# Automating Netflix ML Pipelines With Meson

QCon SF 2017 | Eugen Cepoi, Davis Shepherd





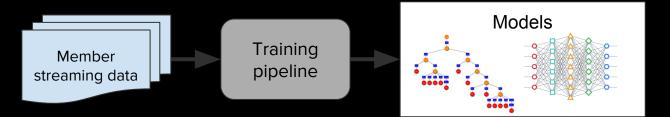




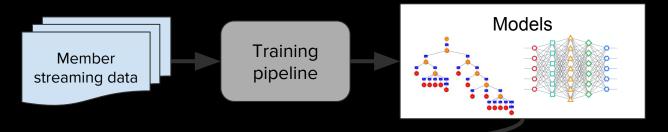
# Create a personalized experience to help members find content to watch and enjoy

Member streaming data



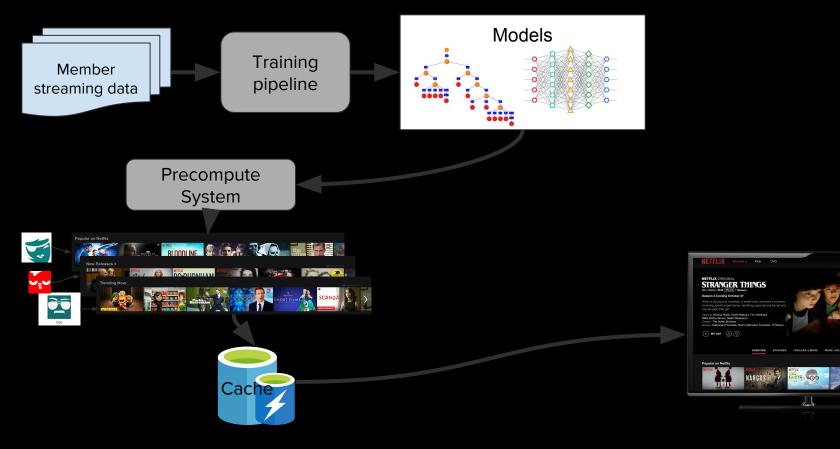




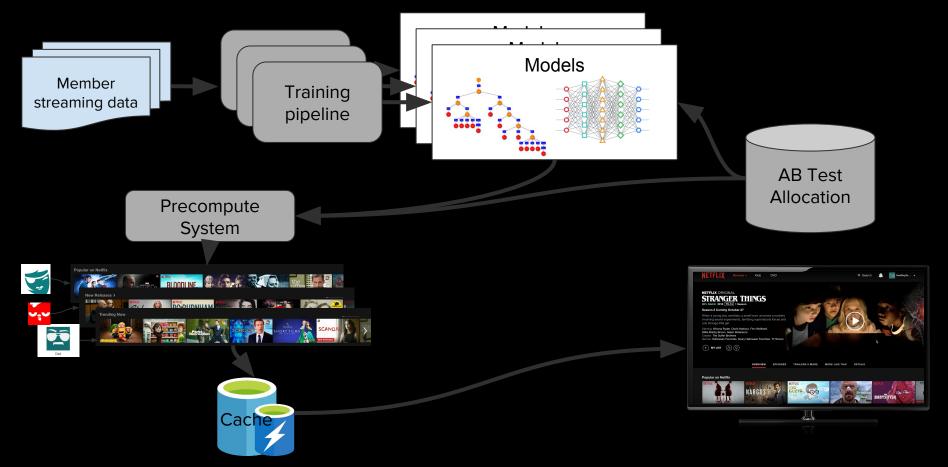








9. Search 🌲 🔛 Overlagte... -



# Innovation is driven by experimentation

## **Training Pipelines**

Data Preparation Spark/Hive/Kafka Stratified Sampling Feature Generation Label Generation Feature Encoders

Model Training Proprietary Algos Spark/ Tensorflow

Model Selection HyperParameter tuning

Scoring / Inference Pre-compute Live-compute Spark and Online Caches



Model Publish S3 Online Caches

### **Before Meson**

- A collection of operators
- Little to no orchestration
- Often limited to single machine

#### **Desired Properties**

- Support Heterogeneous systems
- Highly flexible generic orchestration
- Handle failures
- Provide Reproducibility
- Support Multi-tenancy
- Support External Triggers

### Why didn't an out of the box solution work?

- Spark and Scala support was paramount
- Options available didn't have the flexibility and scalability that we needed

Meson overview

#### **Meson Overview**

General purpose workflow orchestration engine

Delegates execution to Mesos

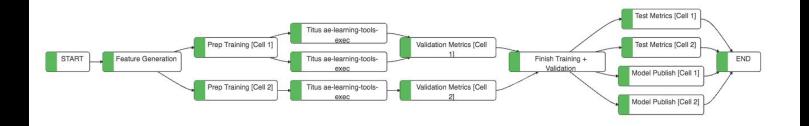
Initially built for Machine Learning pipelines for personalization

Supports complex workflow patterns (branching, loops, foreach)



Workflow Directed Graph of steps, global parameters, triggers...

Step Describes a job and its configuration



### **Defining workflows**

Scala DSL

Python DSL

UI

**REST API** 

```
val sparkStep = Spark(
   jobClass = "netflix.MySparkJob",
   sparkArgs = Seq("--driver-memory" -> "8g")
)
val cpStep = DistCp(from = "...", to = "...")
Workflow(
   id = "my-cool-workflow",
   triggers = Seq(Trigger.CRON("0 0 0 1/1 * ? *")),
   notification = Notification(
      whenStart = true, whenFinish = true, whoCustom = Option("me@gmail.com")
   ),
   parameters = Seq(Parameter.STRING(name = "country", value = "US")),
   definition = sparkStep.sequence(cpStep).end()
```

#### **Parameters**

Used to configure steps, job arguments, and step transitions

MVEL expression to derive parameter values at runtime

Predefined macros

```
Workflow(
    id = "my-cool-workflow",
    parameters = Seq(Parameter.LONG(name = "dateint", expression = Option("tsToDateInt(RUN_TS)"))),
    definition = Step(
        name = "jobX",
        parameters = Seq(
            Parameter.LONG(name = "day", expression = Option("dayFromDateInt(dateint)"))
        )).end()
```

#### **Workflow patterns**

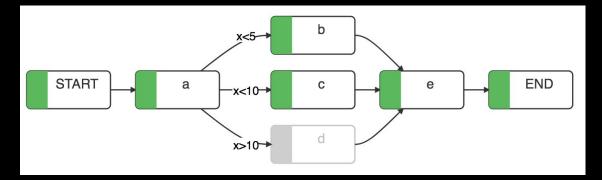
Branches (OR, XOR, AND)

Loops with XOR

Foreach

Step("a").or(
 Condition("x<5").branch(Step("b")),
 Condition("x<10").branch(Step("c")),
 Condition("x>10").branch(Step("d"))
).sequence(Step("e"))

Using parameters and MVEL



#### **Data artifacts**

Data artifact defined by a name and a set of partitions (parameters)

Cross workflow dependencies

External triggers

```
Step(
  name = "someJob".
  inputs = Seq(
    DataArtifact(
      name = "views",
      params = Seg(Parameter.STRING(name = "country", value = "US"))
  outputs = Seq(
    DataArtifact(
      name = "aggViews",
      params = Seq(
        Parameter.STRING(name = "ts", value = "$RUN_TS"),
        Parameter.STRING(name = "country", value = "US")
```

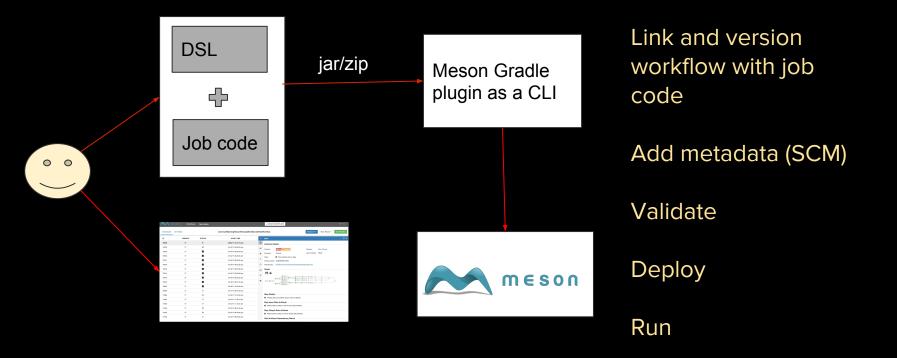
### **Workflow versions**

Workflows have immutable versions

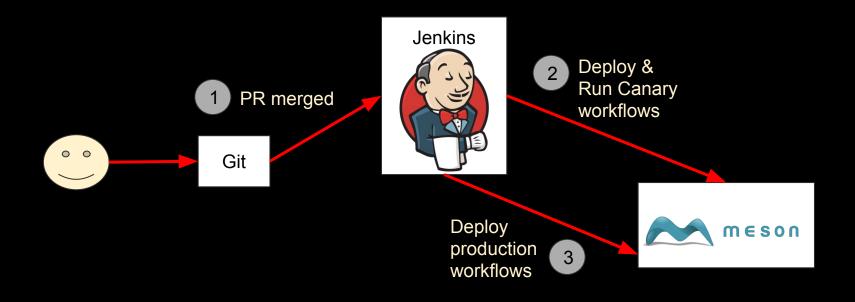
**Enables:** 

- Better collaboration
- Rollbacks
- Reproducibility

#### **Deploying workflows with the Gradle plugin**

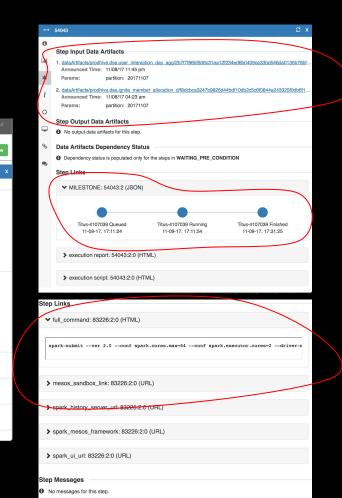


#### Automated releases with canary workflows

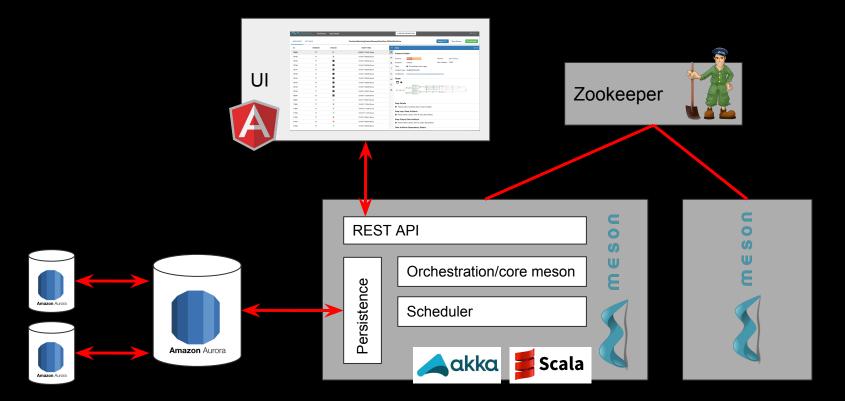


## **Monitor & Debug**

<b>~</b> m	eson Workflows	Data Artifacts			+ CREATE WORKFLOW			🔁 Abo
INSTANCES	SETTINGS		ContinueWatchingFeatureRevampWorkflow-AB	Test	Workflow	Version: 77 -	More Actions -	Run Workfie
ID	VERSION	STATUS	START TIME	<b>↔</b>	78929			2
78929	77	•	10/20/17 12:43:19 pm	0	Instance Details			
78750	77	×	10/19/17 06:32:55 pm		Actions: Stop C Retrigger	Params:	View Params	
78749	77		10/19/17 06:30:04 pm	*	Duration: o hours	Has Artifacts:		
78748	77		10/19/17 06:26:25 pm		Tags: () This workflow has no tags.			
78747	77		10/19/17 06:25:30 pm	Ι	Initiator Type: SUBWORKFLOW			
78746	77		10/19/17 06:25:20 pm	0	Initiated By: /workflows//versions//instances//steps//	stepInstances/		
78745	77		10/19/17 06:24:59 pm	Ģ	Graph			
78744	77		10/19/17 06:23:06 pm	<b>9</b> 0				
78743	77		10/19/17 06:22:55 pm	0				
78742	77		10/19/17 06:21:23 pm	2				
78340	77		10/18/17 10:55:58 am		and the second s			
78094	77	~	10/17/17 09:47:02 am					
77865	77	×	10/16/17 10:24:39 am		Step Details			
77694	77	~	10/15/17 11:36:12 am		Please select a workflow step to view it's details.			
77548	77	~	10/14/17 11:45:45 am		Step Input Data Artifacts  Please select a step to view its input data artifacts.			
77406	77	×	10/13/17 06:51:46 pm					
77404	77	×	10/13/17 06:44:06 pm		Step Output Data Artifacts  Please select a step to view its output data artifacts.			
77402	77	*	10/13/17 06:35:39 pm		Data Artifacts Dependency Status			



#### **Architecture**

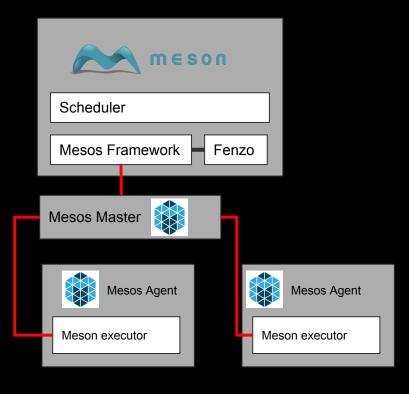


## Scheduling

Meson as a Mesos Framework

Mesos offers resources and runs the steps

<u>Fenzo</u> (Netflix OSS) makes scheduling decisions

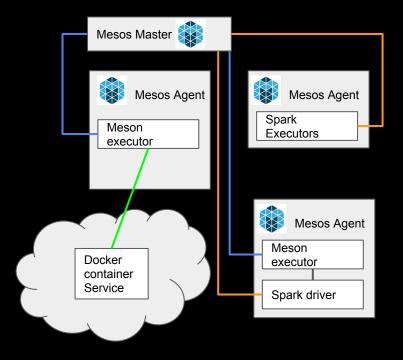


#### **Execution**

Custom executor code for different runtime systems (spark, bash...)

Publish runtime debug infos (logs, url to monitoring Uls...)

Meson executors survive Meson scheduler failure



Looking Forward

#### **Closing the loop**

Interact with Meson from the running job

Output parameters to leverage loops and foreach

Expose debugging information through Artifacts

Progress Milestones, Links, Counters, Images, etc.

#### A day in the life of a workflow...

Backfills, work prioritization and parallelism

Avoid re-doing work after fixing a bug and re-deploying a workflow

Explicit (data) lineage

Looking back



2+ years in production

10+ managed and self-service deployed clusters

**1000+** daily Production and A/B Test ML pipelines

2000+ EC2 instances in Spark/Mesos compute pool

20000+ of steps run per day

#### One Abstraction doesn't fit all

#### Evidenced by the many names:

- Workflow
- ProcessFlow
- Pipeline
- DAG
- DataFlow

# Over specialization will inevitably weaken other use cases.



#### One Abstraction doesn't fit all

With the REST API Meson provides "workflows as a service". Enables many domain specialized abstractions:

- A/B test orchestration
- ML orchestration
- ETL pipelines
- Notebook Automation
- And more..

