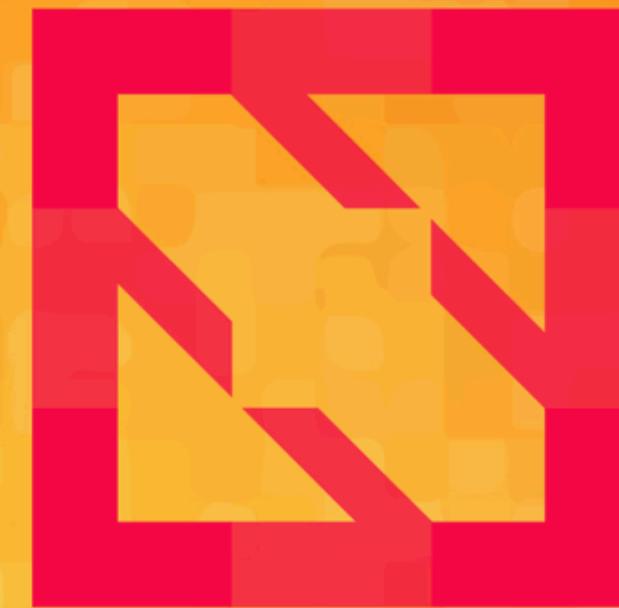




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Building a Medical AI With Kubernetes and Kubeflow

Jeremie Vallee, Cloud Infrastructure Engineer, Babylon Health





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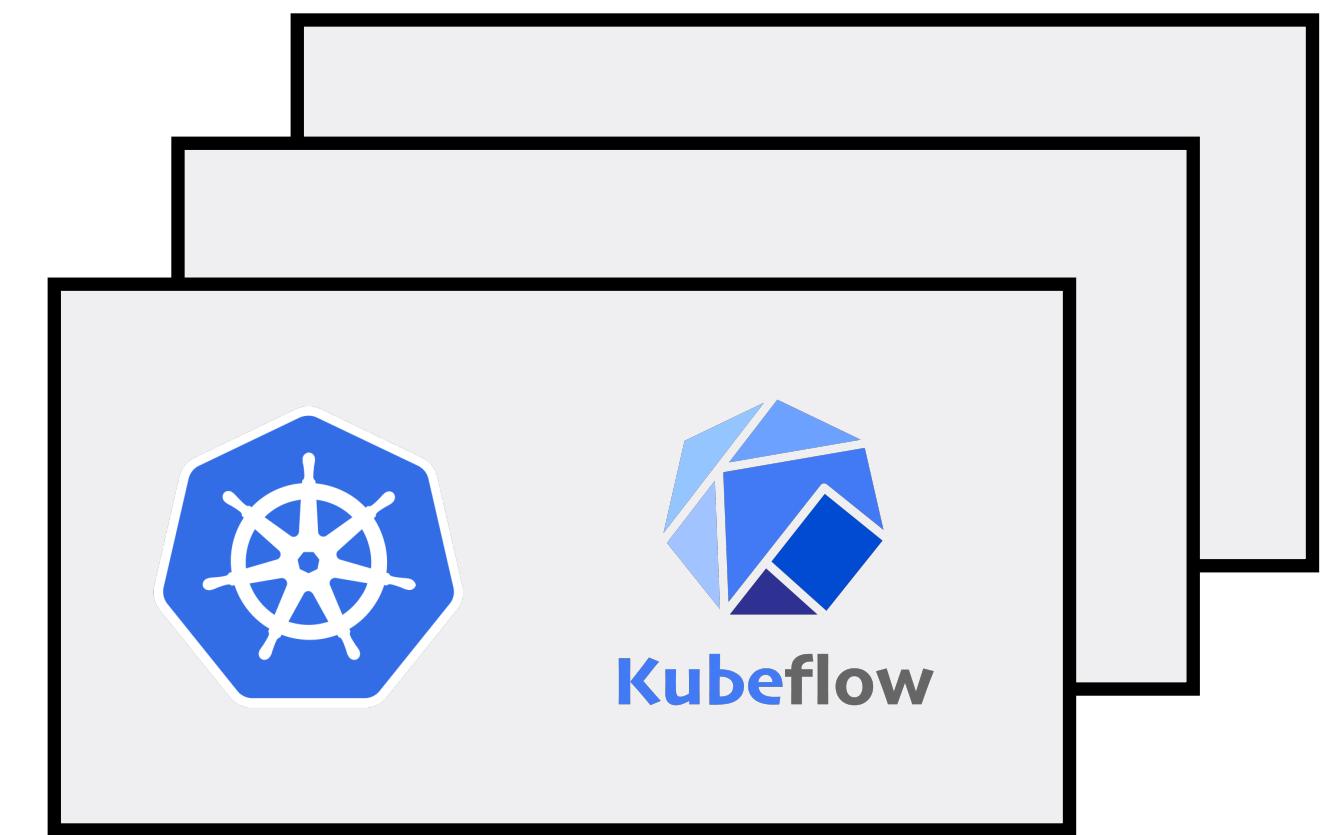
From



“Hi! 🙌

*Is anyone running a training job
on the server? My notebook is super slow...”*

To



*Self-service
Multi Region
Multi Cloud*

Research and Training Platform



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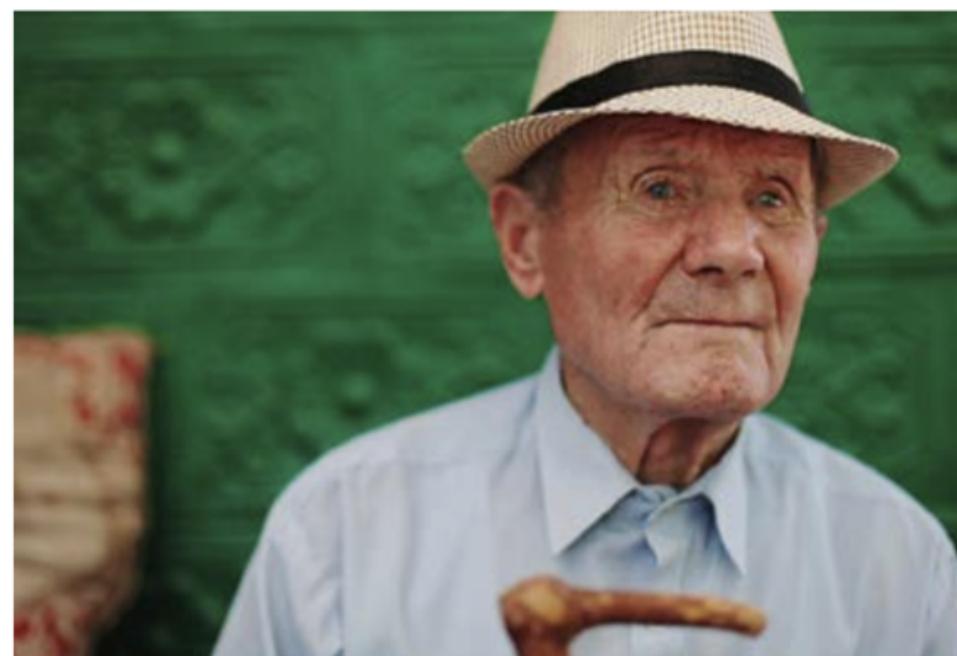
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babylon

We believe it is possible to put
an **accessible** and **affordable**
health service in the hands of
every person on earth.





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Accessibility

50% of the world population lacks access to essential health services, however **67% have access to mobile phones.**

In 2014:

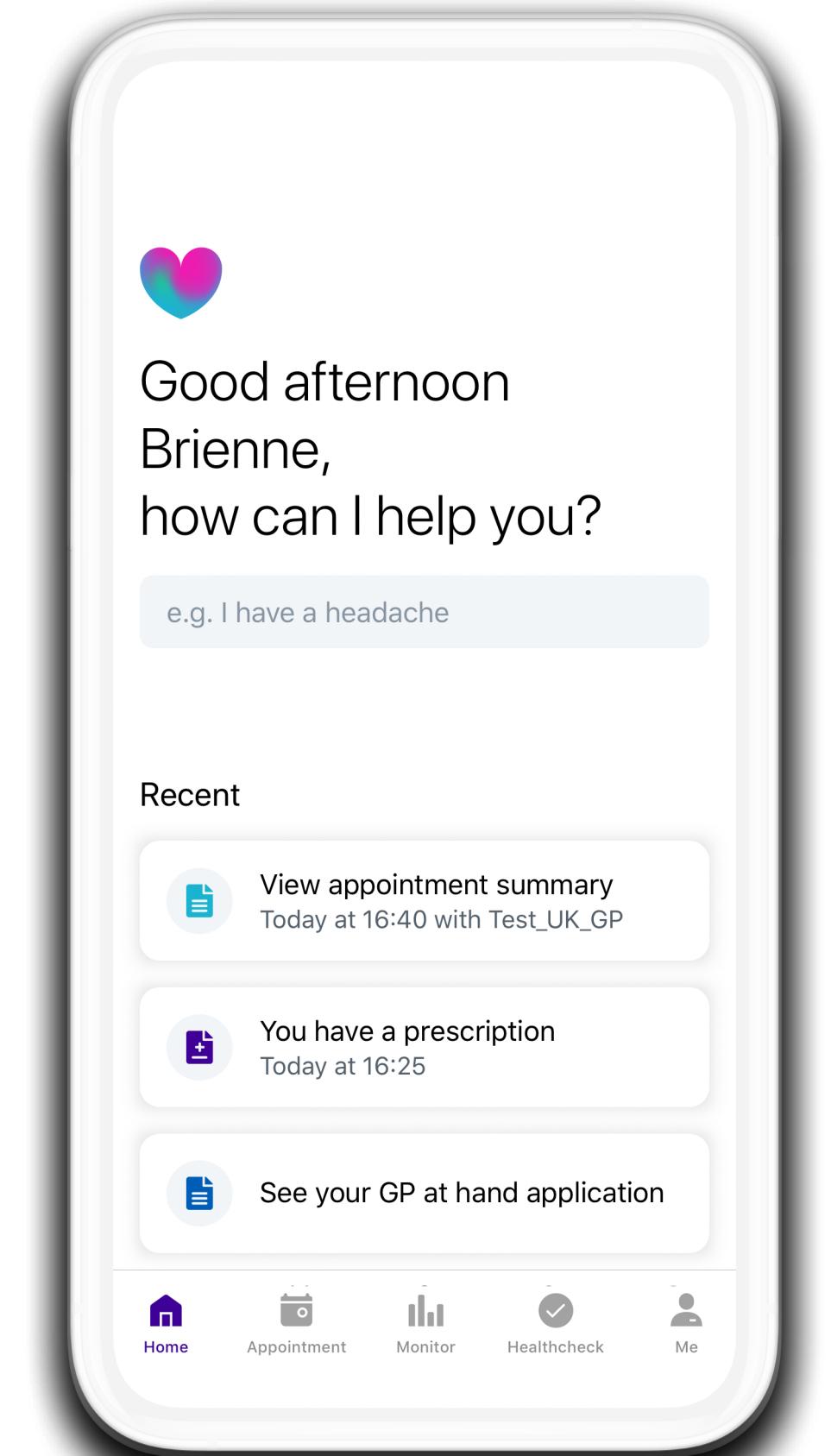
Digital consultation with a doctor from your mobile phone.

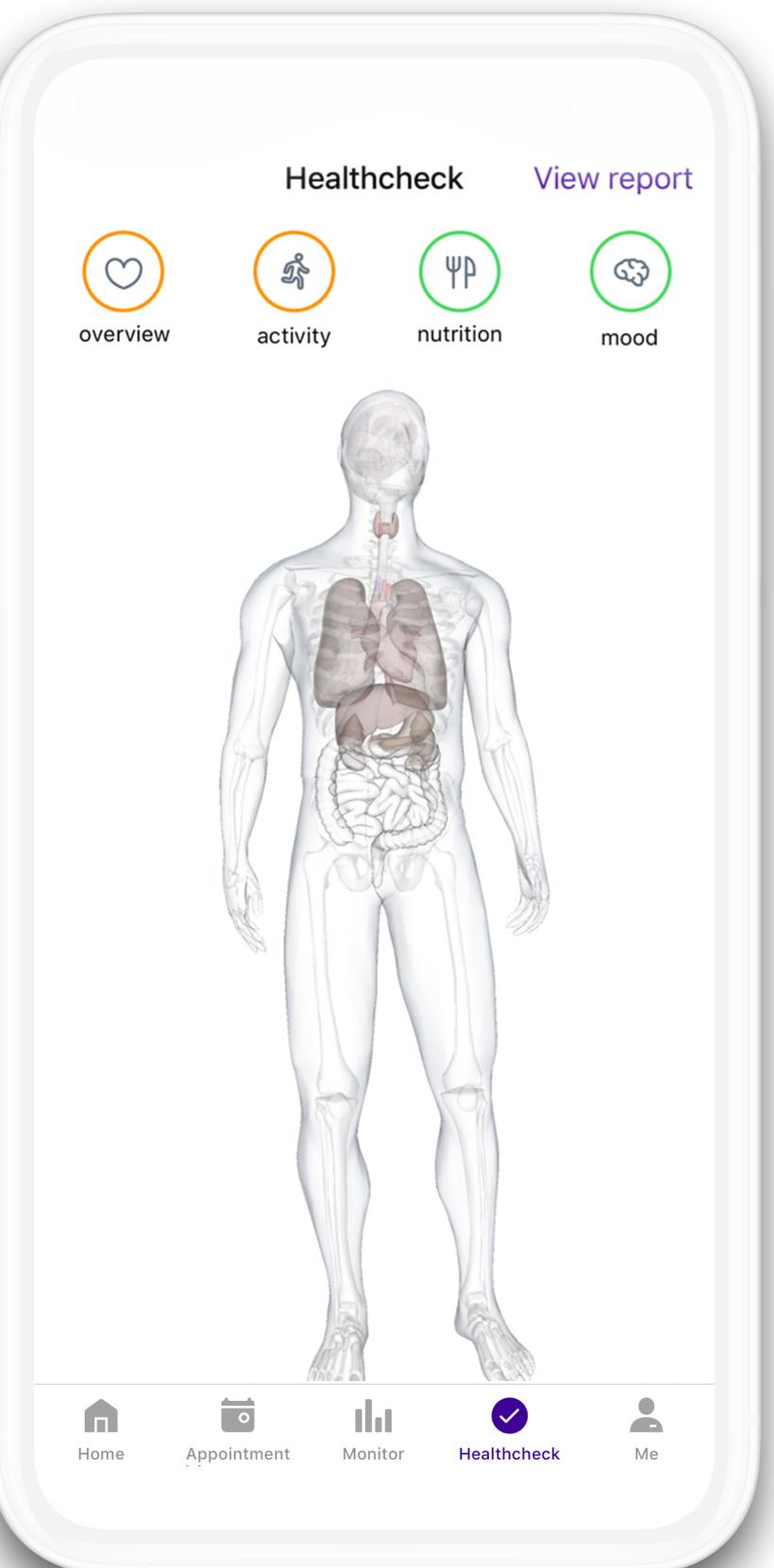
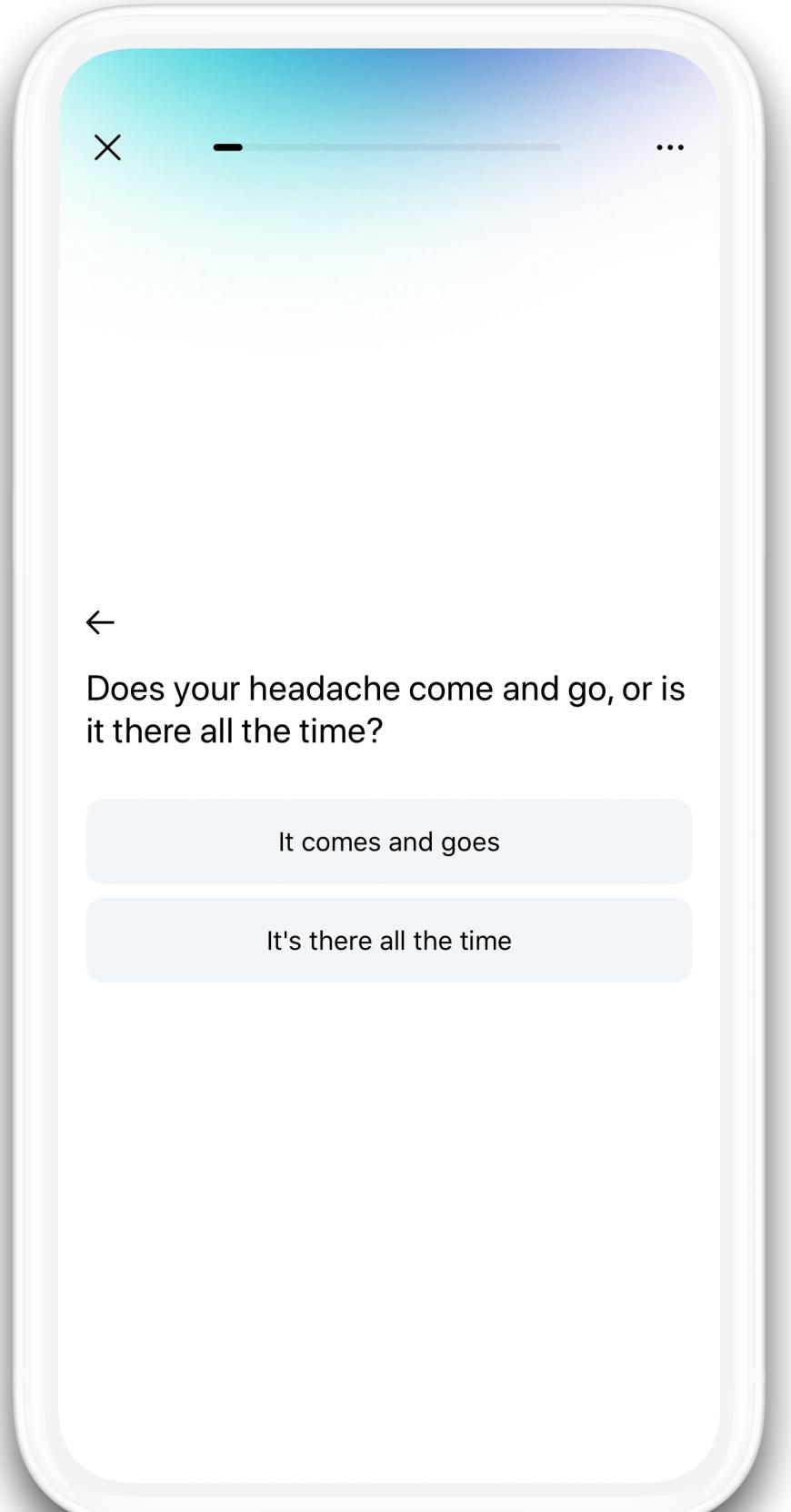
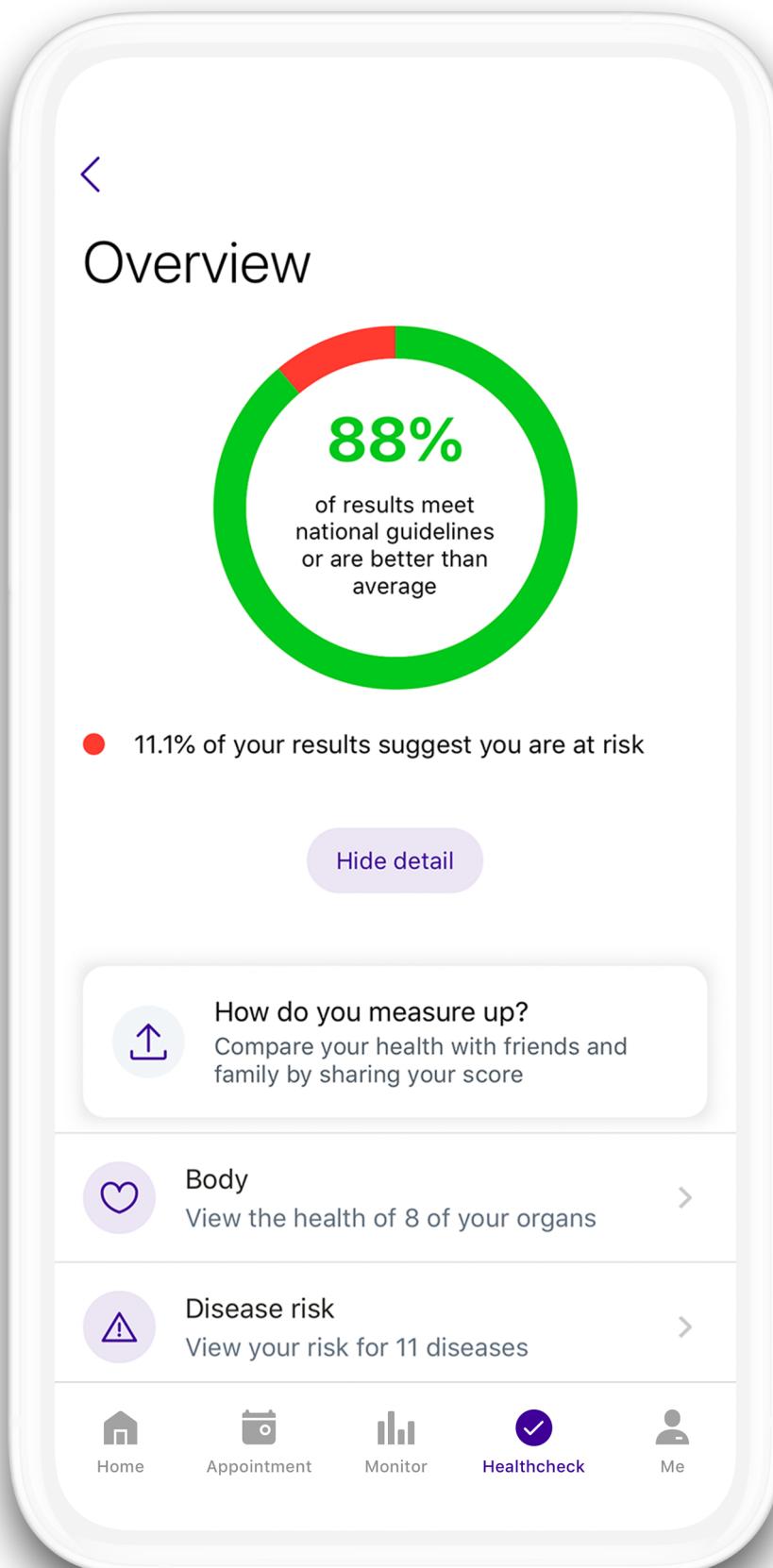
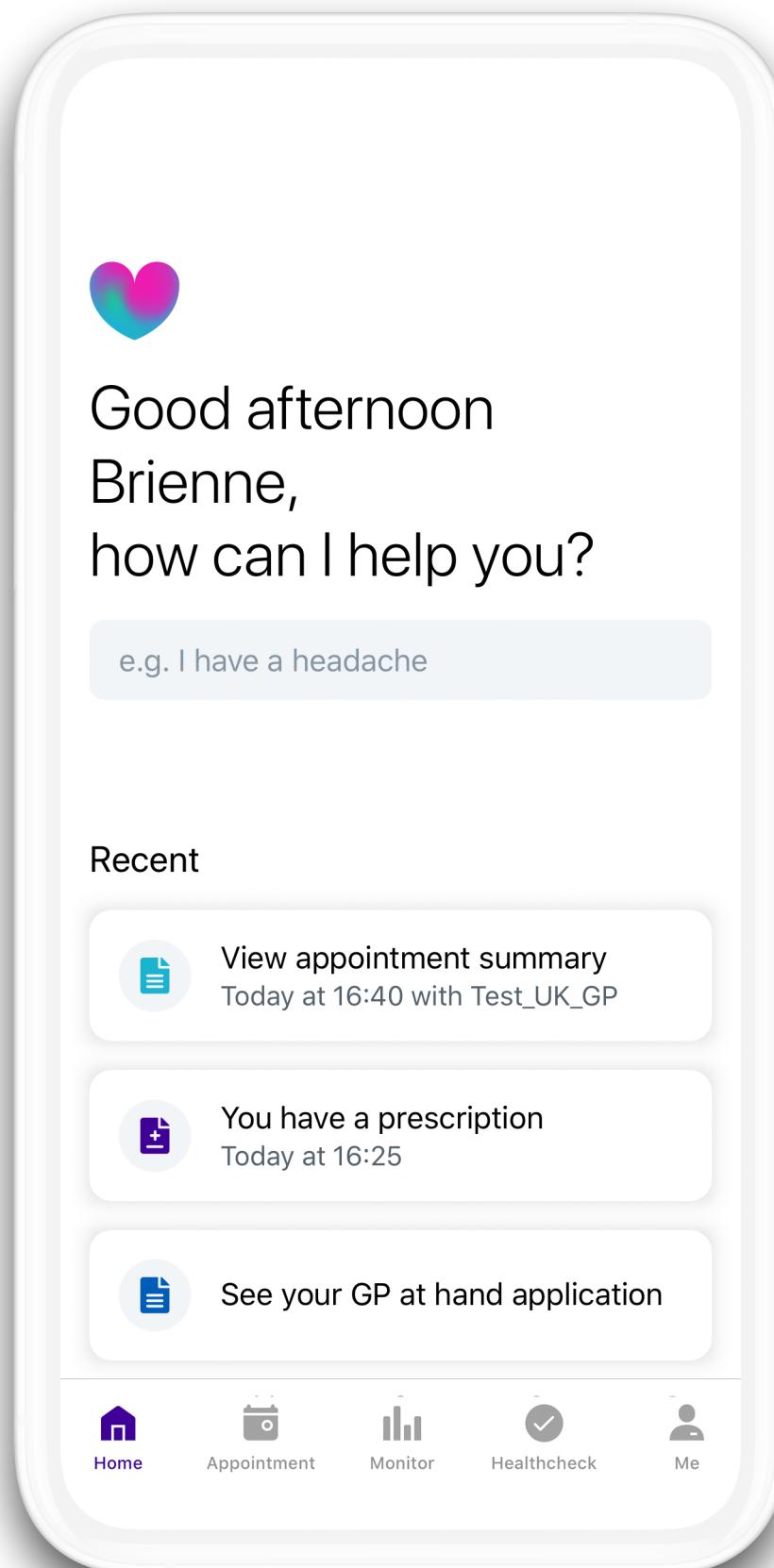
Now:

We've had 2+ million digital consultations in the world.

Affordability?

Affordability





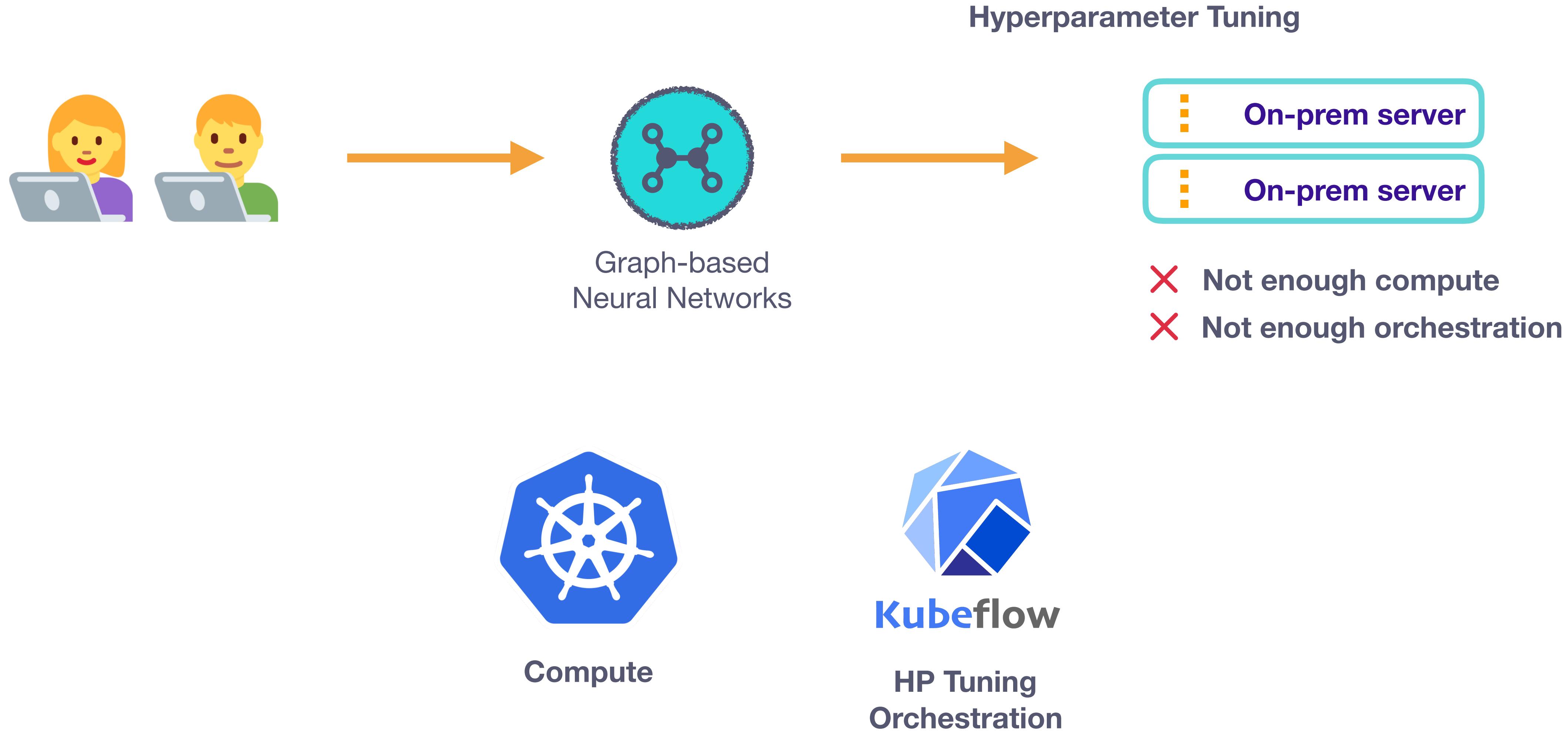
How it started



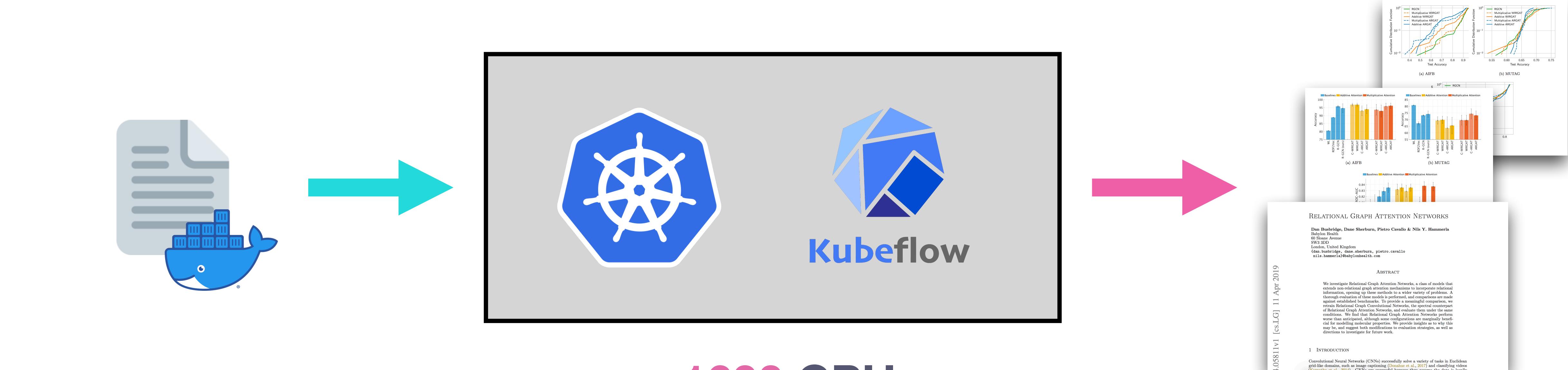
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Hyperparameter Tuning with Katib



1600 CPU
3.2TB RAM

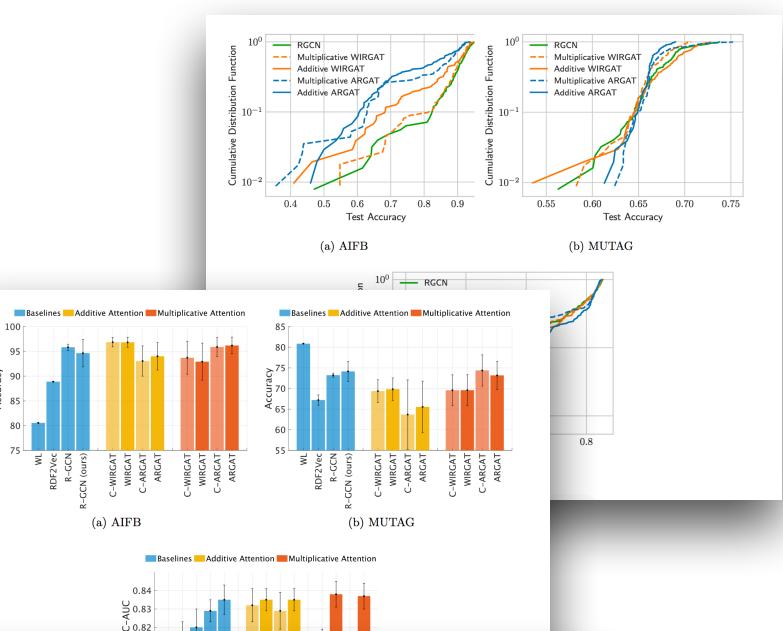
4.05811v1 [cs.I.G] 11 Apr 2019

1 INTRODUCTION
Convolutional Neural Networks (CNNs) successfully solve a variety of tasks in Euclidean grid-like domains, such as image captioning (Donahue et al., 2017) and classifying videos (Carreira and Zisserman, 2014). However, CNNs have difficulty generalizing to non-Euclidean domains, such as graphs.

Dan Busbridge, Dane Sherburn, Pietro Cavallo & Nils Y. Hammerla
Babylon Health
60 Sloane Avenue
SW3 3DP
London, United Kingdom
(dan.busbridge, dane.sherburn, pietro.cavallo,
nils.hammerla)@babylonhealth.com

ABSTRACT

We investigate Relational Graph Attention Networks, a class of models that extends non-relational graph attention mechanisms to incorporate relational information, opening up these methods to a wider variety of problems. A thorough evaluation of these models' performance and generalization against established benchmarks. To provide a meaningful comparison, we retrain Relational Graph Convolutional Networks, the spectral counterpart of RGCNs, under the same experimental conditions and evaluation conditions. We find that Relational Graph Attention Networks perform worse than anticipated, although some configurations are marginally beneficial for certain tasks. This provides insights as to why this may be, and suggest both modifications to evaluation strategies, as well as directions to investigate for future work.



Challenges & motivation



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Growing company



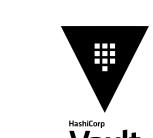
Increasing need for
compute access
AI Research & Engineering tooling



Global footprint



**Self-service
AI Research & Training
Platform**



Let's talk about:



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Kubernetes
Infrastructure

Cluster
Bootstrapping

Networking

Project
Isolation

Kubeflow

Monitoring

Multi Cluster

Use cases

Next steps

Kubernetes Infrastructure



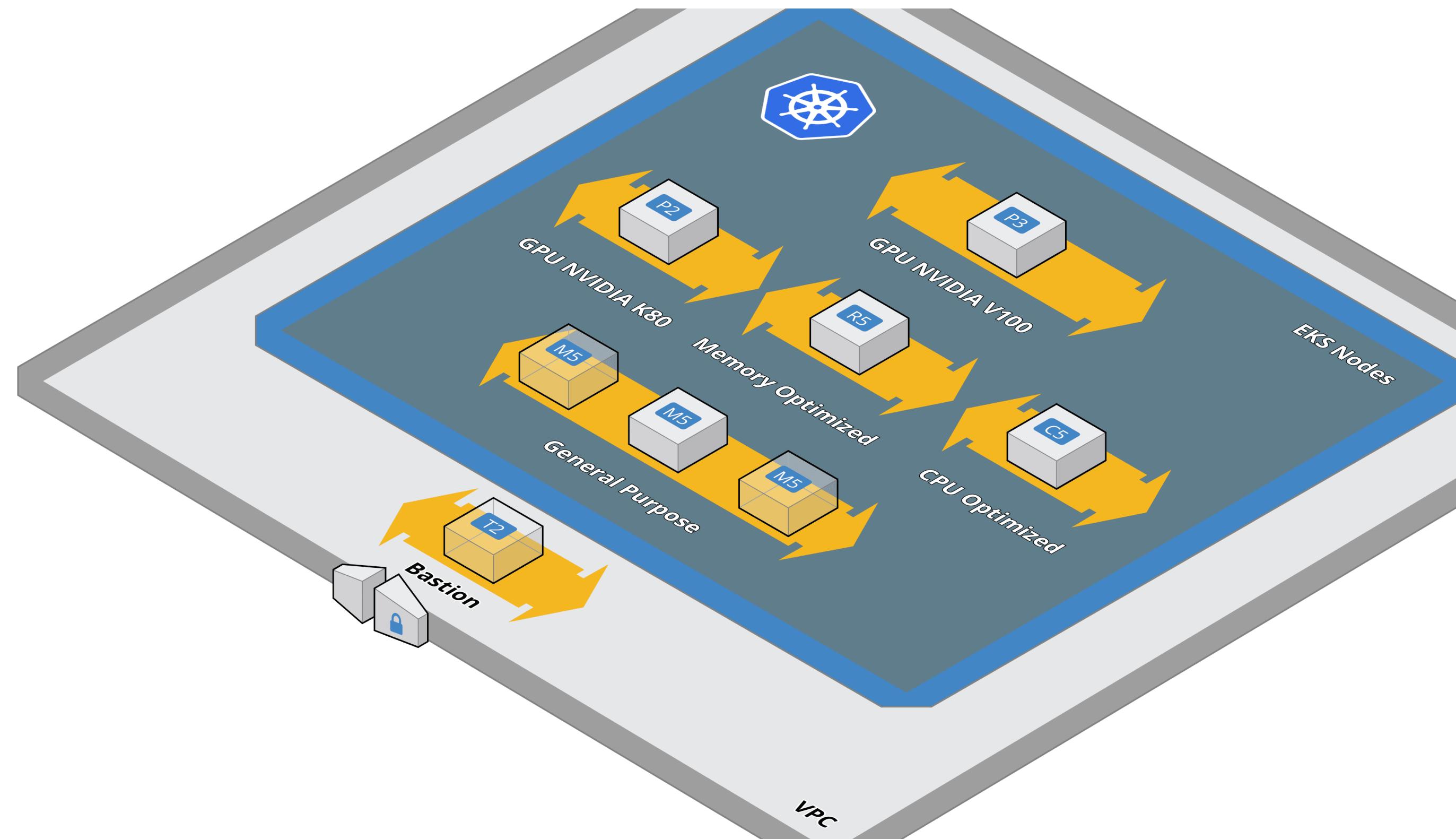
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Prepare for different workload types

Harden your nodes

Make Kubernetes API private

Encrypt your nodes (root volumes + others)



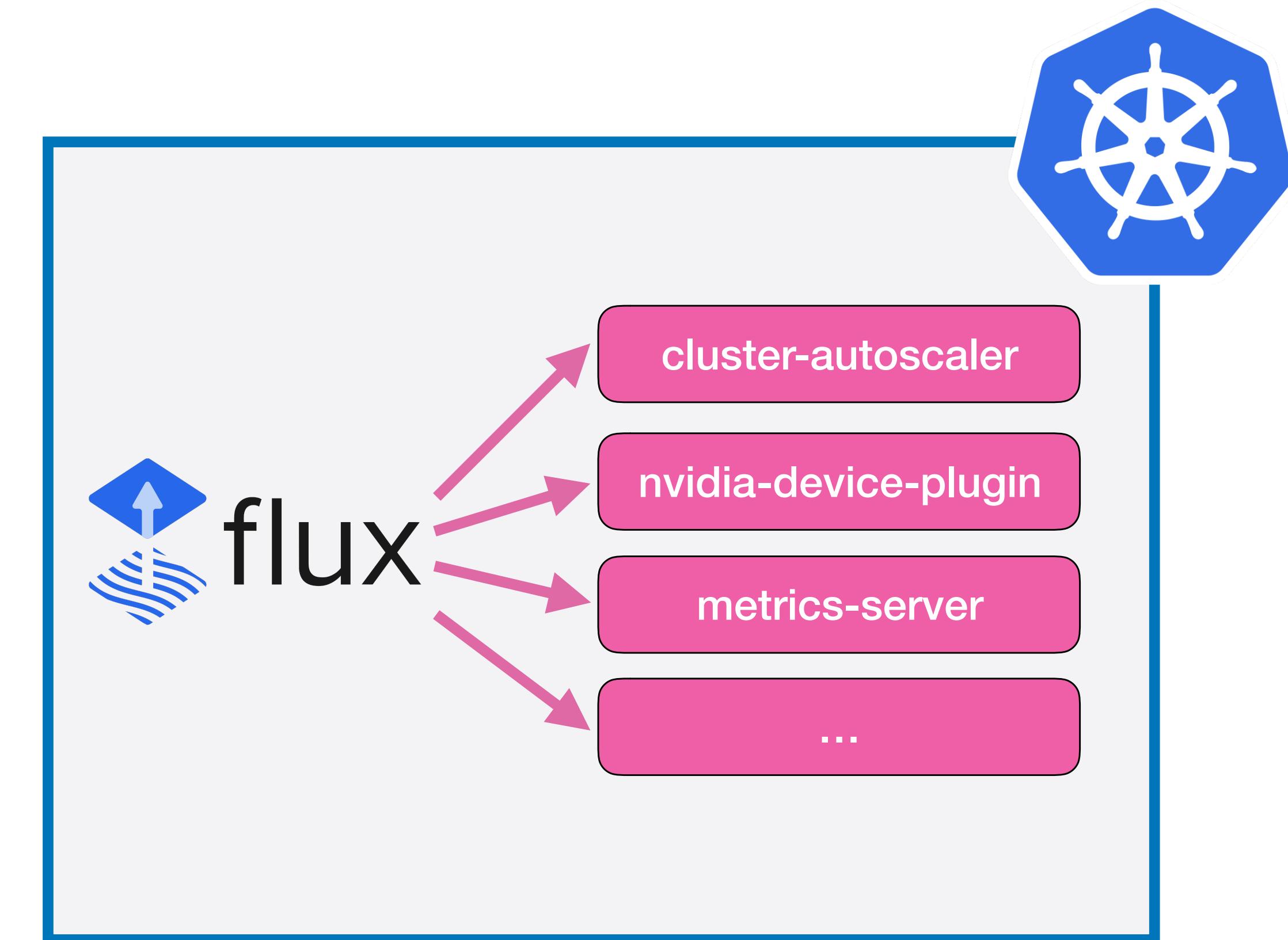
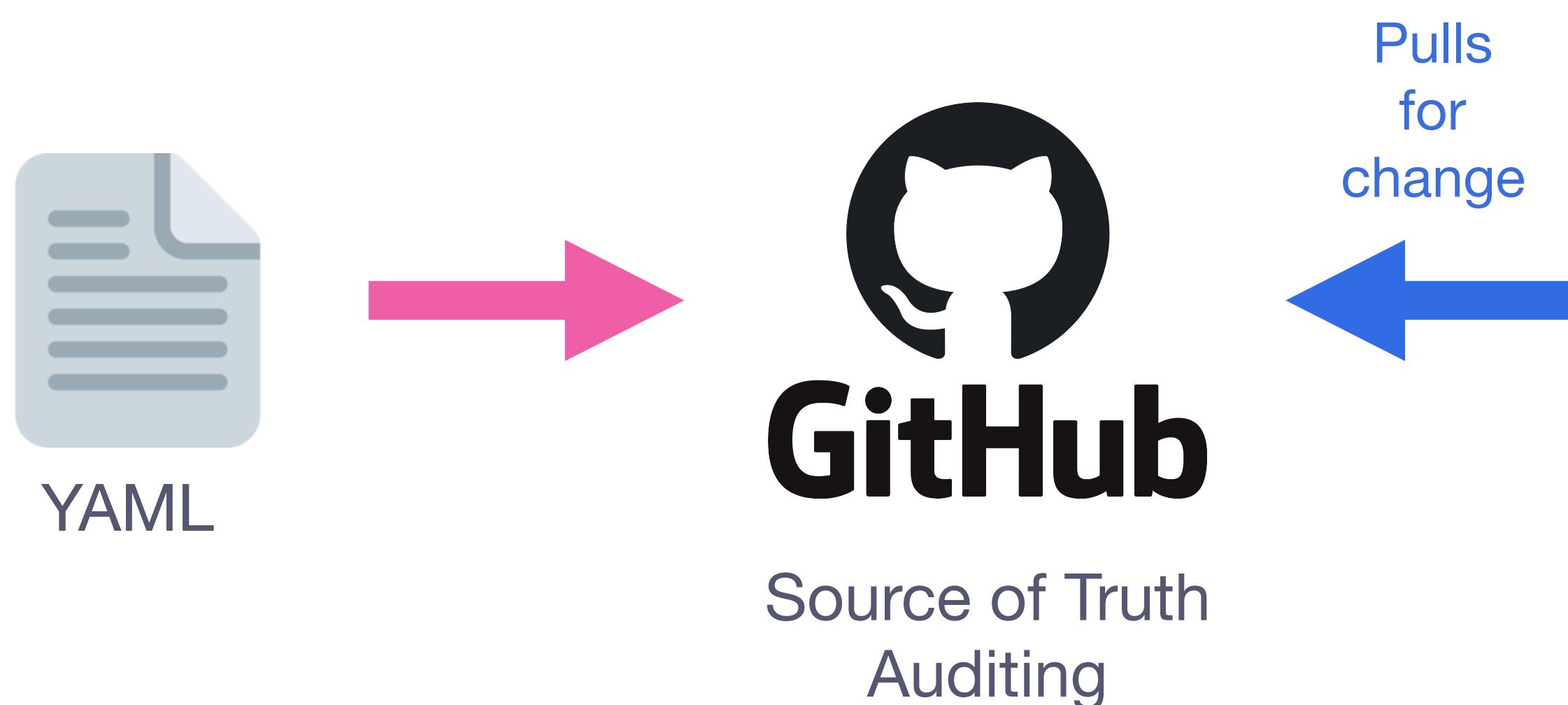
GitOps Bootstrapping



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Reduce human error

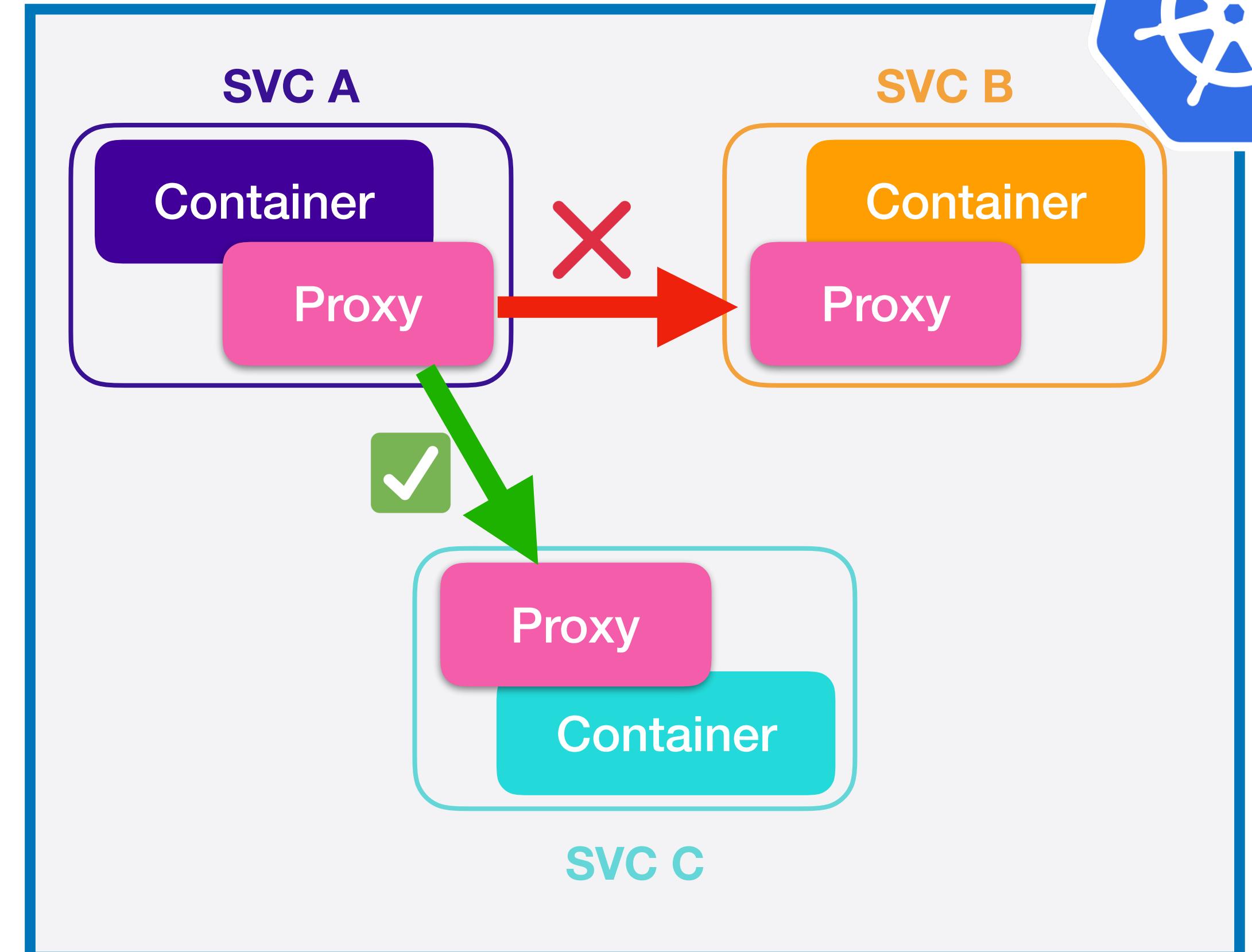
Networking



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```
apiVersion: rbac.istio.io/v1alpha1
kind: ServiceRole
metadata:
  name: c-access-role
  namespace: default
spec:
  rules:
  - methods:
    - '*'
  paths:
  - '*'
  services:
  - c.default.svc.cluster.local
```

```
apiVersion: rbac.istio.io/v1alpha1
kind: ServiceRoleBinding
metadata:
  name: bind-a-to-c
  namespace: default
spec:
  roleRef:
    kind: ServiceRole
    name: c-access-role
  subjects:
  - user: cluster.local/ns/default/sa/a
```



Zero-trust policy

Mutual TLS

JWT
check



Project Custom Resource

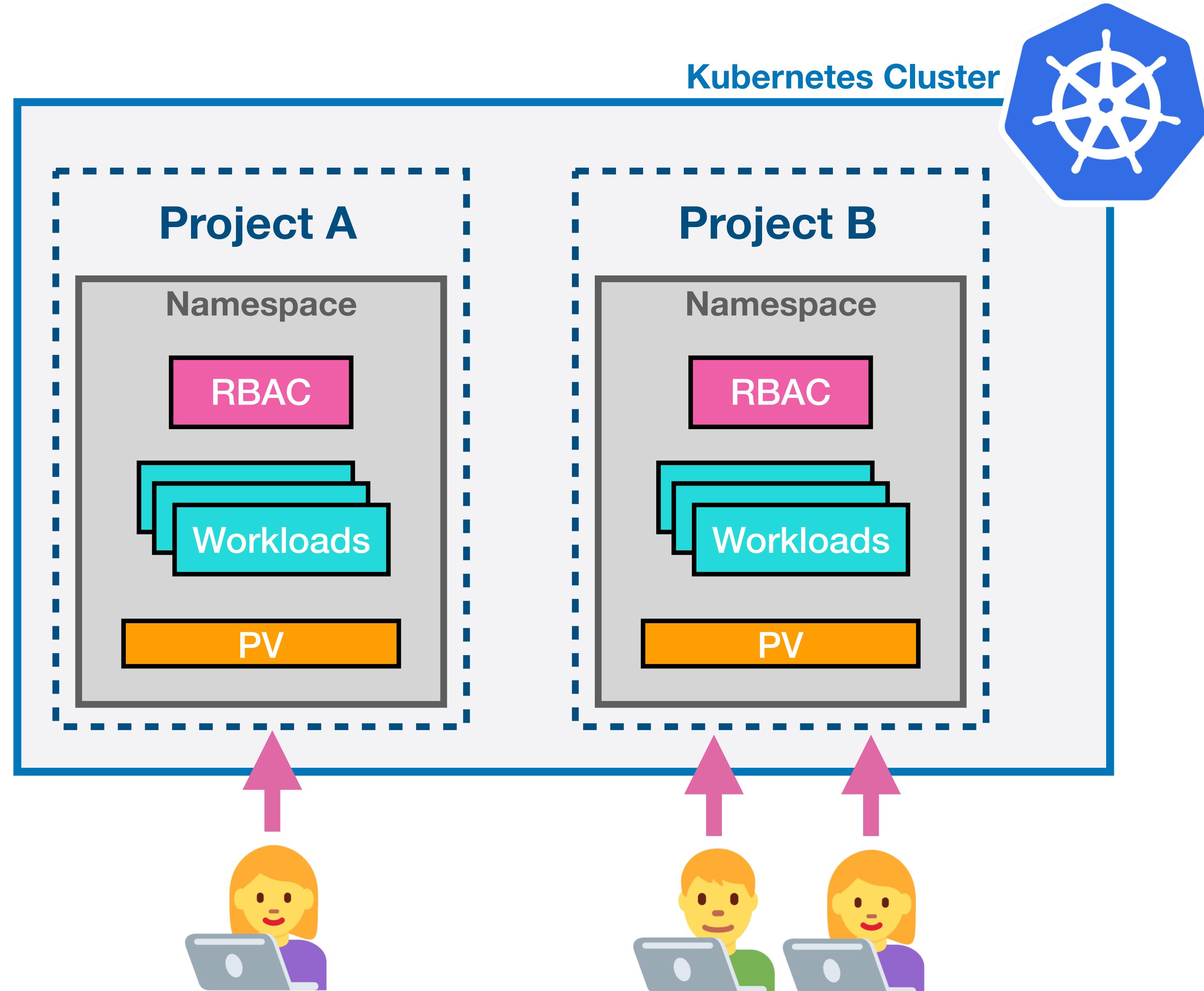


Goal:

- Isolate workloads
- Enable collaboration (multi-user)
- Provide shared volume per project
- Additional business logic and metadata

Solution:

- “Project” CRD and controller
- Integration with Kubeflow “Profile” CRD



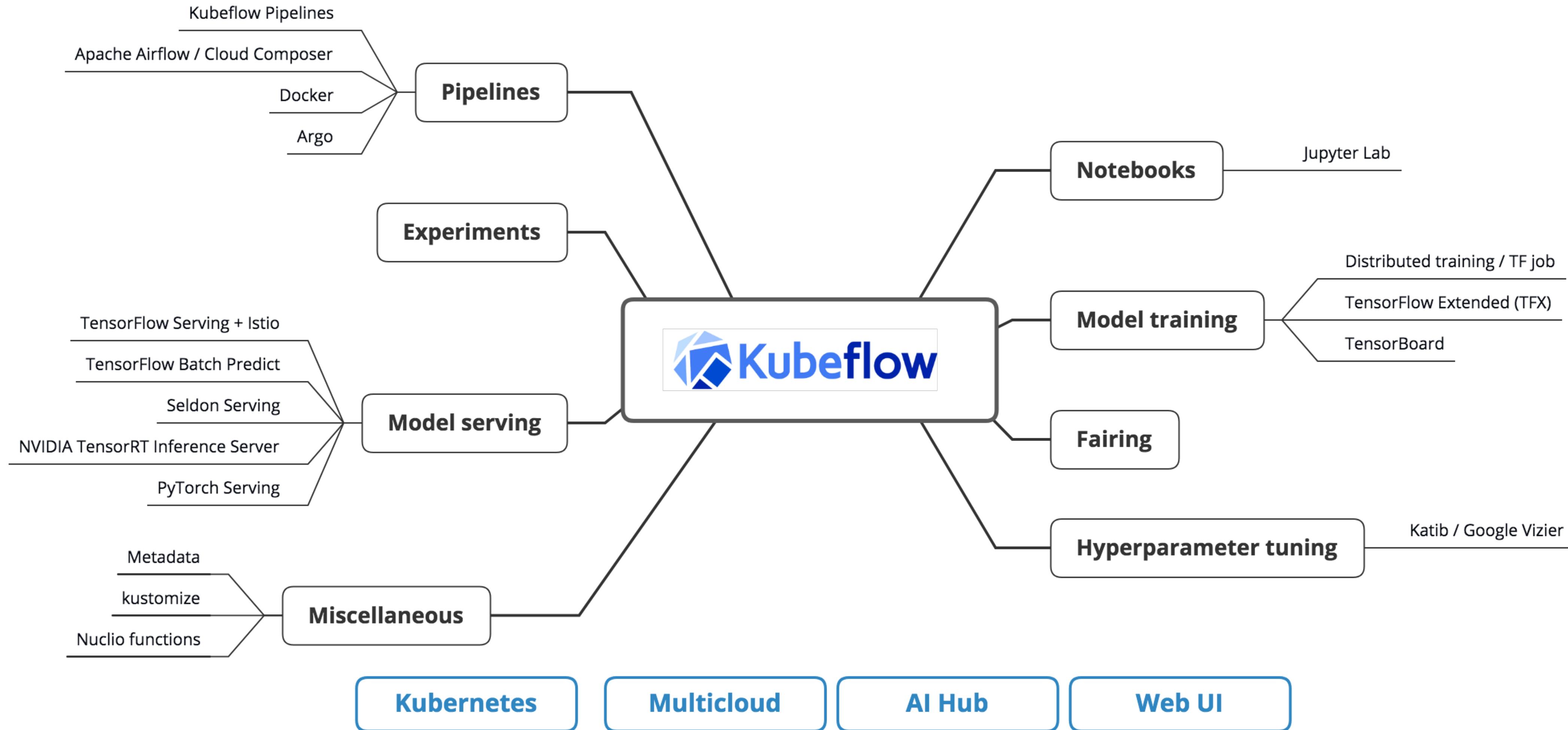
Kubeflow



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Kubeflow



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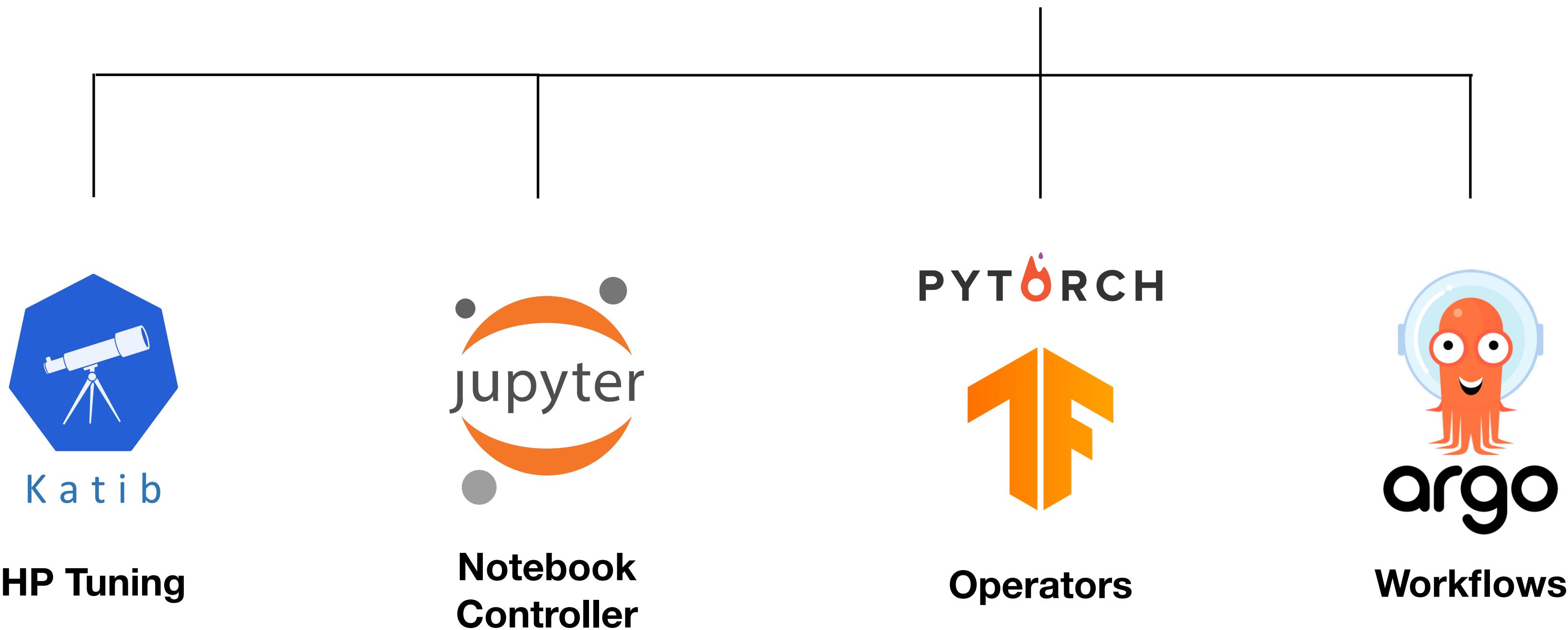


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Modular: install what you need

Deployment: GitOps (Kustomize + Flux)



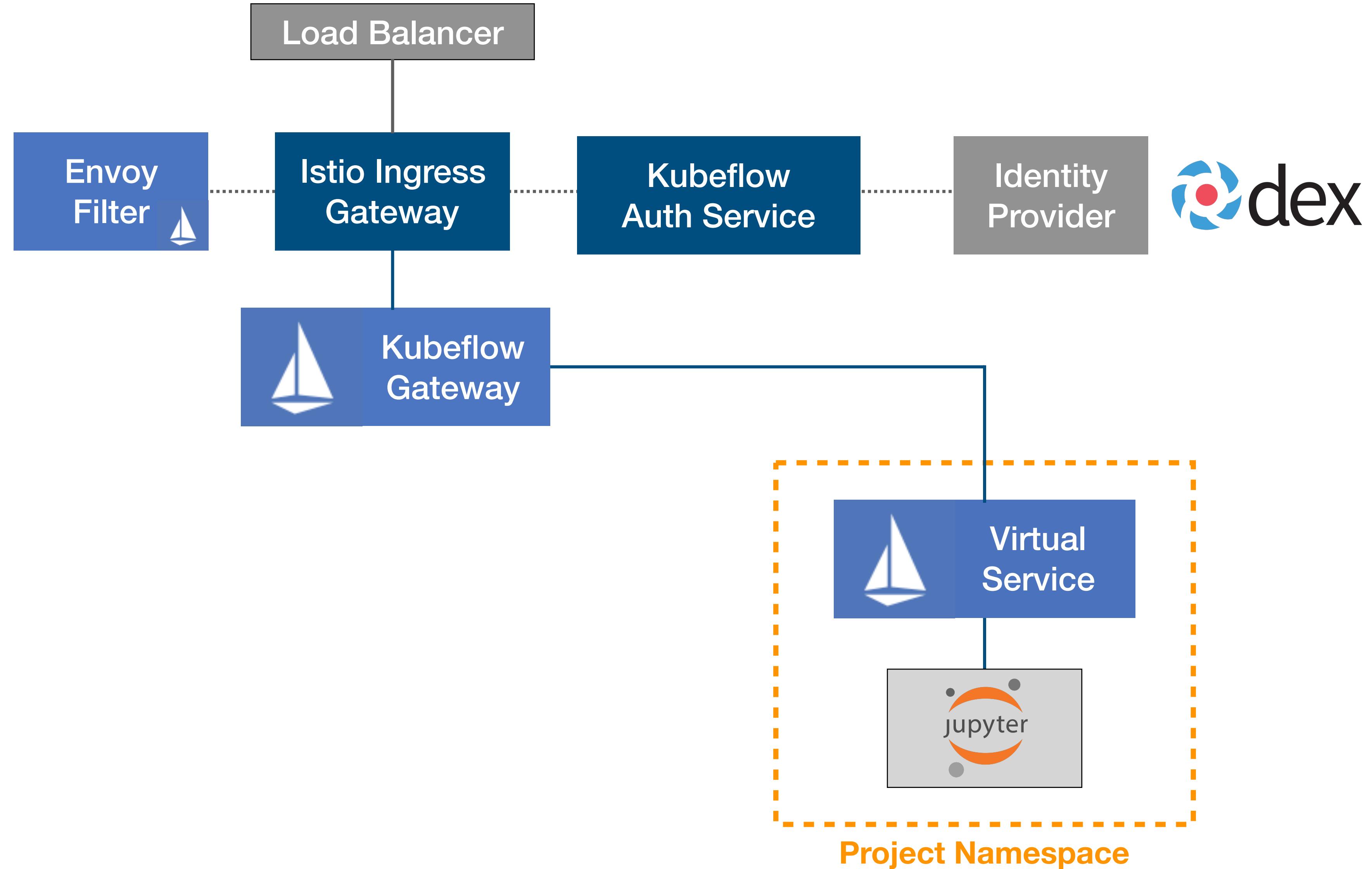
Kubeflow + Istio + AuthN



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Data and Secrets



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Data

Control access to data via an auditable layer

Allow for data discoverability (with tools like [Amundsen](#))

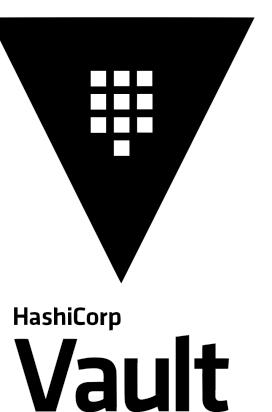
Amundsen

<https://github.com/lyft/amundsen>

Secrets

Use a secret manager

Pods can authenticate to Vault and get secrets loaded in memory



Monitoring

Goal:

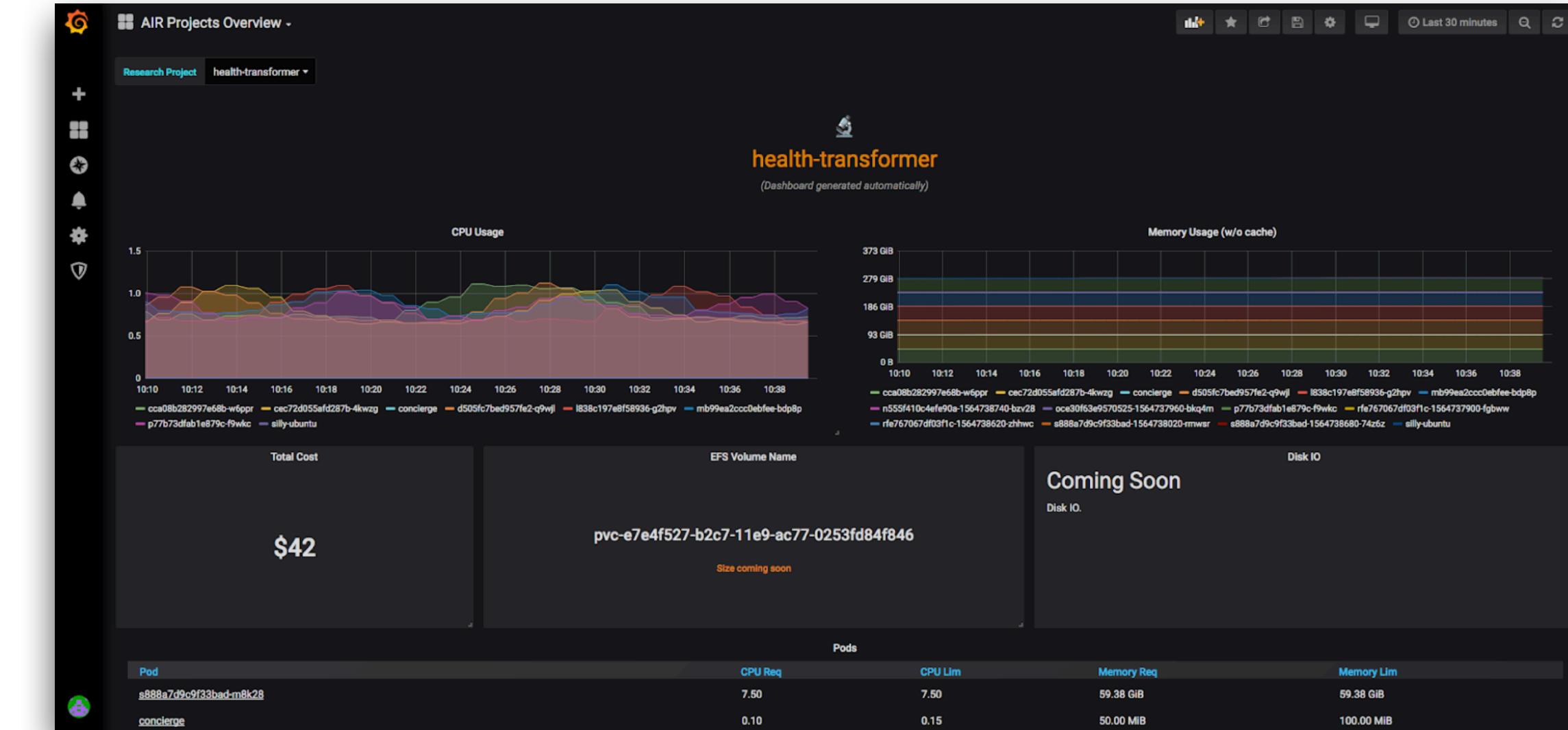
- Monitor projects for users
- Monitor cluster for MLOps
- Collect any metrics from jobs
- Automated dashboards



Prometheus



Grafana



Lessons learned:

- Grafana Dashboards as *ConfigMaps*
- Allow users to submit dashboards via *GitOps*
- Multiple K8s clusters? Use Federation feature
- Need to scale up prometheus storage? Use *Thanos*

Monitoring cost



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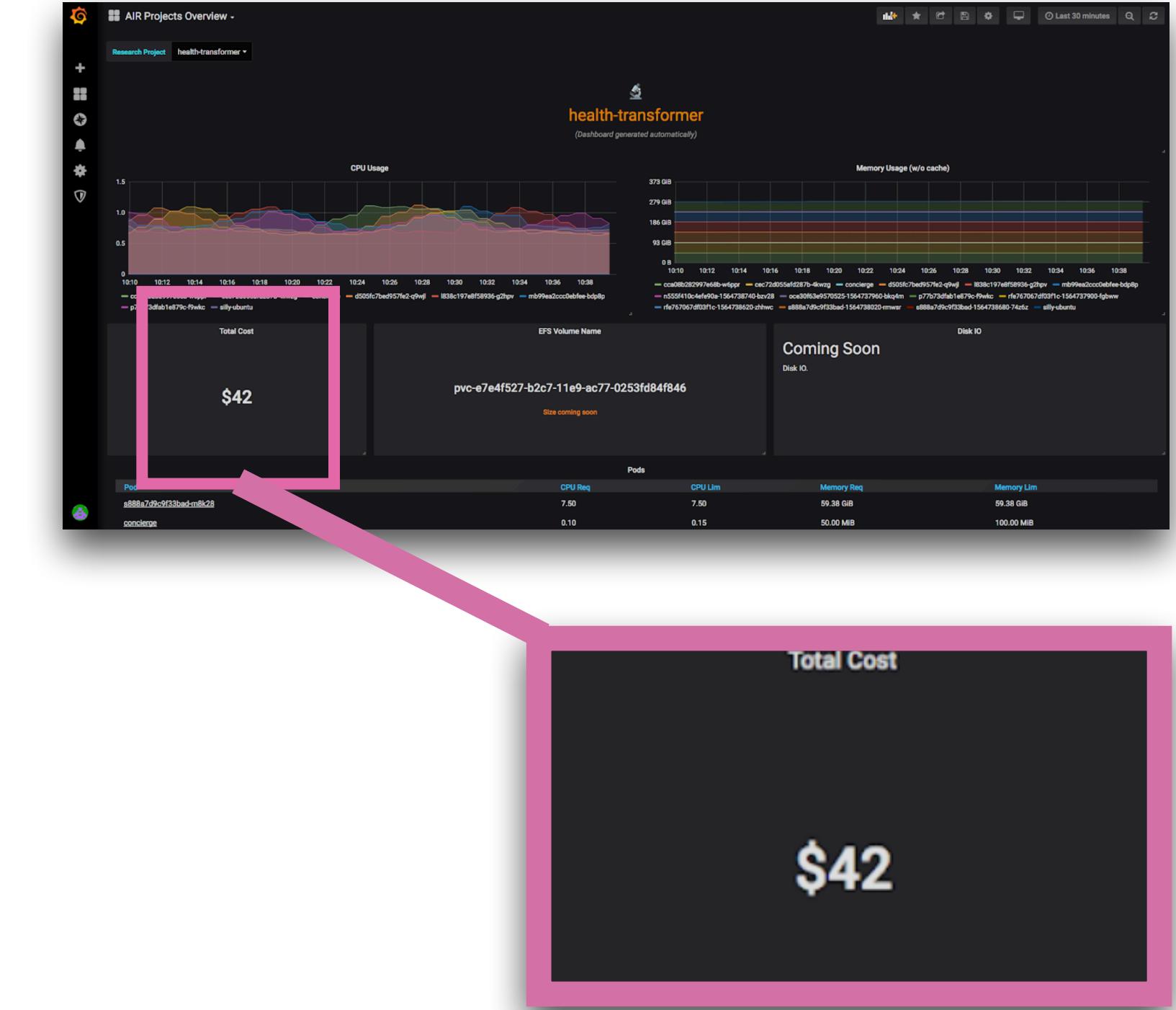
Keeping track of cost in a cloud environment is vital especially in AI/ML

Many proprietary options...

One of them open-sourced their cost model engine:



github.com/kubecost/cost-model



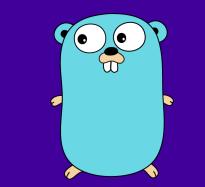
Babylon:

3rd
Party
Tool

Get
cost per
namespace

Expose
cost as metrics

cost-exporter



Pull metrics





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Self-service access to ML Toolkit

CLI-based interface

Monitoring

On-demand compute

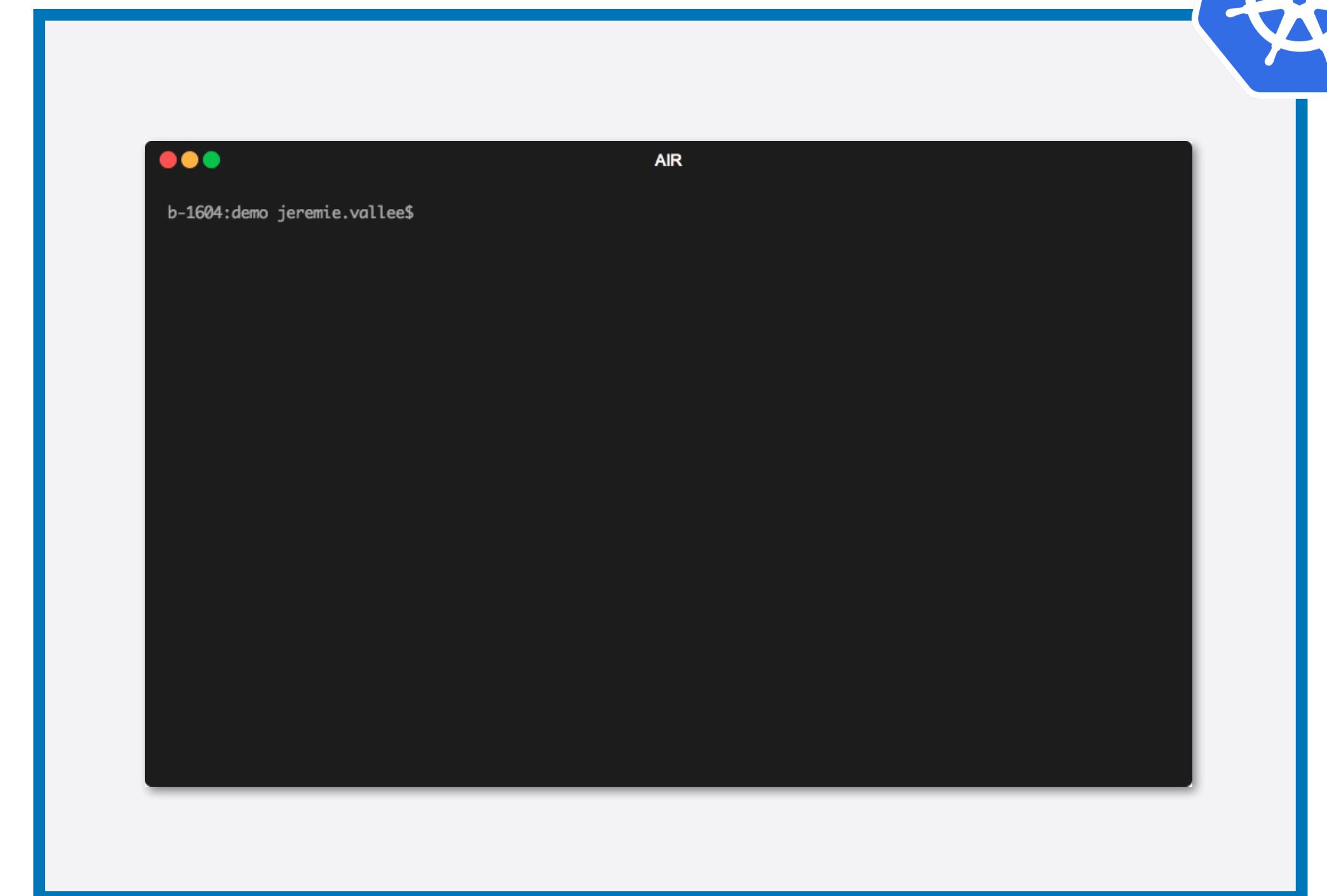
Network RBAC, mTLS



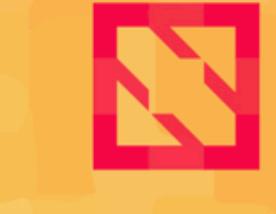
Single cluster

Slow on-boarding (GitOps)

Complex Kubernetes objects



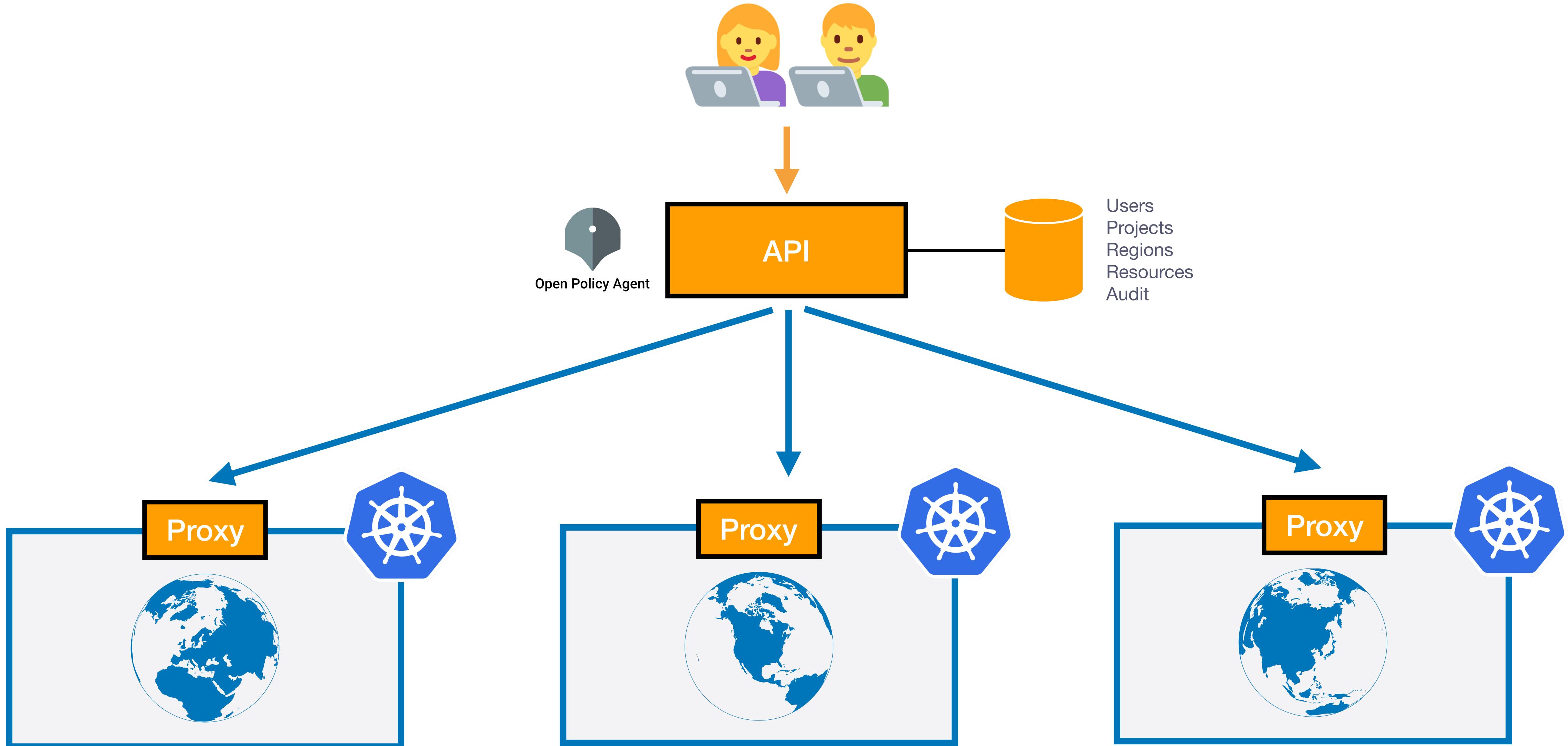
Global platform



KubeCon

CloudNativeCon

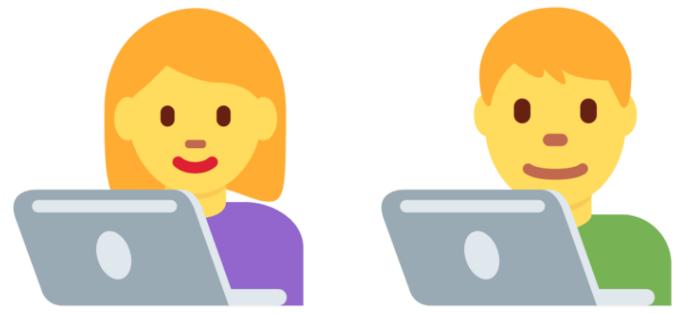
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Simplifying resources



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```
{  
  "kind": "pod",  
  "name": "simple-gpu-example",  
  "image": "nvidia/cuda:8.0-cudnn5-runtime",  
  "command": ["python"],  
  "args": ["script.py"],  
  "resources": "gpu_medium"  
}
```

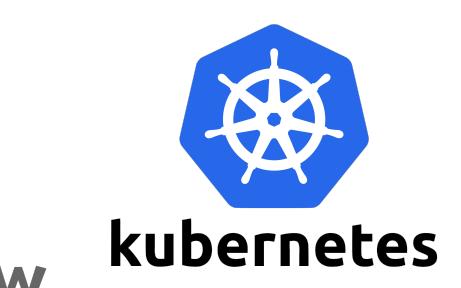
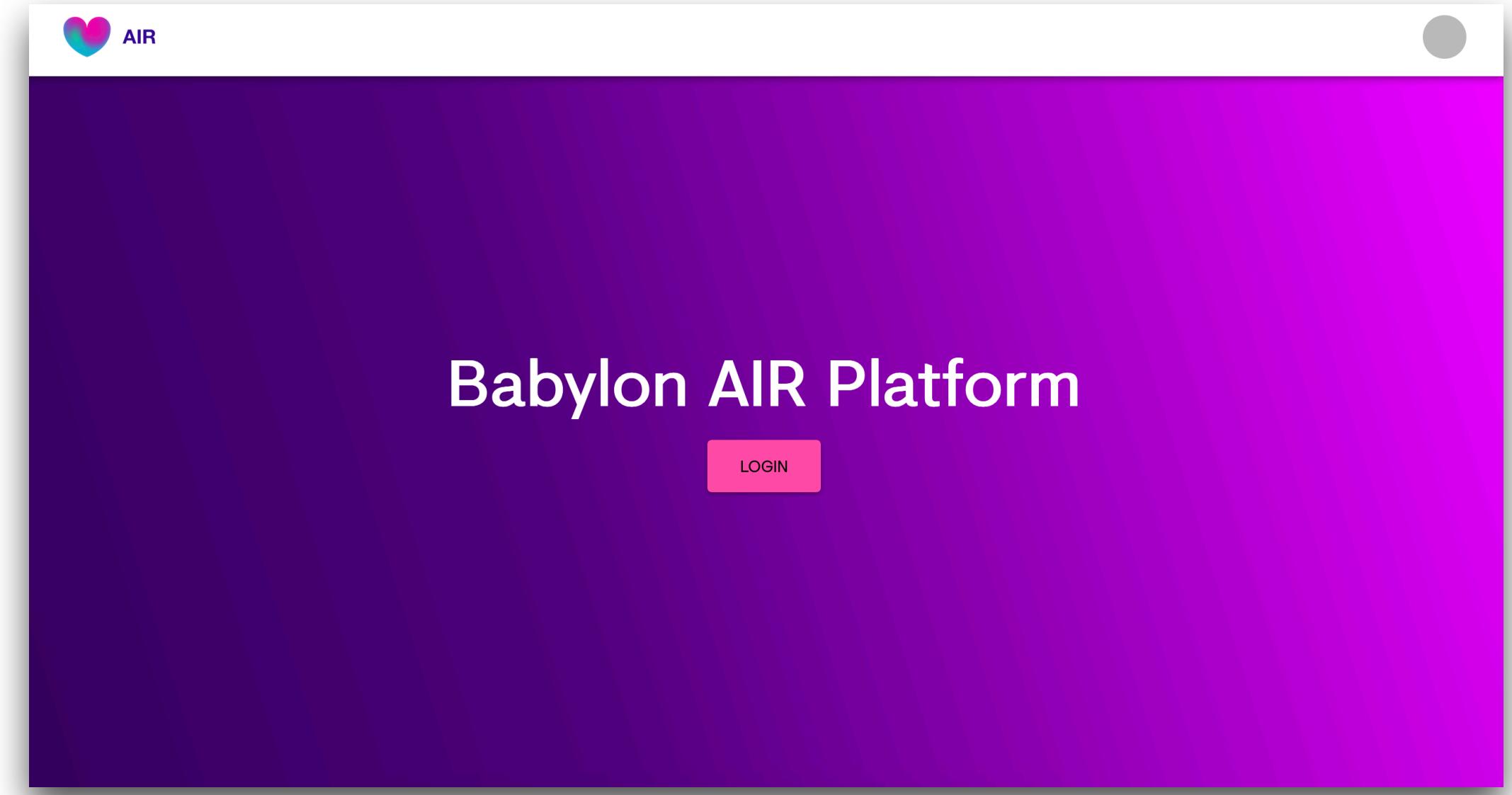


API

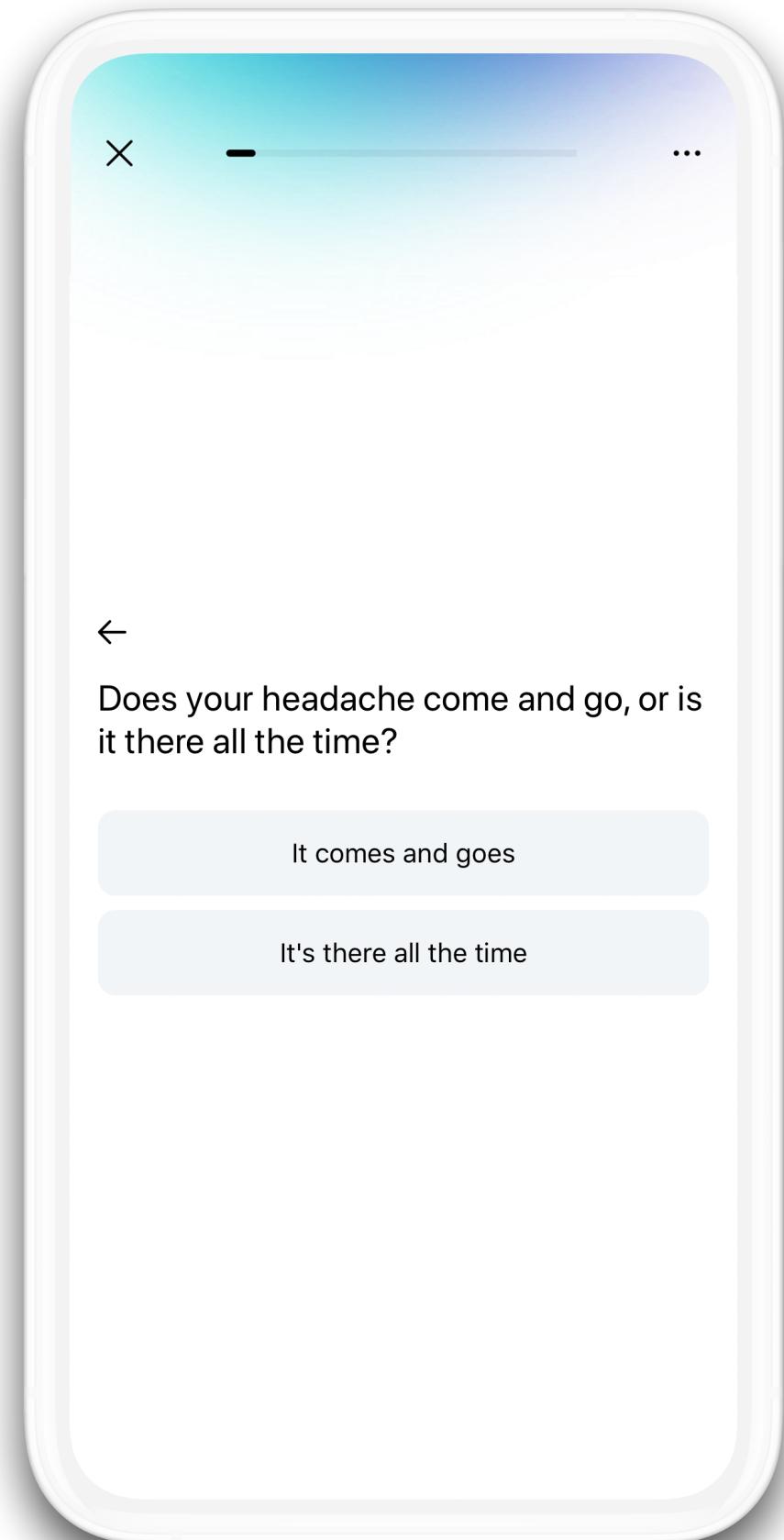


```
---  
apiVersion: v1  
kind: Pod  
metadata:  
  name: simple-gpu-example  
  namespace: my-project  
spec:  
  containers:  
    - image: nvidia/cuda:8.0-cudnn5-runtime  
      command: [ "python" ]  
      args: [ "script.py" ]  
      name: simple-gpu-example  
  resources:  
    limits:  
      memory: "16Gi"  
      cpu: "8000m"  
      nvidia.com/gpu: 1  
    name: "tensorflow"  
  volumeMounts:  
    - mountPath: /mnt  
      name: efs-storage  
  restartPolicy: "OnFailure"  
  volumes:  
    - name: efs-storage  
      persistentVolumeClaim:  
        claimName: efs  
  imagePullSecrets:  
    - name: my-deploy-pull-secret  
  tolerations:  
    - key: "nvidia.com/gpu"  
      operator: "Equal"  
      value: "true"  
      effect: "NoSchedule"  
  nodeSelector:  
    accelerator: nvidia-tesla-k80
```

- 
- Self-service access to ML Toolkit
 - CLI and/or Web interface
 - Monitoring
 - On-demand compute
 - Network RBAC, mTLS
 - Multi region
 - Fast on-boarding (UI or API)
 - Simpler object definitions (heavy-lifting in the backend)
 - Multi user



Use Case: *Clinical Validation of our Symptom Checker*



Symptom Checker: Bayesian network

Has two jobs:

1. Get as much relevant evidence from patient as possible
2. Find most likely disease based on evidence received

$$P(D_i | E)$$

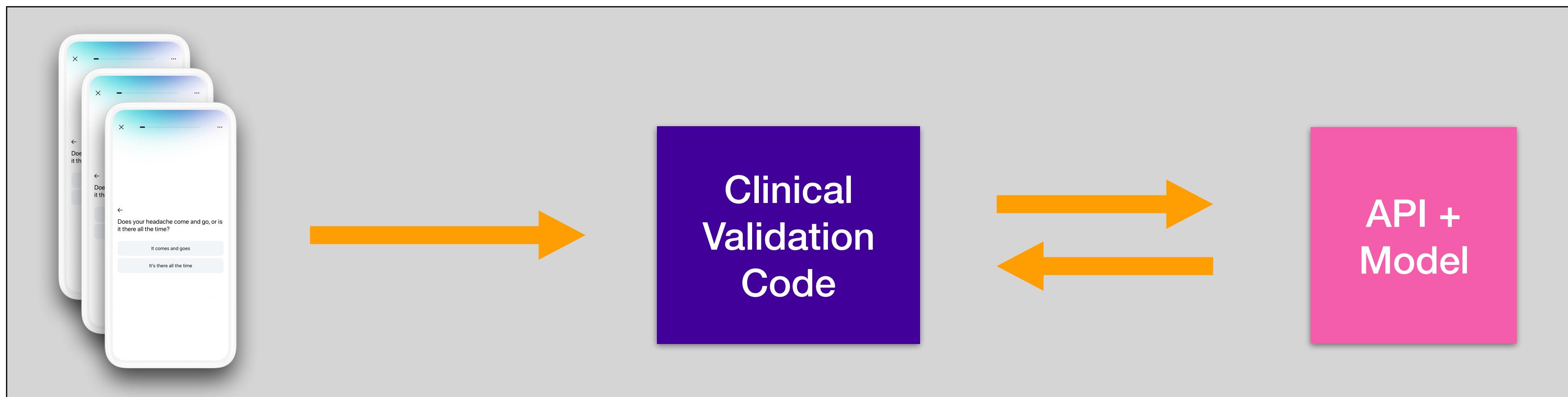
Probability of Disease given Evidence

We want to evaluate how well it performs.

Use Case: *Clinical Validation of our Symptom Checker*

Clinical Validation:

- Simulating patient interaction via use cases generated by doctors
- Evaluating both questions asked by model and outcome
- Original duration: **10 hours (and lots of misery)**

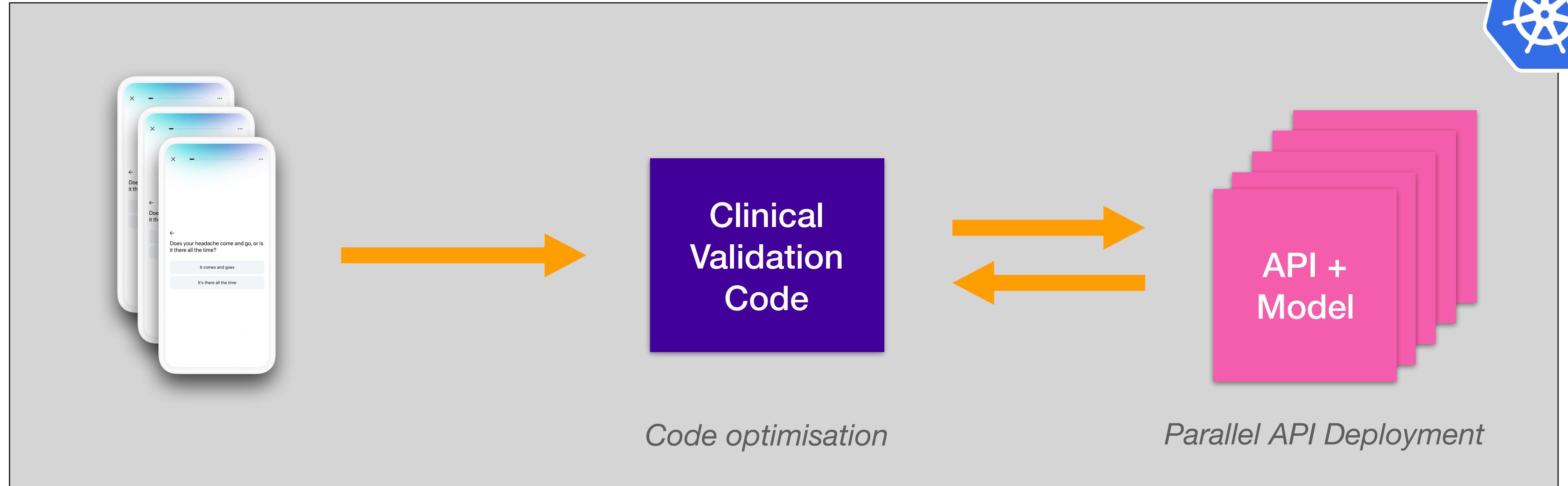


*Thousands
of use cases*

*Managing use cases
+
Assessing results*

Serving Model

Use Case: *Clinical Validation of our Symptom Checker*

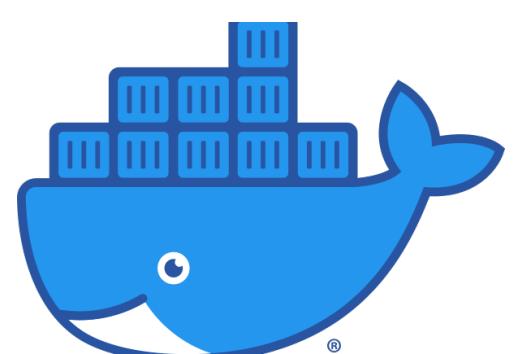


New duration: < 20 minutes

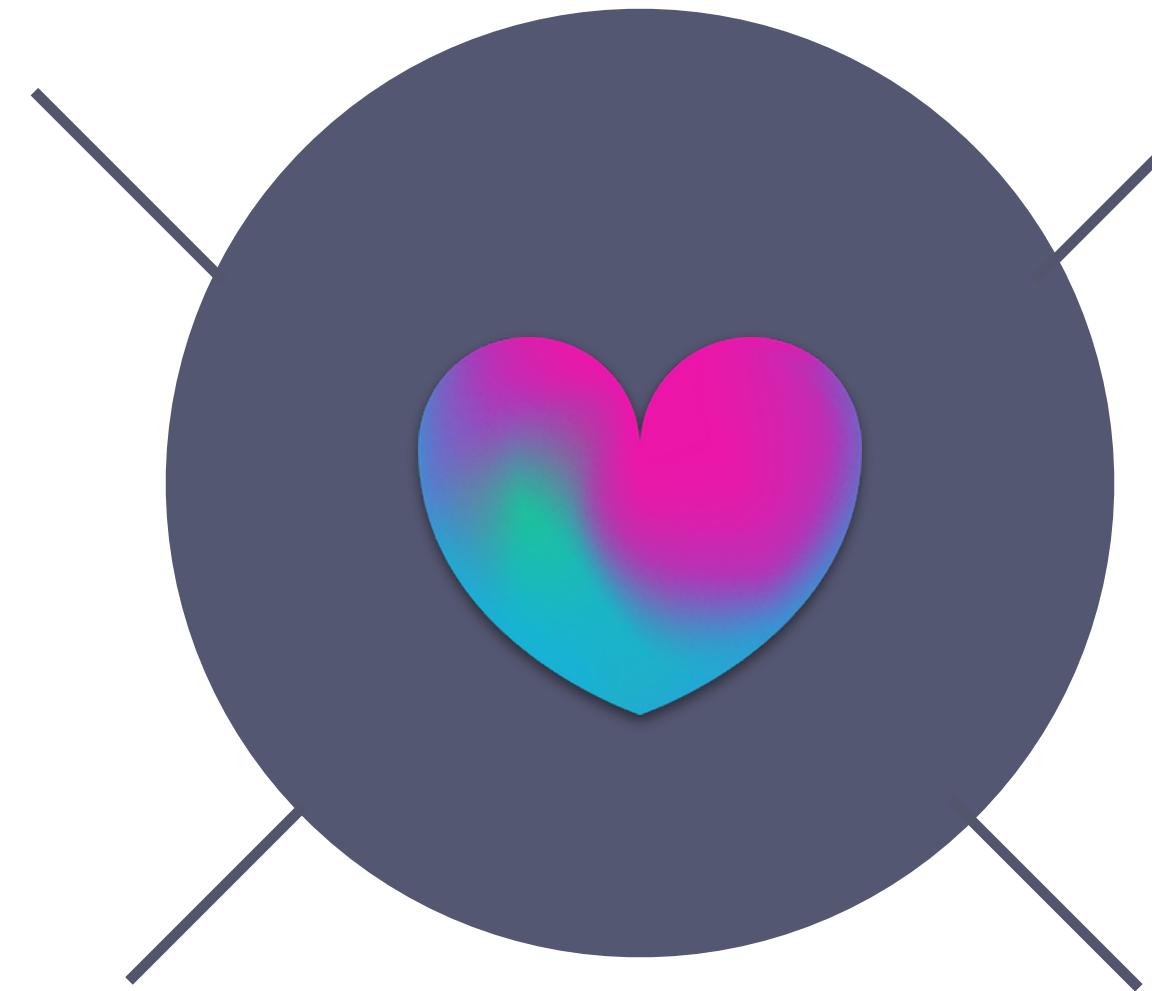
Now running on every Pull Request

Improving feedback loop → faster iterations → **increasing safety and quality of models**

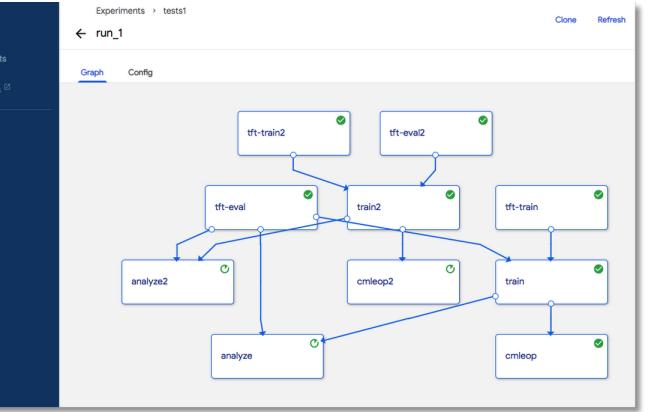
Next steps



Better serving
Improving user experience



Integrating Kubeflow Pipelines



Better Metadata Tracking



Sacred

Wrapping up



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Kubeflow:

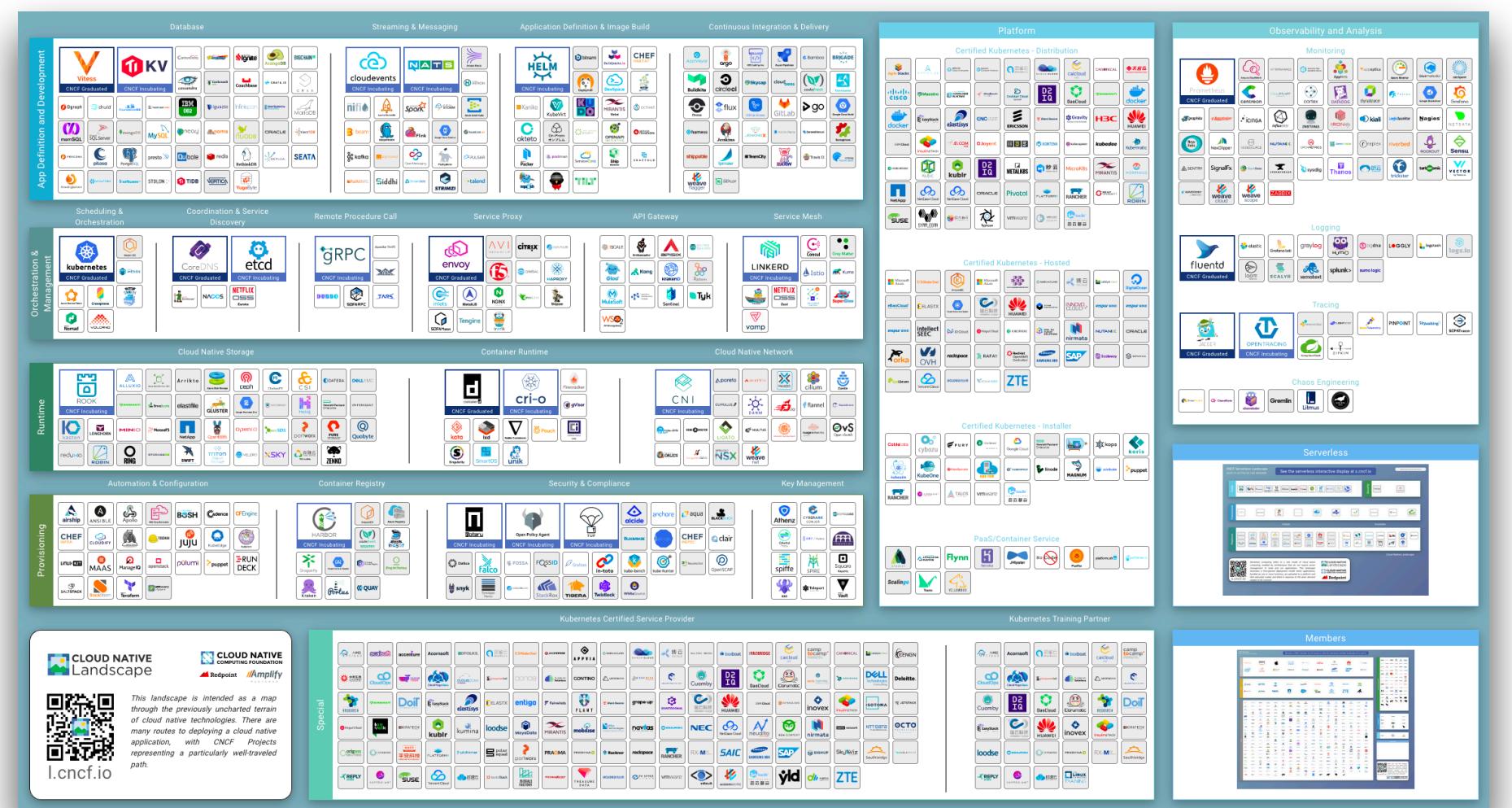
- Most complete ML toolkit for Kubernetes
- Great modularity
- Easy to get started

Security and compliance on K8s:

- Many open-source tools out there can help
- Have a look at the Cloud Native landscape and start from there

MLOps:

- Enabling AI/ML teams with tooling and infrastructure
- Always ask: *what are the big pain points for your AI/ML teams?*
- Focus on 1 pain point, build proof-of-concept, then add as feature



<https://landscape.cncf.io/>

Thank you!



@jeremievallee



jeremie-vallee

*We're hiring in **UK** and **USA**!*



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