DISCOVER®

Next-gen Data Flow Platform for the Enterprise

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Agenda

(1)

Discover's next-gen data ingestion platform built on NiFi

2

What it takes to build an enterprise-ready platform

3

Challenges and how we overcame them

4

Next steps with the platform



Discover is a leading U.S. direct bank & payments partner

\$60Bn in Credit Card Receivables



1 in 4 Households¹

Leading Cash
Rewards



\$37Bn Consumer Deposits \$9Bn Private Student Loans \$7Bn Personal Loans



- \$183Bn Payment Services Volume
- 185+ Countries/Territories

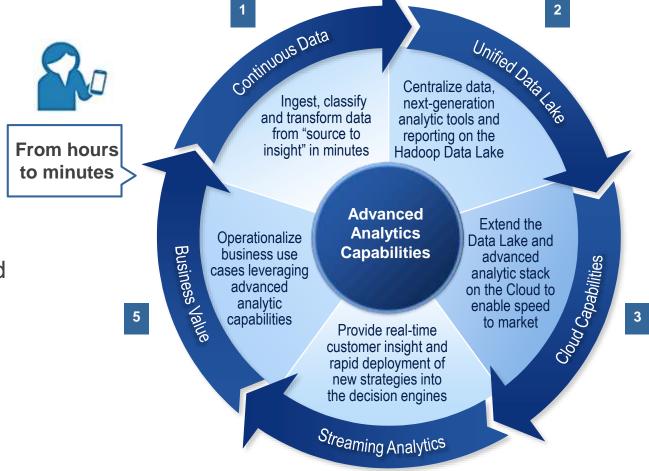
Note(s)

salances as of March 31, 2017; volume based on the trailing four quarters ending 1Q17; direct-to-consumer deposits includes affinity deposits . TNS' Consumer Payment Strategies Study



Advancing our data-analytic capabilities

 Built around a foundation of a continuous data pipeline and hybrid data-analytic lake



Unified data ingestion platform built on NiFi



Unified data ingestion platform

- Ingest data from source systems
- Push to the Enterprise Data Lake
- Governed process leveraging common-reusable templates

What is NiFi?

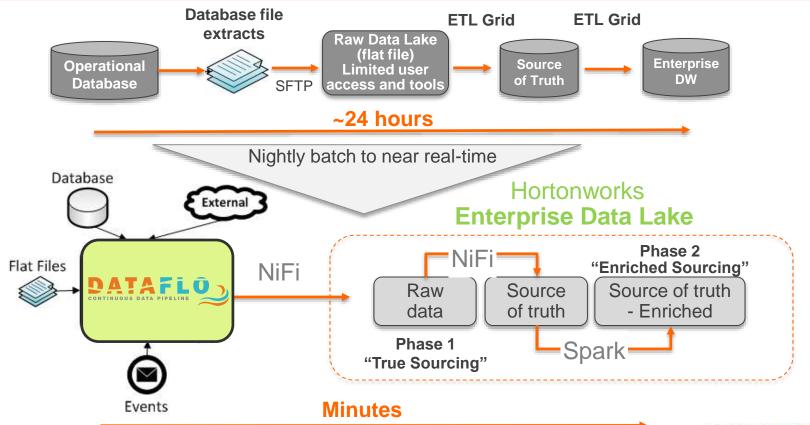
- Enables automated data flow management
- Acquires data from producers
- Delivers to consumers while orchestrating the flow

Why we chose NiFi to build our data ingestion platform

- Scalable and Customizable
- Provenance
- Promotes reuse
- Secure
- ☐ User Interface (drag & drop)

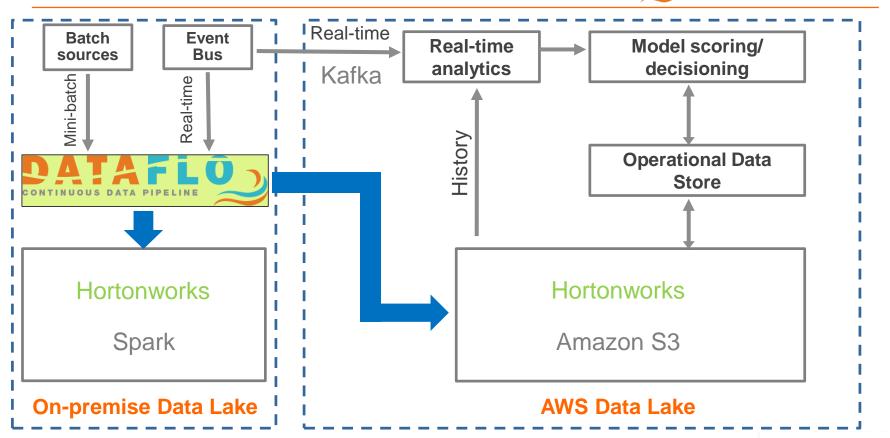


The next-gen platform built on NiFi and Spark is designed to streamline our data pipeline into a near real-time paradigm



We are also extending the capability of PATATE INTO the cloud



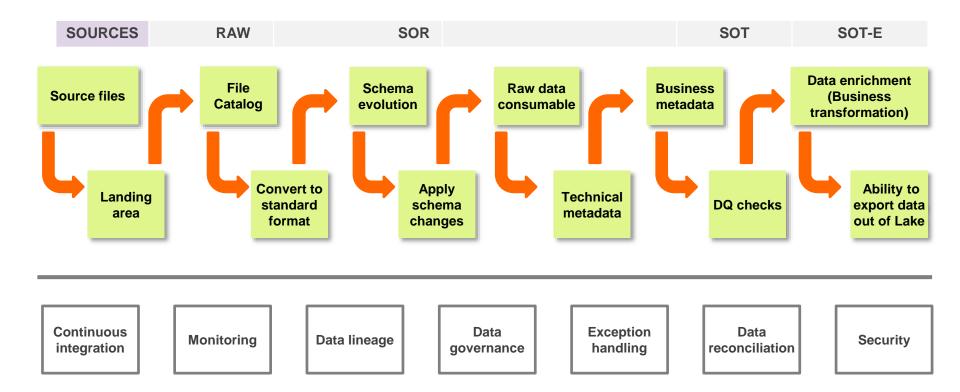


Data Flow Categorization within the Hadoop Data Lake





Detail flow and foundational components





Ingesting complex data - How complex?

Format of files will vary, some are easy to consume, others hard

Example: Records with Dynamic arrays/vectors of primitives or strings

```
Schema: First Name, Last Name, Array_size of Sibling_Name[], Sibling_Name[0-N], City
Data:
```

```
John, Doe, 2, Susie, Chris, Chicago
Mary, Johnston, 3, Ashley, Tom, Mike, Atlanta
Frank, Smith, 1, Ralph, Toronto
```

Example: Records with an array of Struct data types

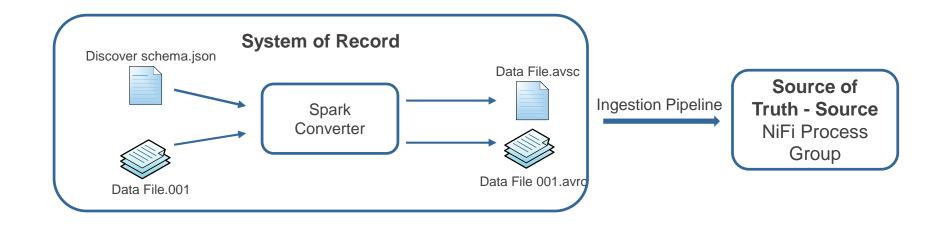
Schema: First Name, Array size of CompanyStruct[], CompanyStruct.Name, CompanyStruct.City, CompanyStruct.YearsWorked, Age

Data:

```
John, 1, Discover, Chicago, 3, 44
Mary, 3, Sales Unlimited, Dallas, 2, Auditors R' Us, Atlanta, 5, Discover, Chicago 4, 35
```



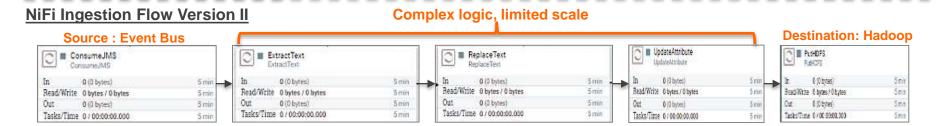
Our solution – A custom NiFi processor to handle complex data types





Continuous improvement of real-time data ingestion using NiFi

NiFi Ingestion Flow Version I 24 hours **Destination: Hadoop** Source: Flat File ■ ConvertCSVToAvro ■ GetFile ■ MergeContent ■ PutHDFS ConvertCSVToAvro GetFile MergeContent PutHDF5 In 0 (0 bytes) 0 (0 bytes) 5 min 0 (0 bytes) 5 min 0 (0 bytes) 5 min Read/Write 0 bytes / 0 bytes Read/Write Obvtes / Obvtes 5 min Read/Write 0 bytes / 0 bytes 5 min Read/Write 0 bytes / 0 bytes 5 mir 5 min 5 min 0 (0 bytes) Tasks/Time 0 / 00:00:00:00.000 5 min. Tasks/Time 0 / 00:00:00 000 5 min Tasks/Time 0 / 00:00:00.000 5 min Tasks/Time 0 / 00:00:00.000 5 min





ETL on Hadoop progression



Data enrichment from SOR to SOT (~600 jobs)

Version I

Traditional ETL tool

Version II

ETL on HiveQL

Version III

ETL on Spark (hand-coded)

Coming soon
Automated
(flow-based)
ETL on Spark

Run time: ~18 hours

~8 hours

~1 hour



Upcoming enhancements to our data pipeline



Integrating data quality, catalog into NiFi flow

Custom processors to parse complex data structures

Enterprise scale ETL on Hadoop using Spark

Selfservice data pipelines Integrating batch and real-time data pipelines



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Hiring Data Engineers

Q&A