

Systems like Apache Kafka, Amazon Kinesis, and Amazon MSK are great for the ingestion, delivery, and real-time analysis of streaming data. Adding a rules engine that detects patterns of events occurring in many systems enables organizations to support a much wider set of use cases. Rules engines also abstract the complexity of defining rules, empowering business users to leverage context alongside their expertise to define the next best action. In this session, learn how to combine the power of Apache Kafka, Amazon Kinesis, or Amazon MSK with business rules engines to automate sophisticated decision-making and actions. This presentation is brought to you by TIBCO Software, an APN Partner.





With traditional data processing, responses are delivered after the fact



Data is collected and stored

Data is placed in a persistent data store (Relational DB, NoSQL, or Hadoop environment)



Analytical processes are executed

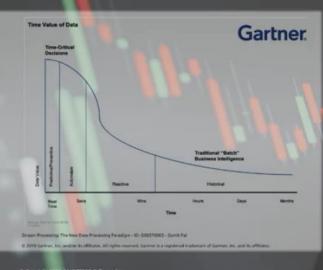
Analysis is performed on the stored data to detect opportunities or threats

Actions are identified

Once actions are identified, they are delivered and executed across various business channels

You cannot make today's decisions based on yesterday's information!

Decisions are more valuable immediately after an event occurs



As time passes, events tend to lose their value

Opportunity to predict a failure is missed

Customer is no longer interacting with your organization

Real-time capability is needed to maximize event value

The way applications are built needs to change



From Deterministic

- · Everything was a deterministic set of rules
- · Predetermined procedural logic (If, Else)



To Smart

- Software needs to become both predictive and prescriptive
- · Understand context and respond accordingly

Machine learning is an answer, but not the only answer



- Sometimes the real-time processing of events is sufficient for your problem!
- · Simplifies definition of the decisions
 - Declarative rules
 - Median, mean, time-weighted averages, variability/robustness



Machine Learning

- Classify large quantities data, such as images, text, etc.
- · Broad set of patterns needs to be detected
- Sufficient data must be available

Project Flogo



Project Flogo

Open-Source Ecosystem for Event-Driven Applications





10-50x lighter than Java, .NET, or Node.js



100% open-source stack for all things event-driven



Common core for all event-driven capabilities



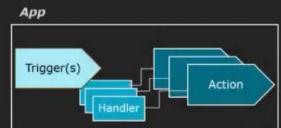
Deploy as serverless functions, containers, or to IoT edge devices



learning

Flogo constructions: Triggers, handlers & actions





App = Trigger(s) + Actions (Activities)

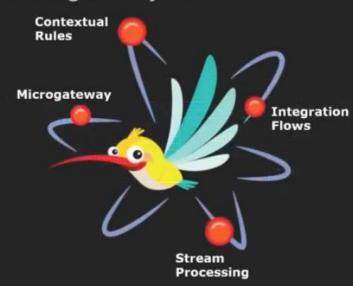
- Receive data from external sources
- Are managed by a configurable threading model
- Have a common interface enabling anyone to build a Flogo trigger

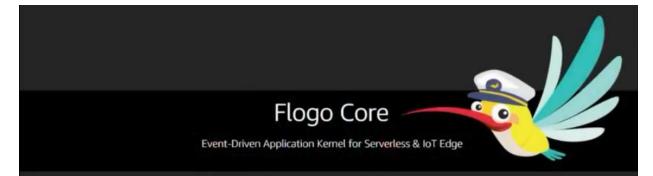
Dispatch events consumed by triggers

Actions

- Process events in a manner suitable with the implementation
- Have a common interface enabling unique event processing capabilities

The Project Flogo ecosystem







Powerful event-driven programming model based on triggers and actions



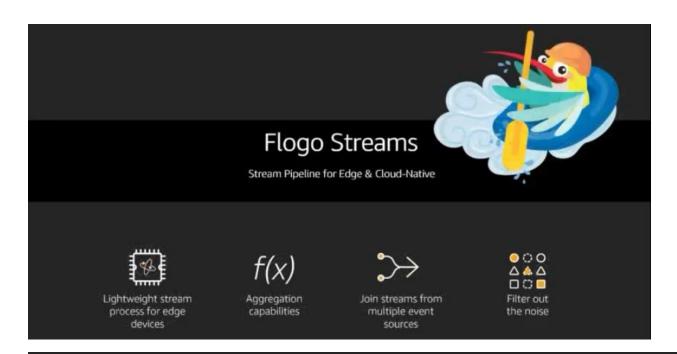
Internal channel for action chaining; bundle all application needs into a single bin



Common contribution model enabling you to leverage 100s of activities & triggers

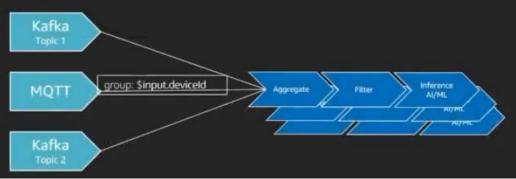


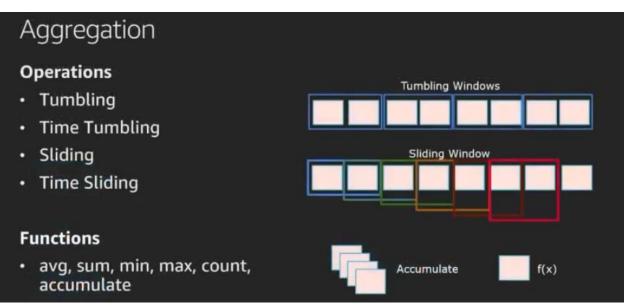
Simple interface to extend event processing capabilities



Grouping & joining events across triggers

- With Flogo Streams, you can group and join events across multiple streams of data
- · Joins occur against pipeline input/output







Flogo Rules

Real-Time Contextual Decisioning



Real-time decisions to determine next best action/outcome and experience



Open source/core complete with extensions, APIs, and interfaces



Declarative rules/rapid application development



Stateful/contextual reasoning across time and space



Embed Machine Learning into Your Real-Time Applications



Execute TensorFlow models



100% open source with zero lock-in

Real-time data transformations for ML should be easy

- 80%* of data science is data cleaning—production has to replicate this
- · Data scientists and developers receive different starting data
- Developers aren't trained for feature creation
- Tools for pre/post-processing data in production unaddressed



CatalystML

Real-time data transformations for machine learning





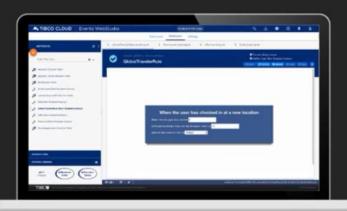


Support for 1:1 mapping into ML models

These capabilities are also available on AWS

TIBCO Cloud Events:

Reduce decision latency by turning streaming data into actions to mitigate risk and improve customer experiences or business operation





Web-based rules authoring to empower the business user

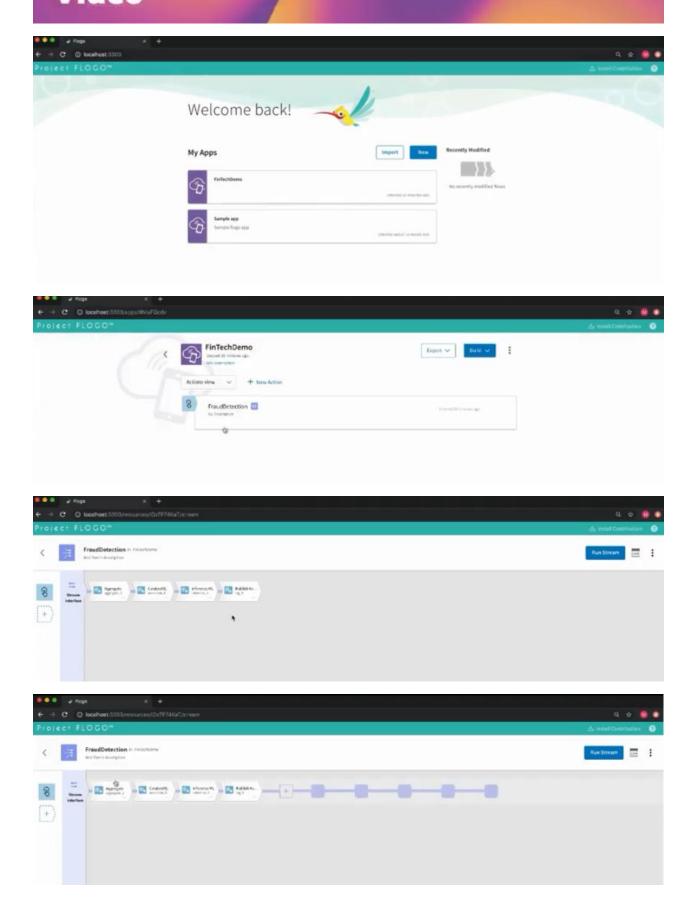


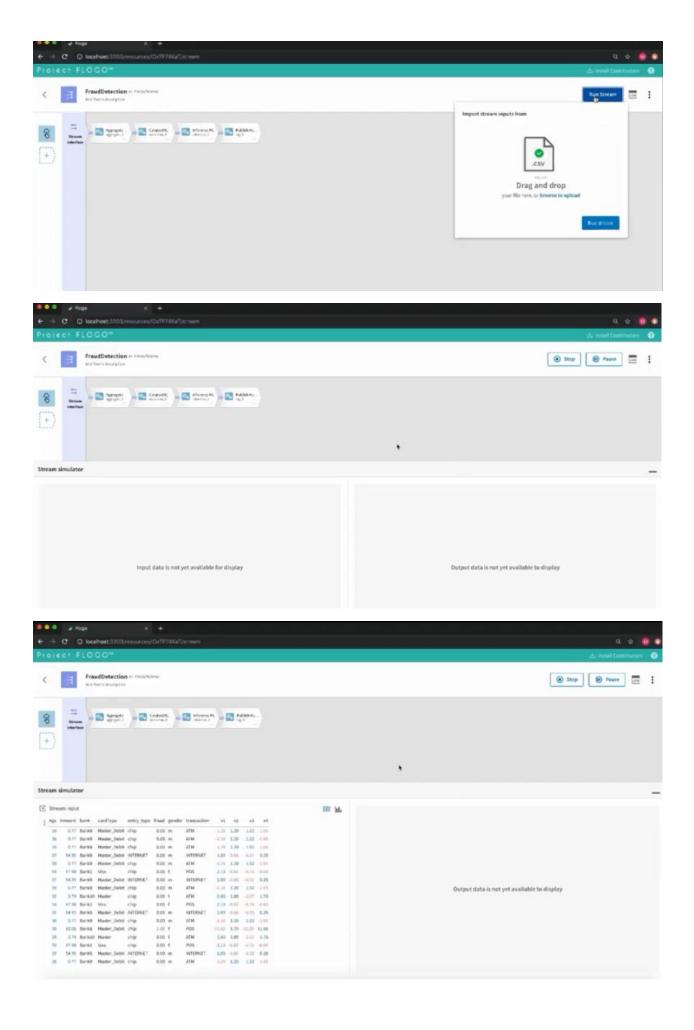
Declarative rules simplify business logic

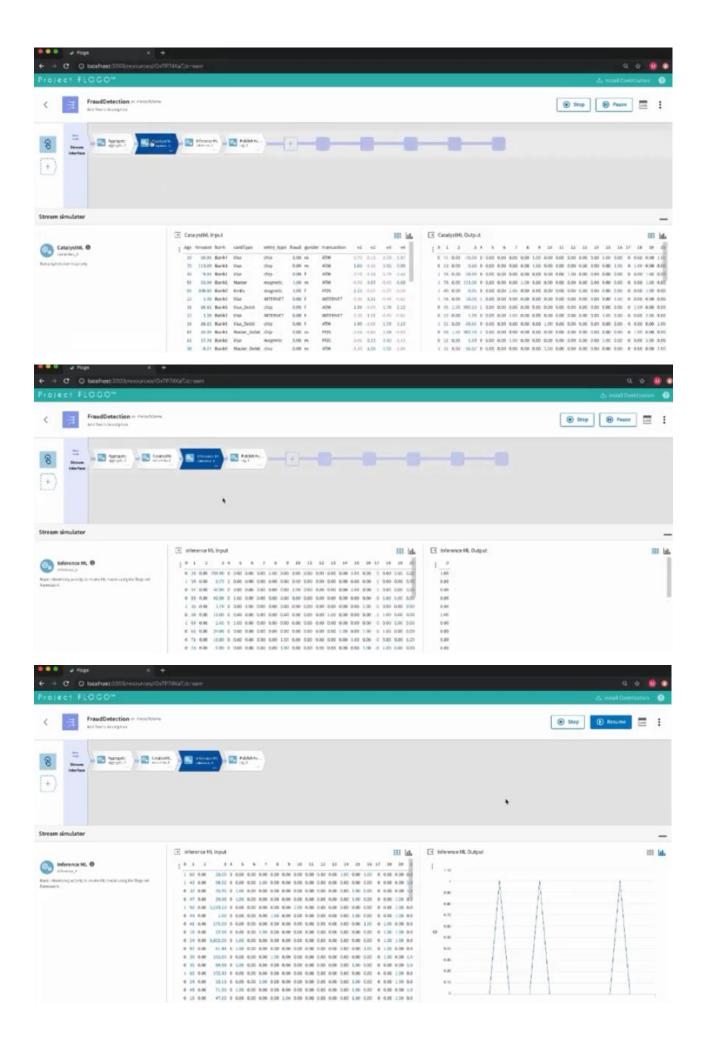


Brings real-time decisions to TIBCO Cloud

Video







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