Lambda architecture design using Azure Databricks for advanced analytics



Agenda

- (1) Introductions and use case
- (2) Who, what, and why of Spark and sdatabricks
- (3) Lambda Architecture: Design Principles
- (4) Demos: ATM Fraud, Azure Airlines
- **5** Wrap-Up: Solution Architectures





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databricks

VISION

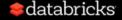
Accelerate innovation by unifying data science, engineering and business

PRODUCT

Unified Analytics Platform powered by Apache Spark

WHO WE ARE

- Founded by the creators of Apache Spark
- Contributes 75% of the open source code, 10x more than any other company
- Trained 100k+ Spark users on the Databricks platform



Who's Using it Today











Amgen adopted Databricks to migrate to a **Modern Big Data Platform on the Cloud**, enabling accelerated drug discovery that involves petabytes of data from multiple biotech field; using Databricks, researchers can use more data during drug development while easily collaborating across teams, leading to greater productivity and new insights

Shell adopted Databricks for Predictive Supply Chain Analytics to optimize their global inventory of spare parts – reducing legacy batch scoring jobs that took longer than 40 hours to less than 45 minutes – and save millions of dollars in operational efficiencies

Viacom adopted Databricks to increase customer retention and loyalty by analyzing petabytes of **Streaming Data** on video quality in real time to improve user experience; this is a machine learning use case on massive amounts of streaming data; the result was higher an increase in customer retention by 3.5x - 7.0x

HP adopted Databricks to support their **Consumer IoT** and **Machine Learning** use cases. HP's data scientists use sensor readings from > 20 M connected devices to create new marketing campaigns for HP Ink; using Databricks the data scientists eliminated the need for DevOps efforts and were able to focus on training more models.

Databricks Customers Across Industries

Financial Services

Healthcare & Pharma

Media & Entertainment

Data & Analytics Services

Technology



BLACKROCK



JPMORGAN Chase & Co.







REGENERON



RADIUS° D





DOW JONES



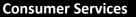






Public Sector

Retail & CPG







Energy & Industrial IoT













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Open source data processing engine built around speed, ease of use, and sophisticated analytics

10-100x faster than MapReduce (Hadoop)

Storage agnostic, allowing federation & simple data access

Easier to program
Python, SQL, R, Java, Scala

More interactive data exploration

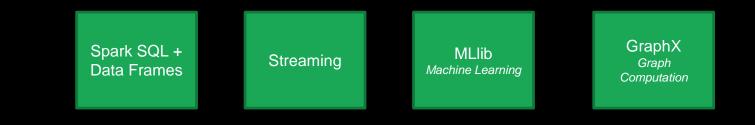
APIs for SQL, machine learning, deep learning, streaming, graph

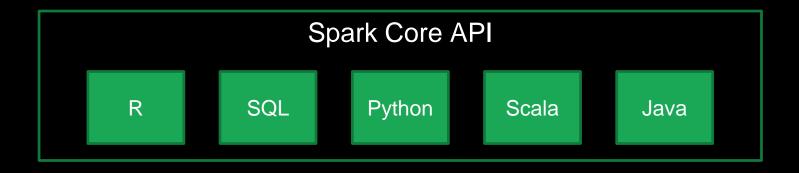
1000+ contributors across 250+ companies



Apache Spark Ecosystem









Do you know Databricks?

Databricks makes building big data and Al applications simple, fast, easy, and collaborative with our **Unified Analytics Platform** powered by Apache Spark™ and built for cloud.

Data Engineering Data Scientists Business Analysts



Apache Spark[™] provides a **single processing engine** for your big data and AI workloads including batch/ETL, streaming, SQL, graph, machine learning and deep learning workloads on petabytes of data on cloud data lakes. The result is higher productivity and faster time to insights and outcomes for your clients.

Databricks' founders are the original creators of Apache Spark™ and we have engineered our **platform as a service** for the cloud to improve elasticity, ease of use, performance, reliability, and cost-effectiveness compared to alternatives.

Get more from Big Data & Al Projects

- Do More -Higher productivity without DevOps or cluster administration

- With More Data At scale (volume and variety)
with better cost/performance and elasticity

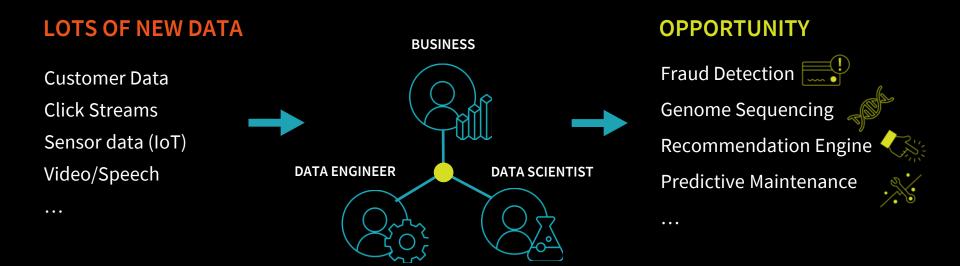
 More Reliably Avoid project delays due to bugs/breaks and simplified data pipelines

 With a Lower TCO Lower cloud and personnel costs and pay only for what you use

- More Secure -Satisfies industry security requirements (e.g. GDPR, HIPAA, and PCI)

- Enable AI & ML - Reach the potential of AI & ML use cases

Al is a Game Changing Opportunity



Machine Learning Requires Collaborative Experimentation on Big Data



Hardest Part of Al isn't Al, it's Data

"Hidden Technical Debt in Machine Learning Systems," Google NIPS 2015

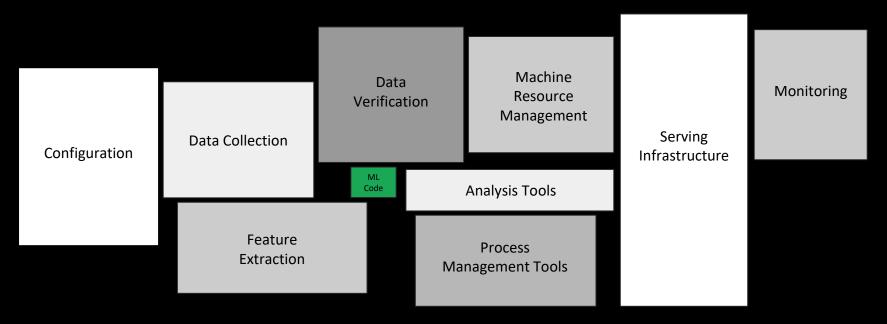


Figure 1: Only a small fraction of real-world ML systems is composed of the ML code, as shown by the small green box in the middle. The required surrounding infrastructure is vast and complex.

Big Data and AI Complexity Slows Innovation

DATA SCIENTIST /ANALYST

BUSINESS SME

ANALYTICS ENGINES

DATA ENGINEER

DATA PREPARATION

Lack of collaboration reduces data science productivity

ETL STREAMING SQL ML & DEEP LEARNING

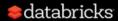
STALE DATA

Multiple analytics engines increase complexity and slows innovation

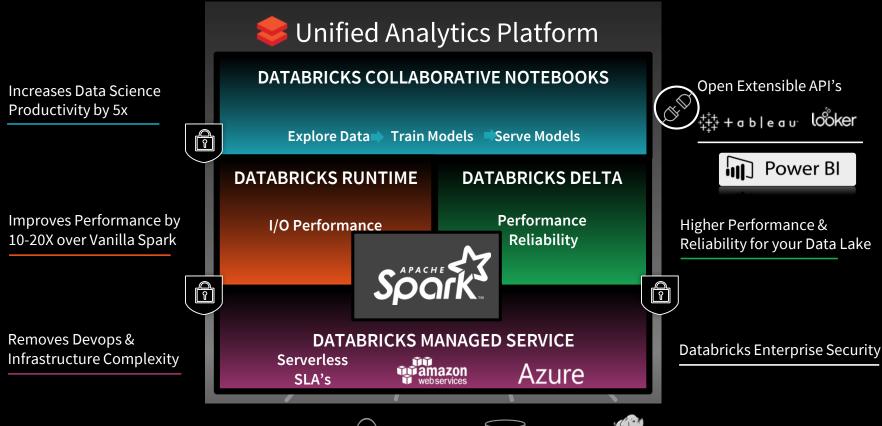
Data pipelines produce stale data with poor performance

Manual operations to manage big data infrastructure





Accelerate Innovation with Databricks













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What is Lambda Architecture

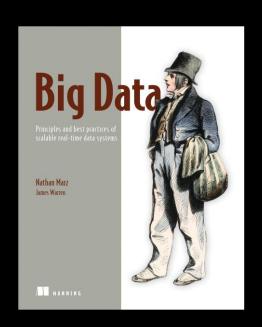
Data processing architecture

Generic, scalable, fault-tolerant

Low-latency reads, updates, ad-hoc queries

Nathan Marz - Apache Storm @ Twitter

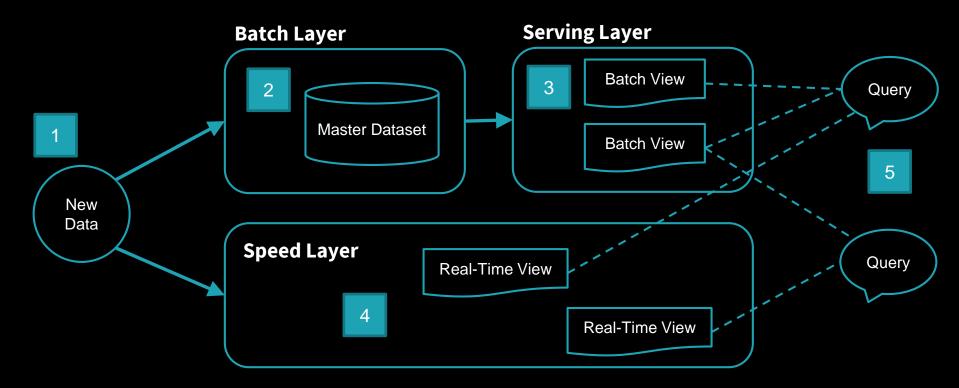
Principles and best practices of scalable realtime data systems

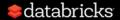




Lambda Architecture

Design Principles

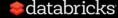




Lambda Architecture

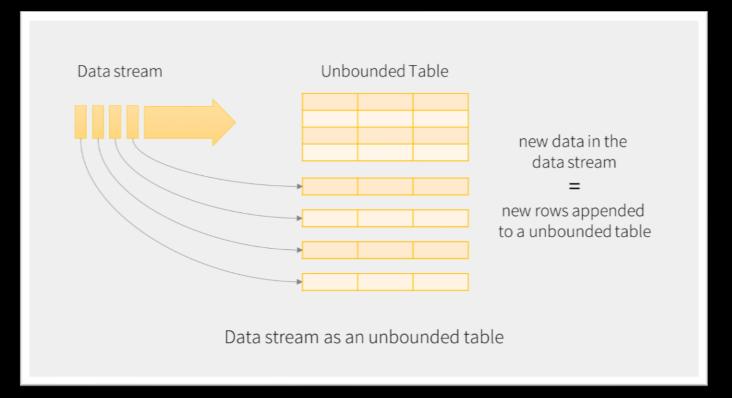
Real World Examples

Fault / Fraud detection
Manufacturing / Machine Logs / Robotics
Network / Security monitoring
Digital Marketing / Websites Clicks / Telemetry
Portfolio Management / Algorithmic Trading
IOT / Connected devices



Lambda Architecture

Databricks Structured Streaming



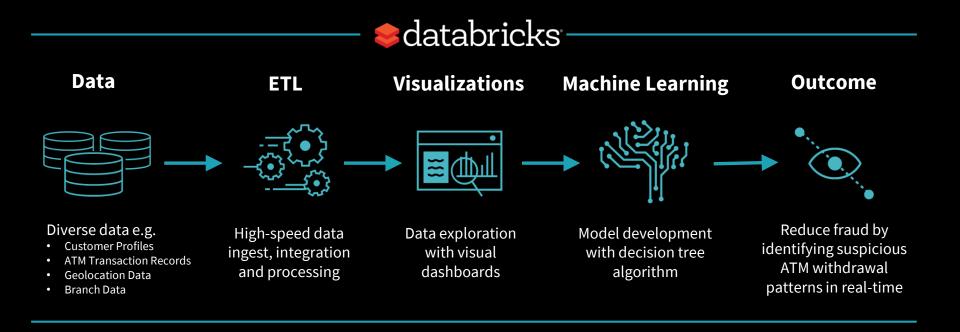


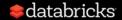
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ATM Fraud Detection

Financial Services

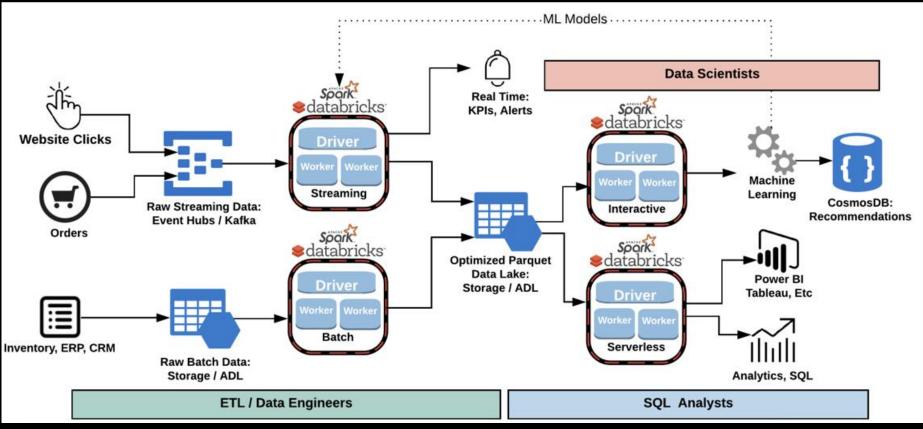




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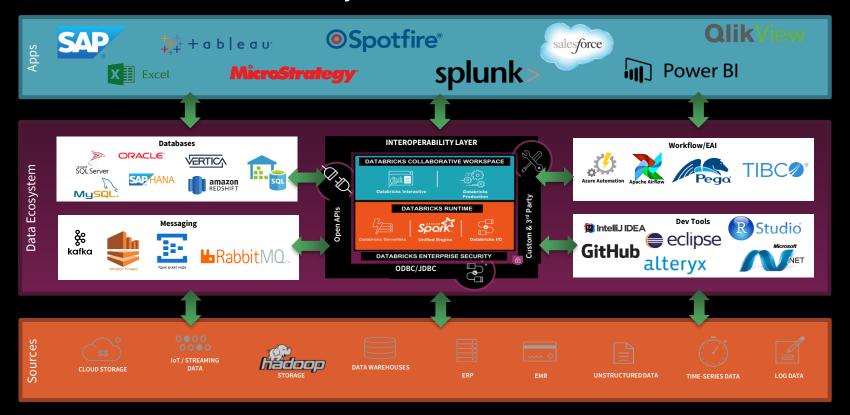
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Databricks 101 Architecture



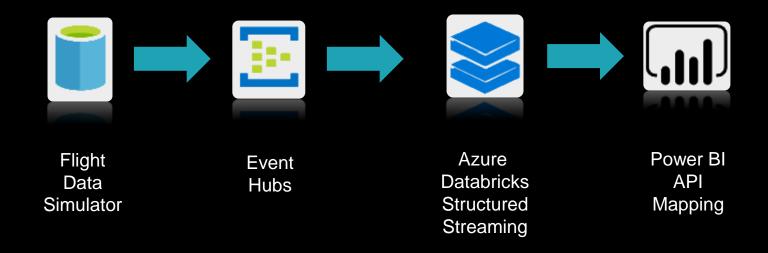


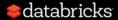
Databricks Ecosystem Reference Architecture





Azure Airlines – Flight Demo





Thank you! Questions?

databricks

Main Links

https://azure.microsoft.com/en-us/services/databricks/

https://spark.apache.org/

https://databricks.com/

https://databricks.com/spark/about

http://lambda-architecture.net/

http://spark.apache.org/powered-by.html

https://www.businesswire.com/news/home/20190131005243/en/Databricks-Named-Visionary-Consecutive-Year-Gartner-Magic

Databricks Community Edition

https://databricks.com/product/faq/community-edition

Free version of the cloud based platform

Hosted on Amazon Web Services

Uses a micro-cluster of one driver with 6 GB of memory

Contains training resources

Great way to get started learning about Apache Spark

Blogs

https://www.desertislesql.com/ - Ginger Grant

https://databricks.com/blog
https://databricks.com/blog/category/engineering
https://databricks.com/blog/category/company
https://curatedsql.com/?s=spark

Videos

https://databricks.com/resources/type/videos

https://www.youtube.com/user/TheApacheSpark/feed

https://www.youtube.com/channel/UC3q8O3Bh2Le8Rj1-Q- UUbA

https://databricks.com/resources/type/product-videos

https://sparkhub.databricks.com/videos/

https://www.youtube.com/watch?v=TJcEP6AX02U

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https://databricks.com/session/jaws-data-warehouse-with-spark-sql

https://databricks.com/azure-databricks-demo

https://databricks.com/resources/type/customer-stories

Projects and Papers

https://cs.stanford.edu/~matei/

https://amplab.cs.berkeley.edu/tag/spark/

https://spark.apache.org/research.html

https://databricks.com/resources/type/research-papers

Training and Certification

https://databricks.com/training

https://docs.databricks.com/

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http://spark.apache.org/docs/latest/building-spark.html

https://www.coursera.org/specializations/big-data

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https://databricks.com/training/certified-spark-developer

Other projects

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https://www.microsoft.com/en-us/research/project/dryadlinq/

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Dayton Gray Sort Record - 2014

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http://sortbenchmark.org/

http://sortbenchmark.org/ApacheSpark2014.pdf

https://docs.microsoft.com/en-us/azure/architecture/reference-architectures/data/stream-processing-databricks

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https://docs.microsoft.com/en-us/azure/stream-analytics/stream-analytics-window-functions

https://docs.databricks.com/spark/latest/structured-streaming/index.html https://databricks.com/blog/2016/07/28/continuous-applications-evolving-streaming-in-apache-spark-2-0.html

https://databricks.com/blog/2018/05/03/benchmarking-apache-spark-on-a-single-node-machine.html

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https://docs.microsoft.com/en-us/azure/machine-learning/team-data-science-process/spark-overview

https://github.com/mspnp/reference-architectures/tree/master/data/streaming_azuredatabricks

https://azure.microsoft.com/en-us/blog/azure-databricks-industry-leading-analytics-platform-powered-by-apache-spark/

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https://towardsdatascience.com/sql-at-scale-with-apache-spark-sql-and-dataframes-concepts-architecture-and-examples-c567853a702f

https://docs.microsoft.com/en-us/sql/big-data-cluster/big-data-<u>cluster-overview?view=sqlallproducts-allversions</u> https://github.com/giulianorapoz/DatabricksStreamingPowerBl https://databricks.com/blog/2017/01/19/real-time-streaming-etlstructured-streaming-apache-spark-2-1.html https://databricks.com/blog/2016/05/23/apache-spark-as-acompiler-joining-a-billion-rows-per-second-on-a-laptop.html