

Machine Learning on Streaming Data

with Apache Kafka, Apache Beam, & TensorFlow

About Us



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Big Thanks to:

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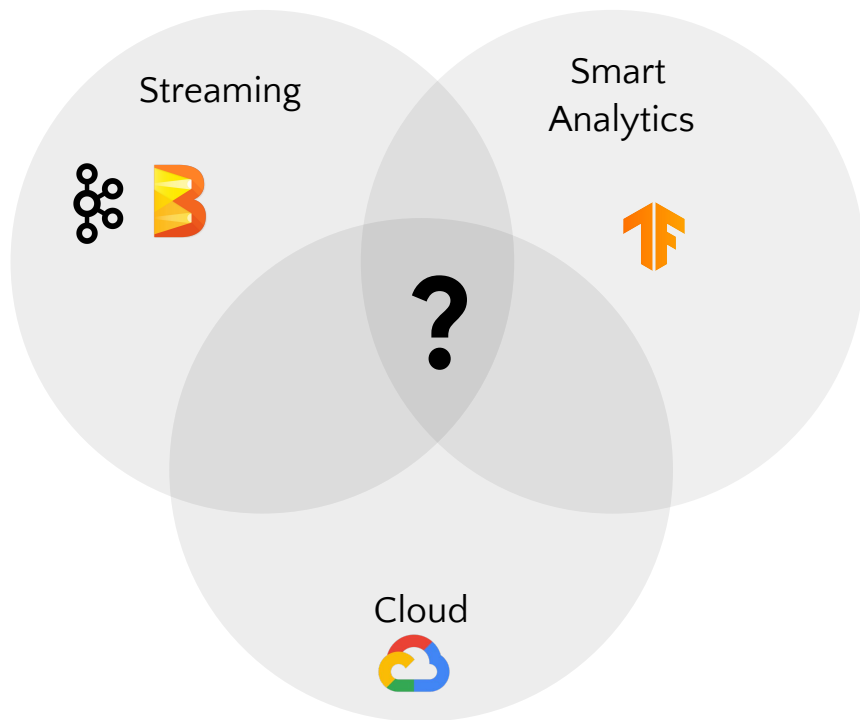
Agenda

1. Motivation
2. Architecture
- 3. Use Case Walk-Through w/ Demo**
4. Summary

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Motivation

Technology Landscape



InfoWorld's 2019 Technology of the Year Award Winners:

- **Apache Beam**
- **Apache Kafka**
- Elastic Stack
- DataStax Enterprise
- Firebase
- Horovod
- H2O Driverless AI
- Keras
- Kubernetes
- LLVM
- .Net Core
- PyTorch
- Redis
- **TensorFlow**
- Visual Studio Code
- XGBoost

ML Platform

Notebook



Data Ingestion



Data Analysis &
Transformation



Trainer



Model Evaluation
& Validation









Serving



Orchestration

ML Framework

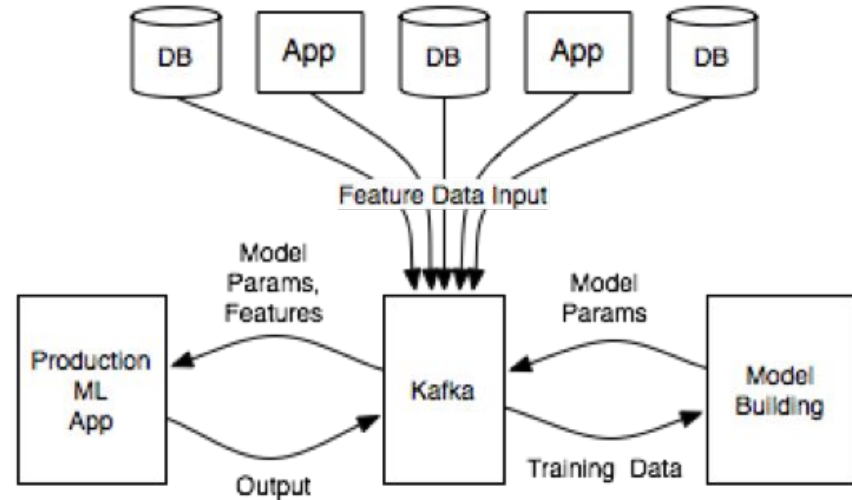
| OSS | Managed Service |
|--|---|
| Apache Kafka Event streaming platform  | Confluent Cloud Monitoring, Replication, Data Balancing  |
| Apache Beam Data processing pipelines Unified batch & streaming  | Dataflow Automated resource management of workers  |
| TensorFlow Robust foundation for machine and deep learning  | Cloud Machine Learning Engine <ul style="list-style-type: none"> • <u>Training</u>: Distributed training infrastructure that supports CPUs, GPUs, and TPUs • <u>Serving</u>: Host models for batch & online prediction  |

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Architecture

Reference Kafka ML Architecture

- Data pipelines are simplified
- Building analytic modules is decoupled from servicing them
- Usage of real time or batch as needed
- Analytic models can be deployed in a performant, scalable and mission-critical environment

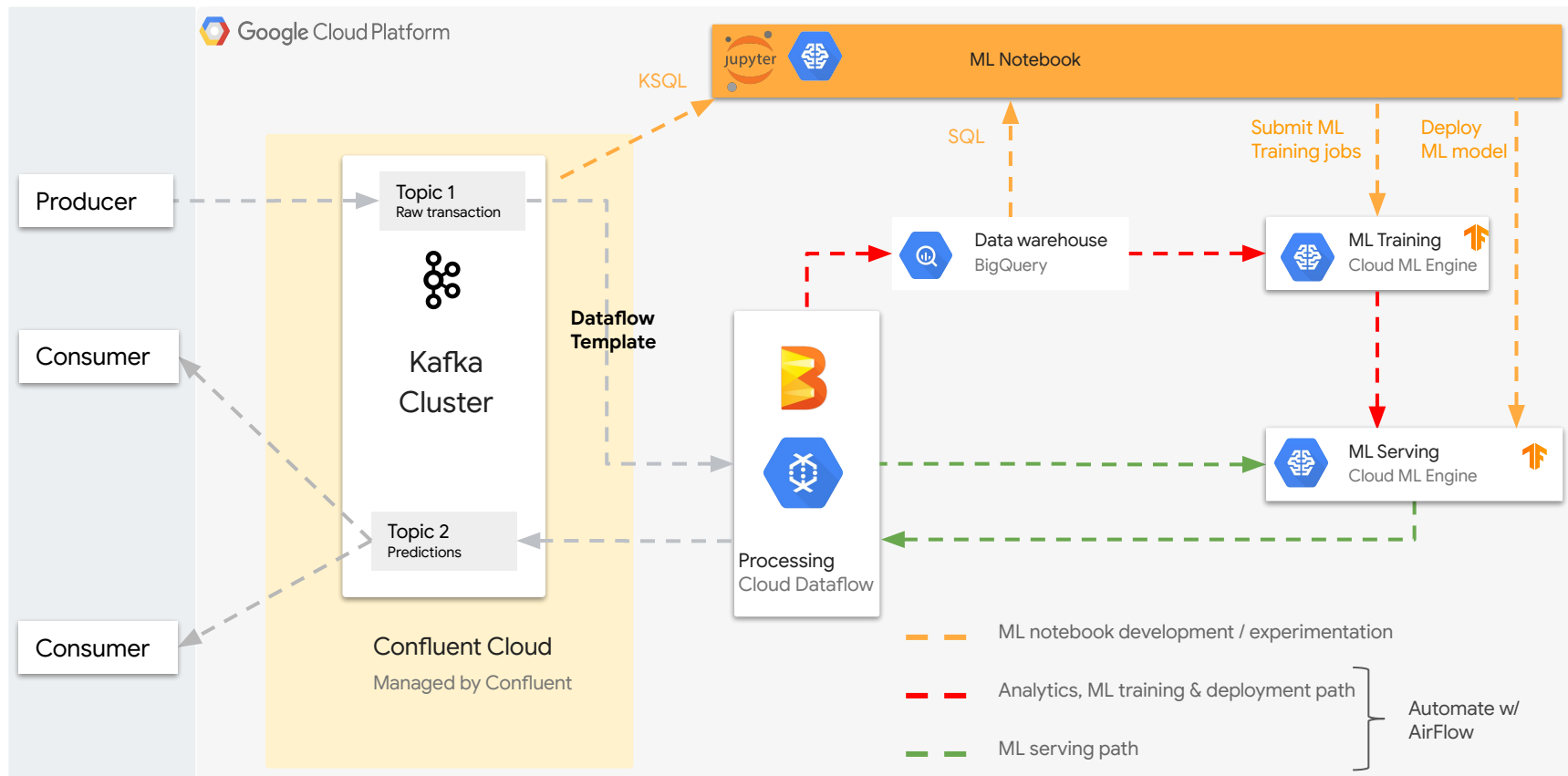


Kai Waehner

Technology Evangelist, Confluent

<https://www.confluent.io/blog/build-deploy-scalable-machine-learning-production-apache-kafka/>

Leverage managed services to simplify & focus on code not infrastructure



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Use Case Walk-Through

Kaggle Case Study

Fraud Detection of Credit Card Transactions

284,807

transactions

492

Fraud (0.172%)

- Collect transaction data
- Analyze historical data
- Train model on historic sample
- Evaluate model based on precision & recall
- Predict fraud on new streaming data

<https://opendatacommons.org/licenses/dbcl/1.0/>

- Andrea Dal Pozzolo, Olivier Caelen, Reid A. Johnson and Gianluca Bontempi, [Calibrating Probability with Undersampling for Unbalanced Classification](#), In Symposium on Computational Intelligence and Data Mining (CIDM), IEEE, 2015
- Dal Pozzolo, Andrea; Caelen, Olivier; Le Borgne, Yann-Aël; Waterschoot, Serge; Bontempi, Gianluca, [Learned lessons in credit card fraud detection from a practitioner perspective](#), Expert systems with applications, 41, 10, 4915-4928, 2014, Pergamon
- Dal Pozzolo, Andrea; Boracchi, Giacomo; Caelen, Olivier; Alippi, Cesare; Bontempi, Gianluca, [Credit card fraud detection: a realistic modeling and a novel learning strategy](#), IEEE transactions on neural networks and learning systems, 29, 8, 3784-3797, 2018, IEEE
 - Dal Pozzolo, Andrea [Adaptive Machine learning for credit card fraud detection](#) ULB MLG PhD thesis (supervised by G. Bontempi)
- Carcillo, Fabrizio; Dal Pozzolo, Andrea; Le Borgne, Yann-Aël; Caelen, Olivier; Mazzer, Yannis; Bontempi, Gianluca, [Scarf: a scalable framework for streaming credit card fraud detection with Spark](#), Information fusion, 41, 182-194, 2018, Elsevier
- Carcillo, Fabrizio; Le Borgne, Yann-Aël; Caelen, Olivier; Bontempi, Gianluca, [Streaming active learning strategies for real-life credit card fraud detection: assessment and visualization](#), International Journal of Data Science and Analytics, 5, 4, 285-300, 2018, Springer International Publishing

DEMO 1 – 5 min

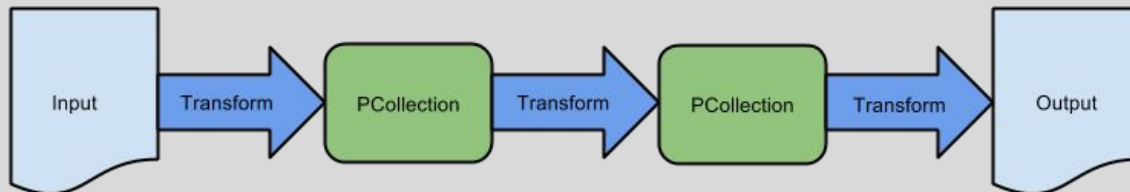
Sending our credit card data

Confluent Cloud, Creating a Topic, Python Script, Security

Kafka to BigQuery



Java Code



```
KafkaIO.<String, String>.read()
```

```
BigQueryIO.writeTableRows()
```



Create a template for easy re-usability by an analyst

Dataflow Template

Additional parameters

Name

Value

bootstrapServers

outputTableSpec

inputTopic

redacted



Explore data & train ML model

Query directly from topic

```
from ksql import KSQLAPI
```

redacted

Query petabytes of data

```
%%bigquery
```

redacted

Submit ML training job

```
gcloud ml-engine jobs  
submit training
```

redacted

DEMO 2 – 5 min

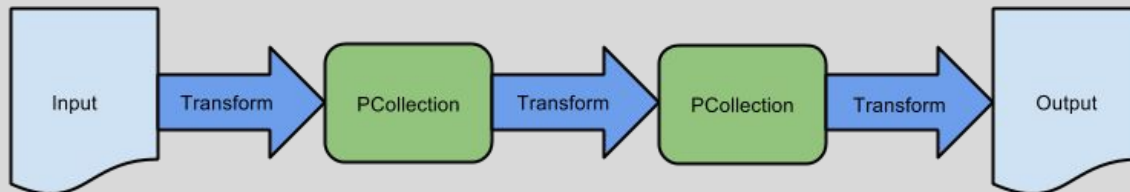
Dataflow template & job

Jupyter: KSQL, BQML, TensorFlow CMLE job

Send Predictions back to Kafka



Java Code



`KafkaIO.<String, String>read()`

`KafkaIO.<String, String>write()`

Request

Response

Train
Model

Publish models

Hosted ML Model

Cloud Machine Learning
Engine

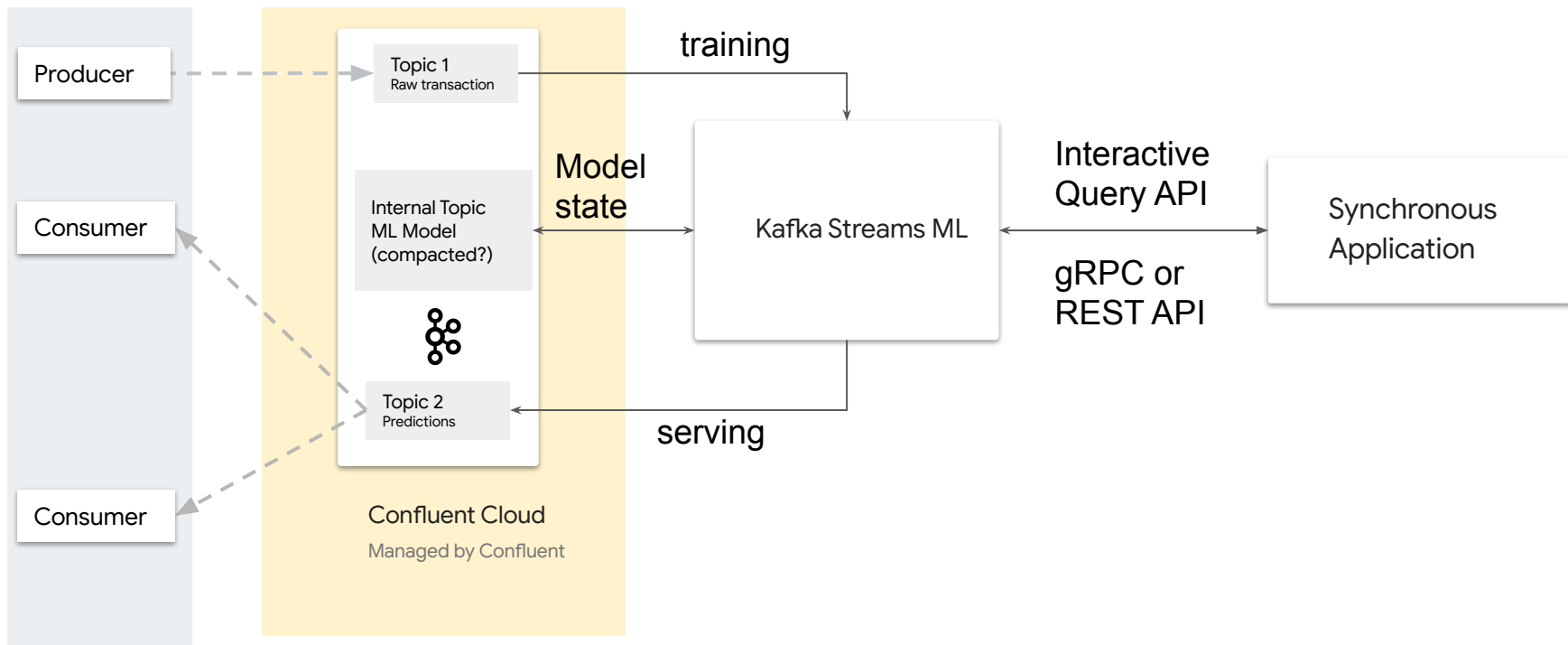


DEMO 3 – 5 min

- (1) Deploy model as an end point
- (2) Prediction sent to Kafka topic to be consumed
- (3) Track models & monitor predictions in CMLE UI

Futuristic Architecture: Pure Kafka-based ML

Resilient, highly available, sync & async

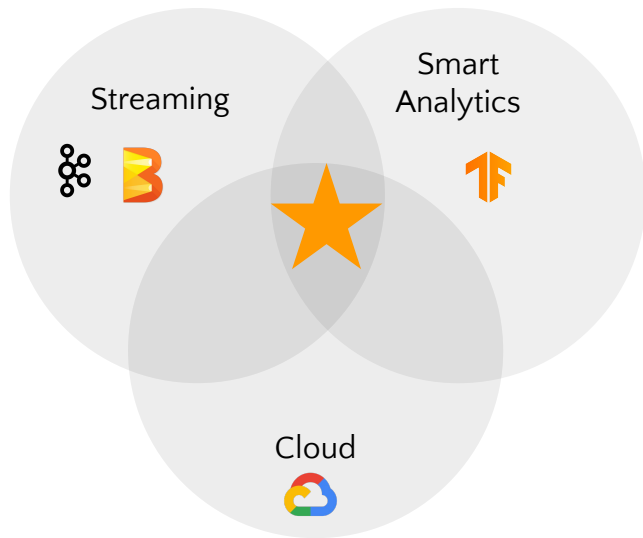


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Summary

Summary

- Kafka + Beam + TensorFlow = Great foundation for future
 - Batch today → streaming tomorrow
 - Small data → big data tomorrow
 - Shallow learning today → deep learning tomorrow
- Make data & ML easier for yourself by using managed services
- Build for many other use cases:
 - Predictive maintenance
 - Logistics routing
 - Image search & recommendations in e-commerce



Talk to Google Cloud



Booth

Learn More

Blog: Enabling connected transformation with Apache Kafka and TensorFlow on Google Cloud Platform

bit.ly/2CHERol

KafkaIO on Beam

bit.ly/2YwL3Jc

KafkaToBigQuery Dataflow Template Example

bit.ly/2HQqVN0

Contact us

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Questions

