

SPARK | BEYOND

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Score: 0.009 | Support: 21%

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## About us

Established in 2013 to accelerate AI-powered problem-solving.

Since then we have **delivered \$Bns in tangible ROI** for our customers **across 100s of use cases**.

## Mission

Unlock AI-driven **'Always Optimized' KPIs** for any organization



### Global Footprint

Presence across Asia, Europe and US with employees spread across **8 countries**



### Industry Validated

100s of success stories across within **Fortune 500 companies globally**



### Partner first DnA

**Partner-first organisation** with global reach with GSIs



### Enterprise Ecosystem



Azure



aws



Google Cloud



ChatGPT



Claude



Gemini

## Industrial IoT Use Cases

Select top-line and bottom-line impact generating use cases

### MINING



#### Fuel/Resource Efficiency

Detect shifts in fuel use and adjusts controls to minimize excess

#### Safety Incident Root Cause Analysis

Predict scenarios of safety incidents from historical data

### DISCRETE



#### OEE Root Cause Analysis

Track OEE issues continuously to address inefficiencies early

#### Operator Copilot

Guide operators to act effectively with automated root causes and SOPs

### CONTINUOUS



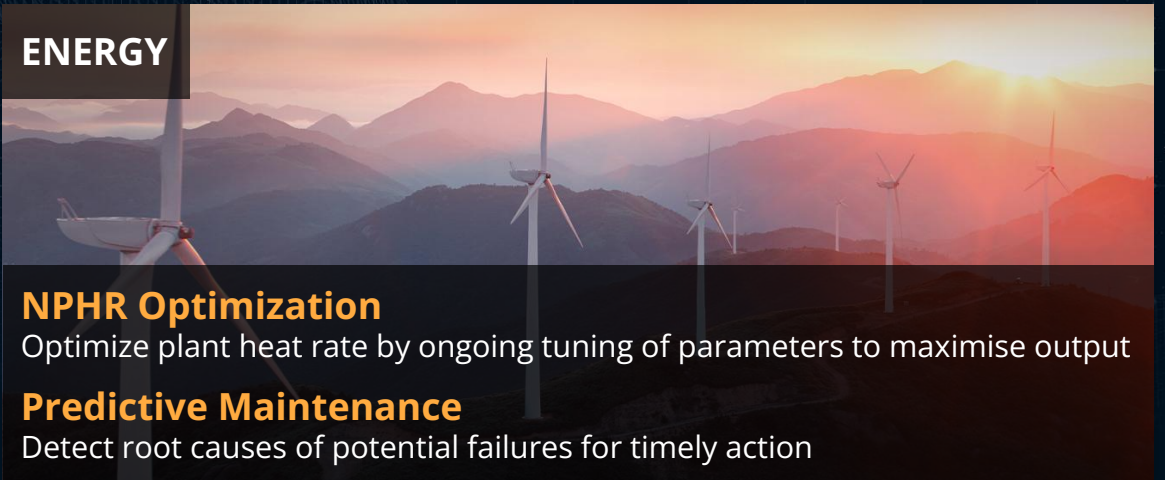
#### Yield Energy Throughput

Maximize yield continuously while minimizing input material/resource use

#### Adaptive QA

Monitor drivers of QA to continuously improve defect rates

### ENERGY



#### NPHR Optimization

Optimize plant heat rate by ongoing tuning of parameters to maximise output

#### Predictive Maintenance

Detect root causes of potential failures for timely action



# Our Technology

# Generative AI doesn't understand **YOUR** business.

For KPI optimization, AI must leverage knowledge from operational data

## Challenges of LLMs

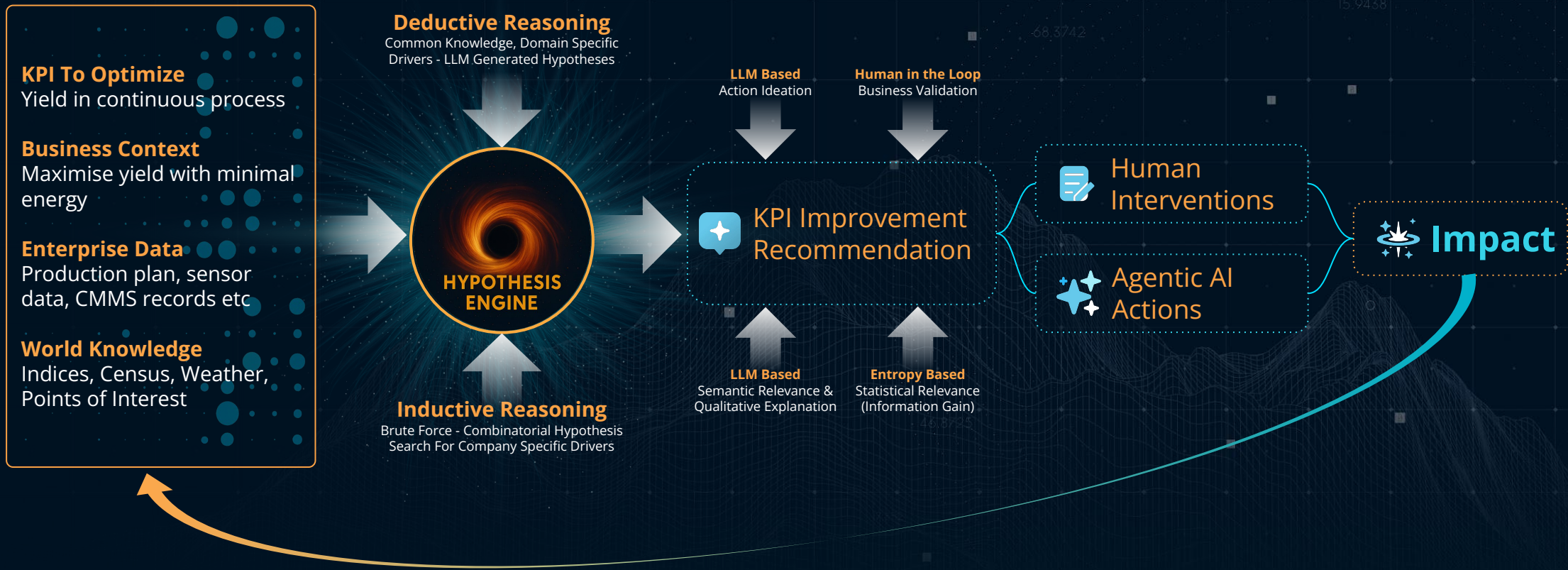
- Limited in understanding patterns hidden in complex operational data
- Unable to ground business reasoning in data.

## Unlocking LLM-powered KPI-optimization for solution-builders

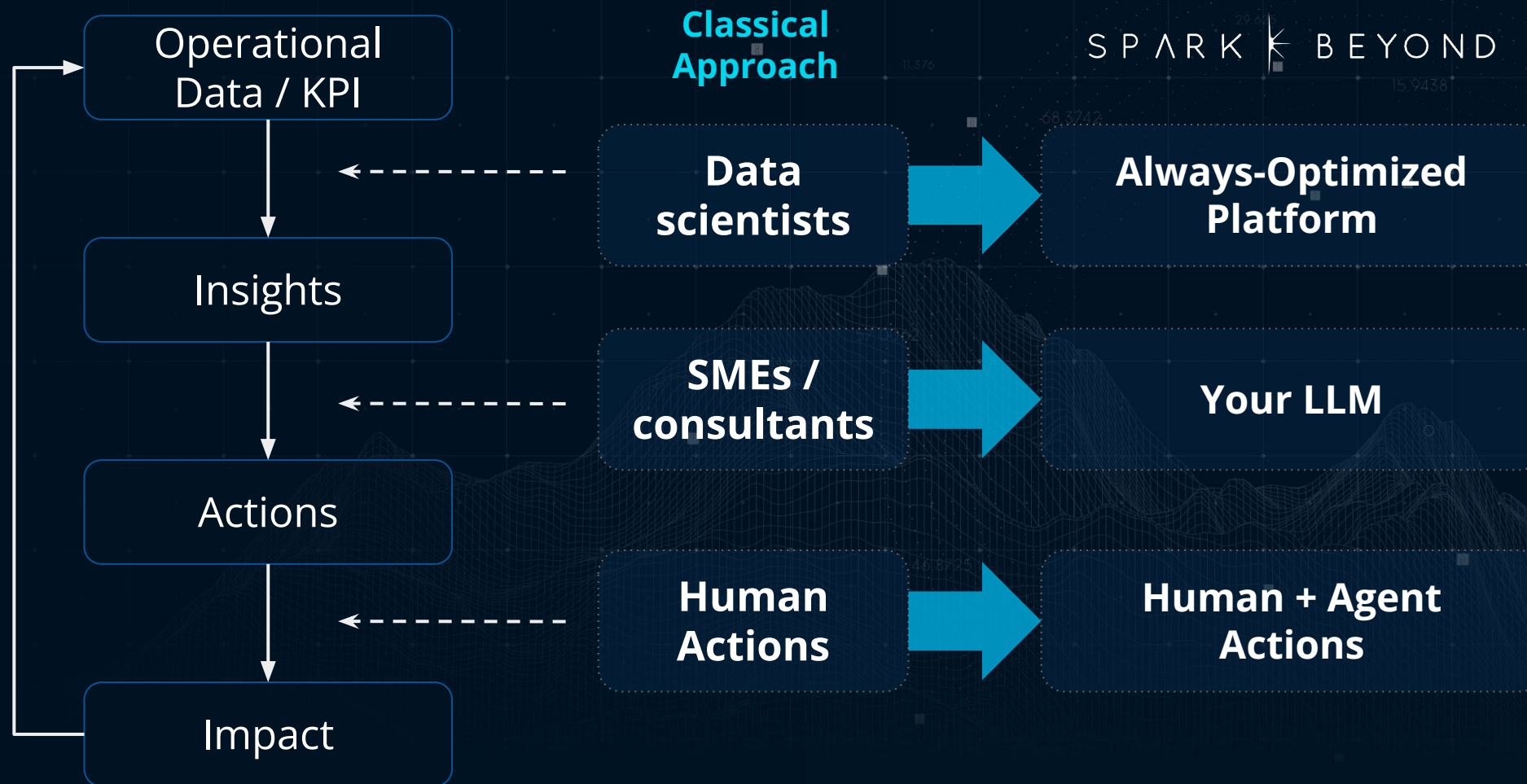


# 'Always Optimized' KPI Architecture

Continuous feedback loop creating impact from enterprise structured data



# Making the paradigm shift to 'Always Optimized' KPI Optimization



# AI CoE Platform

Accelerate Results - No Large Data Teams. No Consulting Roadshows.



## Discover & Prototype in Weeks

Bypass long discovery cycles. We'll help you rapidly prototype AI solutions that uncover the hidden drivers impacting your KPIs.



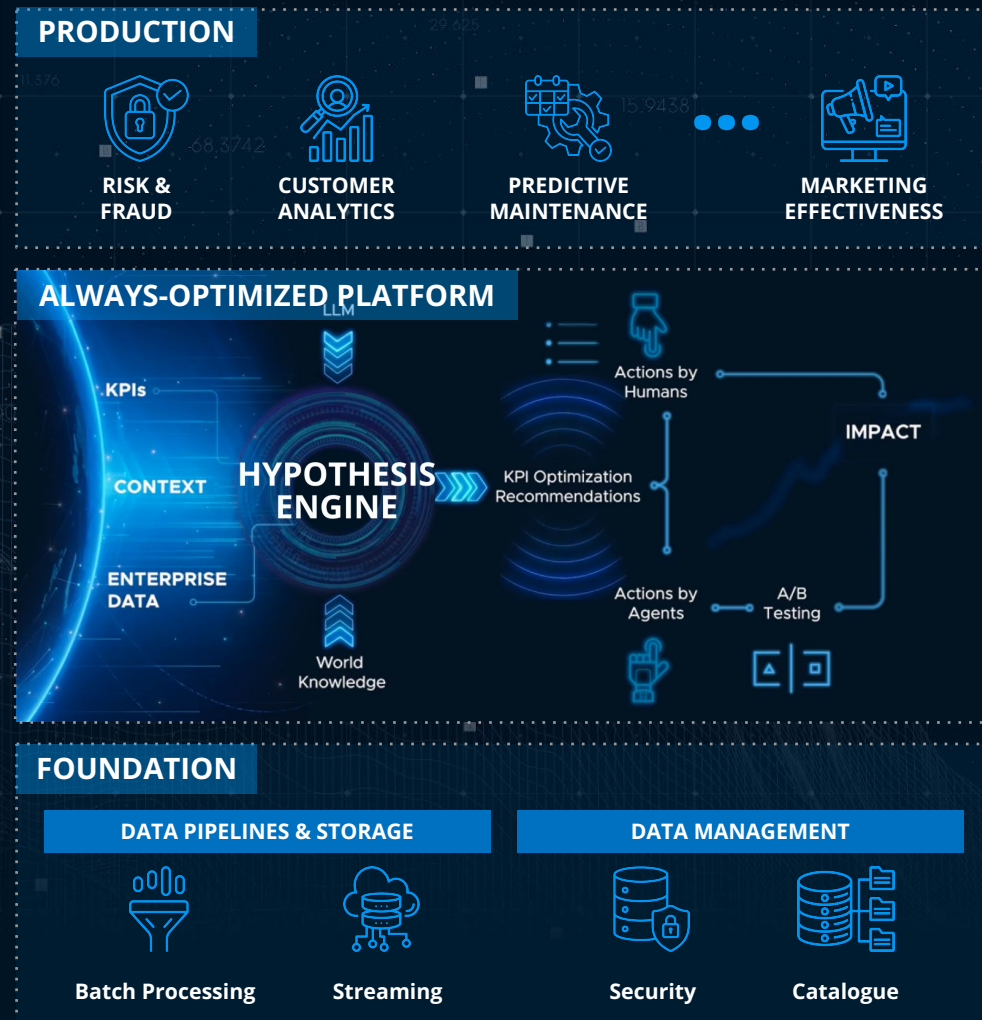
## Prove ROI Before You Scale

Validate the P&L impact with a targeted pilot, delivering a concrete business case and tangible value from day one.



## Scale to an "Always-Optimized" System

Empower a lean team to deploy and expand use cases, creating a continuous optimization loop for your business.



## Why SparkBeyond for Your AI CoE Platform

| Feature               | Impact                                                                                                                                                                                                                                                                    |
|-----------------------|---------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Hypothesis Engine     | Automatically discover and <b>engineer new features</b> from your operational and business data, surfacing <b>hidden drivers</b> of performance. Generate and test millions of hypotheses to identify true root causes and actionable levers for <b>KPI improvement</b> . |
| Explainable AI        | All insights and recommendations are delivered in clear, natural language—enabling business and operations leaders to <b>understand, trust, and act</b> on AI-driven findings. This transparency is critical for adoption and compliance.                                 |
| Open Integration      | SparkBeyond is cloud-agnostic and integrates seamlessly with leading hyper-scaler environments, CRMs, ERPs, and LLMs, ensuring <b>flexibility, scalability, and alignment</b> with your evolving tech stack.                                                              |
| Battle Tested @ Scale | The platform is <b>proven across 100+ Fortune 500 deployments</b> , with a track record of rapid time-to-value and measurable ROI.                                                                                                                                        |
| Unified ML & Gen-AI   | SparkBeyond is a <b>unique combination</b> of advanced machine learning and Gen-AI agent workflows providing a single foundation for a broad spectrum of AI use cases.                                                                                                    |
| Rapid Deployment      | Designed for <b>fast implementation</b> and operation in complex, distributed data environments—delivering actionable insights without requiring lengthy data warehouse projects.                                                                                         |

# Existing approaches to link LLMs to enterprise data are **insufficient to address structured data needs**

## Overview of current approaches (not-exhaustive)

### Pre-Training & Fine Tuning

#### What is it?

Pre-training a model on a selected corpus applicable to your enterprise domain  
Fine-tuning LLMs to answer domain specific questions

#### Limitations

- Expensive to re-train
- Does not address structured data sources
- Fine-tuning is better suited to teaching specialized tasks or styles and less reliable for factual recall.

### Retrieval Augmented Generation

#### What is it?

Retrieve data from outside a foundation model and augment your prompts by adding the relevant retrieved data in context

#### Limitations

- Structured data requires a query for RAG based solution to retrieve
- Retrieved query needs to be LLM compatible
- RAG is largely limited to searchable documents

### Code Interpretation & Generation

#### What is it?

LLM task to translate a query spoken in natural language into SQL/code automatically

#### Limitations

- User needs to define the intent and insights
- Path to using the insight in an LLM use case is several steps away for a user

### In-Context Learning

#### What is it?

One/few-shot learning example to gain new knowledge (e.g. feeding an existing ppt report about a quantitative analysis)

#### Limitations

- Context needs to be textual
- Context document can get easily outdated



# Credentials

## Boosting beverage production line efficiency with root cause analysis

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**Identified 13% potential OEE improvement through root cause analysis of production downtimes**



### CHALLENGE

- A leading global beverage manufacturer wanted to improve **Overall Equipment Effectiveness (OEE)** for its filling and packaging lines
- Despite having **Lean, Six Sigma, and excellence programs** in place, they sought to use data to better understand **root causes of downtime** and to **foresee breakdowns** proactively

### RESULTS

- Sparkbeyond autonomously pinpointed key root causes: volume output spikes, older product batches and failures at upstream machines
- Estimated **13% potential OEE improvement** from addressing these causes

### APPROACH

- Collected **performance data** on OEE shifts, outages, process orders, and machine status
- Identified **“built-back” events** as the biggest negative driver of OEE
- Applied predictive analytics to **identify root causes** and recommended targeted actions

# Achieving **fuel efficiency in mining** with up to 10% savings

SPARK  BEYOND

**Reduced fuel  
consumption by  
10% in 4 months  
across mining  
fleet**



## CHALLENGE

- High fuel costs (30% of OPEX) pressured a mining fleet to cut consumption
- Needed to identify controllable drivers across a fleet of 100+ vehicles to reduce operational expenses
- Manual reviews are costly and time consuming prevent fuel used to be optimized

## RESULTS

- Identified **~300 predictors** of fuel efficiency, including payload, tire pressure, and driver behavior
- Rolled out **dashboard monitoring** for key fuel drivers
- Achieved a **10% fuel consumption reduction** in 4 months, with ongoing optimization

## APPROACH

SparkBeyond created a digital twin for each vehicle using:

- IoT Vehicle Management System (30+ onboard sensor data points)
- Fleet Management System (trip data)
- SAP maintenance logs
- External sources (weather, contextual data)

Each digital twin integrated 130+ data points per truck

## Optimizing the Net Plant Heat Rate by 3.3% based on load conditions

SPARK  BEYOND

**Cutting fuel costs and emissions by optimizing plant heat rate with advanced analytics.**



### CHALLENGE

- Low efficiency due to boiler and combustion losses
- Heat rate varies with operational load
- High volume of sensor data from over 3000 sensors from PI system but low actionable insights
- Need to **reduce coal input costs and emissions**

### APPROACH

- Model NPHR by load condition levels (Low, Medium, High)
- Ran learning experiments to rank thousands of features
- Prioritized initiatives and actions via SME workshops
- Built RCA and ideal parameter setting dashboard for engineers

### RESULTS

- Identified 23 high-impact controllable drivers and predicted NPHR impact per driver
- Estimated **3.3% heat rate improvement** corresponding to **~3% reduction in CO<sub>2</sub> emissions**

## \$3m impact from optimizing alumina/caustic ratio in alumina refining

SPARK  BEYOND

**Reduced  
variability in  
the alumina  
refining  
process, saving  
\$1.8m per year**



### CHALLENGE

- Controlling the A/C ratio in the digester is critical to the alumina refining process and a major production driver
- However, the client was relying on an outdated predictive model, resulting in high variability throughout the process

### APPROACH

The team used SparkBeyond to improve model accuracy through:

- Analyzed internal sensor data from the existing DBO model
- An **iterative process** to identify optimal time windows and interactions between sensor readings
- Built several models, from **linear regression to advanced boosted trees**, to maximize impact

### RESULTS

- SparkBeyond identified **several ways to improve consistency** in the A/C ratio, with one improvement alone potentially driving over **\$3 million** in added production value
- The client implemented changes valued at **\$1.8 million annually**

# Reducing non-conformance rates in shaft annealing process for a bicycle manufacturer

Direct cost improvement from reducing scrap

SPARK ✦ BEYOND

**Identified adjustments for annealing process to reduce non conformance rates by 25%**

## CHALLENGE

- Automatically identify the root causes of **non-conformance** amongst a sample of **annealed metal parts**, augmenting engineering intuition of operators

## APPROACH

- Combined **visual inspection data** with **furnace sensor** in historian, **power consumption, nitrogen, oxygen** levels and **environmental factors** (room temp, humidity)
- Generated hypothesis and **identified root causes**

## RESULTS

- Identified **nitrogen pressure and oxygen content levels** during heating and cooling respectively, which when fixed, help reduce non-conformance rates from **29% to 4%**



# Proactive identification and resolution suggestion for industrial chiller trips at a bicycle manufacturing plant

Avoid throughput losses

SPARK  BEYOND

**Prevent temporary loss of plant capacity by proactively preventing chiller trips**



## CHALLENGE

- Greenfield site with new **industrial chiller system for cooling** tripped frequently and abruptly with no consistent explanation - each trip while lasts a short duration leads to **plant downtime**

## APPROACH

- Generate hypothesis from sensor and other data from **SCADA** and **operational systems**.
- Use LLMs to bring **engineering and physics knowledge** to boost insights.
- Use **LLMs + OEM manuals** to validate actions and recommendations.

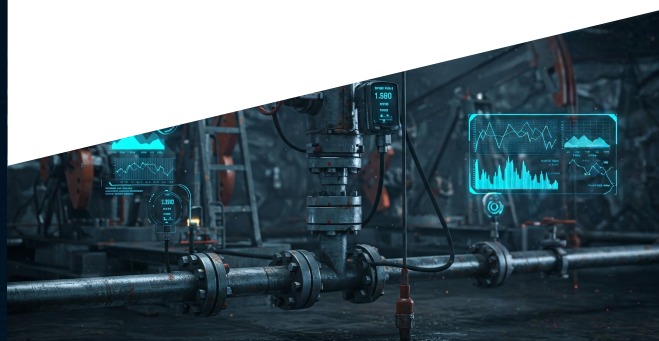
## RESULTS

- Developed a highly accurate predictive model with **85% recall and 80% precision** for identifying trips - \$ value for preventing temporary plant shutdown not known to SparkBeyond

## Predictive Maintenance of ESPs (Electrical Submersible Pumps)

SPARK BEYOND

**\$2M Impact per  
Early Warning  
for Key ESP  
Failures**



### CHALLENGE

- Client needed to anticipate and manage ESP failures to minimize production loss
- Built a model to predict the probability of ESP failure within the next 100 days
- **Data Sets Used:** Past ESP failures, sensor readings, well trajectories, coordinates, completions

### RESULTS

- **\$2M impact per early failure alert**
- Enabled proactive maintenance, minimizing downtime and production loss

### APPROACH

- Reframed task as remaining uptime prediction due to dataset imbalance
- Identified if an ESP is likely to fail within the next 100 days
- Used Discovery Platform with 8 datasets to uncover failure drivers
- Delivered insights as both code and natural language
- Provided daily SHAP-based predictions and explanations
- Outputs shared with maintenance teams to support preventive action

## Proactive Predictive Maintenance for PCP Pump Failures

SPARK  BEYOND

**30+ Days  
Predictive  
Horizon | >50%  
Failure  
Detection  
Accuracy |  
0-15% False  
Positives**



### CHALLENGE

- **Improve the life of wells** to increase production and reduce CapEx
- Build a predictive model despite incomplete data
- Integrate analytics with the client's **IT infrastructure**

### RESULTS

- Identified **>50% of system failures** with a 0-15% false positive rate
- Defined 9 predictive parameters contributing to model power
- Achieved a **30+ day predictive horizon** with additional system data

### APPROACH

- Focused on **Analytics & Insight and IT Integration**
- Built a live model to predict failures in PCP sub-systems from incomplete data
- Delivered an **integrated view of the IT infrastructure** and provided software improvement recommendations

## Predictive Maintenance for Bearing Failure in CMP wafer process

SPARK  BEYOND

**Advanced analytics predicted failures weeks ahead, improving uptime.**



### CHALLENGE

- **Short lead time** limited predictive capacity
- **Manual process** unable to anticipate failures in advance
- **8-hour downtime** affects all 4 chambers

### RESULTS

- **Earlier failure prediction** than manual process
- **Weeks of advance warning** enabled
- Improved **maintenance scheduling**
- Reduced **unplanned downtime**

### APPROACH

- **Connected data** from vibration and maintenance sources
- Used **APC system** to track sensor signals up to 0.5Hz
- Leveraged **Maintenance records** with free-text.
- **Analyzed observations** before bearing failures

# Optimizing wind farm production by continuous temperature adjustment

SPARK  BEYOND

**Improved wind farm throughput by 2% in just two weeks by managing temperature impact**

## CHALLENGE

- A large European energy company faced **unexpected asset downtime** across multiple wind farms, impacting production and IRRs
- Traditional methods were **missing key drivers** behind the downtime

## RESULTS

- SparkBeyond uncovered that **moving from cold to hot weather impacted turbine lubricant viscosity**, and reduced performance over time - so "temperature rate of change" was the most valuable metric
- Initiatives to control the rate of change led to a **2% improvement in throughput** in a two-weeks

## APPROACH

- Automated **time series analysis of sensor data** across various periods (12h, 3 days, 1 week, 1 month)
- Identified **temperature sensors** as critical to performance
- **Augmented internal data with weather datasets** to uncover new drivers

## Maximizing mean annual increment for a plantation

SPARK  BEYOND

**Built a predictive model explaining ~40% of variability in forest growth rates**



### CHALLENGE

- A forestry company in Asia wanted to better identify **predictors of tree growth**
- They struggled with **inconsistent data** from two different MAI measurement methods and sought to **develop a consistent methodology** while **discovering new opportunities** to boost growth rates

### RESULTS

- Identified human-controllable drivers of forest growth, validating expert knowledge
- Built an **elaborate model explaining ~40% of MAI variability**, even after excluding biological factors like tree diameter and height

### APPROACH

- Reduced data preprocessing time and generated **100+ predictive models** across two datasets, including time series models
- Identified **critical features** (e.g., text list variables) improving MAI predictions
- Detected gaps in maintenance and planting methods
- Paved the way for **adaptive and autonomous AI systems** to ensure quality results

# Advanced Analytics for Semiconductor Manufacturing

SPARK  BEYOND

**Boosted  
semiconductor  
line efficiency  
with >40% defect  
reduction and  
predictive  
insights.**



## CHALLENGE

- Identify and reduce production inefficiencies.
- Improve productivity in high-precision semiconductor environment.
- Address recurring issues with long idle times and CPU overload.
- Predict equipment failures to prevent downtime.

## APPROACH

- Analyzed 13 months of I/O data and **20+ parameters**.
- Used RCPS to identify **1,000+ statistical patterns** and top 15 root causes.
- Applied predictive maintenance using vibration sensor data.
- Built failure prediction models with 90% event coverage.

## RESULTS

- **40% reduction in yield detractors** in target workcenter.
- **5 root causes identified** with clear mitigation actions.
- **Predicted failures 4+ weeks** in advance with ~10% false positives.
- Boosted maintenance capacity and line stability.

# Reducing Particle Contamination in CVD Process

SPARK  BEYOND

**Eliminating root causes to reduce particle contamination in CVD.**



## CHALLENGE

- Particle defects drive downtime and wafer scrap.
- CVD issues cause **0.4% yield loss in production.**
- Root causes behind contamination remain unclear.

## APPROACH

- Combined top-down hypothesis testing with bottom-up modeling.
- Automated root cause analysis using process data.
- Tailored diagnostics for high-priority failure points.

## RESULTS

- **Uncovered key root causes** of particle defects.
- Defined actions to **reduce contamination levels.**
- Improved stability and production quality.

# Predicting Failures in Semiconductor Fabrication

SPARK  BEYOND

**Advanced analytics reduced failures and improved fab performance.**



## CHALLENGE

- **High data volume** from advanced fab instrumentation.
- **Limited insight** due to reliance on vendors.
- **Raw data underused** for performance improvement.
- Needed to **identify failure patterns** across systems.

## RESULTS

- **87% accuracy** in predicting failure events.
- **14% false positives**, improving reliability.
- **86% correct** predictions of normal operation.
- Enabled **data-driven operations** and maintenance.

## APPROACH

- **Integrated** and cleaned diverse data sources.
- **Modeled** key performance and failure drivers.
- **Extracted** explanatory variables from large datasets.
- **Validated** predictions and implemented improvements.

A person's hands are shown holding a glowing digital network of nodes and lines, symbolizing technology and connectivity. The background is dark with a grid pattern and various digital elements like arrows and data points. The text "Thank You" is prominently displayed in the center.

# Thank You

SPARK  BEYOND