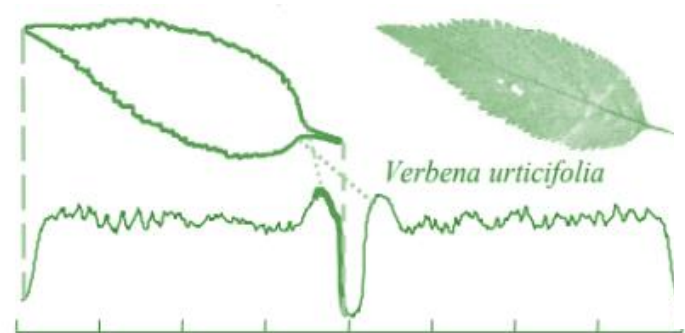
Motifs are recurring patterns that could be mined and leveraged

WHAT DO MOTIFS REALLY REPRESENT?



Motifs are **searchable patterns**, which makes them **monetizable**

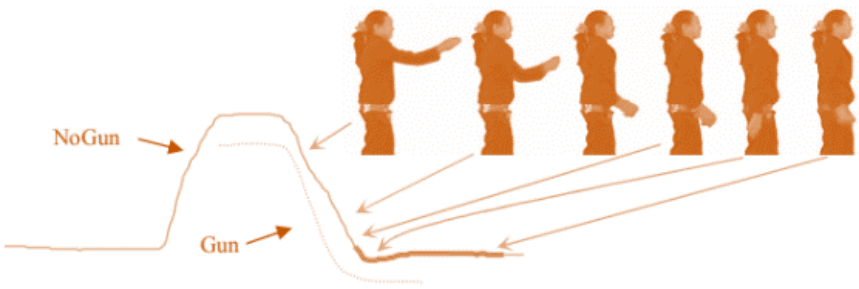
MOTIFS ARE ALL AROUND US



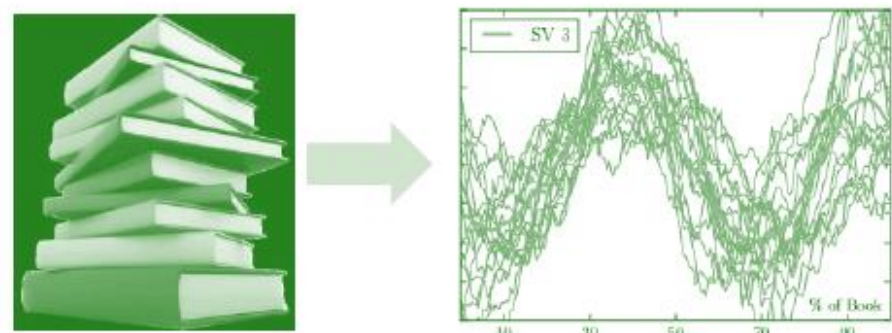
Plant or animal recognition and growth



Language processing



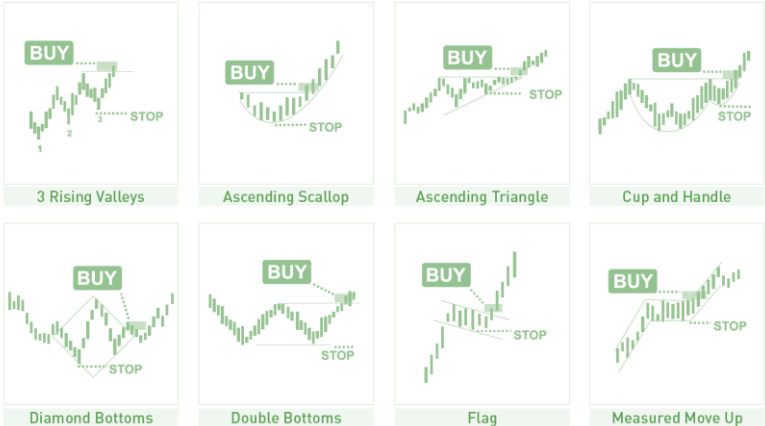
Movement/gesture sensing and identification



Text analysis



Remote medical diagnostic and monitoring



Stock trading and anti-money laundering

HYPE, MAGIC, AND MYTHS



FUNDAMENTAL FLAWS IN MACHINE LEARNING



Machines learn
on their own

Machines are
faster than humans

Machines are
very reliable in
recognizing patterns

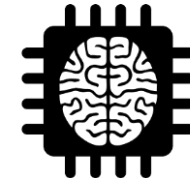
DIFFERENCES BETWEEN HUMAN AND MACHINE LEARNING

HOW DO HUMANS AND MACHINES
LEARN THE DIFFERENCES BETWEEN
THESE TWO SHAPES?



Human Learning:

- Recognize shape differences immediately
- Learn from a few examples the differences
- Identify known shapes instantly



Machine Learning:

- Requires thousands of picture to train the model
- All pictures have to be labeled precisely by humans
- Small shape differences confuse machines

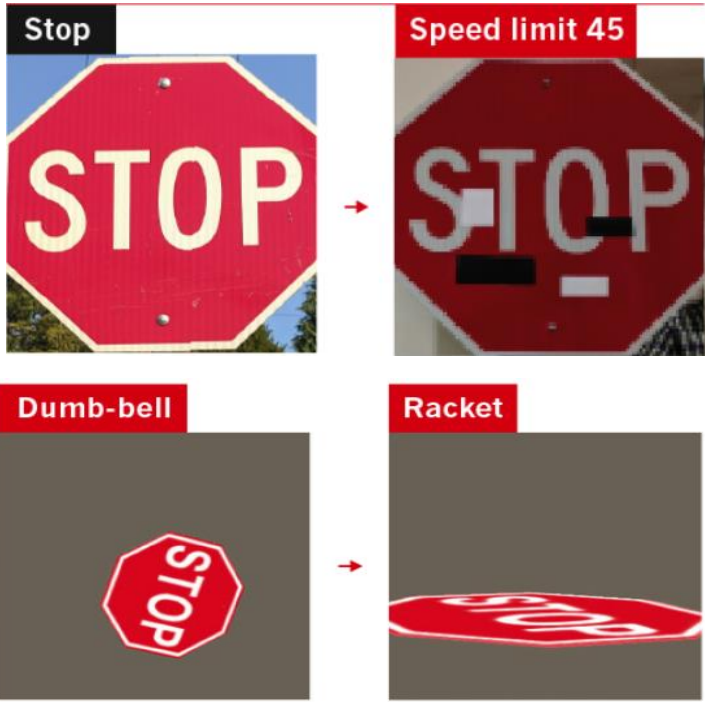
THE CHOICE OF AN AI SOLUTION DEPENDS ON FOUR MAIN FACTORS

Cost

Time

Resources

Reliability



WHAT YOU NEED TO KNOW ABOUT MACHINE LEARNING

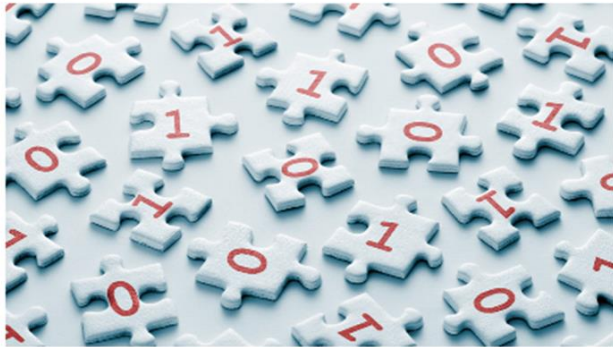
A recent MIT Sloan survey found that only 10% of companies obtain significant financial benefits and achieve ROI with AI, and these companies are not alone.

**Harvard
Business
Review**

Small Data Can Play a Big Role in AI

by H. James Wilson and Paul R. Dougherty
February 17, 2020

Summary Download Print \$8.95 Buy Copy



The way we train AI is fundamentally flawed

The process used to build most of the machine-learning models we use today can't tell if they will work in the real world or not—and that's a problem.



**MIT
Technology
Review**

They trained 50 versions of an image recognition model on [ImageNet](#), a dataset of images of everyday objects. The only difference between training runs were the random values assigned to the neural network at the start. Yet despite all 50 models scoring more or less the same in the training test—suggesting that they were equally accurate—their performance varied wildly in the stress test.

Gartner's research has found that 85% of ML projects fail. Worse yet, Gartner predicts that this trend will continue through 2022 and beyond.

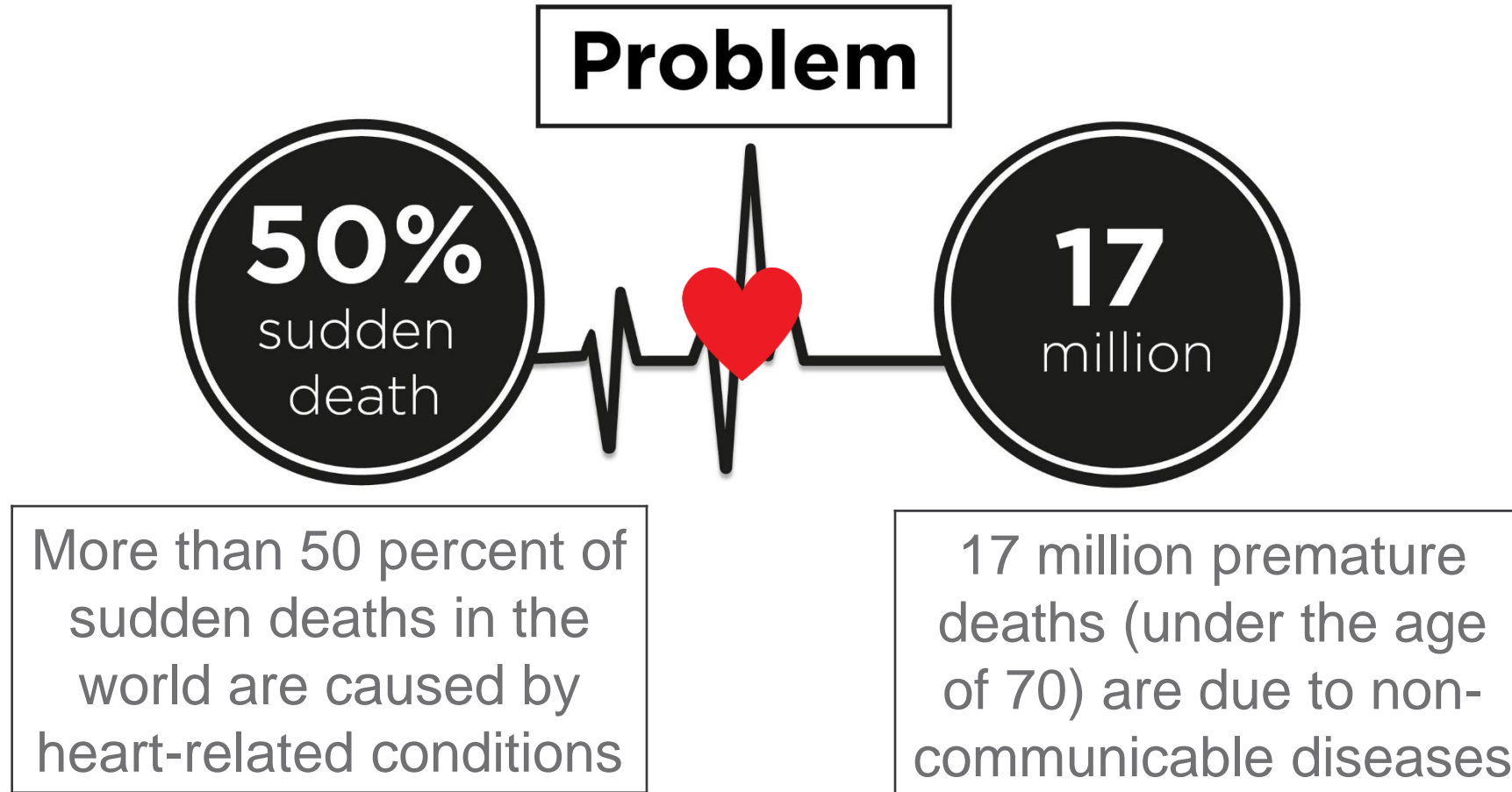
THE RISK OF NOT DOING VS. DOING AI WRONG



CASE STUDIES



CASE STUDY 1: REMOTE CARDIO PATIENT MONITORING



Source: Medical Research Check Point Cardio 2015

CASE STUDY 1: REMOTE CARDIO PATIENT MONITORING (CONTINUED)



Source: Bulgarian Newspaper 24 Hours



The patient was going to the airport



He received a call from the remote diagnostic center

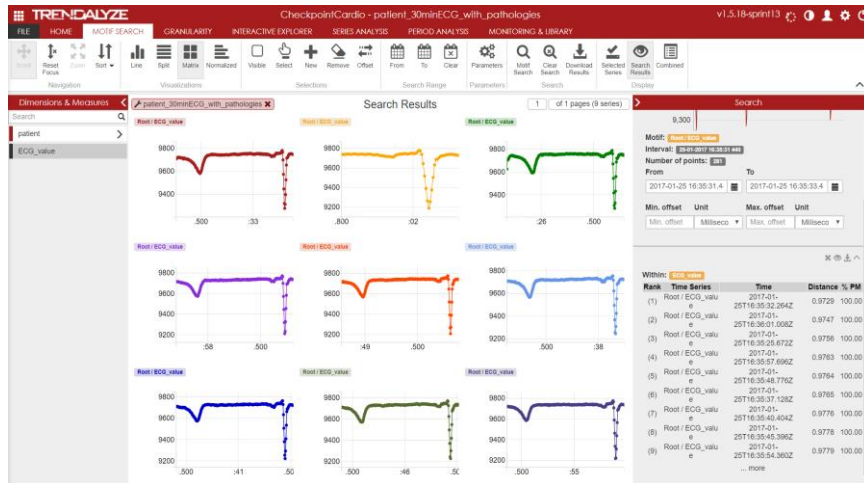


He was taken to hospital



Saved from certain death

CASE STUDY 1: REMOTE CARDIO PATIENT MONITORING (CONTINUED)



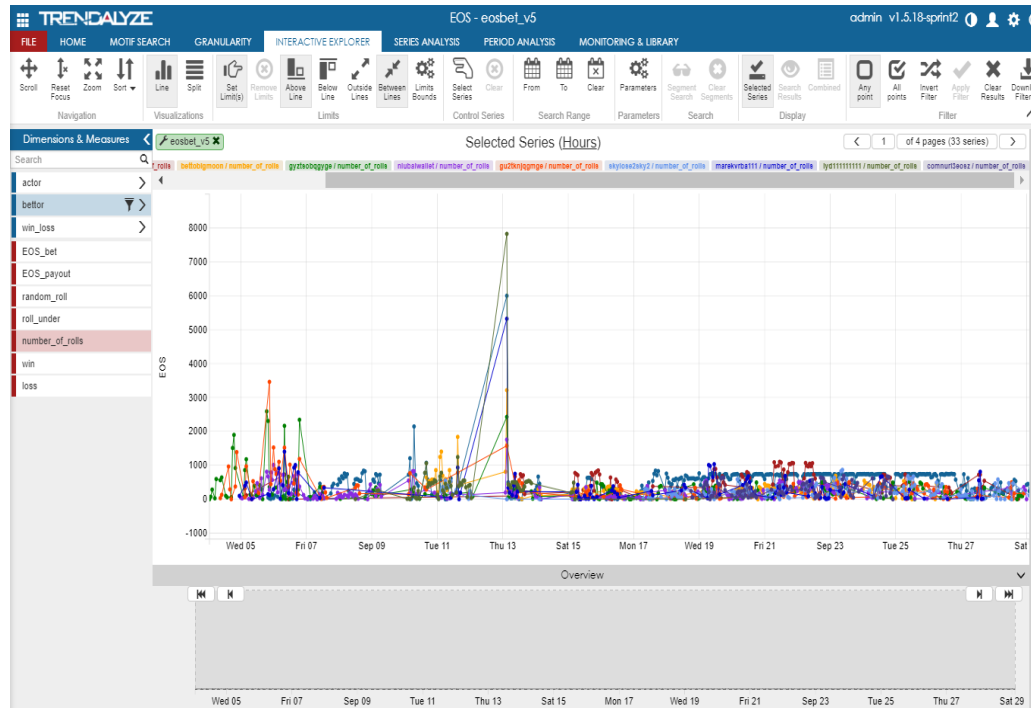
Source: Trendalyze, Inc.



Cardio pathology detection

- Pathology motif (yellow) vs. normal heart beats motifs (green)
- A single heartbeat contains 360 data points
- The pathology motif is detected within a single heartbeat – it has only 180 points

CASE STUDY 2: AUTOMATE BACK OFFICE PATTERNS FOR MONITORING



1

A common ISO scam is to pump the value of an asset using bots to generate fake activities

2

The challenge is to pull the bad patterns from within millions of actively trading wallets

Source: Trendalyze, Inc.

Suspicious Patterns Detection

- The proliferation of digital banking and crypto assets have led to significant increase in the amount and variety of financial crimes
- The cost of monitoring and investigating these crimes has skyrocketed, while clever methods clog the system with false positives
- Like fingerprints, time series patterns reveal price manipulation, AML, and other crimes

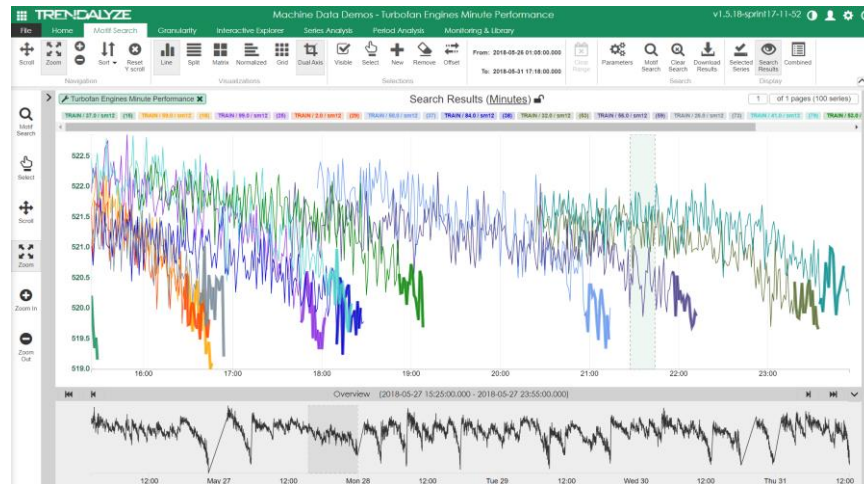
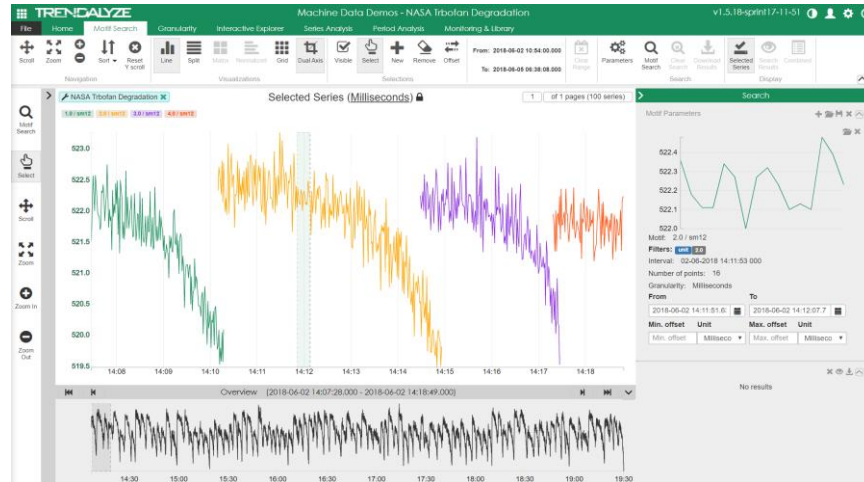
CASE STUDY 3: CONDITION BASED MAINTENANCE

97% Preventive Failure
Detection
without modeling

The two key problems in implementing predictive modeling for Condition Based Maintenance (CBM) are its cost and the accuracy of the results. Traditional machine learning and statistical approaches require highly trained data scientists who are both expensive and in short supply. For CBM, the data scientists need time to familiarize themselves with all the nuances of the specific business cases and to understand them. Hence, they have to discuss these with the subject matter experts (SMEs). Furthermore, all traditional approaches require long histories of equipment failures in order to build accurate models. As a result, money, time and accuracy frequently deter the implementation of CBM.



CASE STUDY 3: CONDITION BASED MAINTENANCE (CONTINUED)



Source: Trendalyze, Inc.



Just-in-time maintenance

- Engineers can select a normal operations motifs or examples of single observations
- Motif search allows to find all similar cases
- Governed libraries of machine operation motifs enable fast monitoring and CBM solutions

CASE STUDY 4: SAMPLE SOLUTION OFFERING

Path to Macro Objectives:

- Proactive monitoring to prevent outages
- Quicker resolution of outages that already occurred
- Predictive maintenance to lower costs

Unified platform for all decisions:

AI and Big Data solution for better utilization and data monetization in streamlining operations

MONITOR



Process information collected by millions of sensors in real time to maintain highly efficient operations

MANAGE



Provide management dashboards, KPIs and reports to ensure resource alignment and goals attainment

MAINTAIN



Predictive maintenance increases the life span of assets and lowers costs by just-in-time equipment repair

MONETIZE



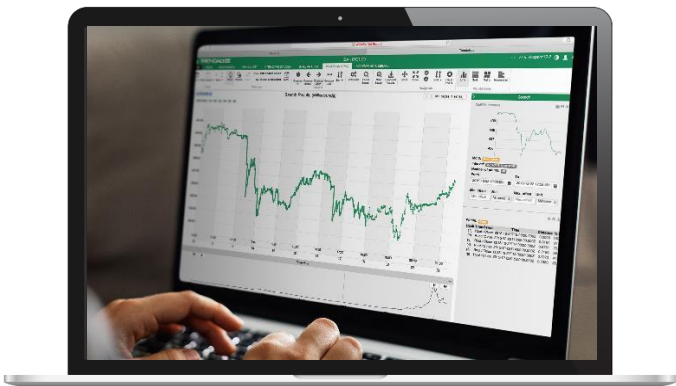
Leverage AI, consumption, market and customer data to optimize pricing and capitalize on newly discovered opportunities

What Producers Need:

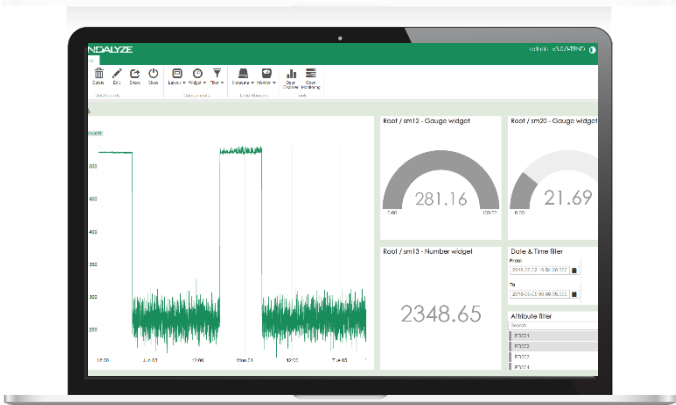
AI-based solutions that are integrated into “Operations and Maintenance” programs, to prevent power outages and lower costs of equipment repairs

CASE STUDY 4: AI2ENERGY SAMPLE SOLUTION OFFERING (CONTINUED)

Self-service IoT analytics tools

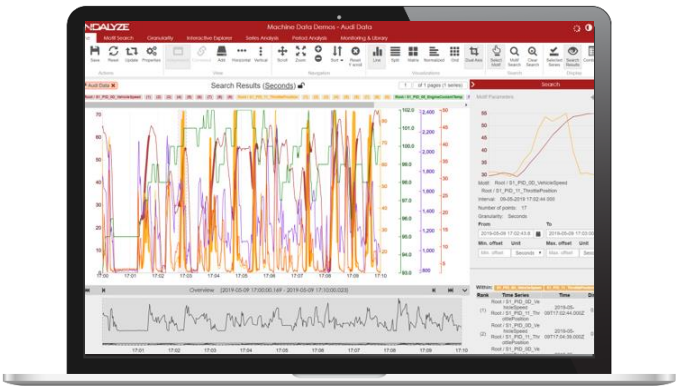


AI identifies patterns and trends for analysis



Search for patterns in real time and get alerts

Search engine for IoT and industrial data



Search for complex patterns in data streams

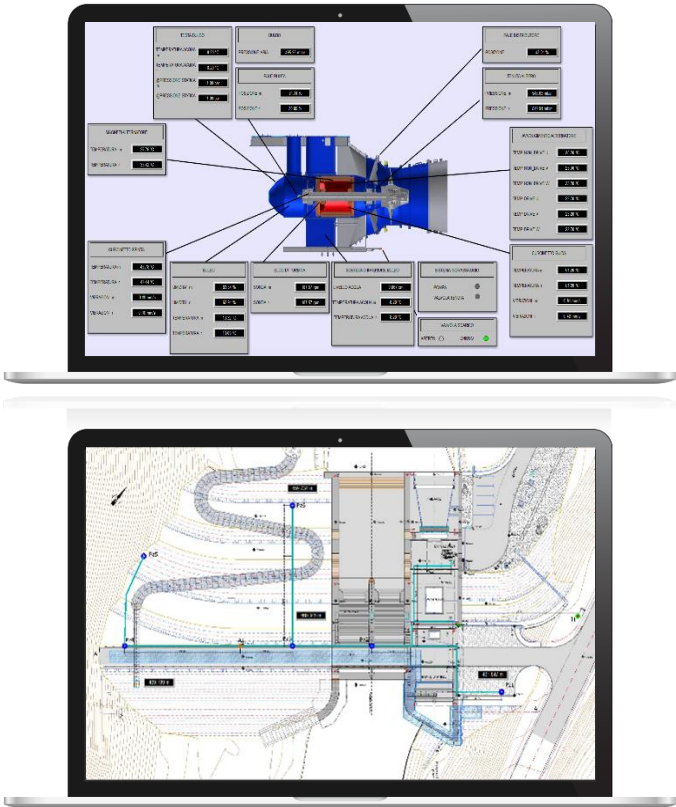


Search for correlated patterns in machine data

Source: Trendalyze, Inc.

CASE STUDY 4: AI2ENERGY SAMPLE SOLUTION OFFERING (CONTINUED)

Real-time operational and environmental monitoring



Operational monitoring and management of every single component

Monitoring of environmental and geographic conditions

Business Intelligence and system monitoring



Real-time BI for power, generation, demand, and distribution

Real-time main units monitoring and control

Source: Trendalyze, Inc.

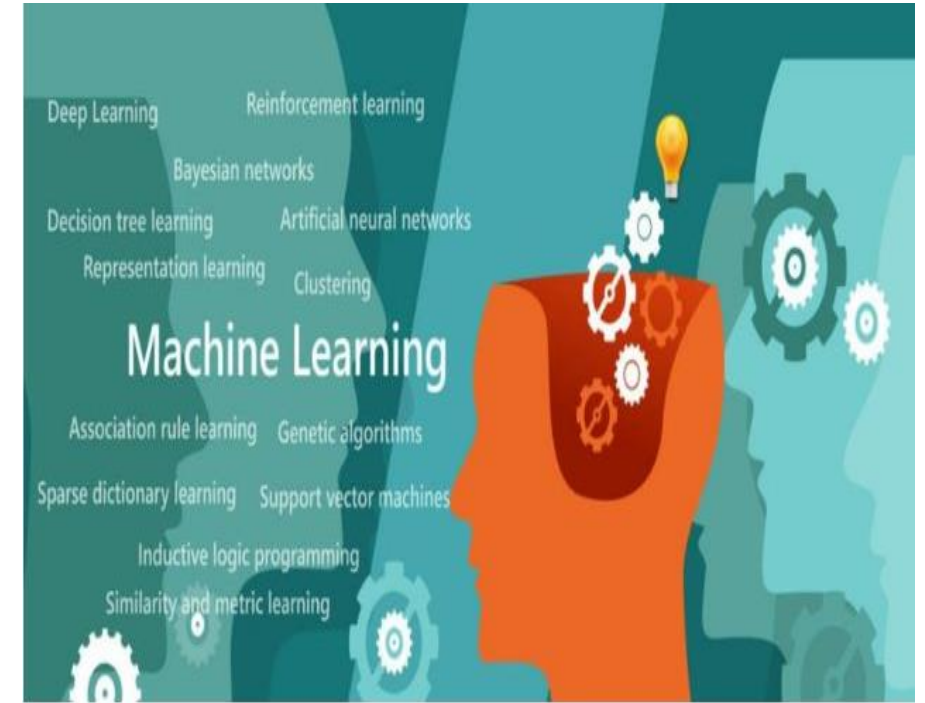
SUMMARY



SHAPE-BASED INTELLIGENCE IS ABOUT CONTEXTUAL ADAPTATION



“Past DARPA* AI investments facilitated the advancement of ‘first wave’ (rule based) and ‘second wave’ (statistical learning based) AI technologies. DARPA-funded R&D enabled some of the first successes in AI, such as expert systems and search, and more recently has advanced machine learning algorithms and hardware. **DARPA is now interested in researching and developing "third wave" AI theory and applications that address the limitations of first and second wave technologies.**”



*Defense Advanced Research Projects Agency (DARPA) is the central research and development organization for the US Department of Defense

WHY SHAPE-BASED INTELLIGENCE TRANSFORMS DATA INSIGHTS

Contextual Adaptation:

Wave 1: “No learning capabilities”

Wave 2: “Statistically impressive, but individually unreliable”

Wave 3: “Systems that construct contextual explanatory models for classes of real-world problems”



Wave 3 Platform’s Advantages:

- Put the “human in the loop” to select contextual patterns for specific problems
- All patterns are completely explainable and can be used for decision automation
- Learn and monitor with just one example or small data sets
- Users can start monitoring applications instantly without model training
- Predictions are based on artificial logical networks with transparent reasoning

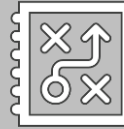
EVOLUTION OF AI-BASED DECISION SUPPORT SYSTEMS



AI 1.0

Expert systems

Expert crafted rules



AI 2.0

**Machine learning
and deep learning**

Statistical learning



AI 3.0

**Shape-based
discovery**

Shape-based learning

QUESTIONS



THANK YOU



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