

#### Vincent Lesierse

# Technical Product Manager - Vamp.io

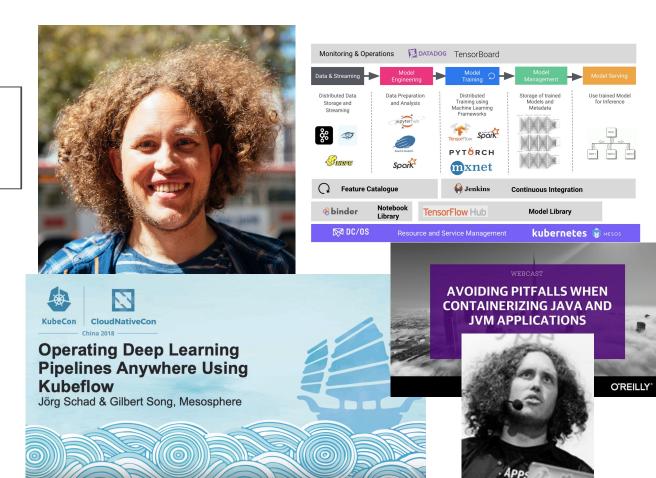
- Previous
  - Software Engineer
  - Software Architect
  - Cloud Architect
- @vlesierse



#### Jörg Schad

Head of Engineering & Machine Learning @ ArangoDB

- Previous
  - Suki.ai
  - Mesosphere
  - PhD DistributedDB Systems
- @joerg\_schad



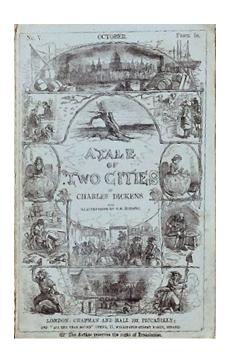
#### A tale of two worlds

It was the best of times, where Microservices allowed to break code in manageable pieces,

It was the worst of times where we actually had to deploy all those pieces,

It was the age of wisdom, where Machine Learning allowed for great discoveries,

it was the age of foolishness, where we suffered to productionize these models...



### A tale of two worlds







**Deployment of Machine Learning Models** 

#### Microservices...



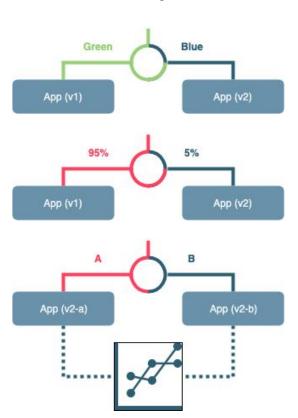
- Independent teams owning their services
- More frequent deployments
- Not more frequent incidents
- Enabling experimentation

# Microservice Deployments

Blue/Green

Canary

A/B Testing Shadow Deployments





Our mission: Provide an easy-to-use and powerful solution to transform software from ideas into value as efficiently as possible, using cloud-native technologies.

Important: Releasing is (much) more than technical deployment. Continuous validation requires observability and "golden metrics". Processes are crucial. Tools and technologies are means to an end.

2014: canary testing & releasing and SLA-based auto-scaling on top of Mesos+Marathon, using HAProxy for networking (service-discovery/load-balancing/mesh/ingress)

2016: added Kubernetes support

Kubecon 2019: added Istio support

# Vamp Stack





















**Amazon Web Services** 





# Machine Learning Model Serving....

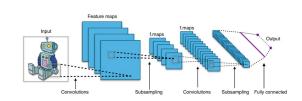


- Productionize trained models
- Retraining and redeployment
- Large number models
- Metadata based Policies
- Testing with live data





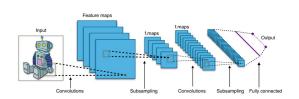




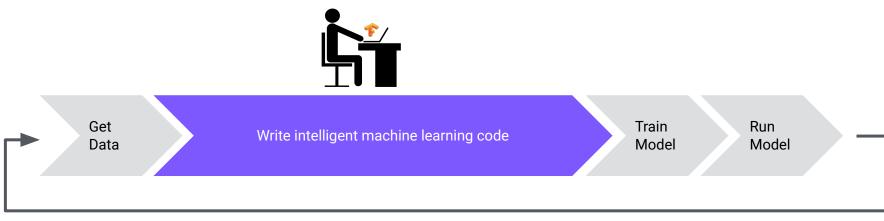








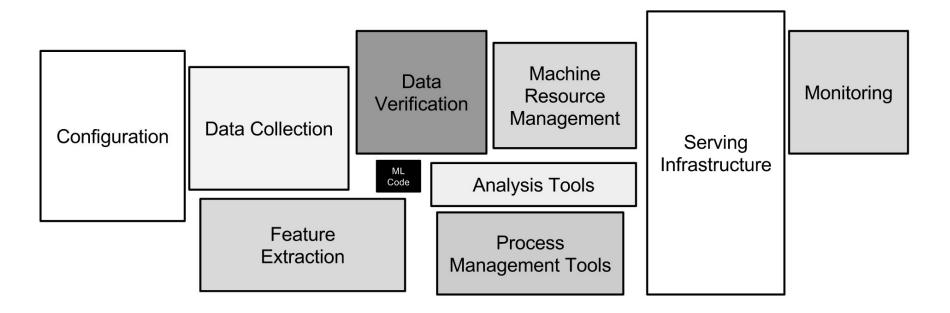
# What you want to be doing







## What you're actually doing



Sculley, D., Holt, G., Golovin, D. et al. Hidden Technical Debt in Machine Learning Systems

#### **Monitoring & Operations**





Distributed Data Storage and Streaming







Data Preparation and Analysis







Distributed Training using Machine Learning Frameworks

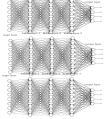




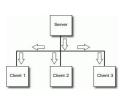


PYTORCH





Use trained Model for Inference







**Jenkins** 

**Continuous Integration** 



Notebook Library



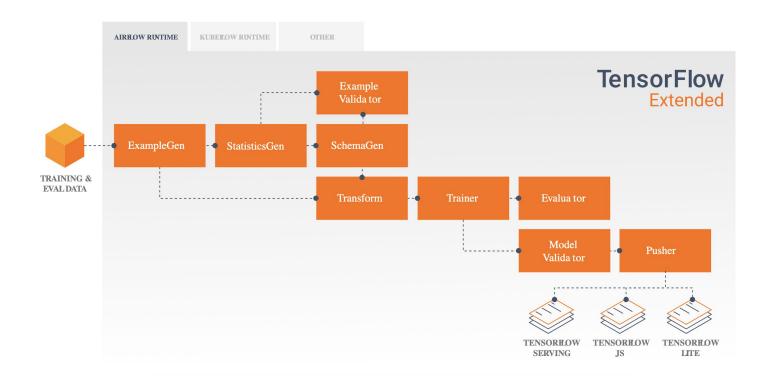
**Model Library** 



Resource and Service Management

kubernetes

## **TensorFlow Extended**



# Challenge: Persona(s)



#### The Rise of the DataOps Engineer

#### Combines two key skills:

- Data science
- Distributed systems engineering

#### The equivalent of DevOps for Data Science

- **Build** automation software to run machine learning systems
- Operate systems so they're available, scalable, and performant
- Evangelize tools and best practices among data scientists

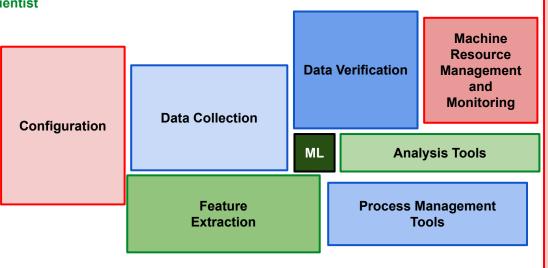


#### **Division of Labor**

**System Admin/ DevOps** 

**Data Engineer/DataOps** 

Data Scientist



Model Monitoring

Serving Infrastructure

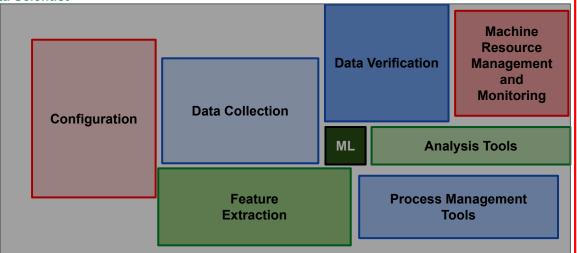
Inspired by "Sculley, D., Holt, G., Golovin, D. et al. Hidden Technical Debt in Machine Learning Systems" article

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Monitoring

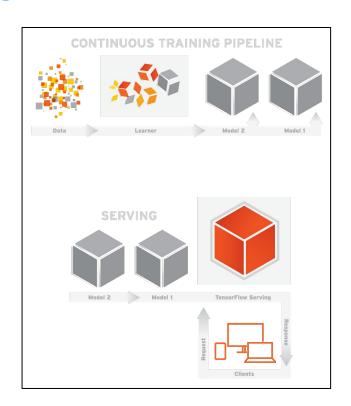
Model

Serving Infrastructure

Inspired by "Sculley, D., Holt, G., Golovin, D. et al. Hidden Technical Debt in Machine Learning Systems" article

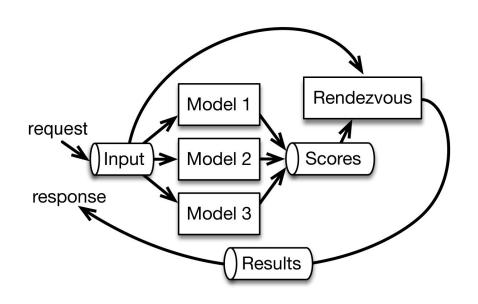
# Machine Learning Model Serving

- Deploying models
  - Choice...
  - Metrics
- Updating models
  - Zero downtime
  - Target environment
- Testing models
  - Test model with live data
- Ensemble Decision
  - Multiple models working together



# Machine Learning Model Serving

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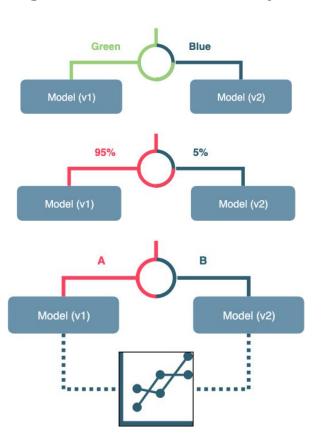


# Machine Learning Model Deployments

Updating model

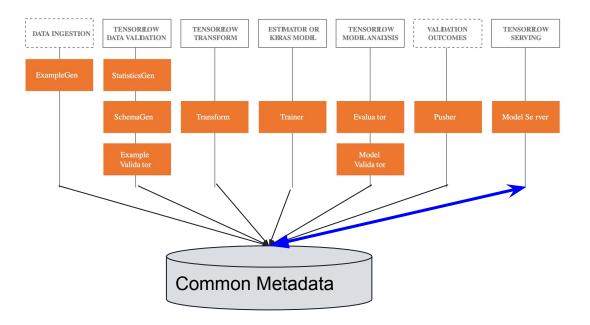
Testing model

Ensemble/
Testing a model

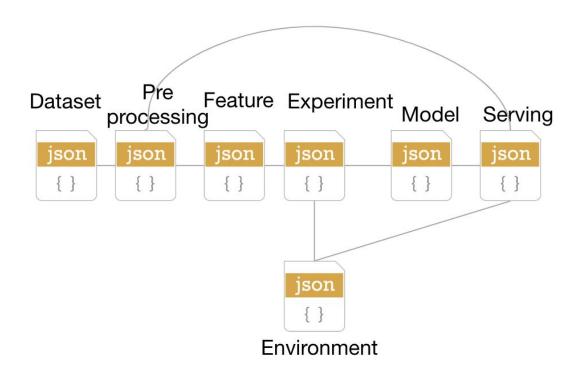


# Metadata... Which model to pick?

- Accuracy
  - Which...
- Latency
- Environments
- Data Privacy
- ....

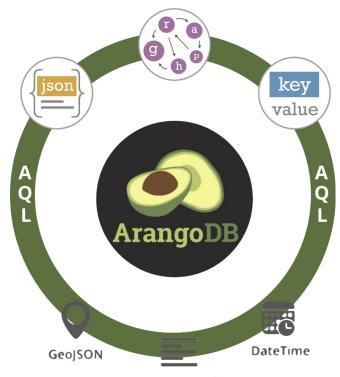


#### Metadata... How to store...





- Native Multi Model Database
  - Stores, K/V, Documents & Graphs
- Distributed
  - Graphs can span multiple nodes
- AQL SQL-like multi-model query language
- ACID Transactions including Multi Collection Transactions

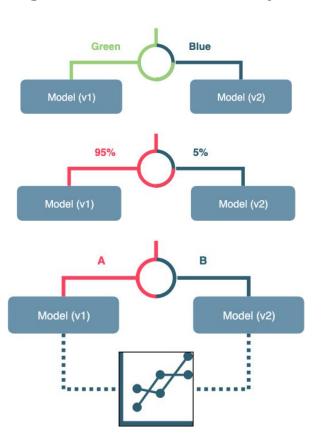


# Machine Learning Model Deployments

Updating model

Testing model

Ensemble/
Testing a model

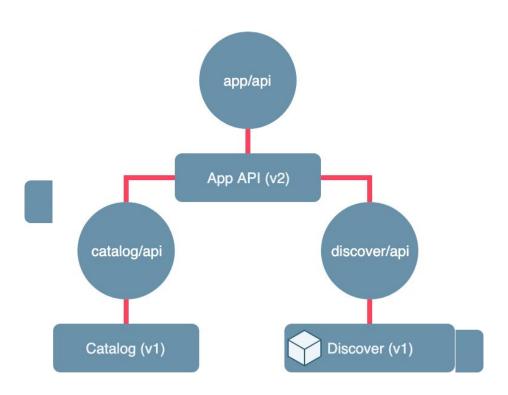


# A tale of two one world

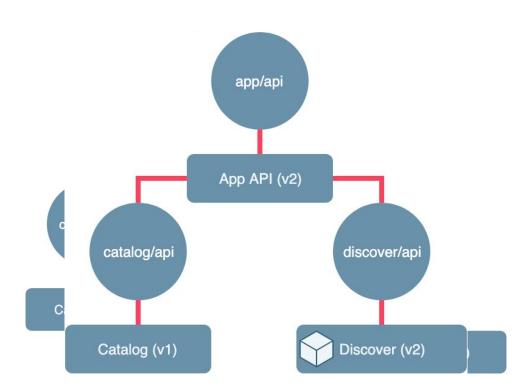


**Deployment of Services** 

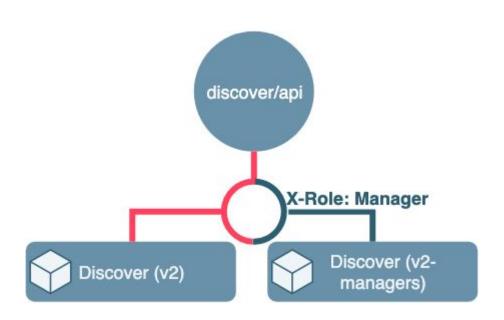
# Demo: Canary Release a Service



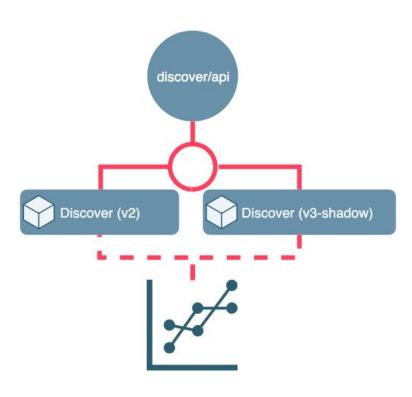
# Demo: Canary Release a Model



# Scenario: Model Segmentation



### Scenario: Shadow Traffic



# What's next for Vamp?

- Integrating Istio for smart networking
- Al/ML based workflows & policies for self-learning and optimisation
- More CI and APM integrations
- Additional release, validation, experimentation and optimisation workflows
- Hybrid containers & serverless support
- Cloud-based version with new & improved core "engine"

#### Engage with us





- Talk to us at Mesosphere stand P13
- Take a look at Vamp: https://github.com/magneticio/vamp
- Istio preview:

https://vamp.io/vamp-istio-preview

- @arangodb
- Take a look at ArangoDB:
- https://www.arangodb.com/
- ArangoML Preview:

https://github.com/arangoml/arangopipe

Download the demo: <a href="https://github.com/magneticio/smartcon">https://github.com/magneticio/smartcon</a>

@vlesierse @joerg\_schad

# A Tale of Two Worlds Canary-Testing for Both ML Models and Microservices



