# **Kubeflow:**Bring your ML project into Production

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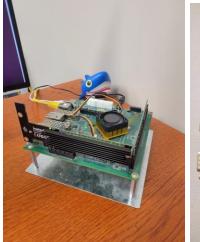
Software Engineer OpenSUSE Member Open-Source Advocator(KYOSS)

Louisville, KY

#### Interests:

- Virtualization
- SysML/Distributed ML
- Infrastructure build and optimization
- ARM64 board Enthusiast
- DevOps

https://xryan.net











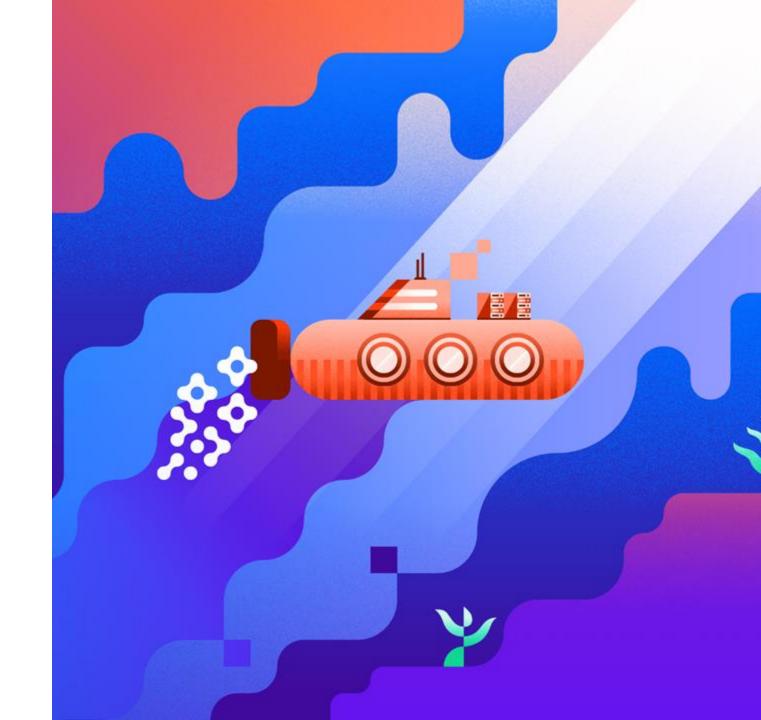






### **Outline**

- Prologue
- Kubeflow
- Kubeflow Components
- Demo
- Beyond
- Q&A













#### Flight Delay Predictor:

https://github.com/xrlyan/Flight-Delay-Prediction-Based-on-Neural-Networks

#### Input:

- Flight no
- Flight date

#### Output:

Delay possibility

#### Features:

- Depart/Arrive airport
- Depart/Arrive time
- Depart/Arrive city weather
- Flight model
- Flight History delay rate

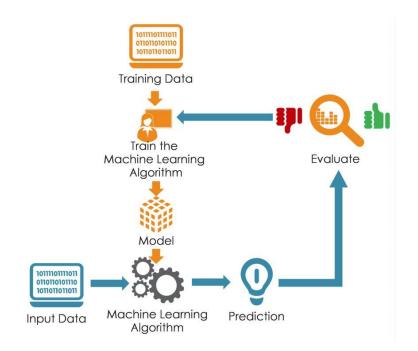




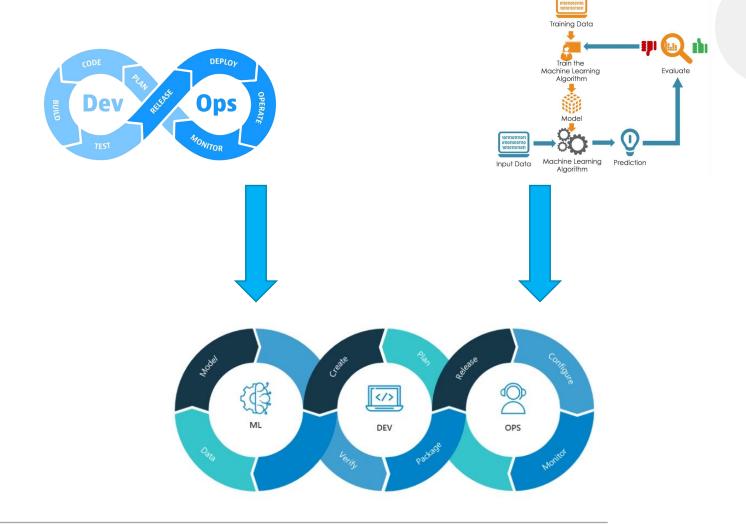
#### As a Software Engineer:



#### As a Data Scientist/Engineer:



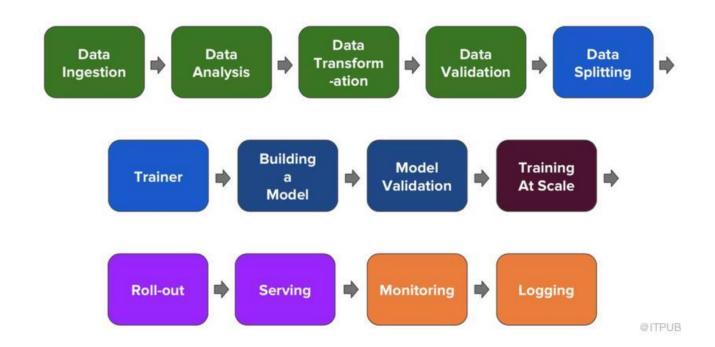






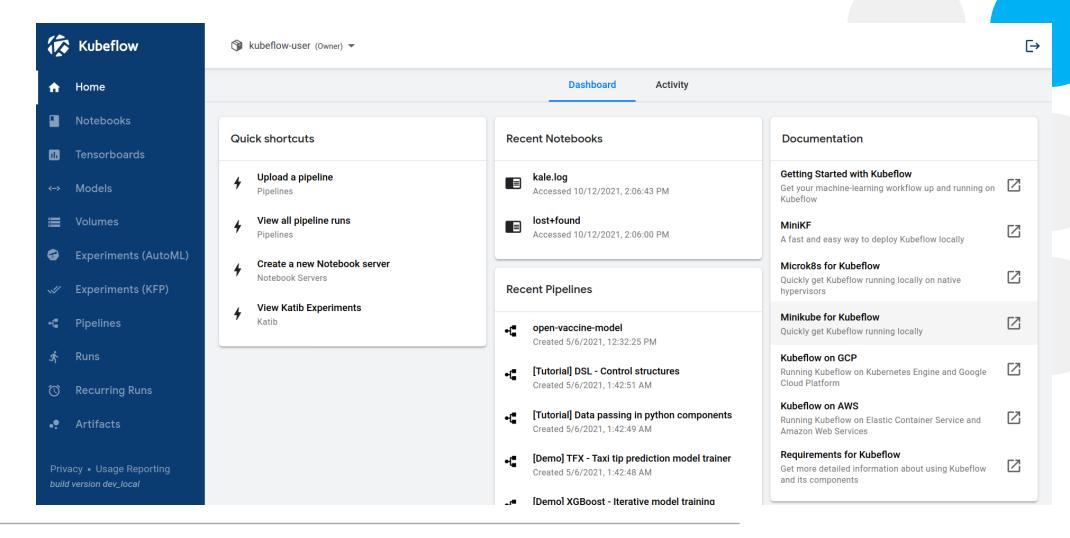
### **Prologue**

#### Eventually, it becomes:





### **Kubeflow-Central-dashboard**





### **Kubeflow**

#### Kubeflow is:

K8S + TensorFlow

**Application Toolkit** 

Orchestration

**Cloud Native** 

DevOps/MLOps



#### Kubeflow is not:

K8S + TensorFlow

Application

Scheduler

Machine Learning Algorithm

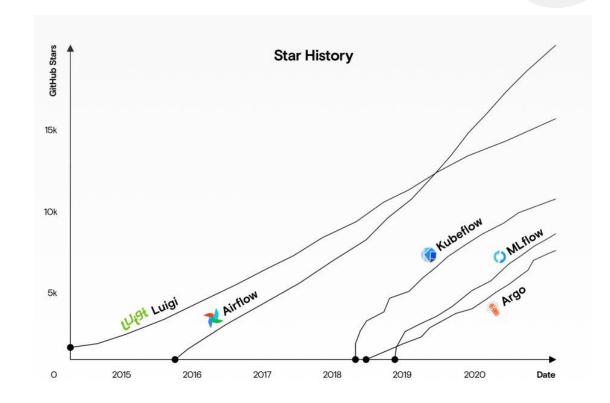
Machine Learning Framework



Machine Learning Orchestration Platform:

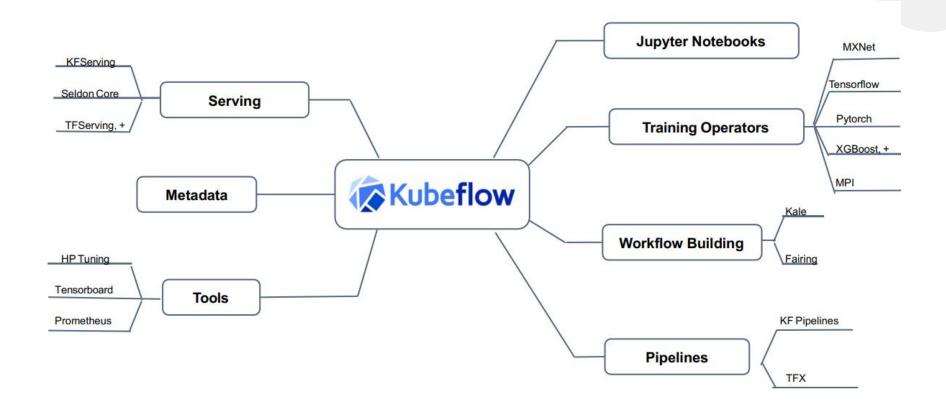
- 1. Orchestrate pipeline
- 2. Orchestrate ML task

Great mind think alike!



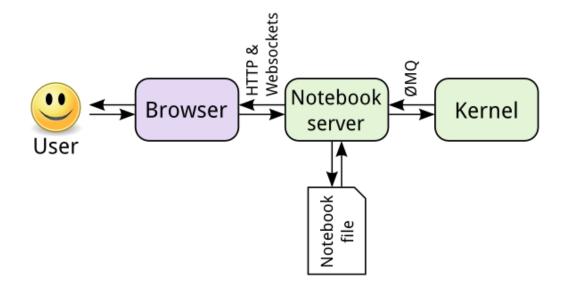


### **Kubeflow components**





### Jupyter-notebook



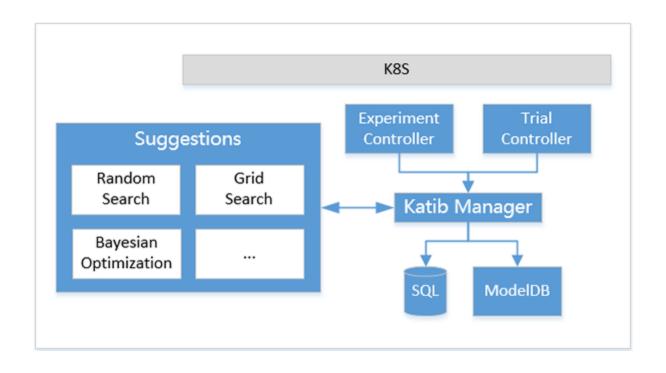


#### **Architecture**

			Tra	in and Predic	t				
Preprocessor				Builder		Deployer			
Notebook	Function	Python	Append	Docker	Cluster	Job	TfJob	Serving	



#### An implementation for AutoML: tune hyperparameter automatically



#### Three CRDs:

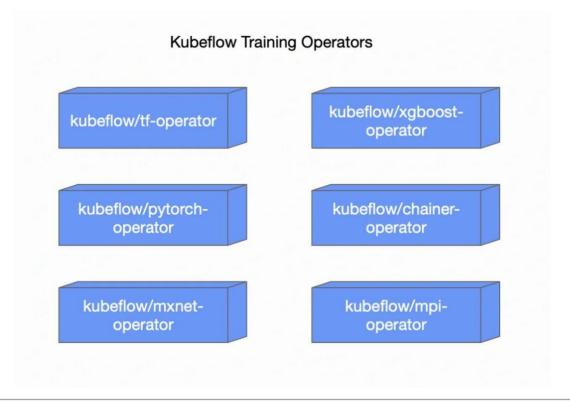
experiment suggestion trial

The experiment creates multiple trials based on different suggestion algorithms.

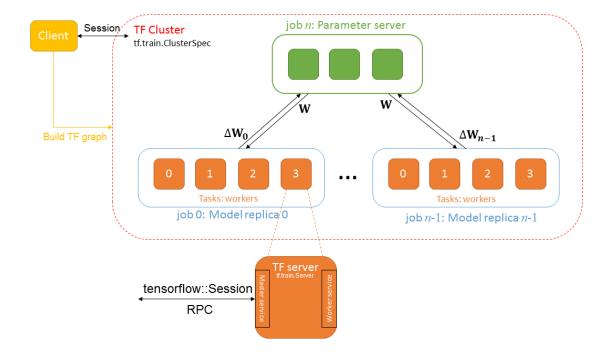


### Training Operator

Operator = Controller + CRD + Webhook Tool: kubebuilder



### **TF-Operator**



Chief coordinate training job

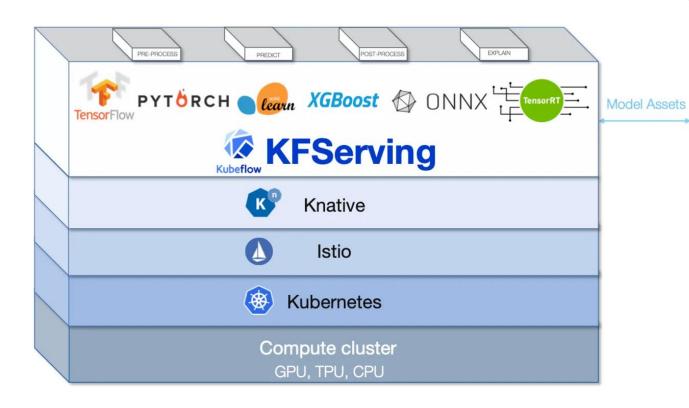
PS server, parameter

Worker

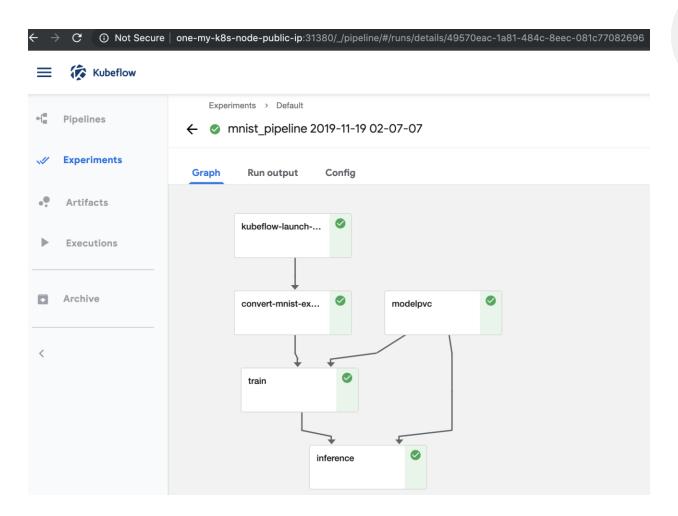
Evaluator



The last mile!

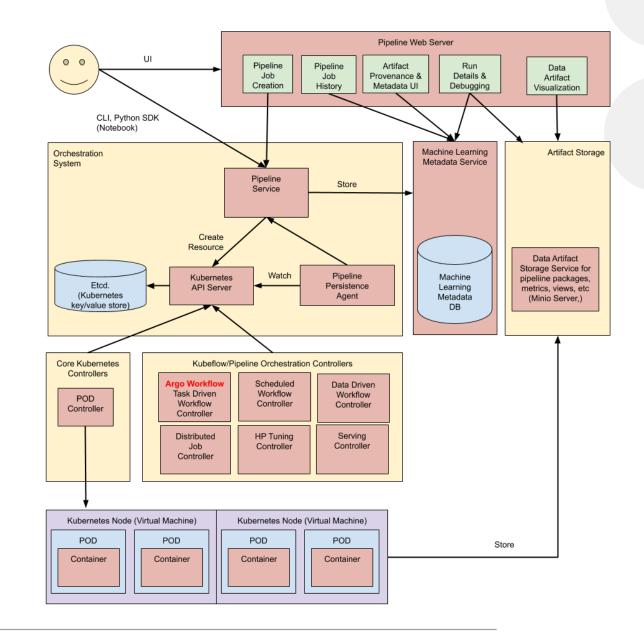


### KFP DAG



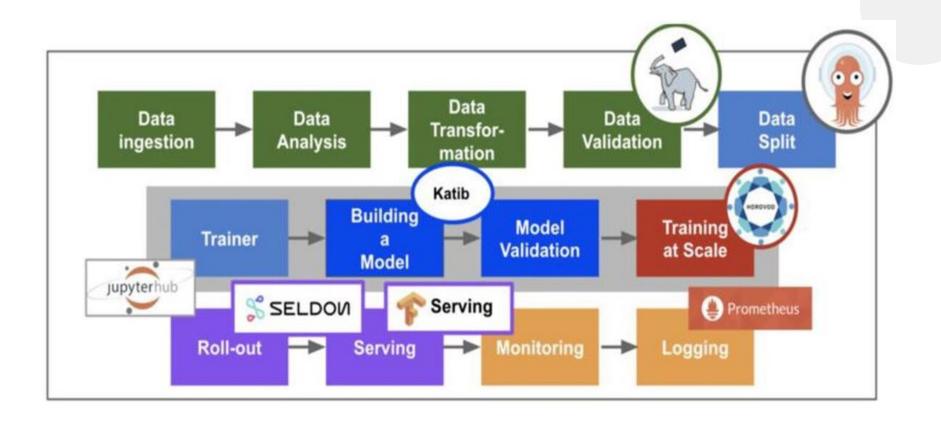


## **KFP Architecture**





### **Kubeflow**





#### Experiment Platform:

DO-DOKS: kubernetes v1.24

Kubeflow: v1.6.1

Linux Distro: Debian 10

#### Demo:

while! kustomize build example | kubectl apply -f -; do echo "Retrying to apply resources"; sleep 10; done kubectl --kubeconfig=/Users/lyan/kubeflow-kubeconfig.yaml port-forward svc/istio-ingressgateway -n istio-system 8080:80

#### Local Setup:

Juju + microk8s: kubernetes v1.22

Kubeflow: v1.6.0

Linux Distro: Ubuntu Jammy

https://charmed-kubeflow.io/docs/quickstart



#### Distributed Machine Learning(\*)

Why?

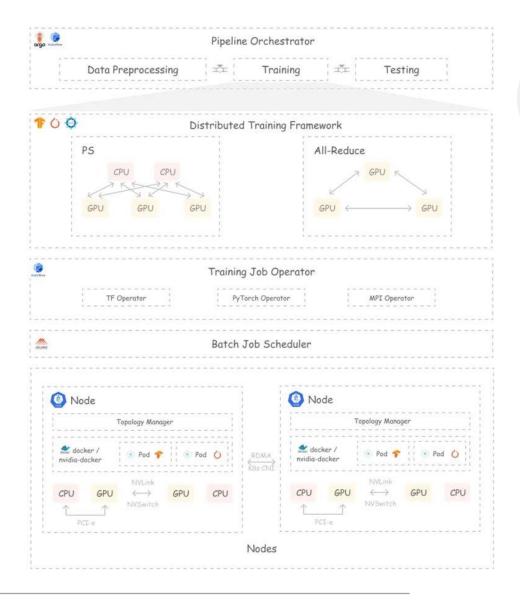
Scalability, we really do not need it if it is a small dataset or model or customer base.

What?

Training Operator Scheduling Inference Model Optimization

How?

Simulate/predict for scheduler Model compiler



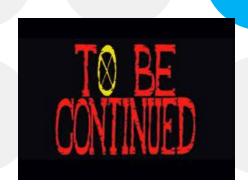




### **Beyond**

#### **Embedded Model**







### Lesson learned

#### 1. Deployment

- 1. Kubevirt 1.22 + kustomerize
- 2. Disable TLS
- 3. Setup/Enable StorageClass

#### 2. Running

- 1. docker runtime re-size
- 2. docker repository setup

#### 3. Training model

- 1. ML training requests a lot resources
- 2. Need to do a lot of experiments
- 3. Setup environment is time consuming
- 4. Needs automation/pipeline

#### 4. System Failure / Efficiency

Monitor large scale machine clusters are difficult Resource Competition

### **Q&A**

#### Thanks!

#### Claim:

All the information is based on personal using experience, no preference or commercial advertising. If there are any conflicts, please refer to the statement from providers.



### **AI Cloud Providers**



























### **Support Matrix**

	M60	P4	P40	P100	T4	RTX 6000	V100	A10	A40	A100	Notes
Aliyun		<b>✓</b>		<b>✓</b>	<b>✓</b>		<b>✓</b>			<b>✓</b>	
AWS	<b>/</b>		<b>/</b>		<b>/</b>		<b>/</b>	<b>✓</b>		<b>✓</b>	
Baidu		<b>✓</b>			<b>/</b>		<b>✓</b>			<b>✓</b>	
Google		<b>/</b>		<b>✓</b>	<b>/</b>		<b>✓</b>			<b>✓</b>	TPU
IBM	<b>✓</b>			<b>/</b>			<b>✓</b>				
Microsoft	<b>✓</b>		<b>✓</b>	<b>/</b>	<b>✓</b>		<b>/</b>	<b>/</b>		<b>/</b>	FPGA/AMD
Oracle				<b>/</b>			<b>/</b>			<b>/</b>	
Tencent		<b>/</b>	<b>✓</b>		<b>✓</b>		<b>✓</b>			<b>✓</b>	
Linode						<b>/</b>					
Paperspace							<b>✓</b>				
Lambda						<b>✓</b>	<b>✓</b>				
Vultr										<b>/</b>	vGPU/MIG

#### https://developer.nvidia.com/cuda-gpus

https://www.nvidia.com/en-us/data-center/gpu-cloud-computing/

#### **NVIDIA Data Center Products**

GPU	Compute Capability
NVIDIA A100	8.0
NVIDIA A40	8.6
NVIDIA A30	8.0
NVIDIA A10	8.6
NVIDIA A16	8.6
NVIDIA A2	8.6
NVIDIA T4	7.5
NVIDIA V100	7.0
Tesla P100	6.0
Tesla P40	6.1
Tesla P4	6.1
Tesla M60	5.2
Tesla M40	5.2
Tesla K80	3.7
Tesla K40	3.5
Tesla K20	3.5
Tesla K10	3.0



### **Al Cloud Service**

#### •IAAS

- ML VM Image
- Container:
  - Docker
  - NGC
- Conda/pip3

#### PaaS

Help manage data and model (paperspace, Colaboratory)

#### •SaaS

Help consume AI solution (IBM Watson, Google voice)