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# Smart **Green Computing** **Cloud-Native** Operations



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# Smart Green Computing Cloud-Native Operations

The European Union and the United States have set up a target of at least 50% - 55% net reduction in greenhouse gas emissions by 2030. But, with the sprawling of the cloud-native workloads and the increased demand for resources: are we doing enough?

Many community efforts and open source projects enable the observability of the power consumption from software resources to hardware resources. How can we combine the visibility provided by these tools to achieve the organization's sustainability goals?

In this talk, we combine CNCF projects and other open source communities tools to create and continuously improve Machine Learning models for cluster operations. These ML models consider a holistic view of a system: from application runtimes, node metrics, cluster metrics, and network metrics to the tracing of the interactions among the distributed components. These ML models are used for the "smart operations" of the distributed systems aligning to the organization's carbon and power optimization goals.

# Smart **Green** Computing **Cloud-Native** Operations



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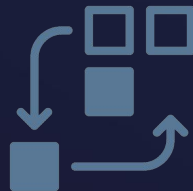


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# Smart **Green** Computing **Cloud-Native** Operations





# Goal-driven Workload Management (1)

## CO2/Energy Aware Workload Placement & Scheduling

### Goal-driven Scheduling

E.g. scheduling workloads to achieve organization's **goal of reduction of CO2 emissions, or reduction of power, or achieve higher energy efficiency**, or to maintain certain latency, etc.

### Governance

Workload placement following local or cross-border regulations

### Continuous Re-evaluation

E.g. Near-Real Time reassessment of placement decision and proactively schedule/deschedule workload.

## Traditional Scheduling



CO2: 1 Units  
Replicas: 3

Workload A



CO2: 3 units  
Replicas: 3

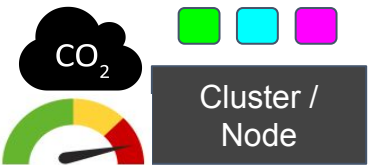
Workload B



CO2: 5 Units  
Replicas: 3

Workload C

~~Goal: < 12 CO2 units per node/cluster~~



Source: **Carbon**  
CO2: 2.23 lb p/kWh

$T\_CO2: 9 \times 2.23 = 20.07$



Source: **Petroleum**  
CO2: 2.13 lb p/kWh

$T\_CO2: 9 \times 2.13 = 19.17$



Source: **Natural Gas**  
CO2: 0.91 lb p/kWh

$T\_CO2: 9 \times 0.91 = 8.19$

# Goal-driven Workload Management (2)

## CO2/Energy Aware Workload Placement & Scheduling

### Goal-driven Scheduling

E.g. scheduling workloads to achieve organization's **goal of reduction of CO2 emissions, or reduction of power, or achieve higher energy efficiency**, or to maintain certain latency, etc.

### Governance

Workload placement following local or cross-border regulations

### Continuous Re-evaluation

E.g. Near-Real Time reassessment of placement decision and proactively schedule/deschedule workload.

## Goal-driven (e.g. CO2) scheduling/rescheduling



CO2: 1 Units  
Replicas: 3

Workload A



CO2: 3 units  
Replicas: 3

Workload B

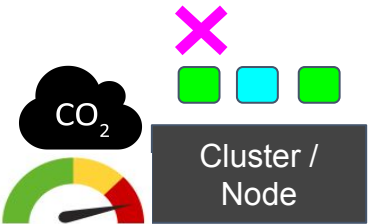


CO2: 5 Units  
Replicas: 3

Workload C

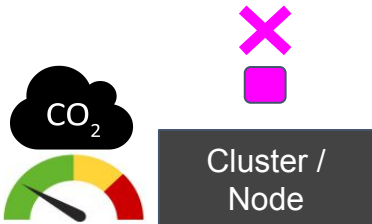


Goal: **< 12 CO2** units per node/cluster



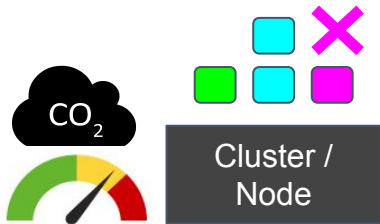
Source: **Carbon**  
CO2: 2.23 lb p/kWh

$$T\_CO2: 5 \times 2.23 = 11.15$$



Source: **Petroleum**  
CO2: 2.13 lb p/kWh

$$T\_CO2: 5 \times 2.13 = 10.65$$



Source: **Natural Gas**  
CO2: 0.91 lb p/kWh

$$T\_CO2: 12 \times 0.91 = 10.92$$

$$(Max) 13 \times 0.91 = 11.83$$

# Domain-aware Workload Management

## Domain Aware Workload Placement & Scheduling

### Governance

Workload placement following local or cross-border regulations

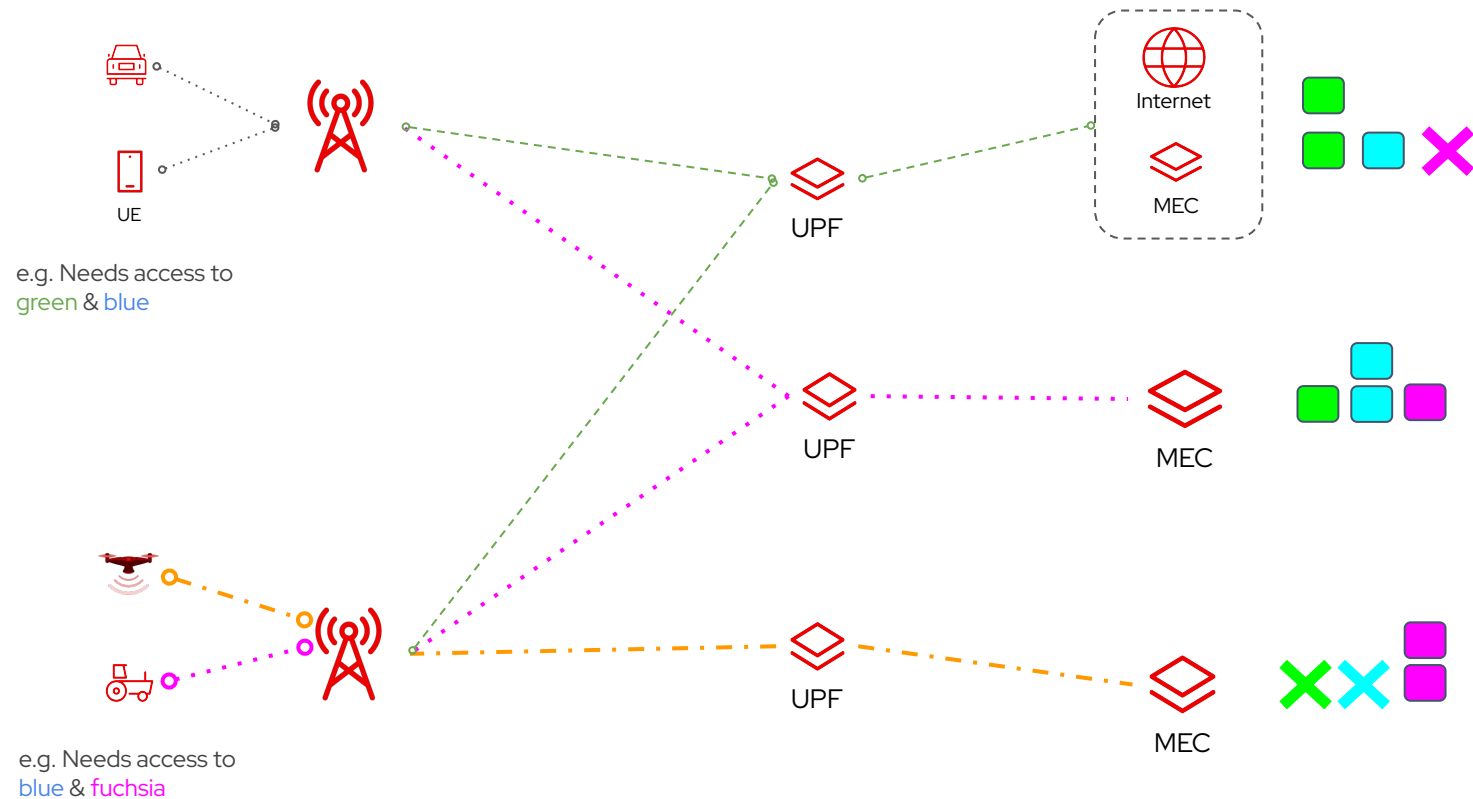
### Domain-aware Scheduling

E.g. scheduling specialized workloads (e.g. 5G UPF, MEC apps) to achieve organization's goal while maintaining the strict constraints imposed by the app or service requirements (e.g. latency, throughput, number of sessions)

### Continuous Re-evaluation

E.g. Near-Real Time reassessment of placement decision and proactively schedule/deschedule workload.

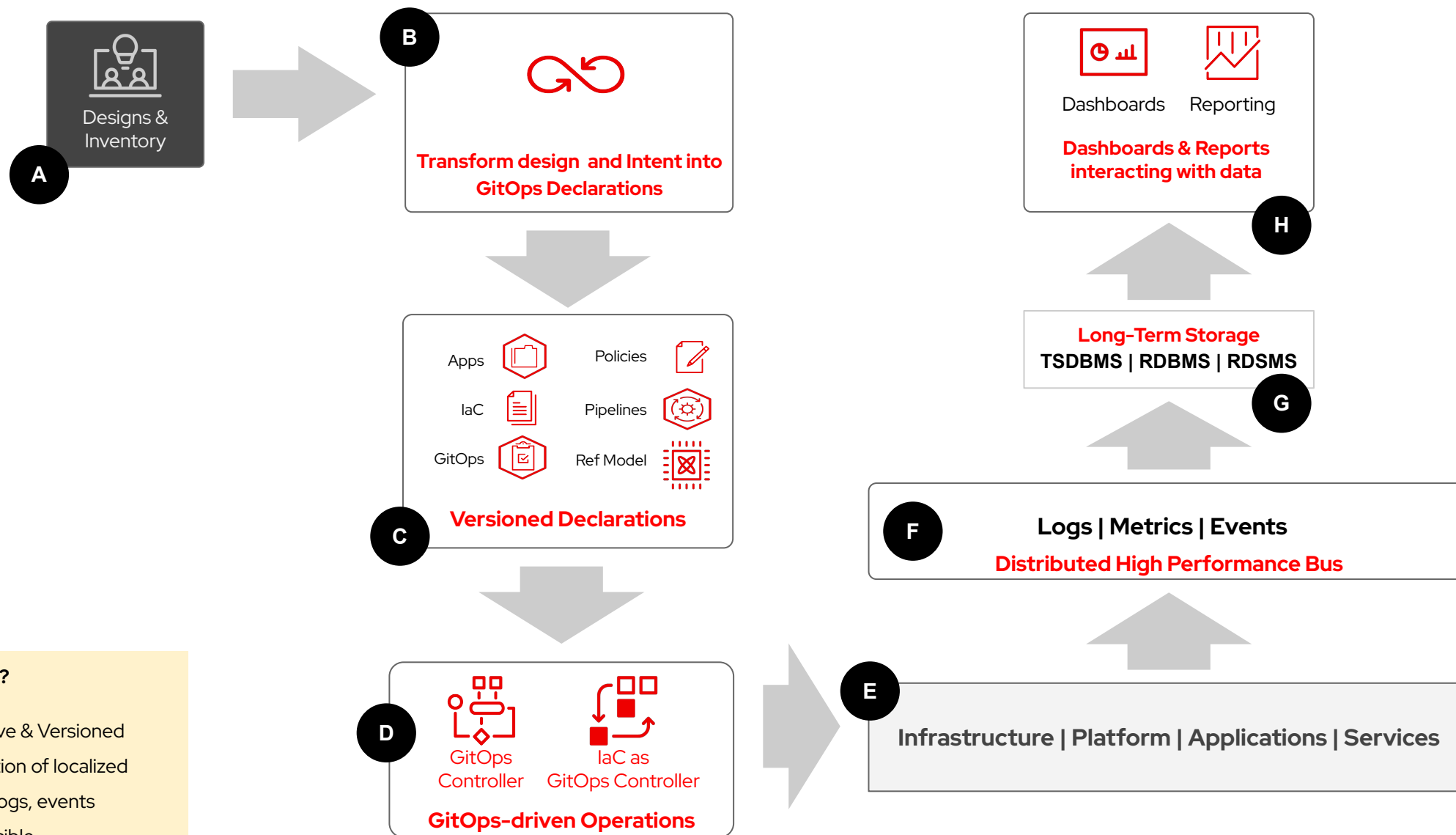
## Workload Placement for 5G Networks





# Composable Experimentation & Operations Stack Design

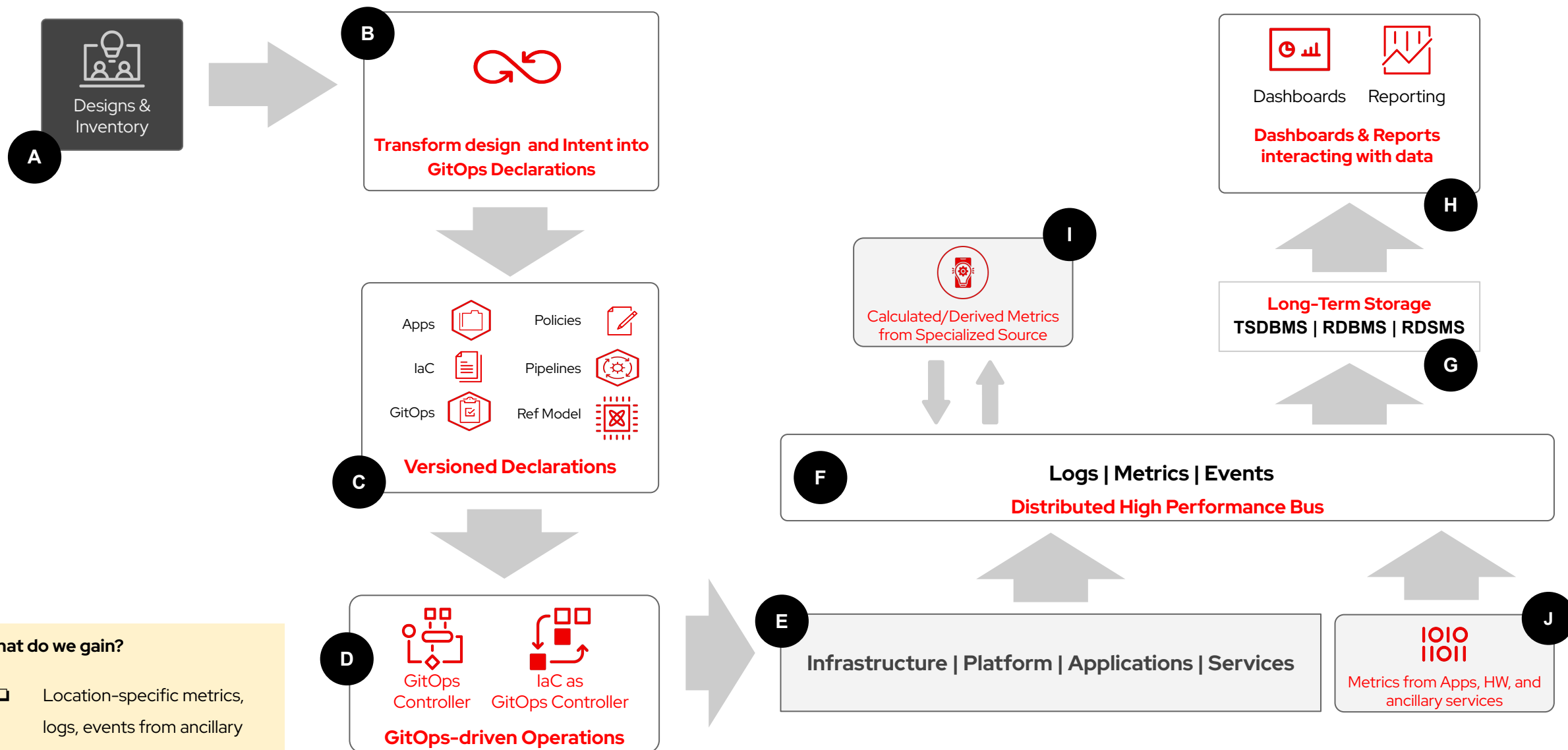
# Step 1: Adopt GitOps Model



## What do we gain?

- ❑ Declarative & Versioned
- ❑ Aggregation of localized metrics, logs, events
- ❑ Reproducible

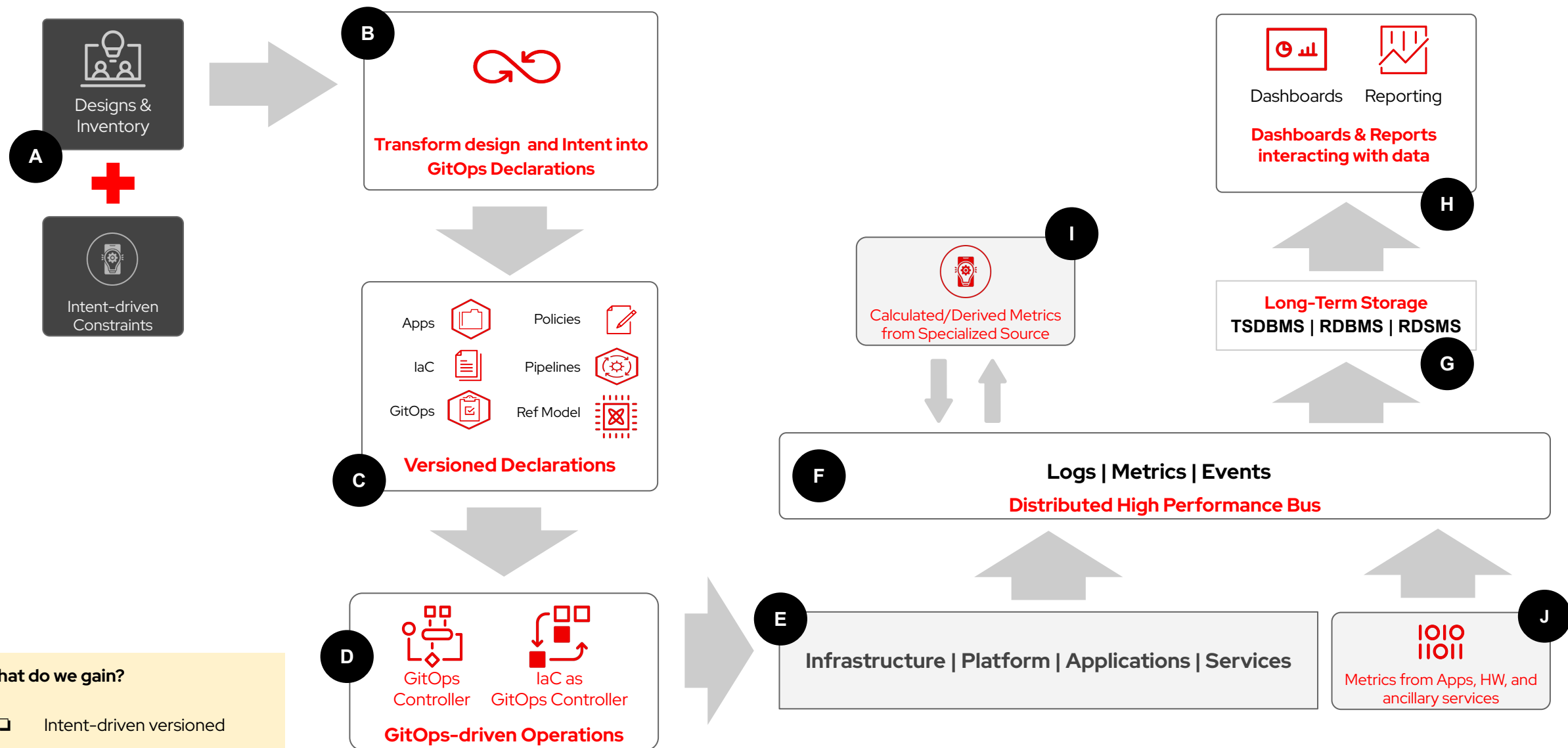
# Step 2: Adopt Enhanced Visibility



## What do we gain?

- Location-specific metrics, logs, events from ancillary hardware and services

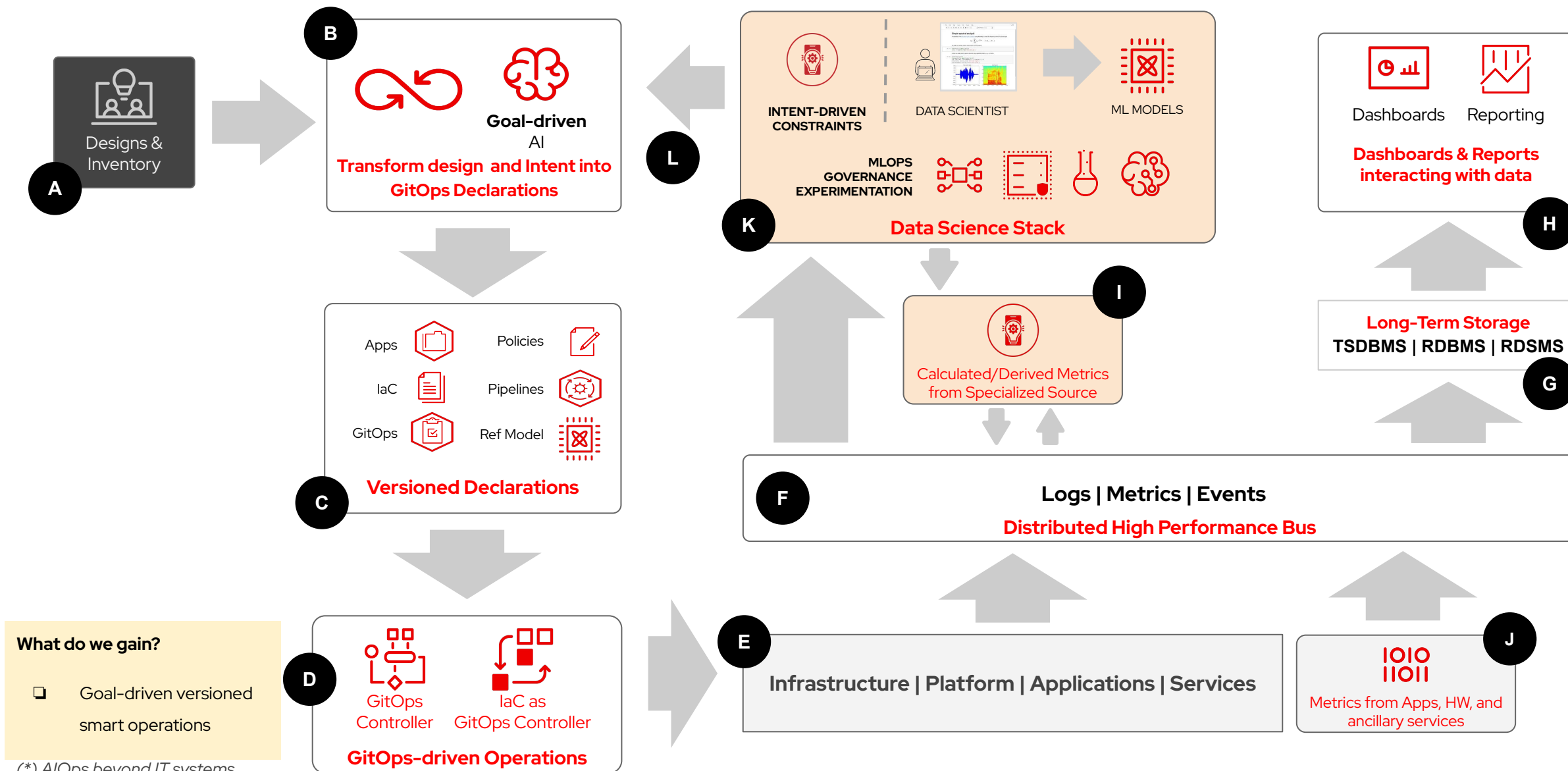
# Step 3: Adopt Intent Driven Orchestration



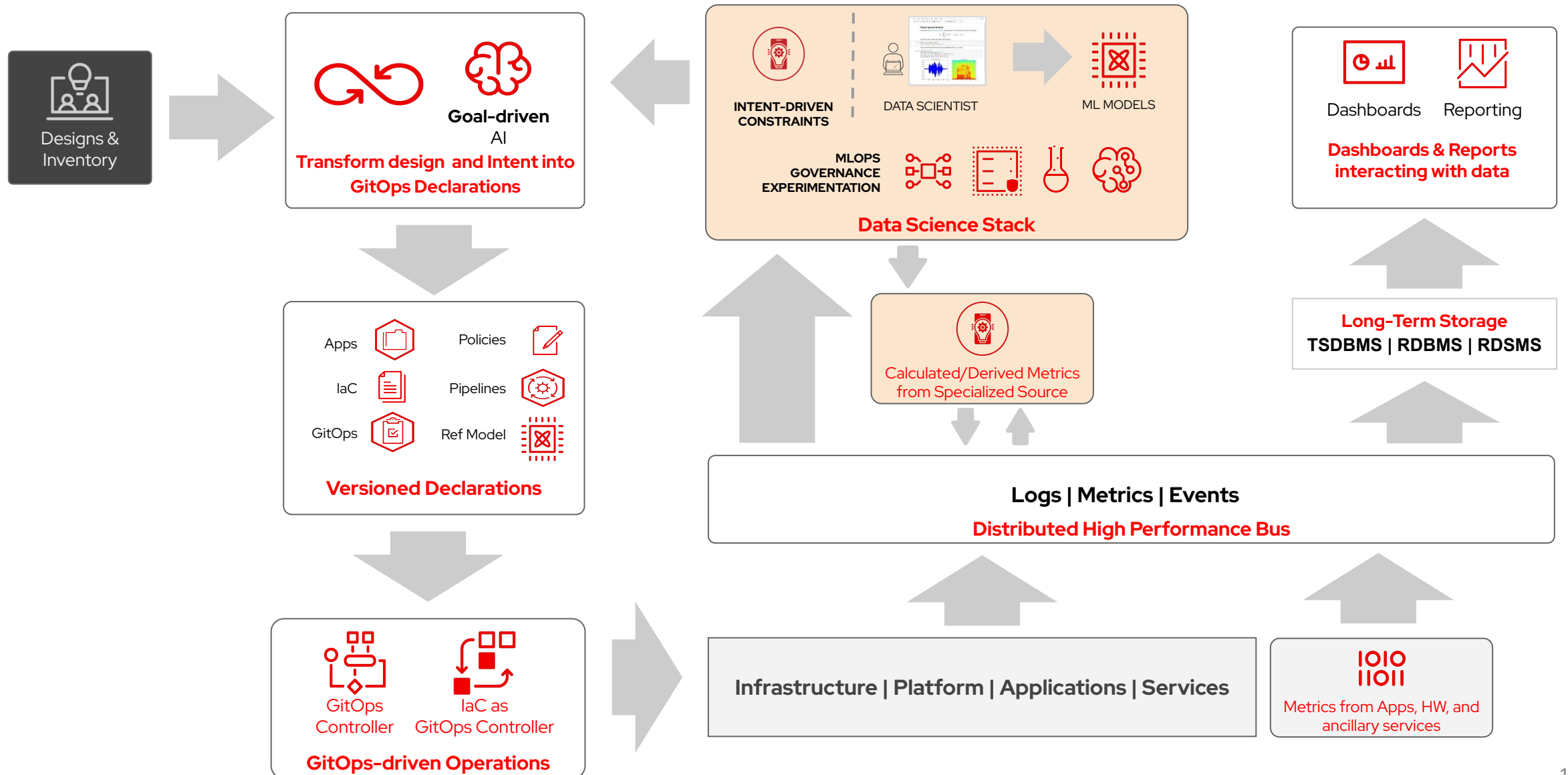
## What do we gain?

- Intent-driven versioned orchestration

# Step 4: Adopt **Goal-Driven AI** (Intent + AIOps\*)



# The Complete Stack





# Example of Use Cases for Stack

- **Goals Aware Workload Management**

- **Maximum CO2 emission per-node or cluster**
- **Energy aware** service-to-service **path selection**
- **Predictive in-time** workload **provisioning**



This is today's POC!!!

- **Domain-aware autonomous orchestration**

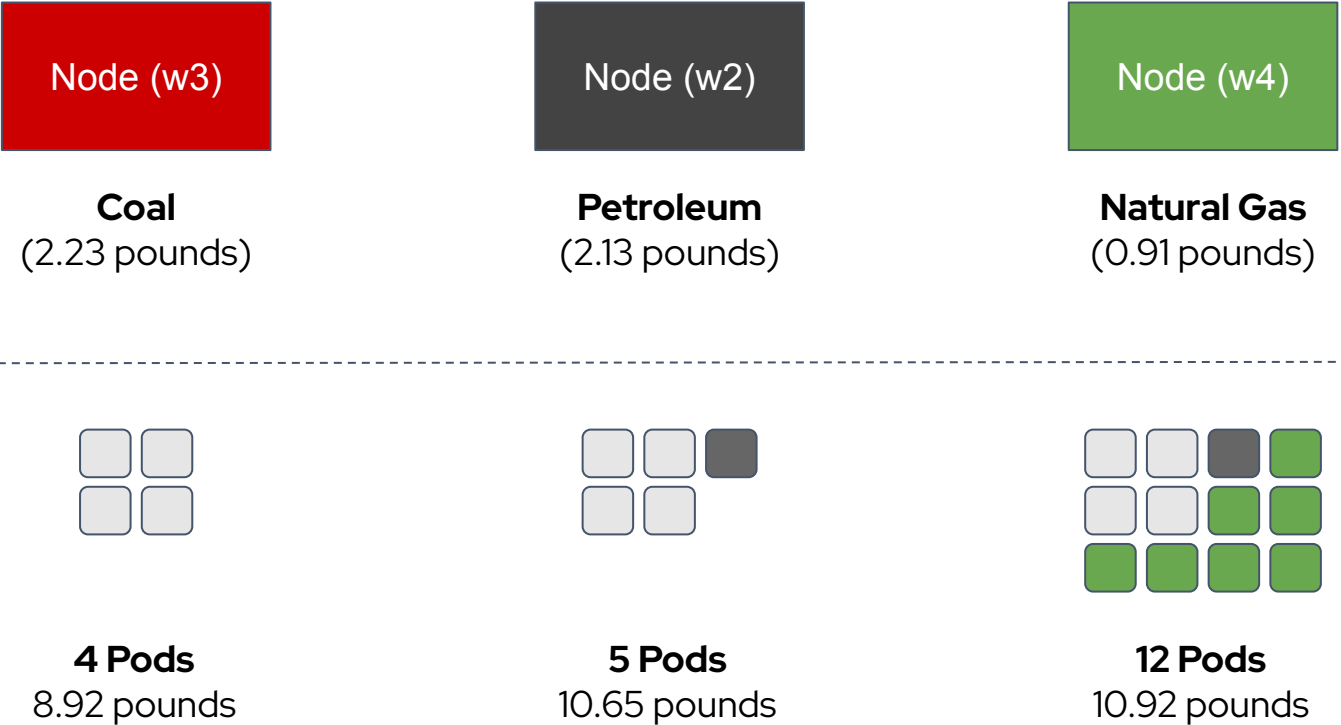
- Autonomous operations of services to **reduce or optimize power** consumption, or **CO2 emissions**
  - **Auto-tiering of services** (cloud, virtualization, bare-metal) based on multi-dimensional factors (financial models, energy sources, geo location, time, bandwidth, latency, QoE)
  - **Smart maintenance** operations (auto-scheduling and execution of maintenance windows)

# CO2 & Energy Aware Scheduling POC

# CO<sub>2</sub> Emission Aware Placement & Scheduling

1 Watt = 1W = 1 J/s (1 Joule p/sec)  
1kW = 1000 J/s

## CO<sub>2</sub> Emissions per pounds per kWh



Goal max 11 pounds per kWh (per Node)

□ = 1 kWh Pod

(\*) US Energy Information Administration How much carbon dioxide is produced per kilowatt hour of U.S. electricity generation? (<https://www.eia.gov/tools/faqs/faq.php?id=74&t=11>)

# PoC High-Level Architecture

## COMPONENTS

### Governance

Policy engine to enforce scheduler usage for emission aware placement.

### Scheduler Policy

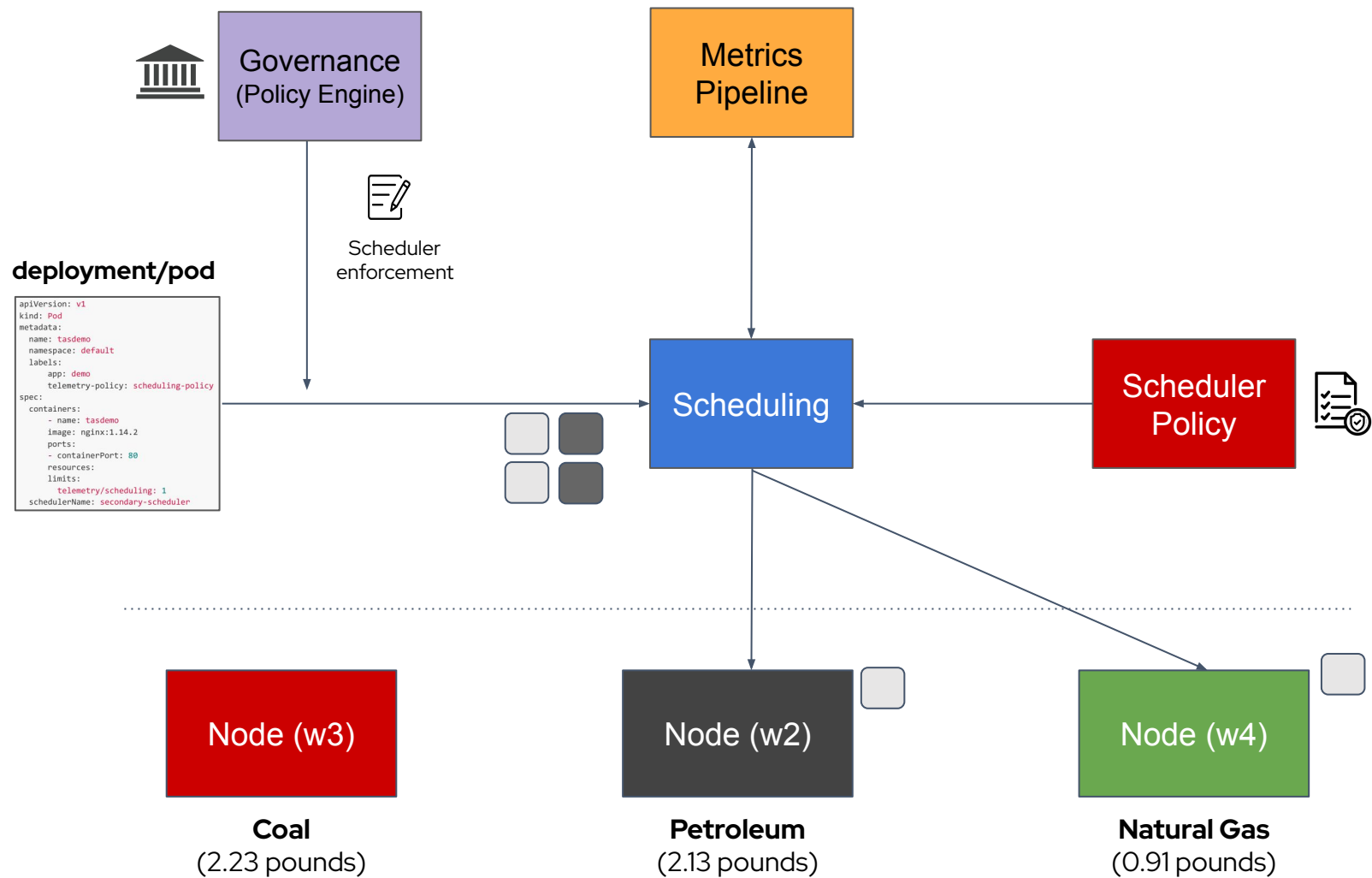
Define strategies and rules for scheduling, descheduling and "dont schedule". Policies are consumed by the Scheduling component.

### Scheduling

Policy driven scheduling. Consumes policies, filters nodes in violation and makes placement/deschedule decision.

### Metrics Pipeline

Provides metrics to be used in the policies and consumed by the scheduler.



# Metrics Pipeline

## METRICS PIPELINE

**KEPLER** (K8s Efficient Power Level Exporter)  
Uses eBPF to probe energy related system stats and export as Prometheus metrics.

### Telegraf

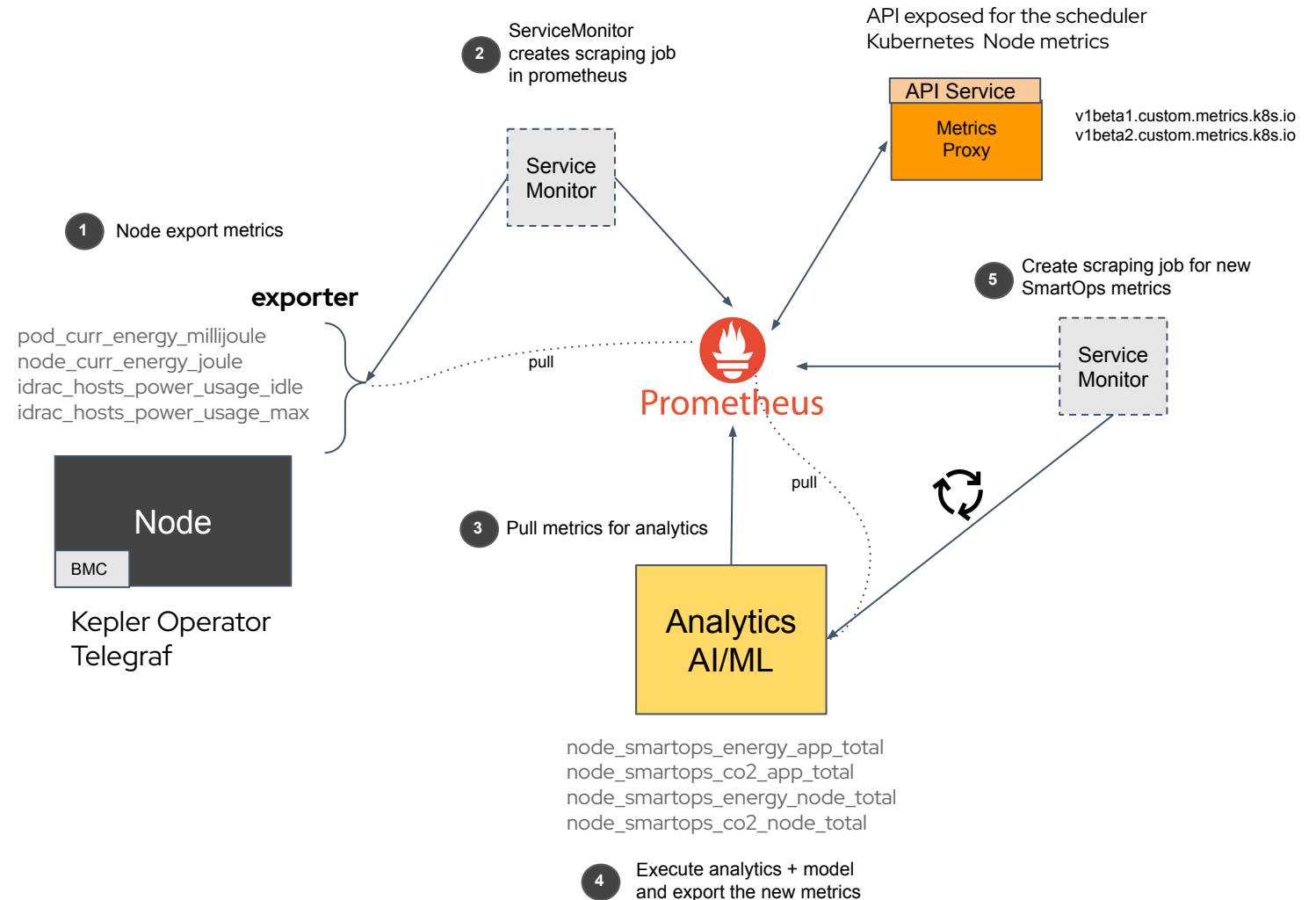
Server agent for collecting and reporting metrics. Uses custom SNMP plugin to pull power usage metrics from iDRAC BMC.

### Analytics AI/ML Agent

Custom analytics agent. Pulls metrics from Prometheus. Add penalties based on node type. Feeds AI/ML model and export SmartOPS metrics.

### Metrics Proxy

Convert prometheus metrics into Kubernetes Node Metrics to be consumed by the scheduler.



## GOVERNANCE / SCHEDULING

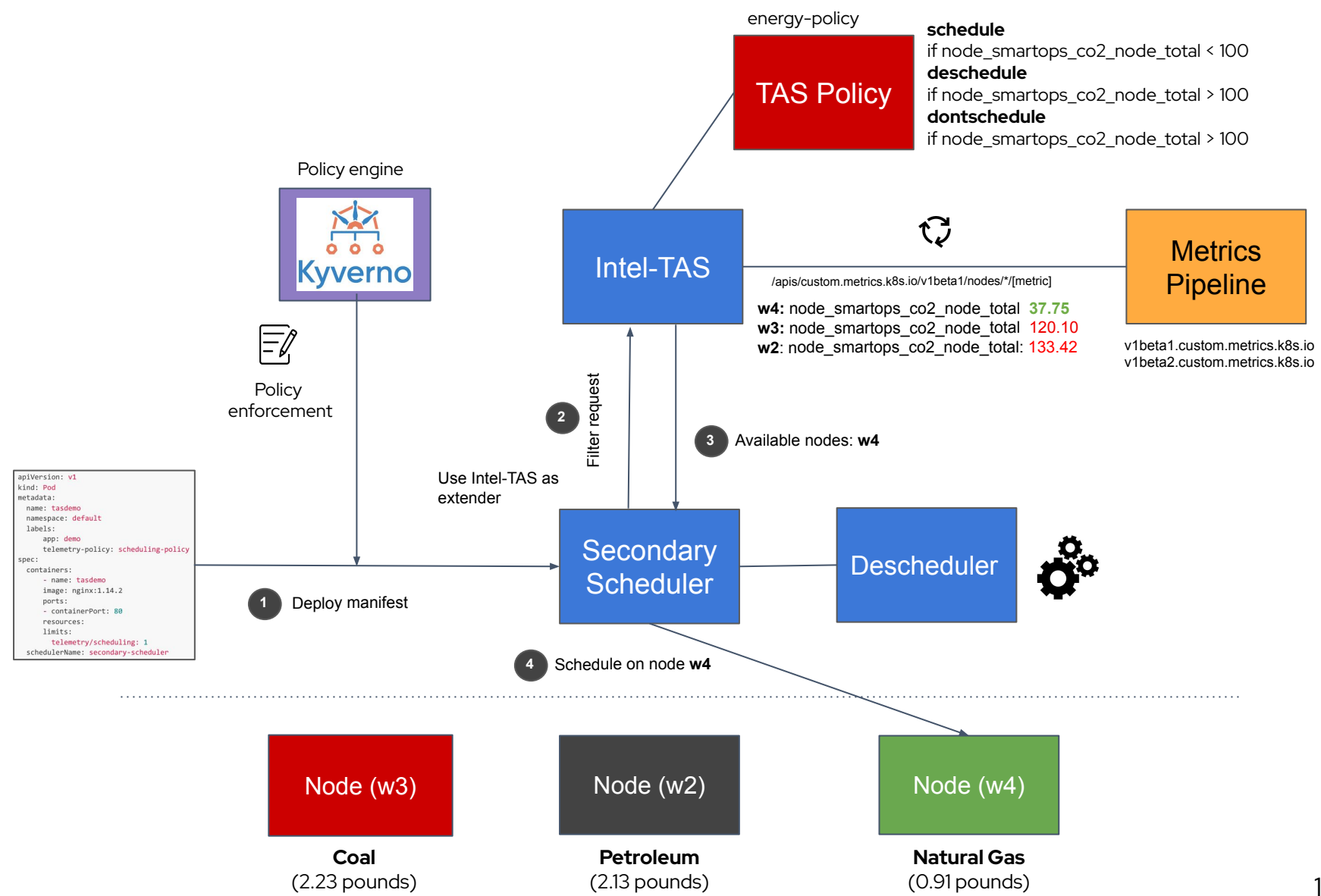
**Kyverno**  
Policy engine. Used to enforce usage of secondary-scheduler and the energy-policy by injecting additional specs on pod manifest

**Secondary Scheduler Operator**  
Customized scheduler image using scheduler plugin KubeSchedulerConfiguration

**Descheduler Operator**  
Deschedule based on affinity/anti-affinity rules

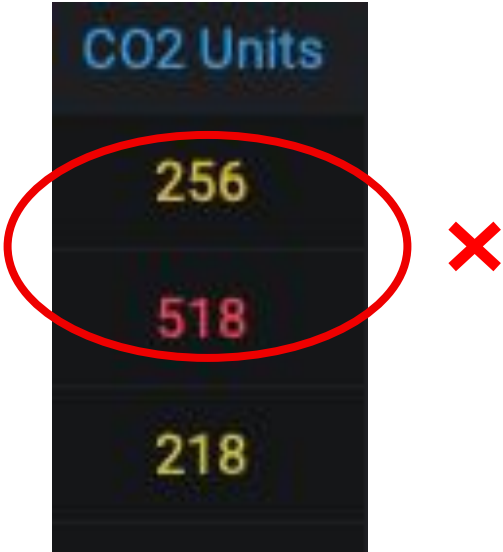
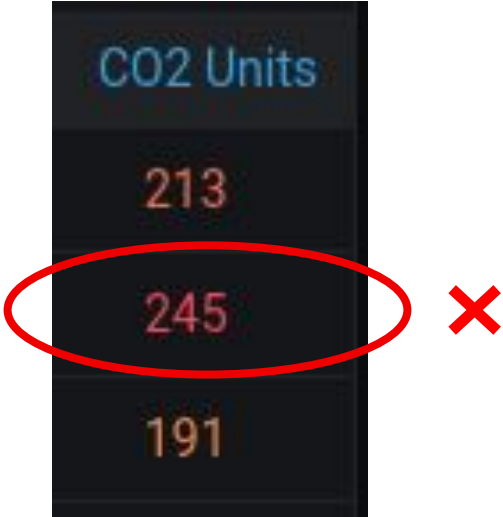
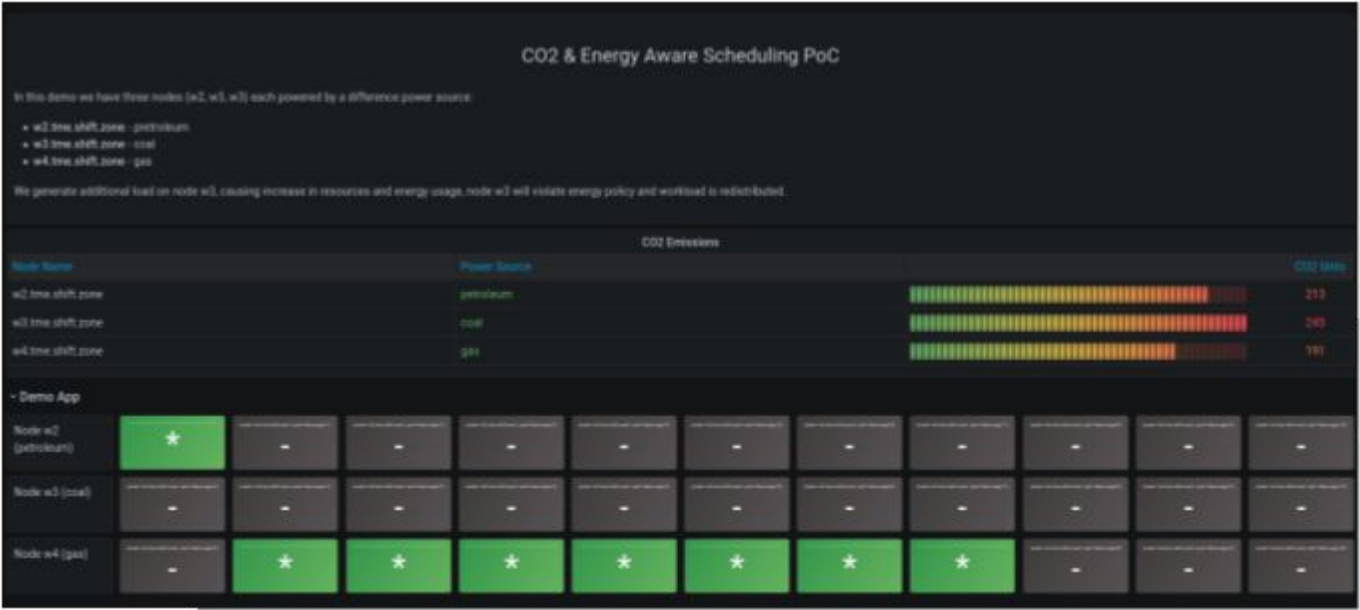
**Intel-TAS** (Telemetry-Aware Scheduler)  
Extender implementation for K8S scheduler that consumes platform metrics and makes intelligent scheduling/descheduling decisions based on defined policies

**TAS Policy**  
Policy to define rules for scheduling or de-scheduling decisions based on metrics.





# POC w/Goal of CO2 < 220 units per Node



# Referenced Projects

- **Kubernetes-based Efficient Power Level Exporter (Kepler)**

- <https://sustainable-computing.io>

- **Kubernetes Event-driven Autoscaling (KEDA)**

- <https://keda.sh>
- **Carbon Aware Scaling POC** (KEDA + Kepler)
  - <https://www.youtube.com/playlist?list=PLhiQtL6gPrYrdpvhCoxD9ydT6TqNQBA4A>

- **Intel Telemetry Aware Scheduling (TAS)**

- <https://github.com/intel/platform-aware-scheduling>
- **Intel Intent Driven Orchestration Planner** (for future integrations)
  - <https://github.com/intel/Intent-Driven-Orchestration>

- **Open DataHub (ODH)** (a Data Science & ML Platform)

- <https://opendatahub.io>
- *Kubeflow, Kafka, Prometheus, Grafana, ArgoCD, Tekton, Knative, Jupyter Notebooks, Elyra, **sklearn**, Seldon, Tensorflow, Airflow, Apache Superset, Hue, and many others.*

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**SUSTAINABILITY TAG**



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