

apiVersion: slide/v1

author:

- - affiliation: "IBM Center for Open Source Data and AI Technologies (CODAIT)"
  - name: "Romeo Kienzler"
  - role: "Data Scientist"
- - affiliation: "University Hospital Basel"
  - name: "Ivan Nestic"
  - role: "ML Engineer"
- - name: "The Elyra, JupyterLab, Kubernetes and KubeFlow authors"

kind: Titlepage

published:

- - type: video
  - url: "https://www.youtube.com/watch?v=H8WskMEUI74"
- - type: conference
  - url: "http://conference.scipy.org/proceedings/scipy2021/pdfs/ivan\_nestic.pdf"

title:

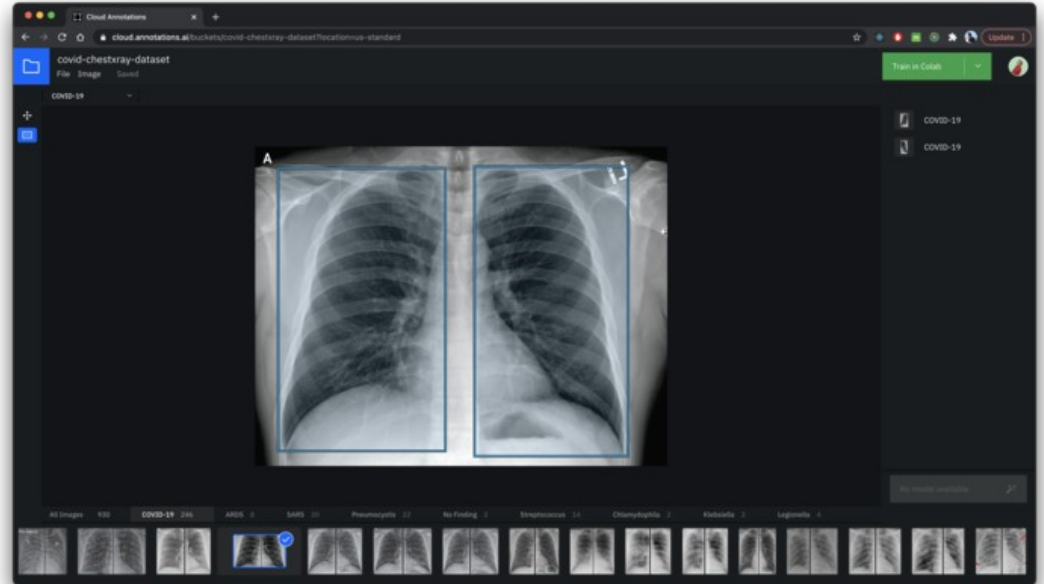
- "CLAIMED – Component Library for AI, Machine Learning, ETL and Data Science"
- "A Kubeflow no code / low code architecture"

# Requirements

- Rapid prototyping using visual editing and notebooks
- Seamless scaling during development and deployment
- GPU support
- ML tools: PyData stack, TensorFlow, PyTorch, ...
- life science tools: DICOM input, DICOM output, ...
- Reproducibility
- Data lineage
- Reference implementation in open source
- Collaboration support

# *Cloud annotations provides...*

*Browser based  
image labeling:  
Classification /  
Object recognition  
training data*



**IBM Cloud Annotations**

*...in Open Source*

# *Docker provides...*

*OS-level “virtualization” to deliver  
and run software in packages called  
containers*

*Lightweight “virtualization”*

*Security and isolation*

*Super-fast startup/teardown*



*...on top of Linux<sup>4</sup>*

# *Kubernetes provides...*

*Some prominent users: Adidas, Booking.com,  
Box, Google, Huawei, IBM, The New York  
Times, ING, ricardo.ch, Spotify, Wikimedia,  
Zalando*

## *Container Orchestration*

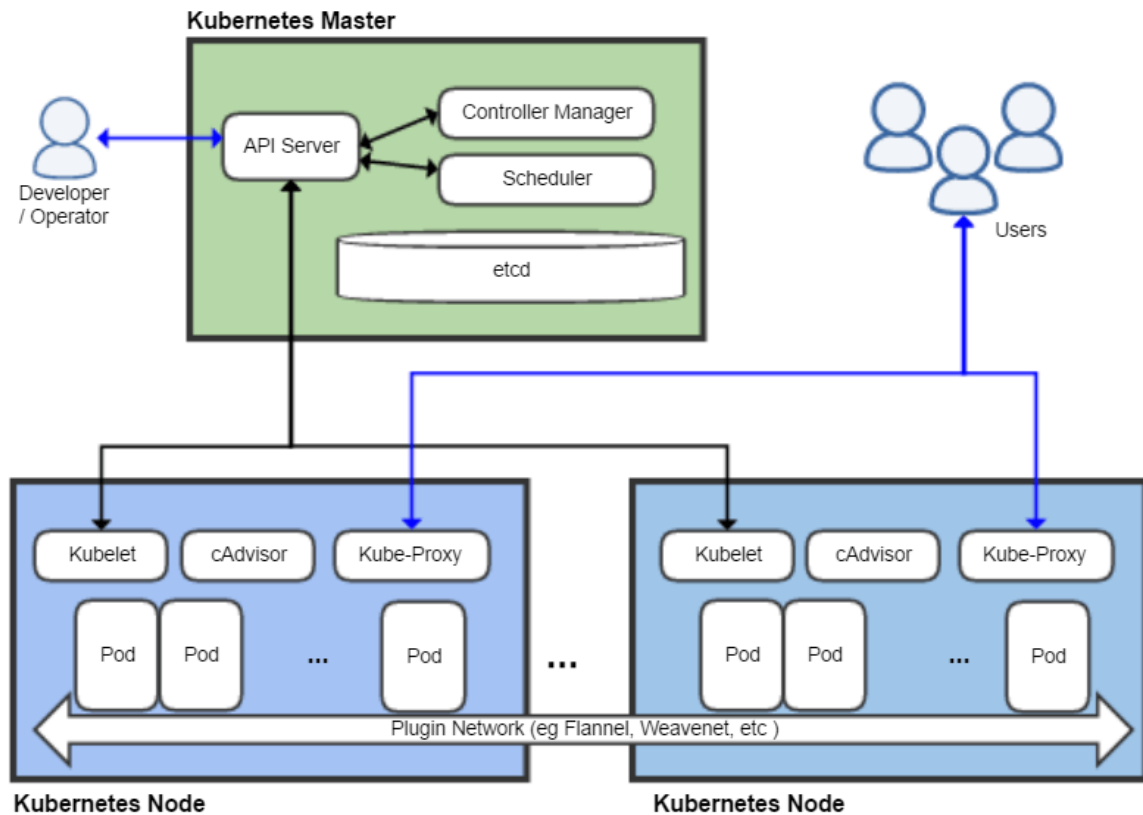
### *Deployment, scaling and management*

### *High availability*



## *...on top of Linux Clusters*

# Kubernetes architecture...



# Kubeflow provides...

Some prominent users:

Amazon Web Services

IBM Watson Services

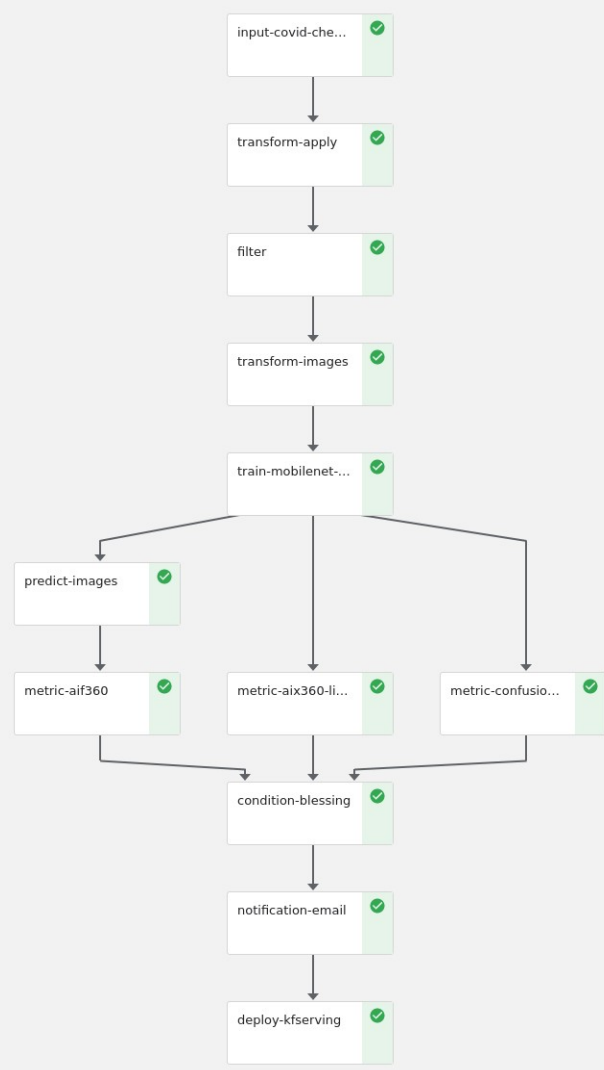
IBM's top clients



**Kubeflow**

...on top of Kubernetes

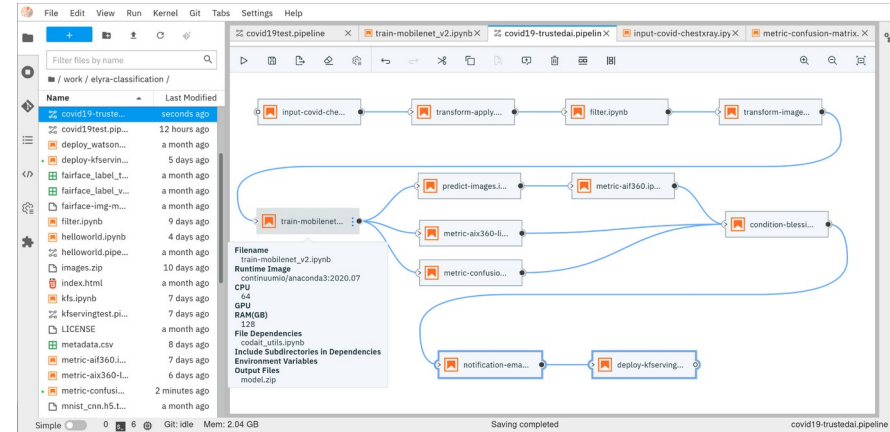
AutoML,  
Deployment,  
Reproducibility  
Notebooks,  
Pipelines, Serving,  
Training, Scale



# *Elyra provides...*

*Some prominent users: Three IBM clients, one Fortune 500 company*

*No Code / Low Code ML Pipeline Design*  
*Re-usable pipeline components*  
*Interchangeability of Engines*  
*(Kubeflow, Airflow, ...)*



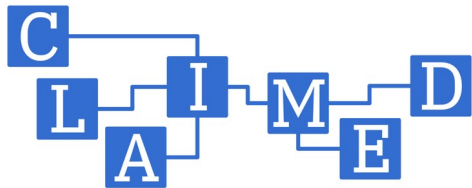
*...on top of JupyterLab, VSCode, ...*



# CLAIMED...

**Component Library for AI,  
Machine Learning, ETL and  
Data Science**

*Portability  
No Code / Low Code  
Pipeline Components  
Jupyter Notebooks  
Sample Pipelines*



*Current users:  
University Hospital Basel  
Motion.AI*

```
component-library
├── analyze
│   ├── spark-ts-trends.ipynb
│   └── checkpoint
│       ├── store_asset.ipynb
│       ├── claimed_utils.py
│       └── CONTRIBUTING.md
├── deploy
│   ├── condition-blessing.ipynb
│   ├── deploy-kfserving.ipynb
│   ├── deploy_watson_machine_learning.ipynb
│   └── README.md
├── filter
│   ├── filter.ipynb
│   ├── README.md
│   └── spark-sample.ipynb
├── input
│   ├── claimed_utils (1).py
│   ├── claimed_utils.py
│   ├── defunct-ray-input-climate-copernicus.ipynb
│   ├── input-climate-copernicus.ipynb
│   ├── input-covid-chestxray.ipynb
│   ├── pycache
│   │   └── claimed_utils.cpython-38.pyc
│   └── README.md
├── LICENSE
├── metric
│   ├── metric-aif360.ipynb
│   ├── metric-aix360-lime.ipynb
│   ├── metric-confusion-matrix.ipynb
│   └── README.md
├── monitoring
│   ├── notification-email.ipynb
│   └── README.md
├── predict
│   ├── predict-images.ipynb
│   └── README.md
├── README.md
├── train
│   ├── README.md
│   └── train-mobilenet_v2.ipynb
├── transform
│   ├── README.md
│   ├── spark-csv-to-parquet.ipynb
│   ├── spark-parquet-to-csv.ipynb
│   ├── transform-apply.ipynb
│   └── transform-images.ipynb
└── visualize
    └── map-from-coordinates.ipynb
```

*...on top of Elyra and KubeFlow*

# Other workflow managers..

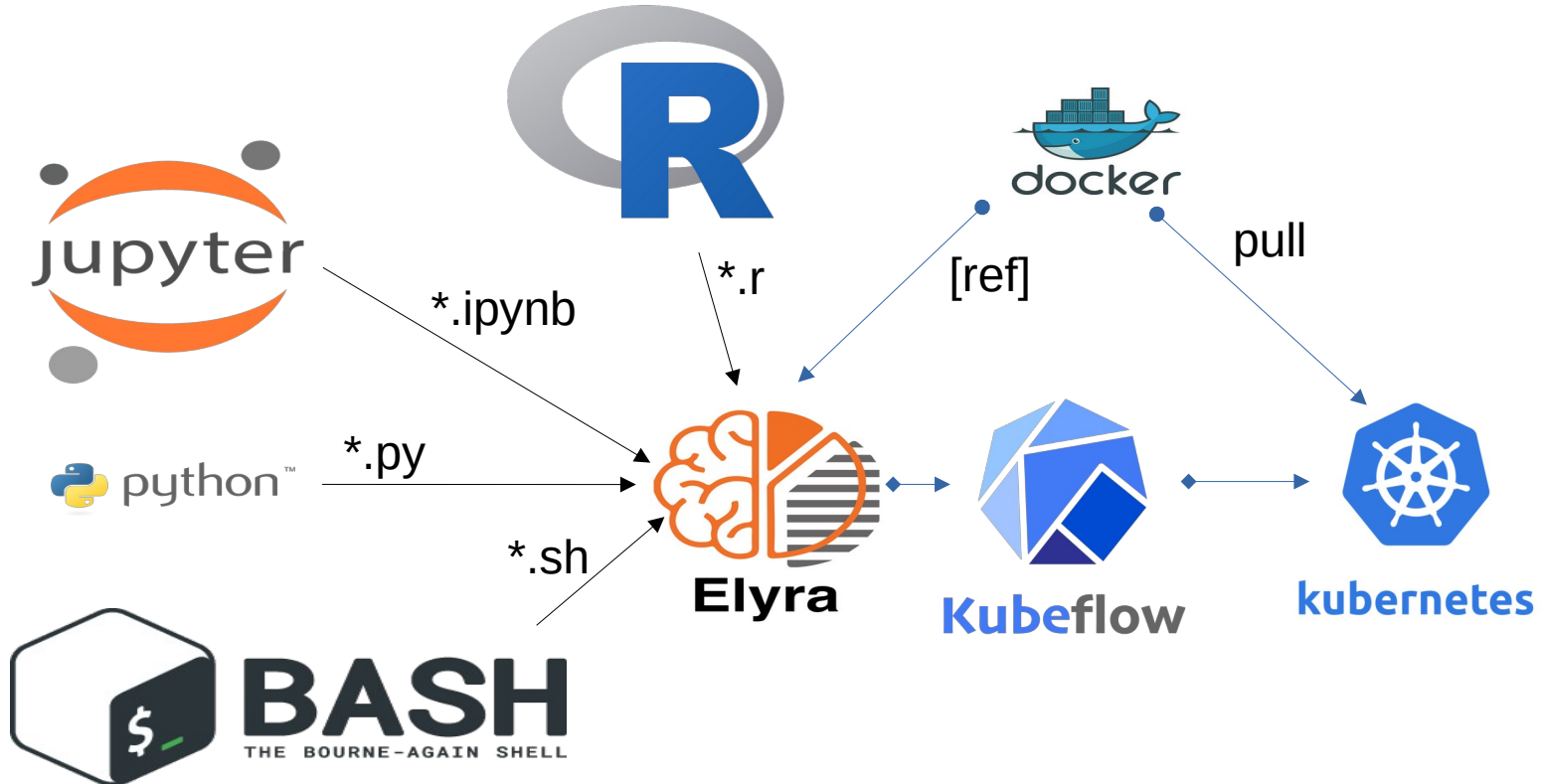
Requirement		KF	AF	Slurm	SM	Qsub	HTCondor	Reana
Kubernetes	Sup- port	X	X		X		X	X
GPU support		X	X	X	X	X	X	X
Component Library		X						
Reproducibility		X	X		X		X	X
Data Lineage		X						X

# Other low code / no code tools..

Requirement	Nifi	NodeRED	KNIME	Galaxy	Elyra
Kubernetes Support				X	X
GPU support				X	X
Component Library	X	X	X	X	X
Reproducibility	X		X	X	X
Data Lineage	X			X	X
Visual Editing	X	X	X	X	X
Jupyter Notebooks					X

*vs. Elyra*<sub>11</sub>

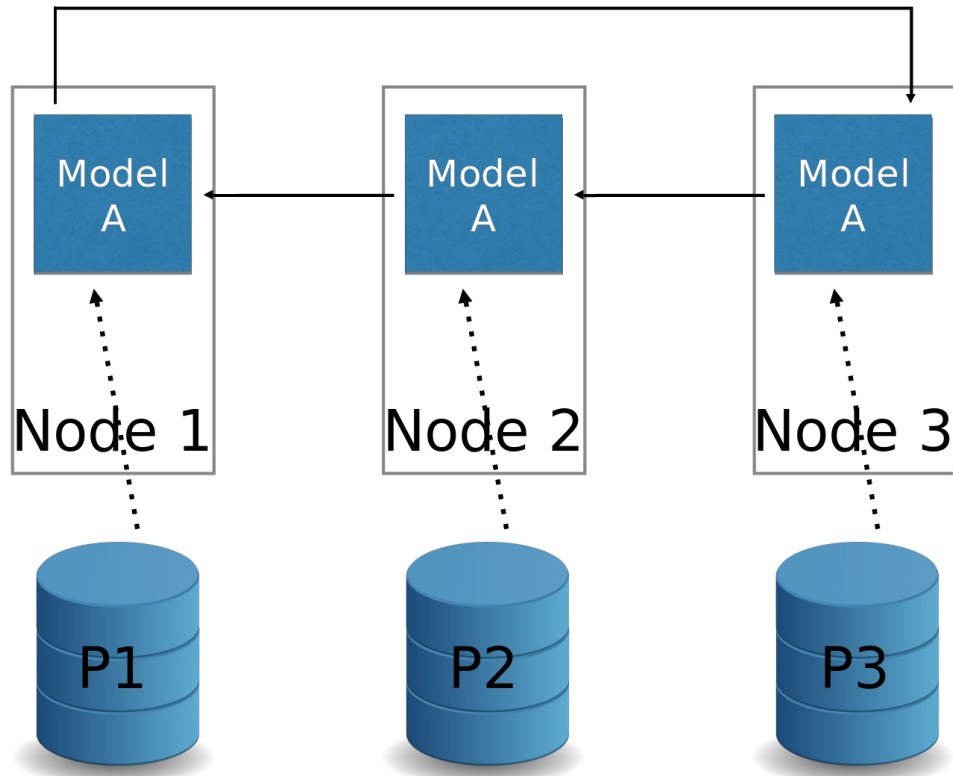
*Full (source code) round-trip visual editing...*



*... and Kubeflow deployment using the Elyra Pipeline Editor*

# Example Pipeline Components

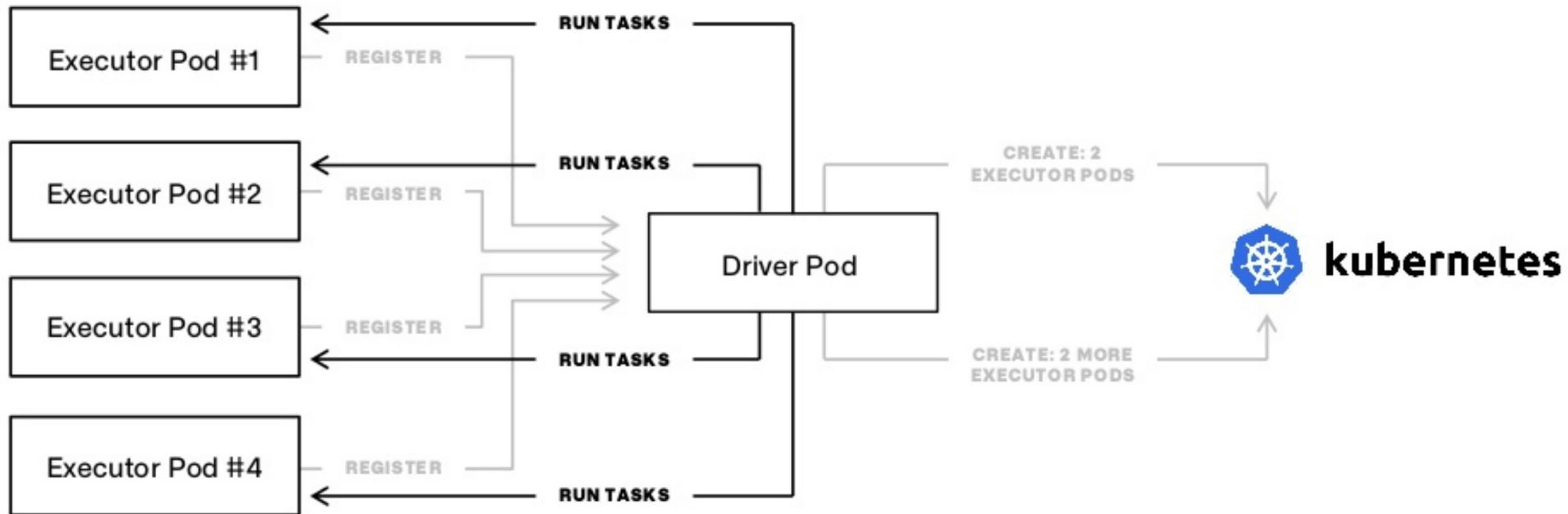
**Category:** Training **Group:** Distributed **Name:** TFJob



The TFJob operator supports parallel training on multiple nodes and GPUs

# Example Pipeline Components

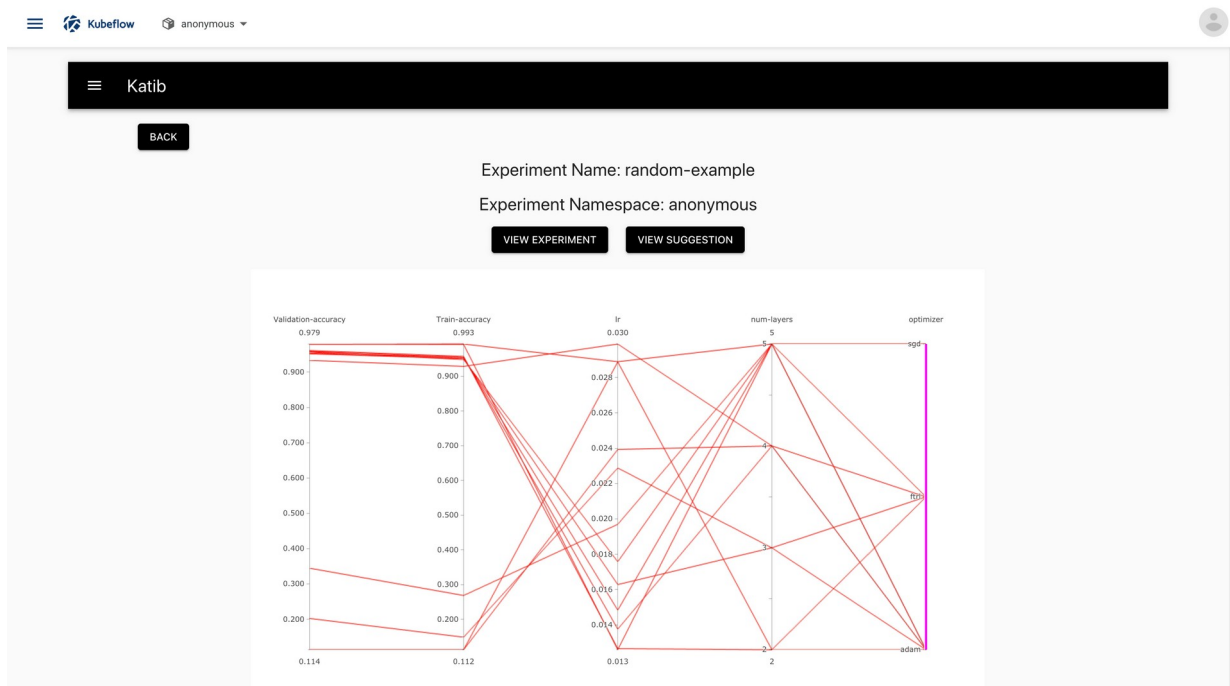
**Category:** Processing **Group:** Distributed **Name:** SparkJob



The SparkJob operator supports parallel processing on multiple nodes

# Example Pipeline Components

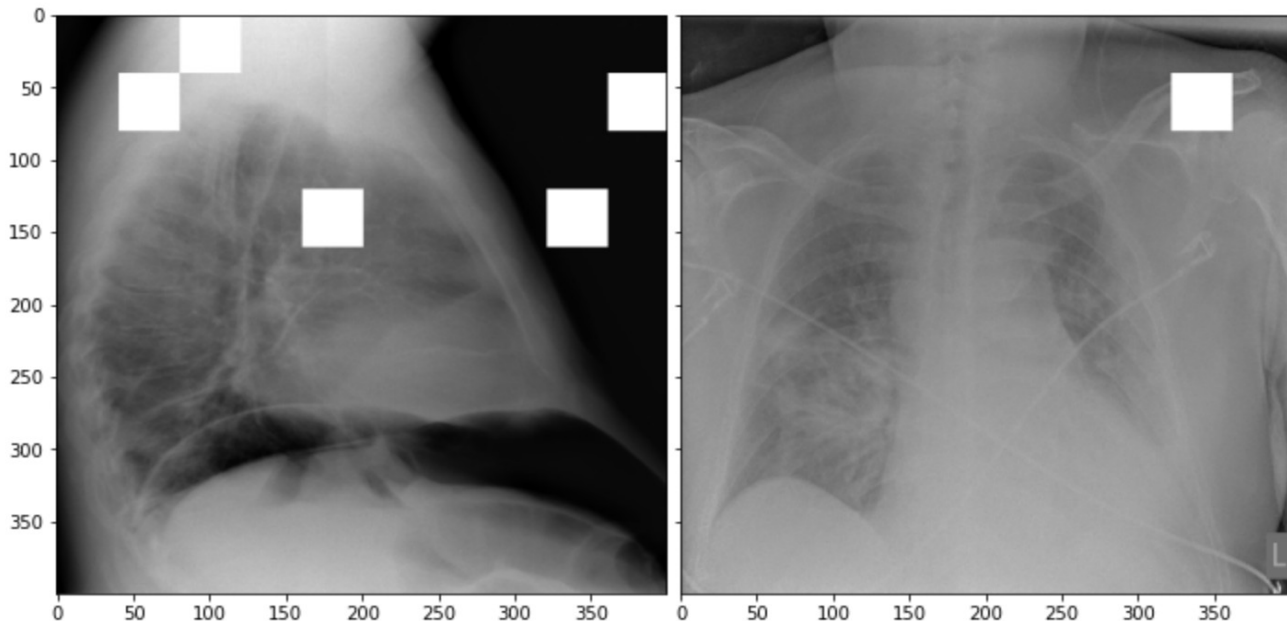
Category: Tuning Group: **Hyperopt** Name: Katib



Visualization of a hyper parameter optimization result

# Example Pipeline Components

**Category:** Metric **Group:** Explainability **Name:** AIX360/LIME

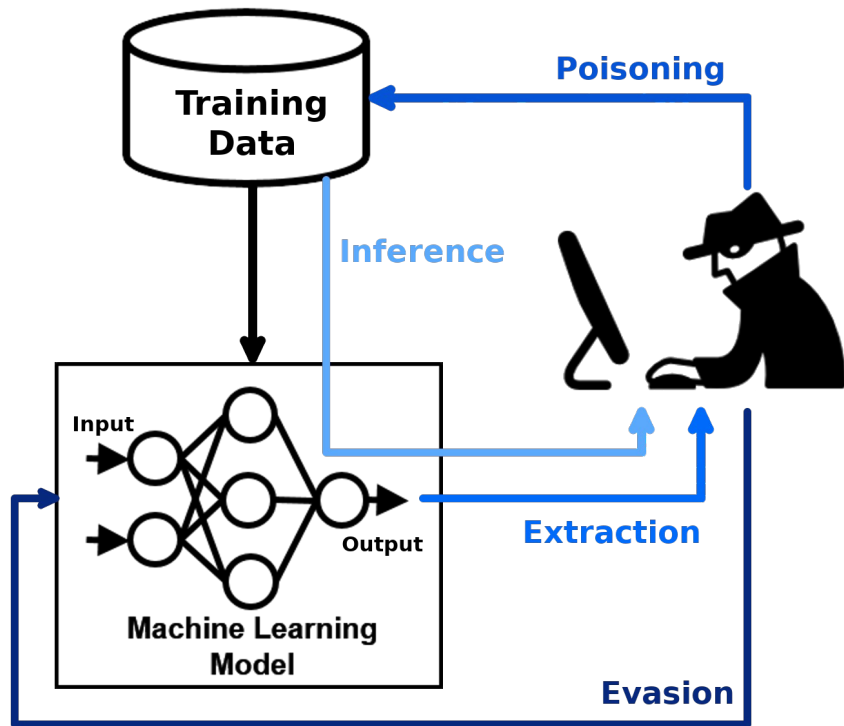


Example on how LIME helps to identify classification relevant areas of an image



# Example Pipeline Components

**Category:** Metric Group: **Adversarial Robustness** Name: ART



Example on how Adversarial Attacks happen

# Example Pipeline Components

## Category: Metric Group: AI Fairness Name: AIF360

Dataset: German credit scoring

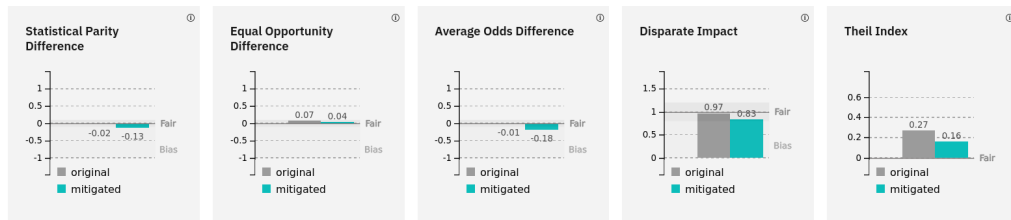
Mitigation: [Adversarial Debiasing algorithm applied](#)

### Protected Attribute: Sex

Privileged Group: **Male**, Unprivileged Group: **Female**

Accuracy after mitigation changed from 75% to 70%

Bias against unprivileged group unchanged after mitigation (0 of 5 metrics indicate bias)

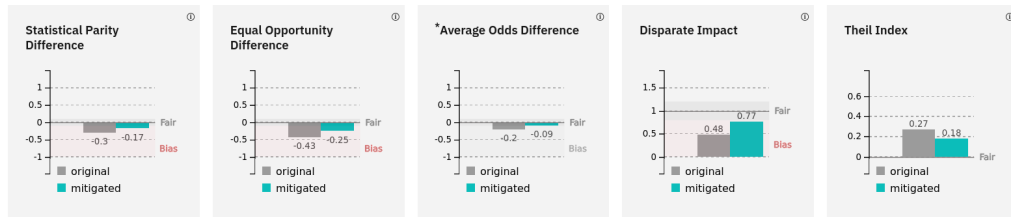


### Protected Attribute: Age

Privileged Group: **Old**, Unprivileged Group: **Young**

Accuracy after mitigation changed from 75% to 69%

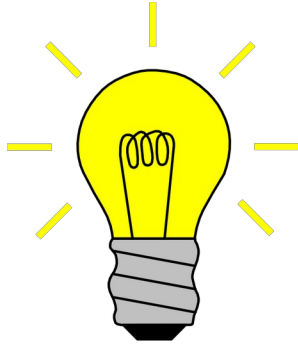
Bias against unprivileged group was reduced to acceptable levels\* for 1 of 4 previously biased metrics (3 of 5 metrics still indicate bias for unprivileged group)



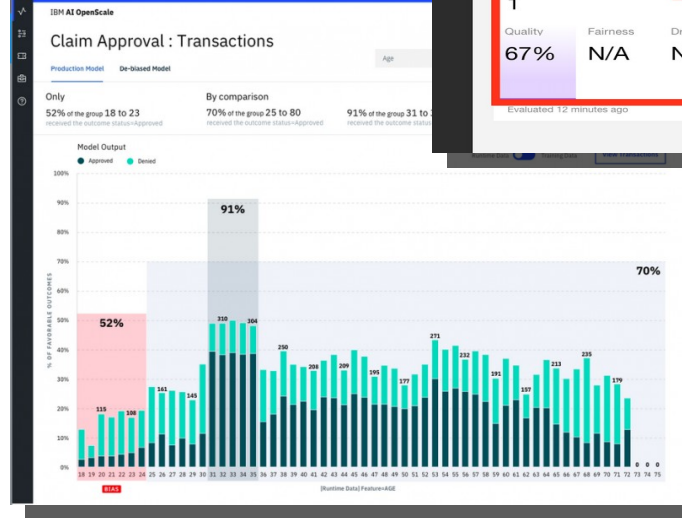
Example on how the AIF360 toolkit computes fairness metrics and mitigates bias

# IBM Watson OpenScale

IBM Watson OpenScale uses the same  
Open Source components on top of  
Kubernetes, Kubeflow and KFServing

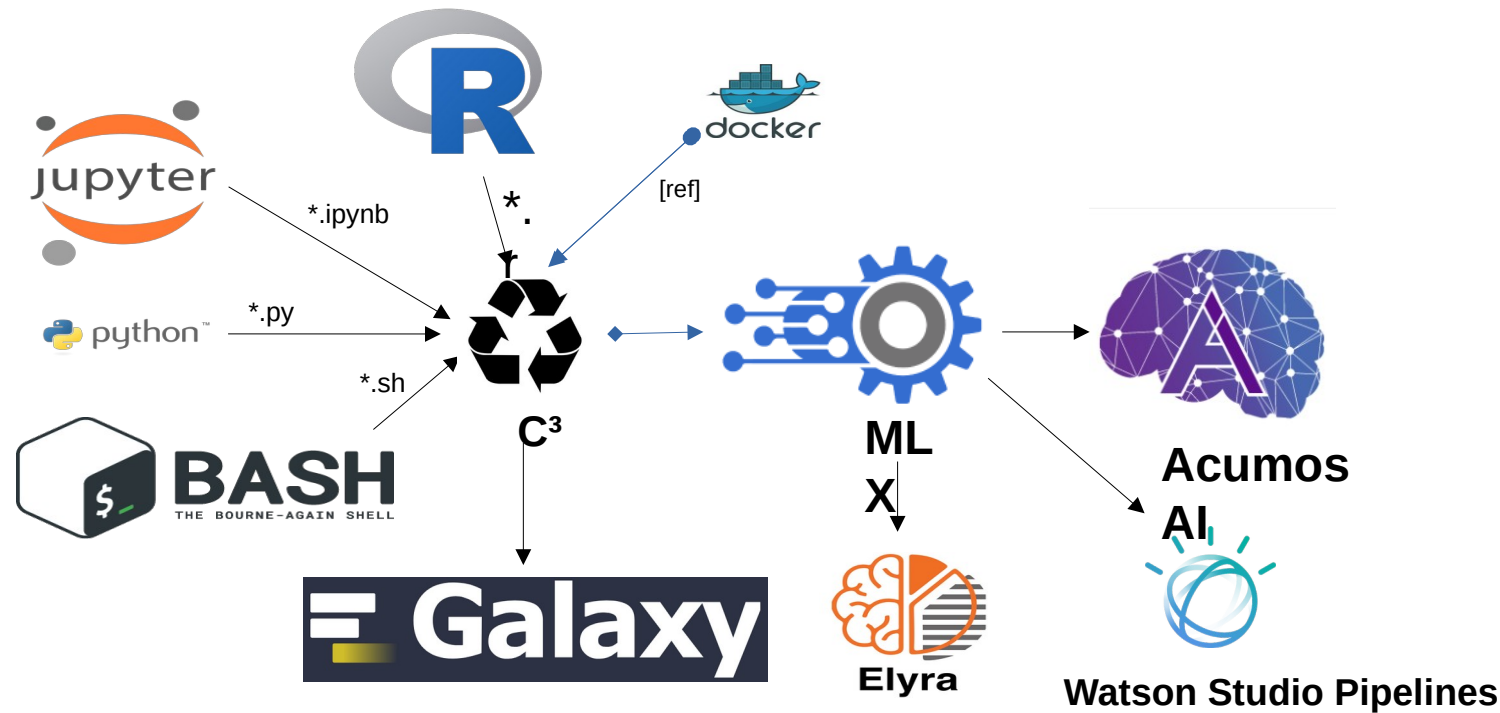


*Did you know?*



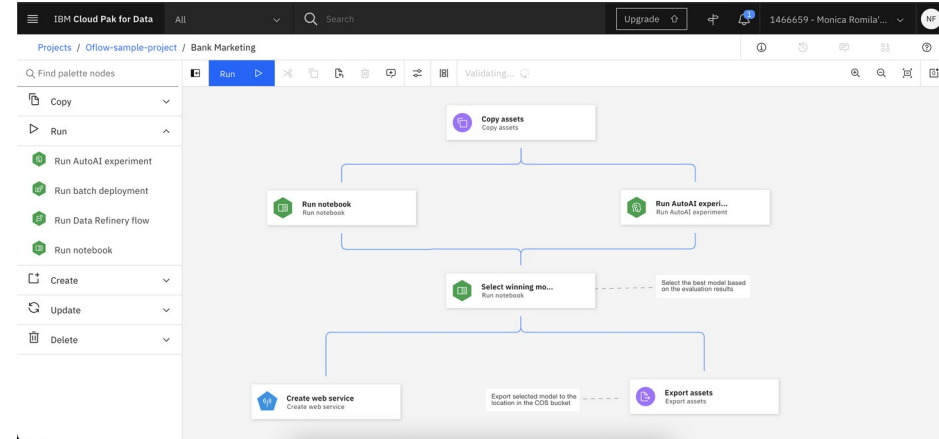
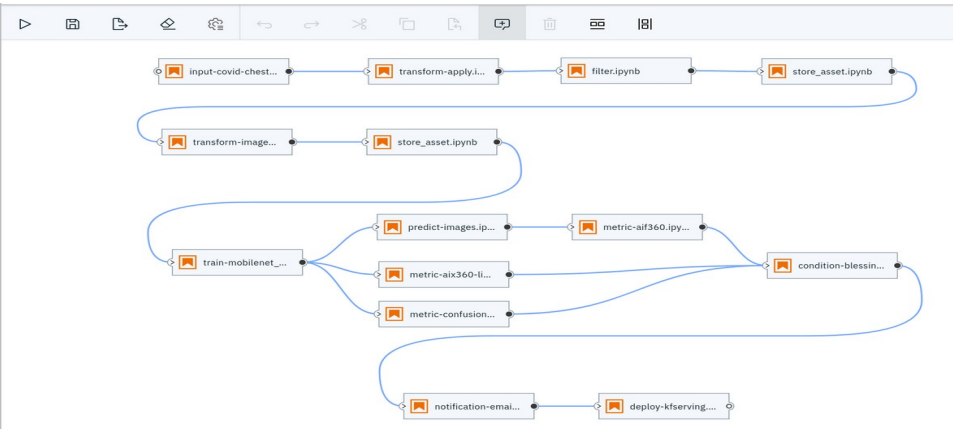
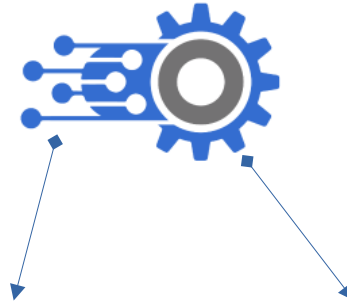
Model Monitoring and visualization of bias  
(IBM Watson OpenScale)

# Component creation and cataloging...



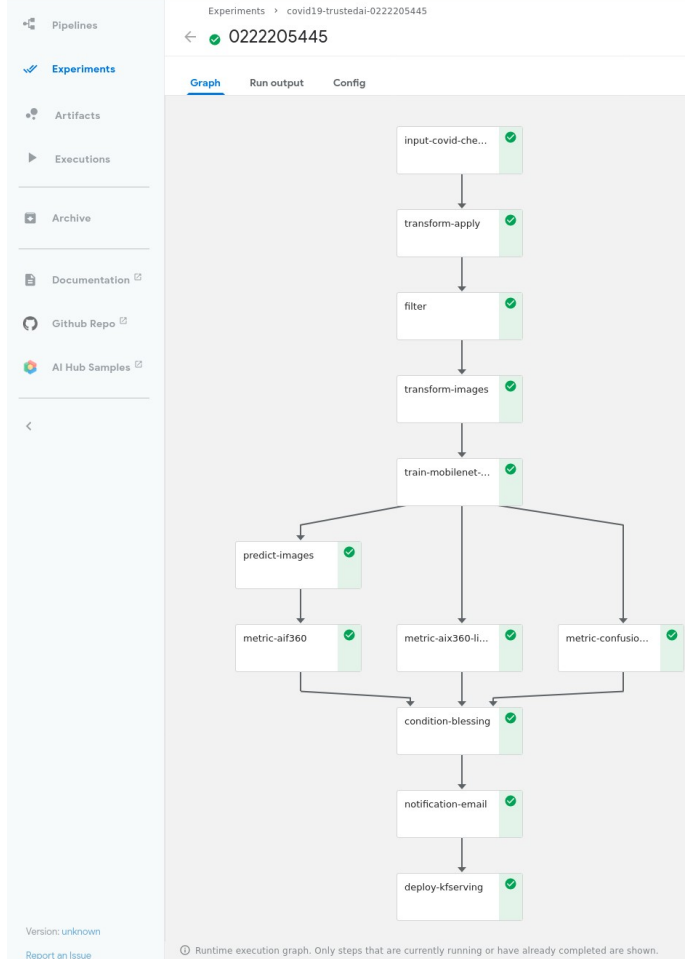
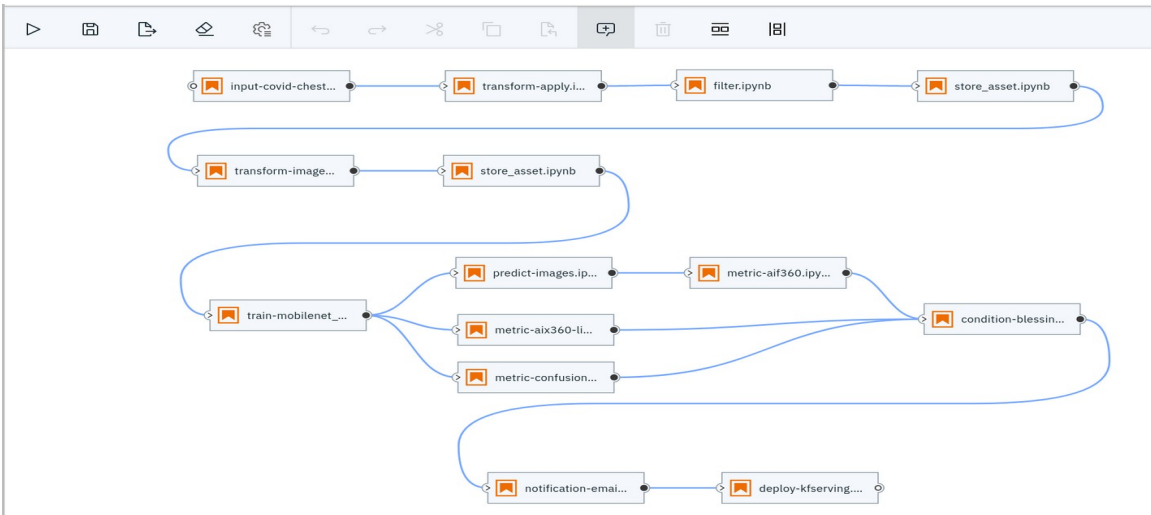
...using Elyra and ML Exchange (MLX)

# Component consumption...



...using Elyra or Watson Studio Pipelines

# The final pipeline in Elyra..



...and KubeFlow

# Summary

- Rapid prototyping using visual editing and notebooks ✓
- Seamless scaling during development and deployment ✓
- GPU support ✓
- ML tools: PyData stack, TensorFlow, PyTorch, ... ✓
- life science tools: DICOM input, DICOM output, ... ✓
- Reproducibility ✓
- Data lineage ✓
- Reference implementation in open source ✓
- Collaboration support ✓