


Enterprise-Scale Machine Learning with IBM Db2 v11.5.4

Shaikh Quader, *AI Architect, IBM Db2*

 LinkedIn: [/in/shaikhquader](https://www.linkedin.com/company/shaikhquader)

 Medium: [@shaikhquader](https://medium.com/@shaikhquader)

 Twitter: [@ShaikhQuader](https://twitter.com/ShaiKhQuader)

66%

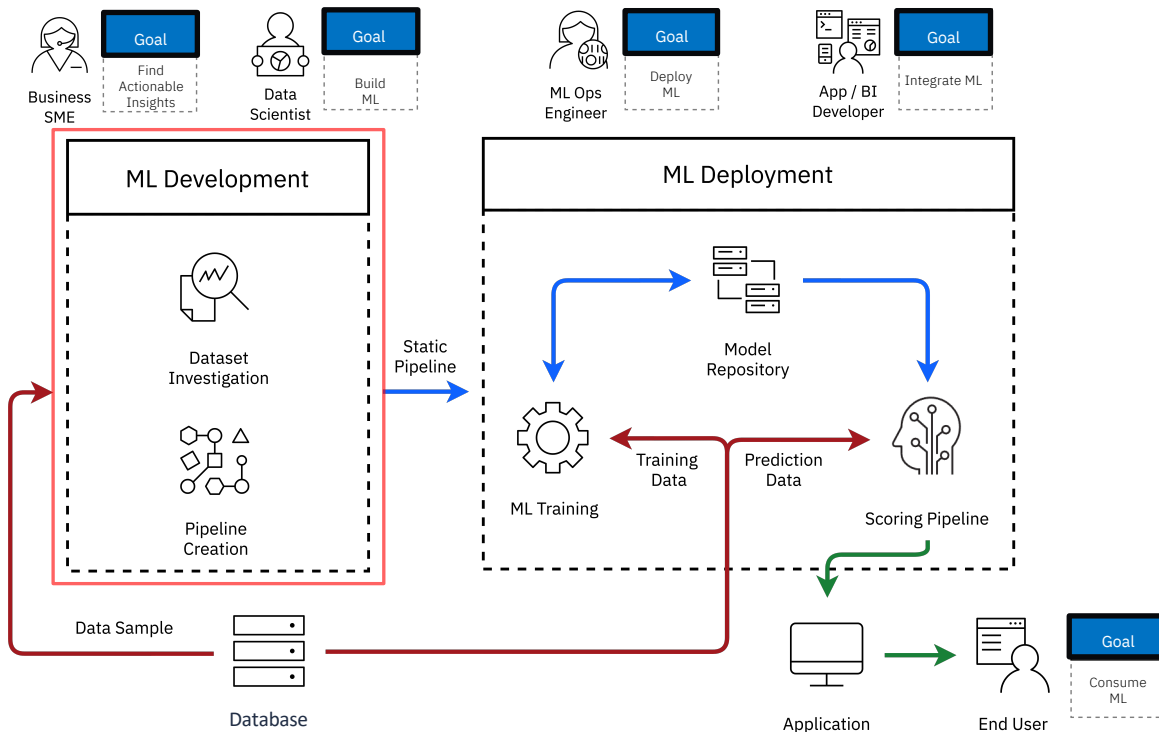
ML projects use Relational data

Source: The State of Data Science & Machine Learning 2017, Kaggle, October 2017
(based on 2017 Kaggle survey of 16,000 ML practitioners)

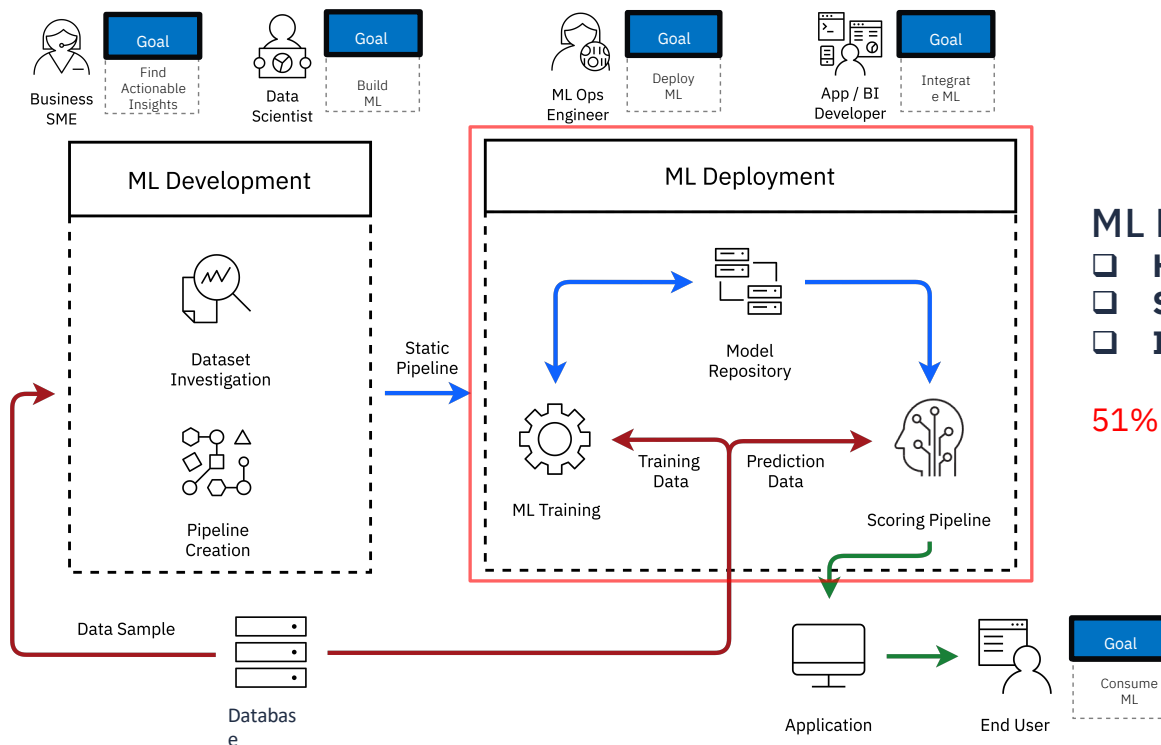
Machine Learning Workflow in an Enterprise

ML Dev Challenges:

- ☐ **Regulated data** (e.g., GDPR)
- ☐ **Data Volume** (data >> Dev env)
- ☐ **Noisy data** (25% data are noisy)



Machine Learning Workflow in an Enterprise



ML Deployment Challenges:

- ☐ Hosting
- ☐ Speed
- ☐ Integration

51% AI projects don't go beyond experiments

Accelerating and Optimizing AI Lifecycle with IBM Db2

01

Integrating Open-source Models with Db2

02

Developing and Deploying Db2-Native ML Models

Solution 1:

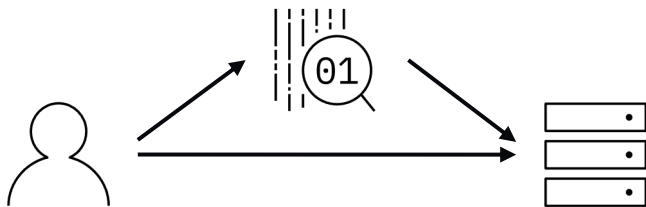
Bring your Open-Source Models to Db2

While the models may be centrally trained, the resulting inference pipelines will be deployed everywhere...to make inferences (“scoring”) **where the data is.**

Agrawal et al.

Cloudy with high chance of DBMS: A 10-year prediction for Enterprise-Grade ML (CIDR '20)

Use Cases for in-database Scoring

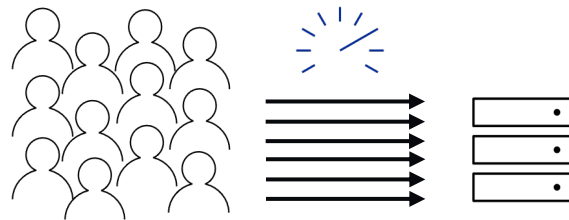


Latency-sensitive Decisions

Instantaneous predictions

Examples:

- Payment processing
- Fraud detection
- Loan/claim pre-approval



Large Batch Predictions

Real-time prediction using “fresh” and large operational data

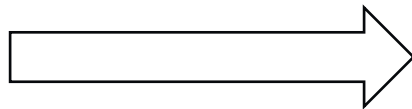
Examples:

- Anomaly detection
- Escalation risk prediction
- Dynamic price optimization

Python UDF: Scoring Python Models via Db2



Watson
or
Open Source



Export the ML pipeline by
serializing *python joblib*

Db2 Server

Host OS

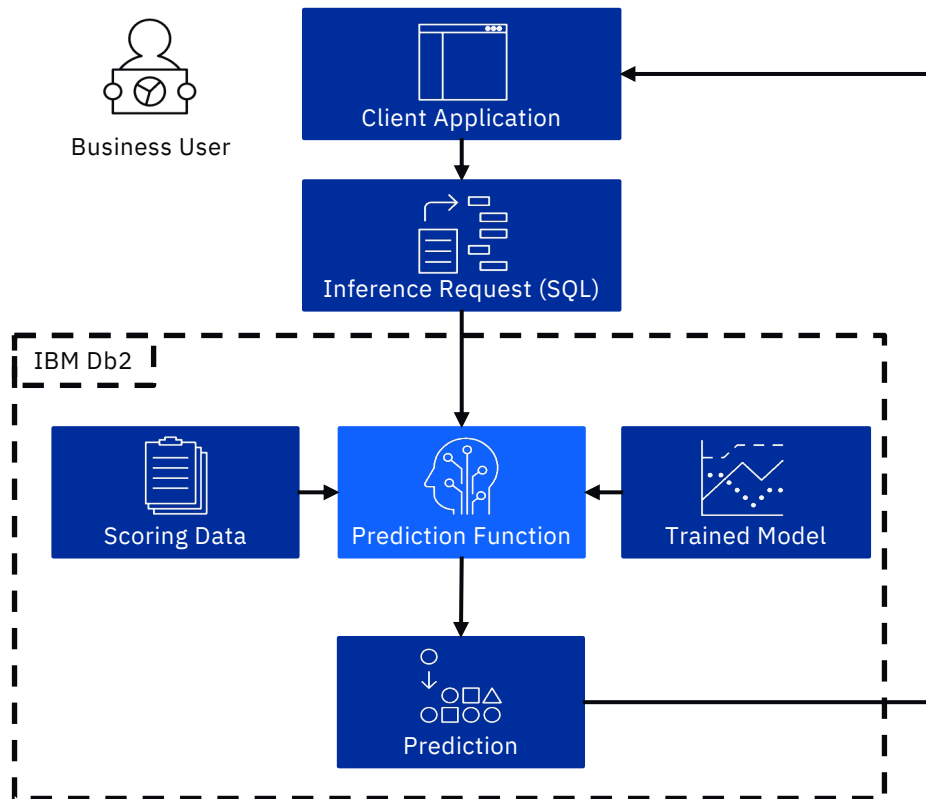
Db2 Instance

Python
Runtime

In-db Inferencing

Benefits:

- ML Infrastructure
- Low-latency
- High-throughput
- Simpler Integration



5x Speed up of Inferencing with IBM Db2

Model & Dataset

Logistic Regression (scikit-learn) Model

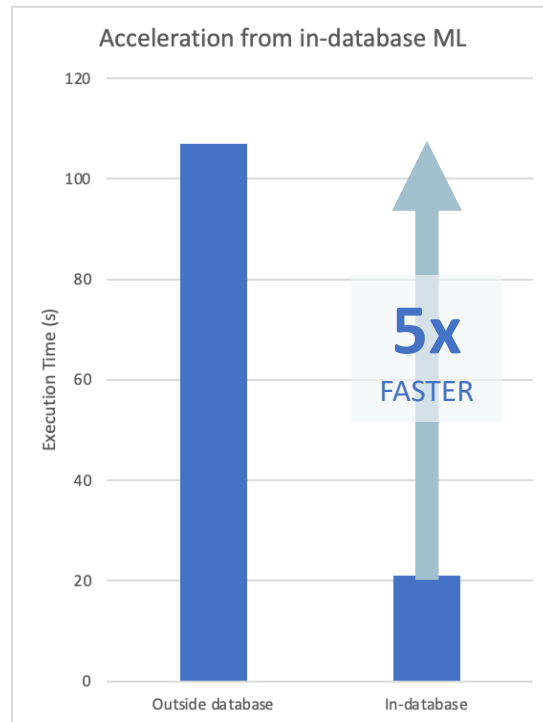
Batch size: 800k rows

Scoring on a Separate System

Data retrieved (over network) from Db2, scored, and written back to Db2: **1m47s**

Scoring with Db2

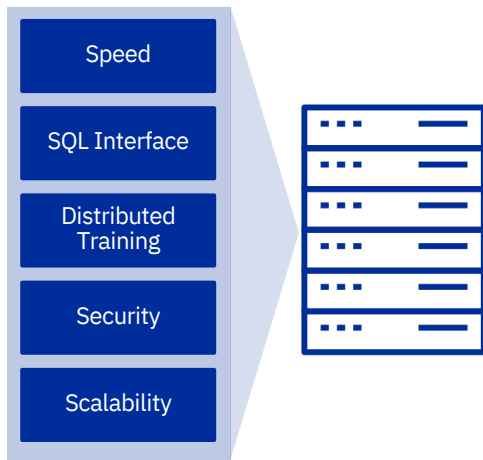
Data retrieved, scored, and written back inside Db2: **21s**



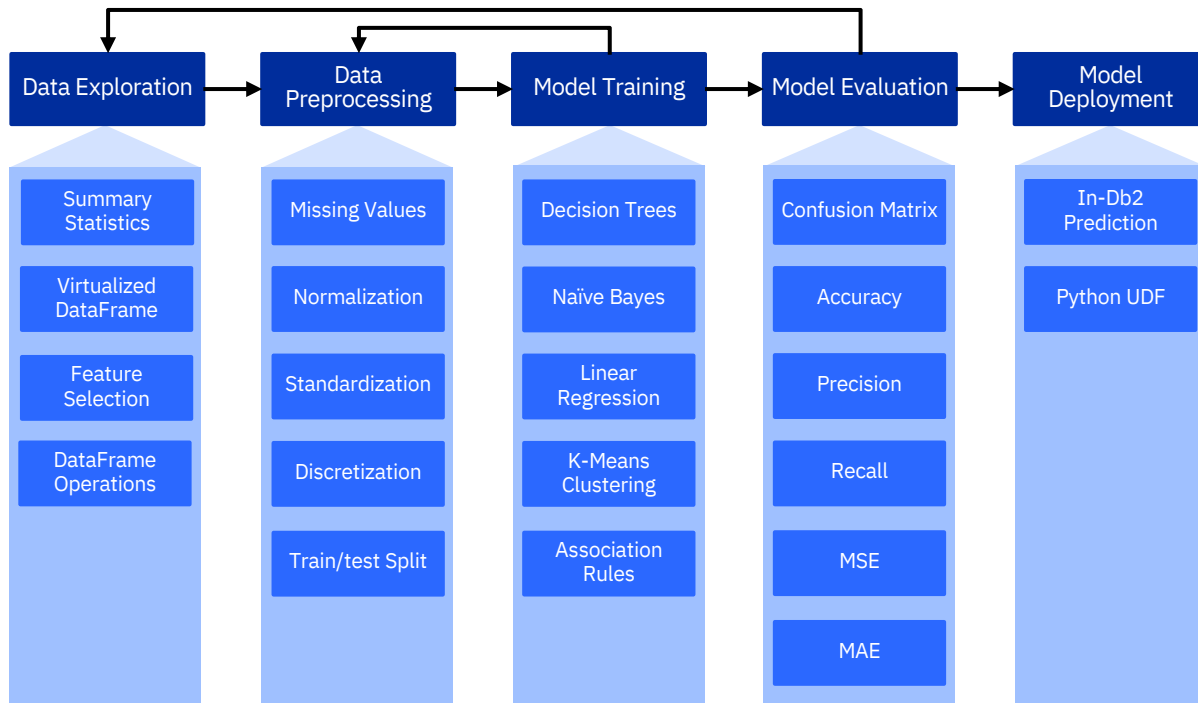
Solution 2:

Build and Deploy ML Models inside the Database

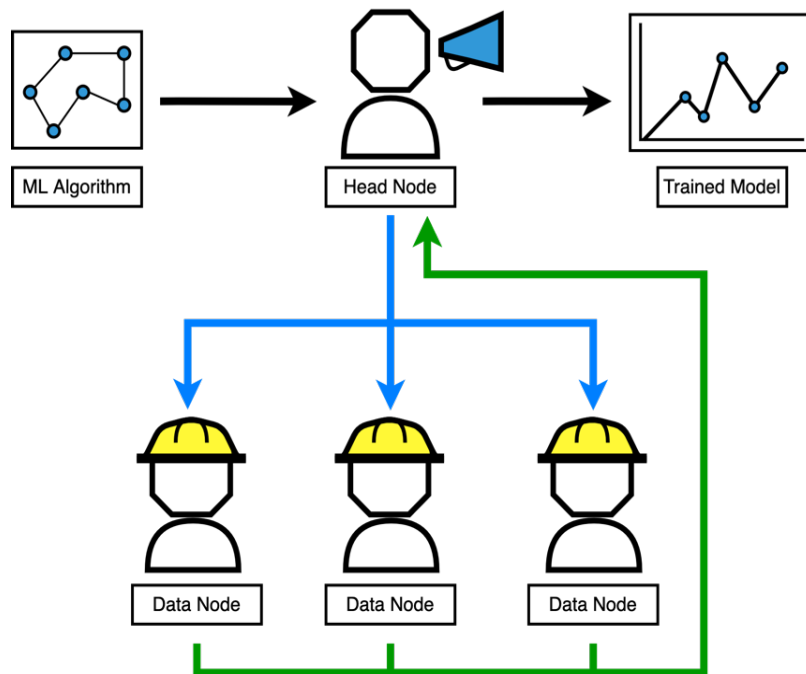
Db2 11.5.4 ML Capabilities



Platforms
 x86 Linux
 PPCLE
 zLinux (Linear Reg. unsupported)



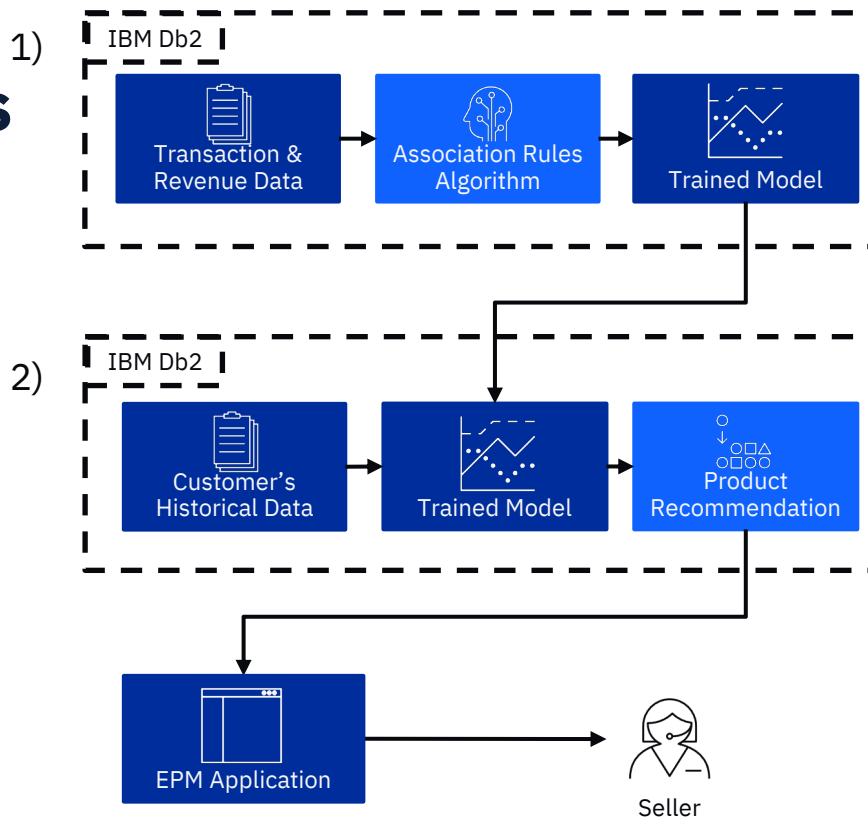
Distributed Model Training



Real Use Case: Product Recommendations

IBM Sales team is finding product recommendations for customers using Db2's ML algorithms on the data stored in Db2 :

- Segmenting customers using k-means clustering algorithm
- Finding product recommendations using association rules mining algorithm



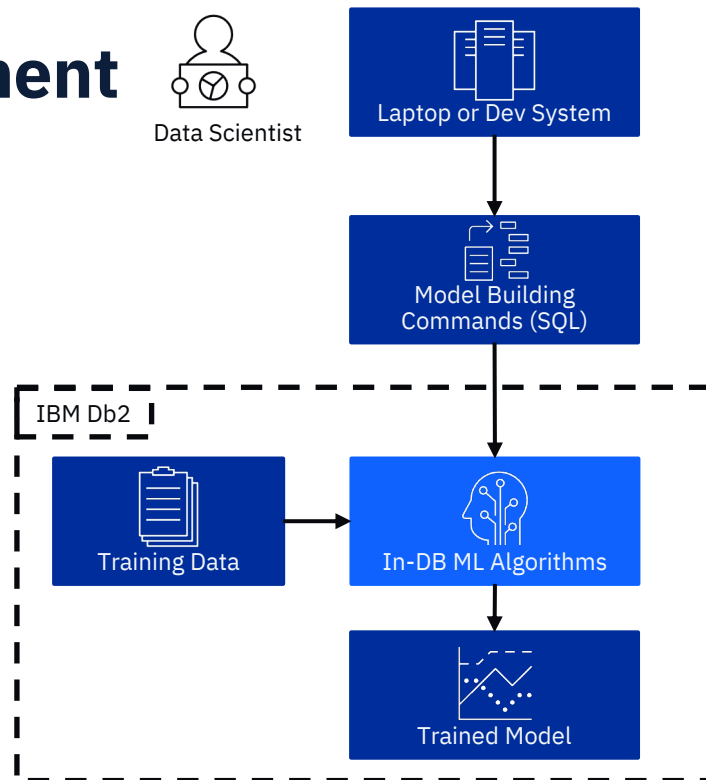
Demo:

Developing and Deploying Db2- Native ML Models

In-Db2 Model Building & Automated Model Deployment

Benefits:

- Speed
- IT cost savings
- Simpler Architecture
- Simplified Governance



Challenges

Talent Gap

Sensitive data

Data transfer costs

Infrastructure

Inference performance

Integration of ML

Benefits of ML with Db2

SQL interface for ML Deployment

Virtualized data access and model building

In-Db2 ML – no data transfer

Secure and scalable Db2 Infrastructure

In-Db2 scoring of Db2-native and open source models

SQL-based Inference Queries

Demos and Tutorials

Demos:

- [Build a Customer Segmentation Model with Db2](#) (K-Means Clustering)
- [Build a Classification Model with Db2](#) (Decision Tree)
- [Build a Regression Model with Db2](#) (Linear Regression)
- [Integrate a Db2-native model with a Cognos Dashboard](#)

Hands-On:

- [Tutorials and Jupyter Notebooks](#)

Documentation:

- [Db2 11.5 Knowledge Center](#)