

Transforming Your Business With Data and AI



James Kobielus
Senior Director Of Research
for Data Management



SPONSOR



databricks



STARDOG



JAMES KOBIELUS

Senior Research
Director,
Data Management
TDWI

Digital business transformation rides on a new generation of applications—automated, intelligent, adaptive, and self-optimizing--that are powered by data and artificial intelligence.

Data is central to digital transformation

- Data that makes processes more agile, automated, predictive, and contextual
- Data for running processes in the cloud
- Data for orchestrating processes 24x7
- Data for enforcing strong governance and control of complex processes
- Data for accelerating process insights
- Data for cross-domain sharing of process intelligence

How would you rate your organization's current data strategy for reaching each of the following digital transformation objectives?

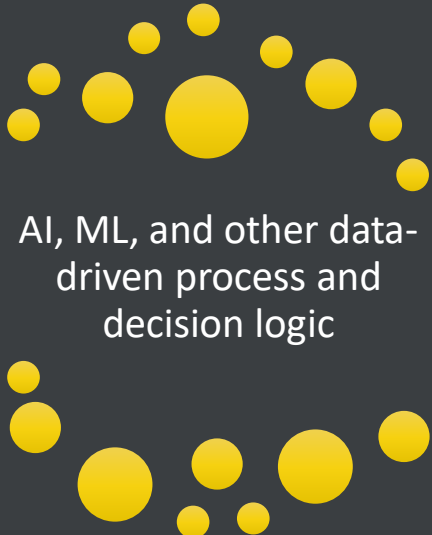
Very successful
Somewhat successful
Neither successful nor unsuccessful
Somewhat unsuccessful
Very unsuccessful
Don't know or N/A

Based on answers from 375 respondents. Ordered by combined "very successful" and "somewhat successful" responses.



Source: TDWI Best Practices Report "Maximizing Business Value with Data Platforms, Data Integration, and Data Management," Q3 2022

Data, analytics, & AI drive continuous transformation



Process automation



Autonomous operation



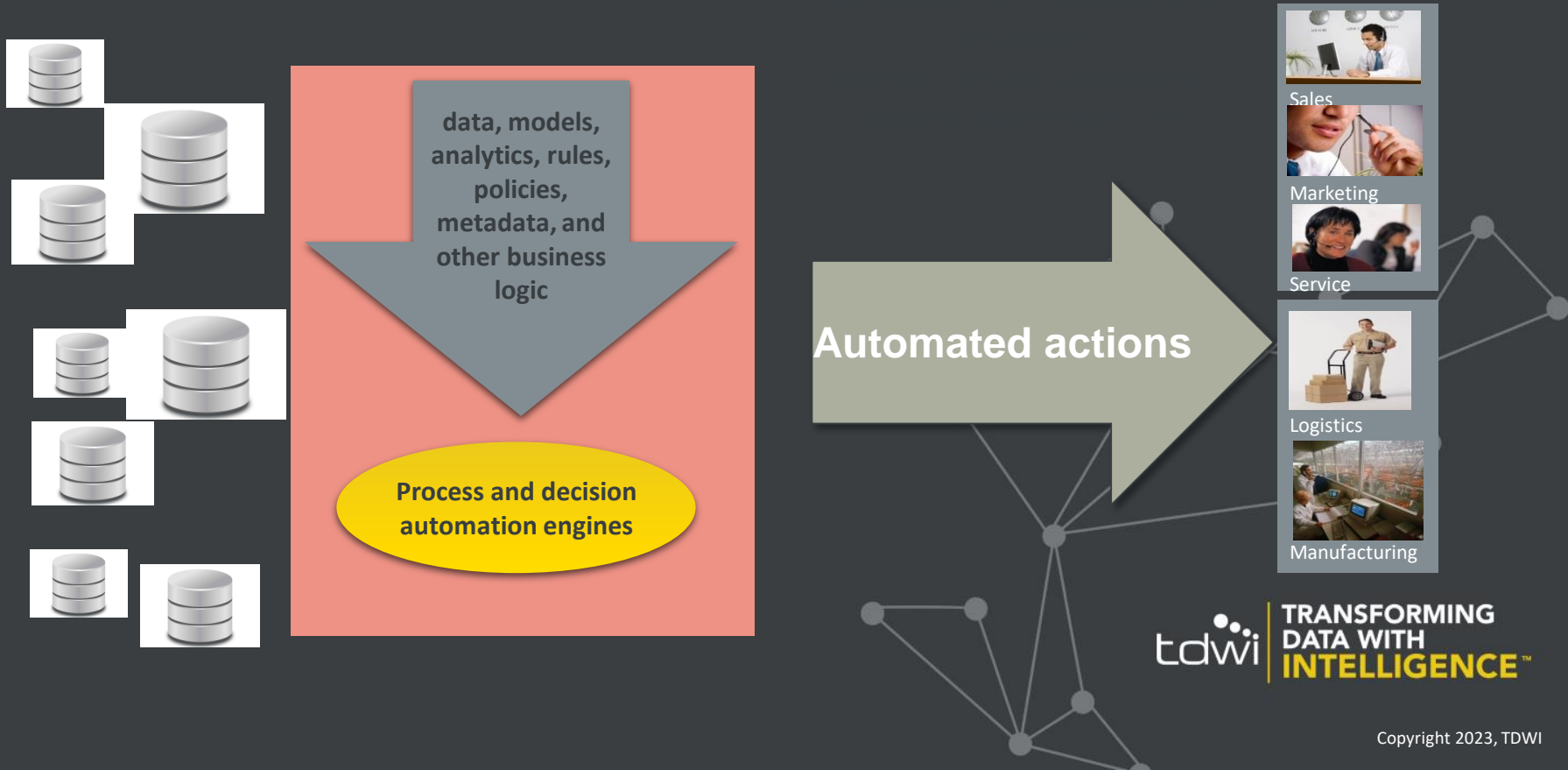
Sensor-driven actuation

SaaS, cloud, mobile, edge, embedded, IoT, and other application platforms

Augmented human decisioning

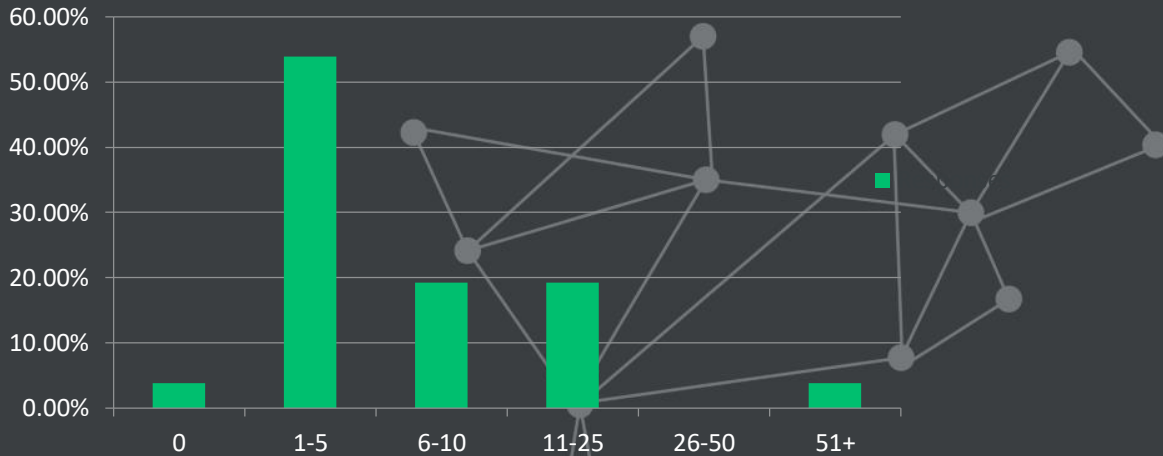


AI-driven transformation depends on adaptive automation



Enterprises are putting AI models to work in their organizations.

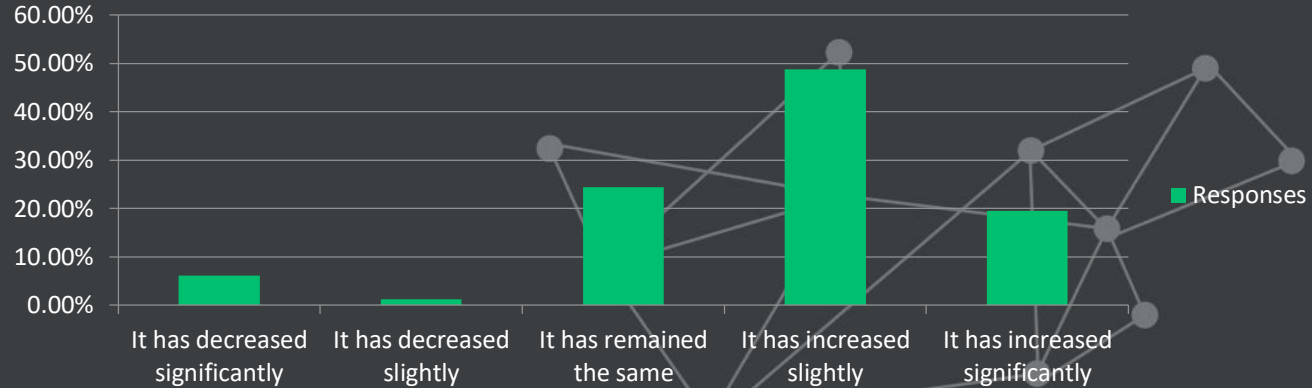
How many models (e.g., predictive analytics/machine learning) does your organization currently have in production?



Source: TDWI Data & Analytics Survey 2022

AI/ML will continue to grow in footprint in enterprise application environments.

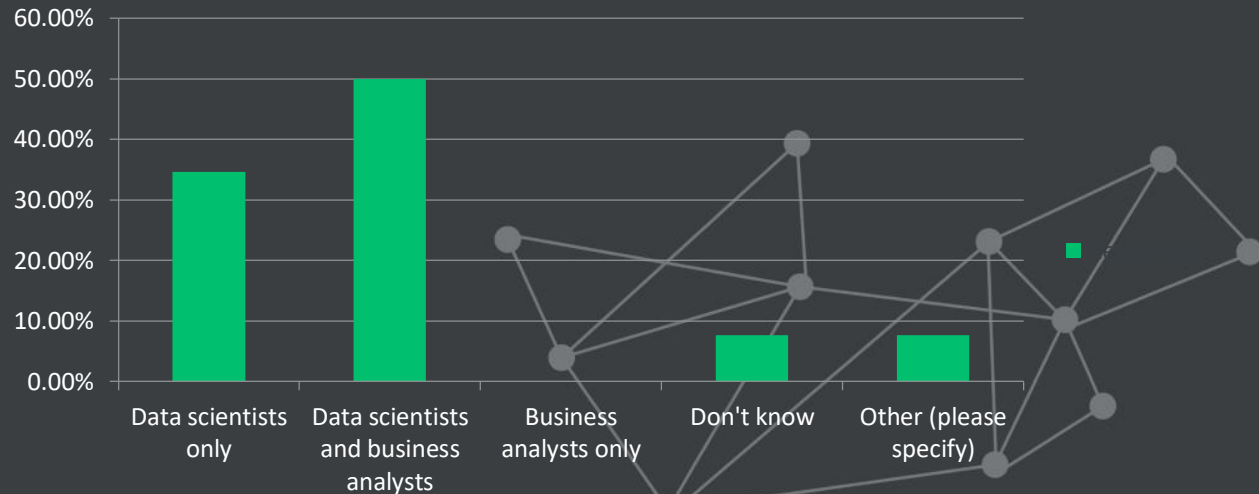
How has demand for more analytics such as machine learning or text analytics changed in your organization over the past year?



Source: TDWI Data & Analytics Survey 2022

Enterprises are relying on professional data scientists to develop AI/ML apps, often in collaboration with business analysts and subject matter experts.

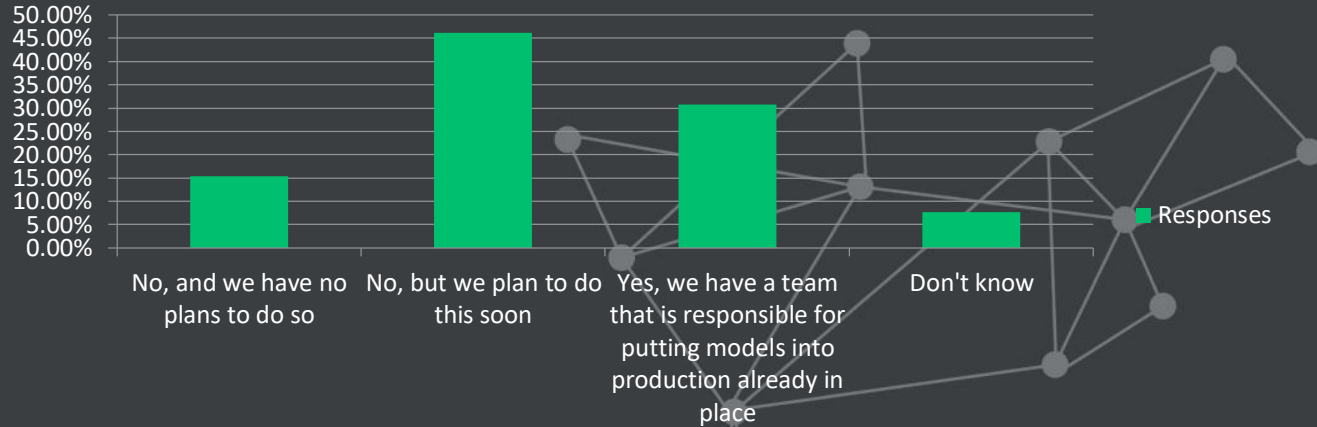
Who is building predictive analytics/machine learning models in your organization?



Source: TDWI Data & Analytics Survey 2022

Operationalizing the deployment and management of AI is a high enterprise priority.

Does your company have an MLOps team that deals with putting predictive analytics/machine learning models into production and monitoring them?



Source: TDWI Data & Analytics Survey 2022



BALA AMAVASAI

Global Technical Director
Databricks

6,000+
global employees

Inventor and pioneer
of the **data lakehouse**

Creator of

10,000+
Customers



databricks

The Lakehouse Company



\$3B
in investment

Gartner-recognized Leader
Cloud Database Management Systems
&
Data Science and Machine Learning Platforms



Lakehouse - one platform for Data and AI

All use cases + personas in one platform



SQL



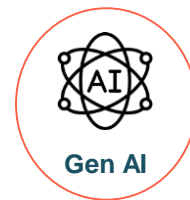
Orchestration



Streaming



ML



Gen AI

Secured data governance

Data in an open format to avoid lock-in

One copy of your data



Databricks helps the biggest manufacturing brands build data-driven businesses

Automotive



Electronics & Hi-Tech



Industrials



Aerospace



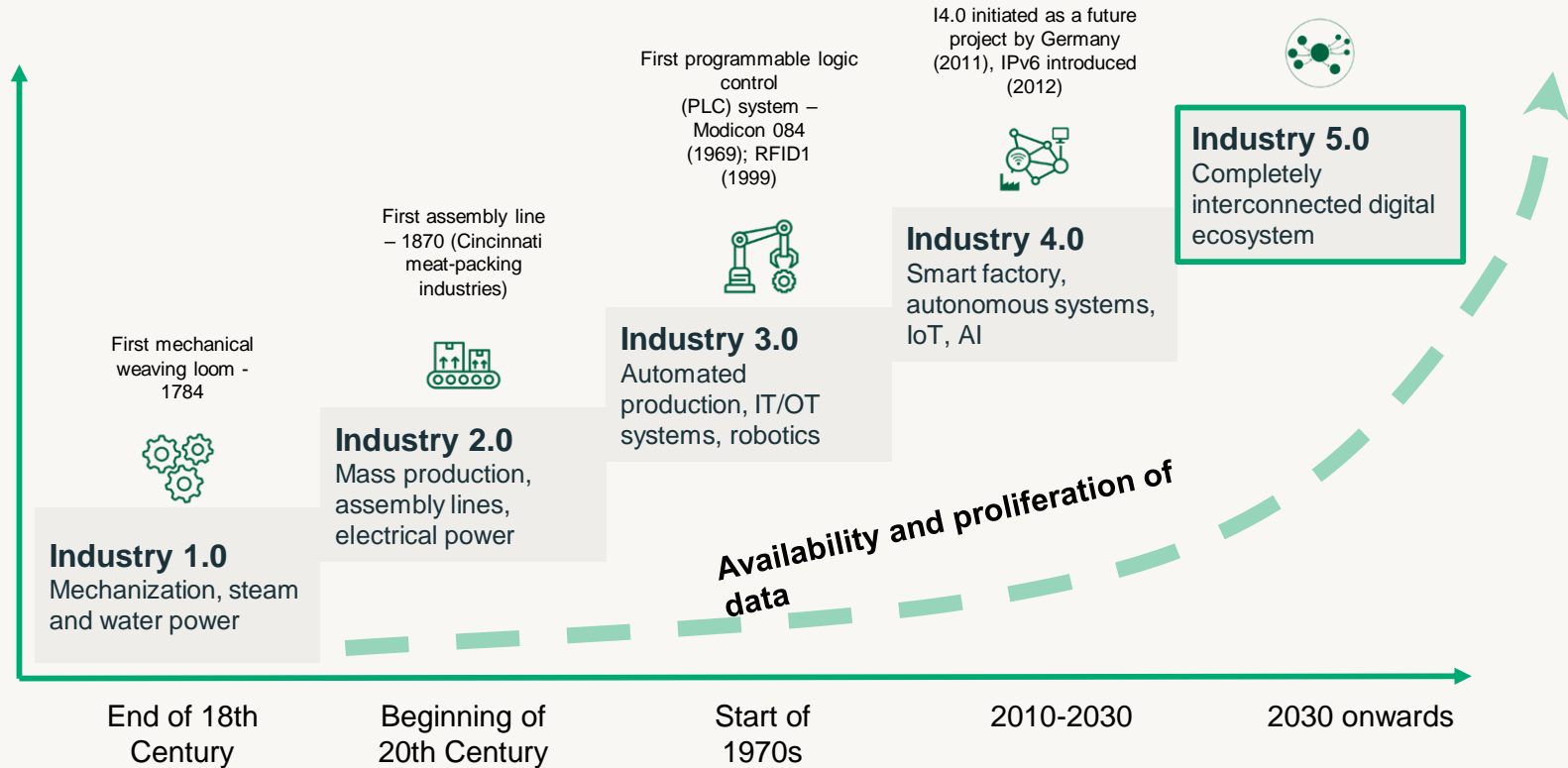
Transportation & Logistics



Basic, Specialty & Finished



Accelerating Manufacturing towards Industry 4.0 and beyond





Databricks enables analytics processing of large-scale data using Apache Spark™, Delta, and MLflow.

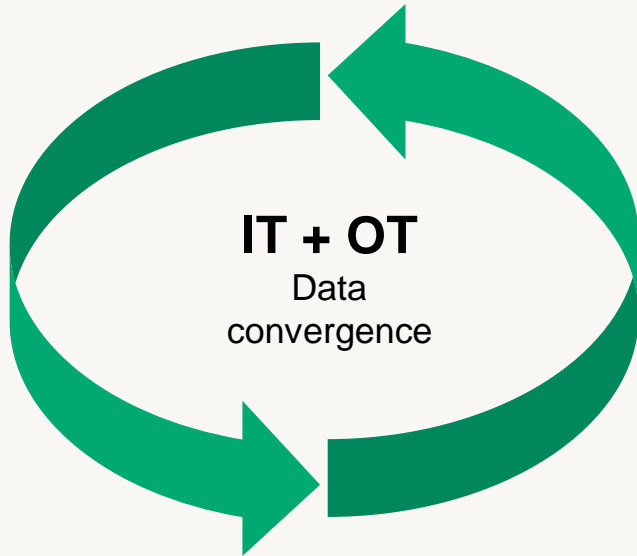
- Best-in-class support from original creators of Spark
- 10-100x faster than Open Source Spark when using Databricks as ETL engine
- Lowest TCO through auto-scaling and auto-configuration capabilities
- Delta Lake provides the ability to specify your schema and enforce it, making it ready for analytics at petabyte scale

The Stardog Enterprise Knowledge Graph platform unifies data based on its meaning, powering a semantic data layer to unleash valuable insights and accelerate your investment in a Databricks lakehouse.

- Richest, most flexible semantic layer
- Most advanced graph data virtualization
- Easiest to develop and reuse data models
- Most intuitive search and discovery of data insights
- Designed for enterprise data and analytics work

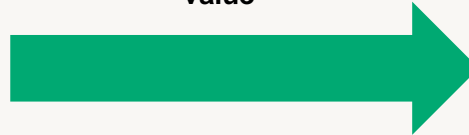


Democratizing Data and AI across the Manufacturing stack



Data from ERP, MES, SCADA, PLCs & Sensors

Data democratization
unlocks business
value



Monitoring and Analytics







Tableau PowerBI



Redash Qlik

Data Products



Key Use Cases

Asset & Operations Optimization

- Predictive Maintenance with Logistics
- Production Line Optimization
- Overall Equipment Effectiveness
- Digital Twin

Production Control

- Yield Optimization
- Automated Quality Control
- Automated Visual Inspection
- Process Control

Supply Chain & Inventory Management

- Inventory Planning & Optimization
- Material Tracking
- Part Forecasting
- Supply Chain Visibility

Demand Planning

- Demand Forecasting
- Demand Sensing
- Inventory Optimization

Finance

- Reporting: P&L, Profitability, Balance Sheet
- Forecasting: Revenue, Cashflow





AL BAKER

VP, Enterprise Solutions
Stardog

Stardog

Company Overview

Unite Data, Unleash Insight.

Stardog's **Enterprise Knowledge Graph platform** connects data based on business meaning into a flexible, reusable semantic data layer to get better insight faster.

We help **customers across many industries** empower data citizens to make knowledge-informed decisions.

“Our primary objective is to provide data at a higher quality and relieve the heavy lifting up front so our data scientists can actually work with the data.”

— Head of IT Research, Top Global Pharma

Select Stardog Customers:



BOSCH

Schneider
Electric



EXFO



Boehringer
Ingelheim



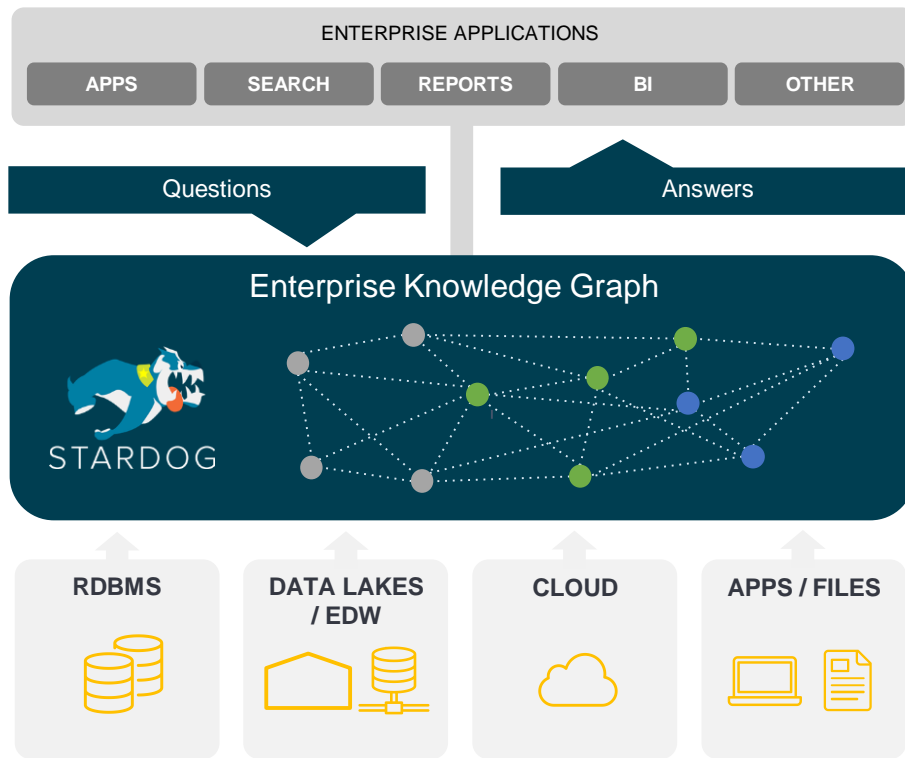
CIS Center for
Internet Security
Creating Confidence in the Connected World.



What is an Enterprise Knowledge Graph?

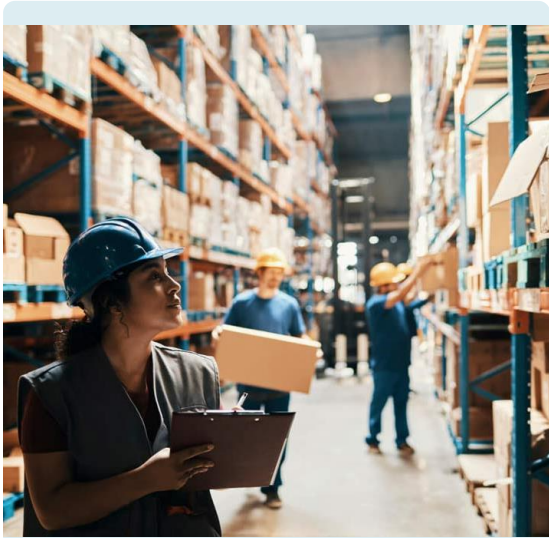
A flexible, semantic data layer for answering complex queries across data silos.

- **Connects** any data source / location and **virtualizes** access
- **Enriches** real-world context into data
- **Infers** new relationships, patterns and insights in data
- **Semantically searches** data by meaning

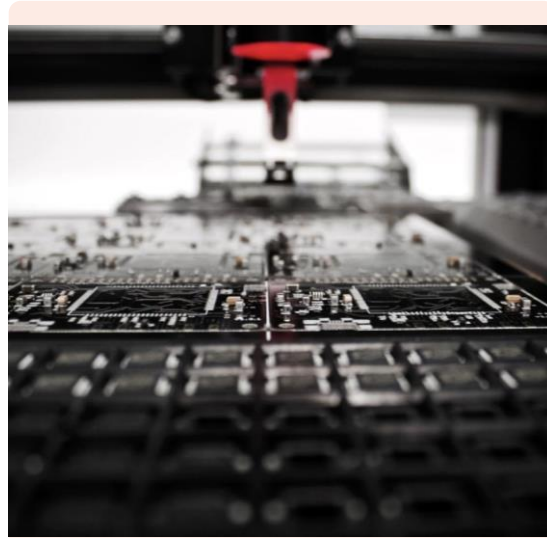


How Industrial Organizations Use Stardog

Supply Chain & Logistics

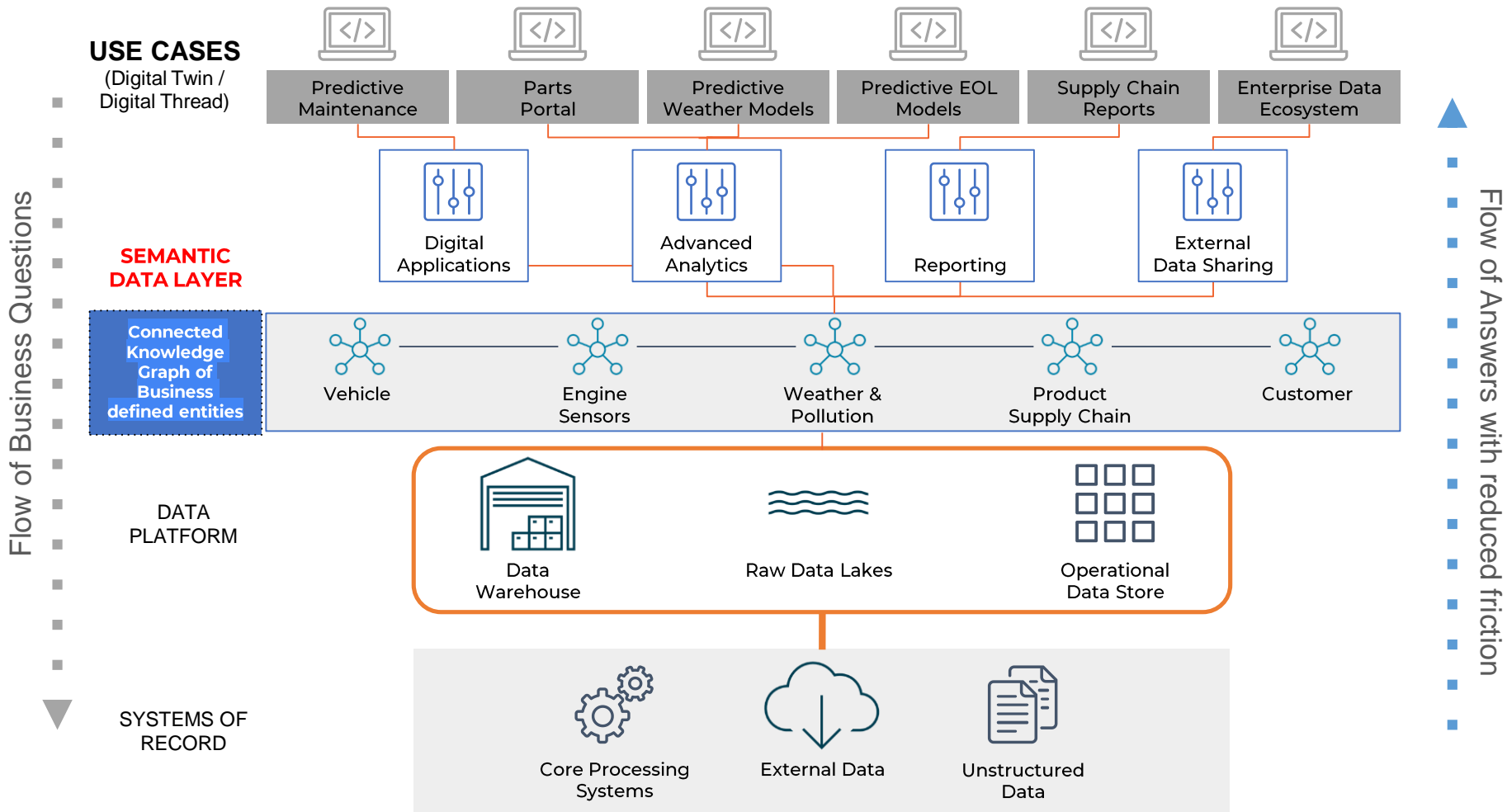


Product 360



Modernize MBSE





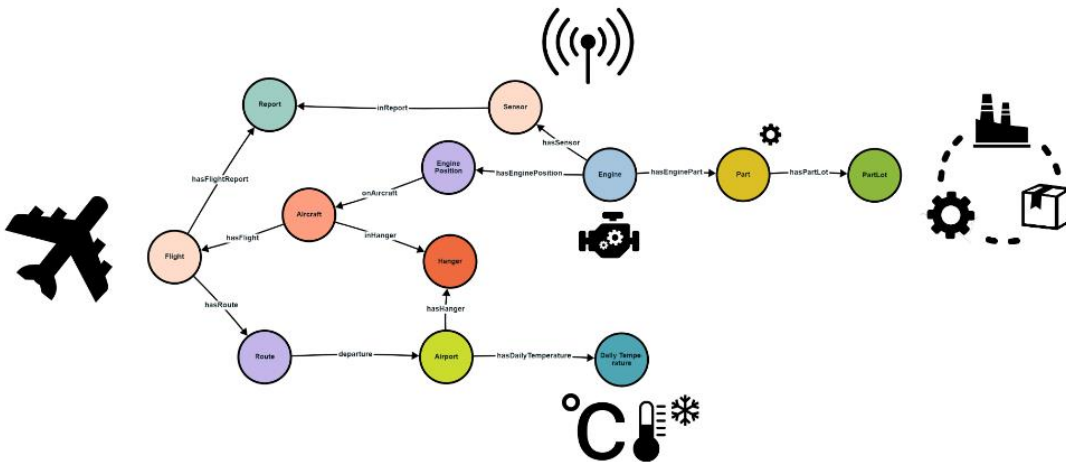
Example: Root Cause Analysis

Markdown    

Root cause analysis from flight to factory

Analysts at the Acme Aircraft Company have noticed an uptick in warning alerts triggered by engine sensors, but it's not immediately clear why only some engines of the same model are affected or why the effect is stronger on some flights than others.

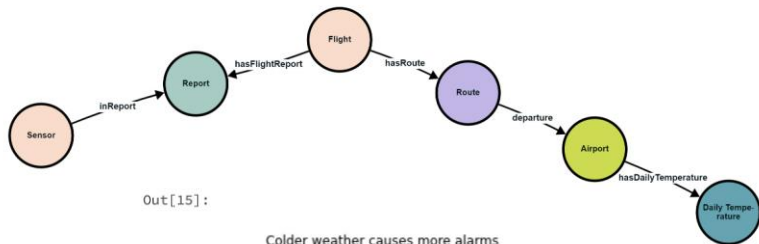
In this demo, we'll use a knowledge graph to discover what common factors are associated with the warnings and track the cause all the way back to a specific badly manufactured lot of a component used in the engines.



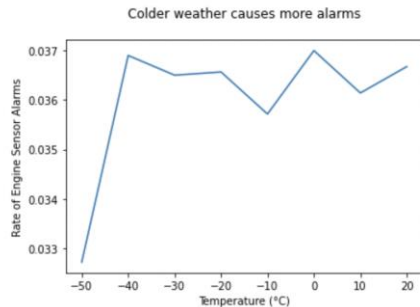
Example: Databricks and Stardog Notebook

Follow a pilot's hunch: does temperature influence the rate of sensor alarms?

To find out if the pilot's hunch was correct, we'll follow a path through the knowledge graph to connect sensor data with weather reports at the departure airport.



Out[15]:



```
1 q = """
2 select ?bad_part_serial_no ?engine_serial_no ?position ?aircraft_tail_no {
3   ?part acme:hasPartLot ?partLot ;
4   | acme:partSN ?bad_part_serial_no .
5   values ?partLot {acme:lot2b-22}
6   ?engine acme:hasEnginePart ?part ;
7   | acme:engineSN ?engine_serial_no ;
8   | acme:hasEnginePosition ?pos .
9   ?pos acme:onAircraft ?aircraft ;
10  | acme:position ?position .
11  ?aircraft acme:tailNumber ?aircraft_tail_no.
12 }
13 ORDER BY ?aircraft_tail_no
14 """
15 bad_parts_df = sg.query(q)
16 bad_parts_df.head()
```

Out[14]:

	bad_part_serial_no	engine_serial_no	position	aircraft_tail_no
0	P2B-ph5	E-a8f	Tail	B-af
1	P2B-no9	E-h8o	Left	B-ag
2	P2B-tn8	E-k5v	Tail	B-bc
3	P2B-ix9	E-l8v	Right	B-cw
4	P2B-dg8	E-p4a	Left	B-dq

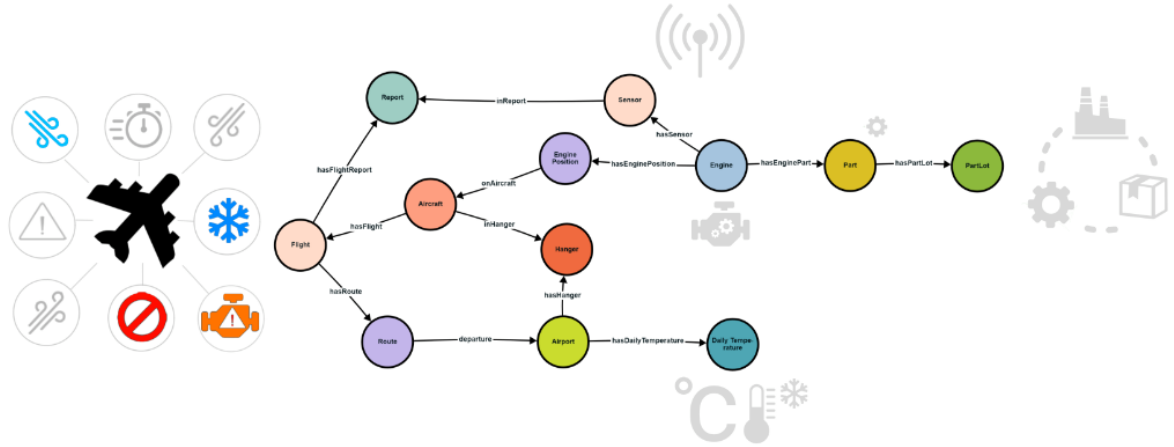
Data Analysis:

Data Segmentation
into Knowledge for
accelerating ML

Inferring Classifications for Different Types of Flights

The engines that Acme manufactures are expected to perform differently under different flight conditions

In this demo, we'll use a knowledge graph to classify types of flights and engine performance according to significant features



Our data scientists need the ability to classify flights or flight segments into analytical categories, for example:

- cold / warm weather
- headwind / tailwind / crosswind
- high / low altitude
- long / short duration
- long / short distance

Our goal: Formalize our technical and business logic as inference rules to classify flights

- Add inference rules to our ontology that apply flight classifications and helpful shortcuts
- Create a chain of inferences for a cascading effect that simplifies what would otherwise be complex queries

Using Datalog with Stardog and Databricks

Rule Detail

Rule Name
flown with part from lot

When the following conditions are met:

Flight — hasFlightReport → Report

Report ← inReport — Sensor

Sensor — fromEngine → Engine

Engine — hasEnginePart → Part

Part — hasPartLot → PartLot

AND

Then infer:

Flight — flown with part from lot → PartLot

Delete Cancel Save

The simple query below has to do a lot!

It will...

- follow a hierarchy linking flight -> aircraft -> engines -> engine parts -> manufacturing lots
- coordinate that to another hierarchy of flight -> report -> engine -> sensor -> alarms
- calculate a ratio of observed alarms per minute of flight for each manufactured lot of each component part in the aircraft engines
- compare that ratio to a predetermined parameter
- classify the manufacturing lot according to whether the observed rate of alarms exceeded the parameter
- classify the flight by whether the aircraft contains a part from a bad lot
- cross-reference flight departure dates against each airport's daily weather data to classify especially cold weather flights
- and finally, classify the flight as `Grounded` if it both contains a bad part and is scheduled for takeoff on a very cold day

Now let's run the query.

Cmd 18

```
1 q = ""select ?flight { ?flight a <http://acme.stardog.com/Grounded_Flight> } limit 10""
2 df = sg.query(q, reasoning = True)
3 df.head(5)
```

Data Analysis:

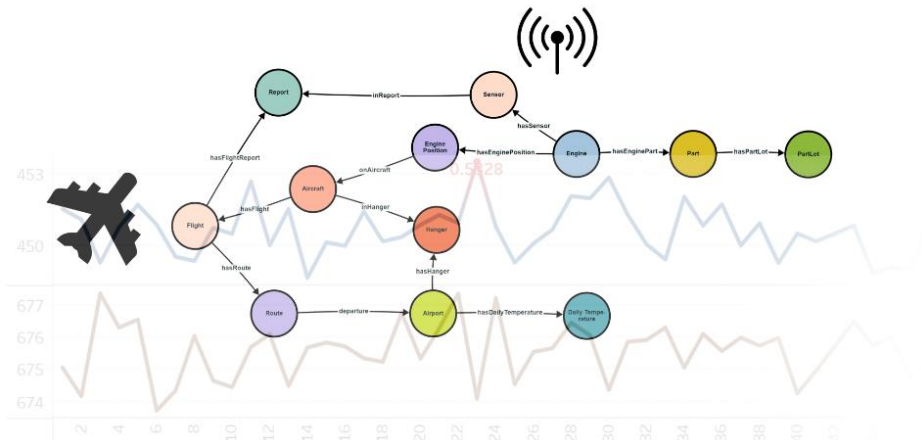
Use Knowledge Graph
logic over Time Series
tables in Databricks

Time Series Analysis on Aircraft Engine Sensor Data

Markdown

Data scientists on Acme Aircraft's quality assurance team regularly analyze real world in-flight sensor data to improve the precision of design specifications for various types of flights.

In this demo, we'll show how to use a knowledge graph to record, analyze, and interpret time series data.

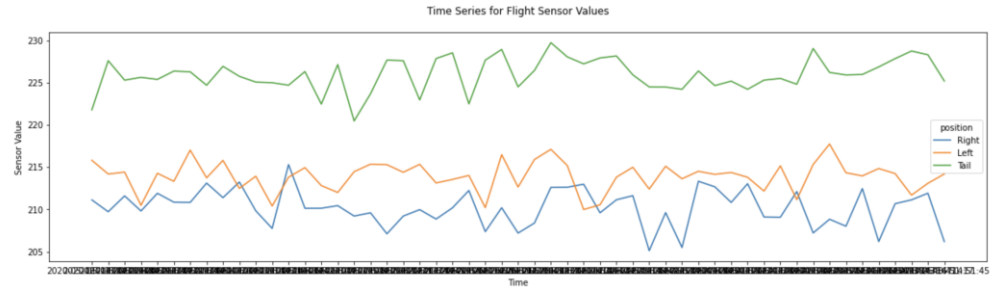
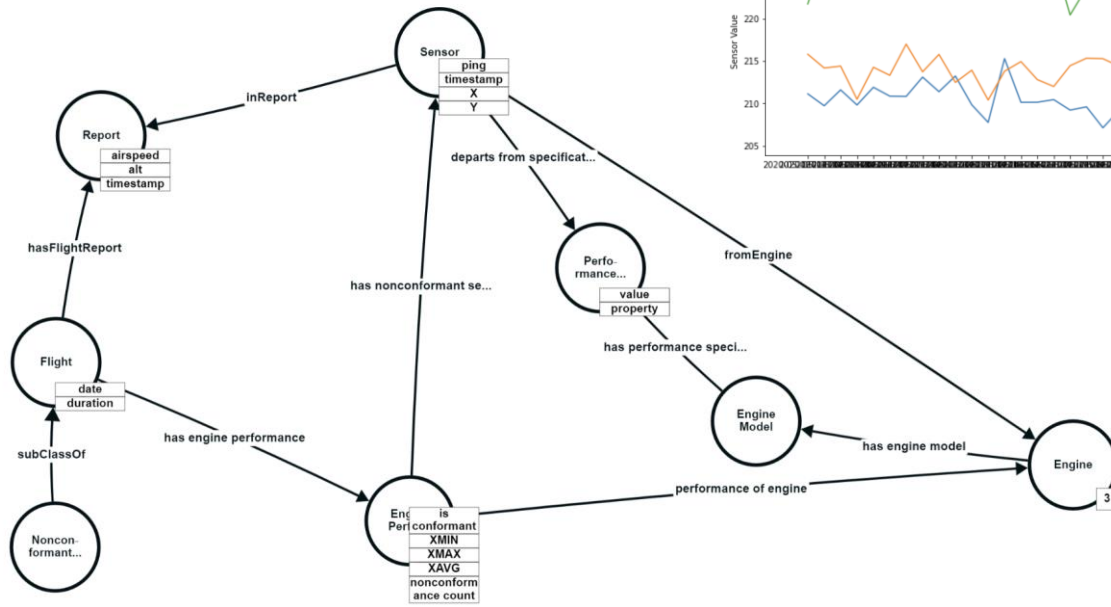


Scenario - Discover what patterns the sensor data can reveal about in-flight engine performance for different types of flights

- Acme corporation manufactures aircraft and publishes expected performance metrics in a design specification document
- Sensors in the engines record measurements at every minute of a flight
- The sensor data can be compared with the expected metrics to evaluate real-world engine performance

Our goal: Analyze time series data in the graph to determine whether engine performance was within the parameters of Acme's design specifications or was nonconformant

Graph Queries over Time Series



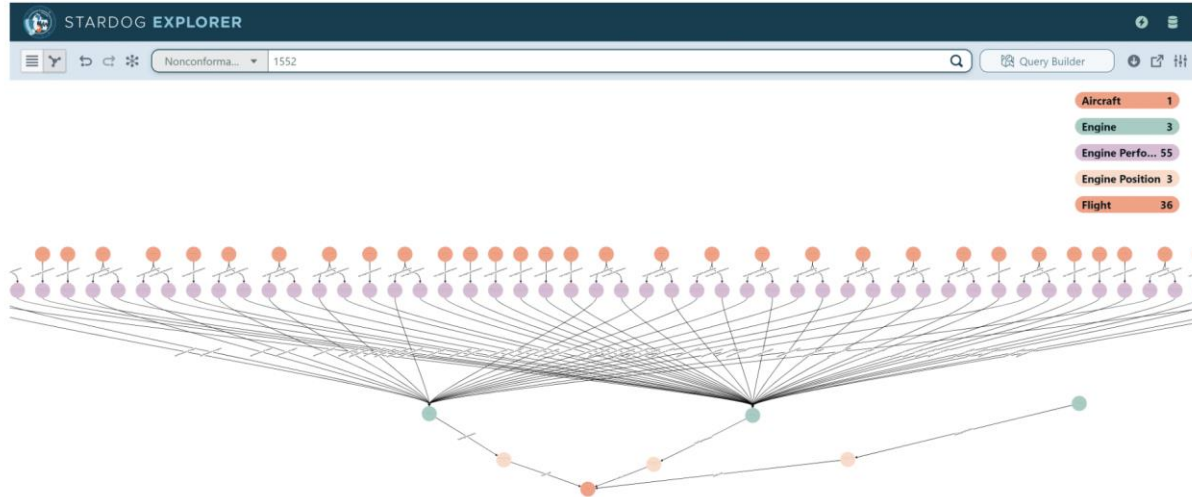
Visualize Graph Results in Stardog Explorer

The screenshot displays the Stardog Explorer interface. On the left, a graph visualization shows a network of nodes and relationships. A central node, 'Engine_Performance_engine_E-n5s_flight_36', is highlighted. To its right, a detailed view of this node is shown, including a list of related sensor IDs and various performance metrics.

Engine Performance	
sensor_36_E-n5s_16	
sensor_36_E-n5s_18	
sensor_36_E-n5s_23	
sensor_36_E-n5s_25	
sensor_36_E-n5s_29	
sensor_36_E-n5s_30	
sensor_36_E-n5s_32	
sensor_36_E-n5s_35	
is conformant	false
lot total alarms	0
nonconformance count	19
performance of engine	E-n5s
type	Engine Performance
XAVG	227.562547459488852105263158
XMAX	230.01032951706765
XMIN	226.61754772429546

Visualize Graph Results in Stardog Explorer

...Or visualize other important information, for example that this aircraft flew 36 flights with 2/3 of its engines in nonconformance!



Quick-start Knowledge Kits



Find this kit at:
<https://cloud.stardog.com/kits>



[Aviation] Flight Planning

Demo

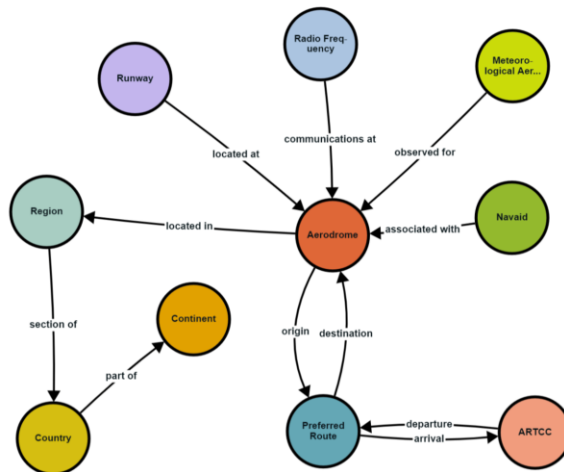
An aviation focused dataset and flight planning demo

Overview

This is a knowledge graph focused on flight planning for commercial aviation. You will see concepts in this graph like `Aerodrome` and `Runway`.

The focus of this graph involves concepts that would be required for planning a flight between two airports. Typically, prior to flight, pilots file flight plans, monitor the weather and make notes of active runways & radio frequencies.

Pictured below is the basic data model for this use case:



What is Industry 4.0?

How does Industry 4.0 support
manufacturers' digital transformation
initiatives?

What does it mean for manufacturing organizations to leverage data-driven insights to open up new business opportunities and accelerate positive outcomes within their supply chains?

How does digital transformation of manufacturers' supply chains depend on enterprise investments in modern data and AI technologies?

How can manufacturers boost the return on their digital transformation initiatives through investments in customer 360, semantic knowledge layers, predictive maintenance, edge computing, digital twins, the Internet of Things, and other pillars of “Industry 4.0”?



How is the payoff from investments in digital transformation boosted by providing self-service access to data and analytics for everyone in the organization regardless of their role or technical experience?

Questions?



CONTACT INFORMATION

If you have further questions or comments:

James Kobiellus, TDWI

jkobiellus@tdwi.org

Bala Amavasai, Databricks

bala.amavasai@databricks.com

Al Baker, Stardog

al.baker@stardog.com

SPONSOR



databricks



STARDOG

Thank you for attending!

