

Security made us do it:

Airflow's new Task Execution Architecture

Who are we?



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Struggles with Airflow 2

- Tasks can talk to Database.
- Nervous while upgrading.
- Hard to Scale.
- Tasks have to be on same network as DB

What happens if I run this DAG in Airflow 2?

```
from airflow.decorators import dag, task

@dag(
    start_date=None,
    schedule=None,
    catchup=False,
)
def danger_dag():
    @task
    def access_db():
        from airflow.utils.db import provide_session
        from sqlalchemy import text

        @provide_session
        def get_dag_runs_directly(session=None):
            session.execute(text("DROP TABLE dag_run CASCADE;"))

            get_dag_runs_directly()

        access_db()

danger_dag()
```

Disclaimer: Astronomer does not accept any responsibility if you try this at home! Please don't.

Upgrade Challenges (2.x → 2.y)

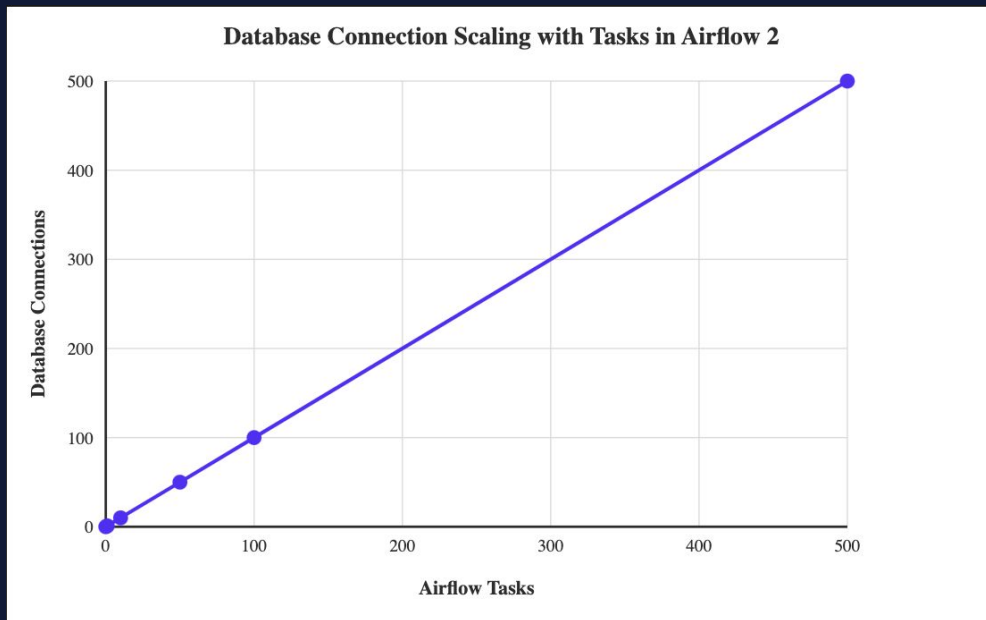
Will this upgrade break my jobs?

- DAGs authored using Airflow's codebase (tight coupling)
- Workers use Airflow's shared codebase to run tasks
- Worry that upgrade will force provider upgrade too

Change to codebase -

- DAG authors impacted! Update *most* DAGs now
- Workers and scheduler must be on identical version

Scaling Challenges in Airflow 2

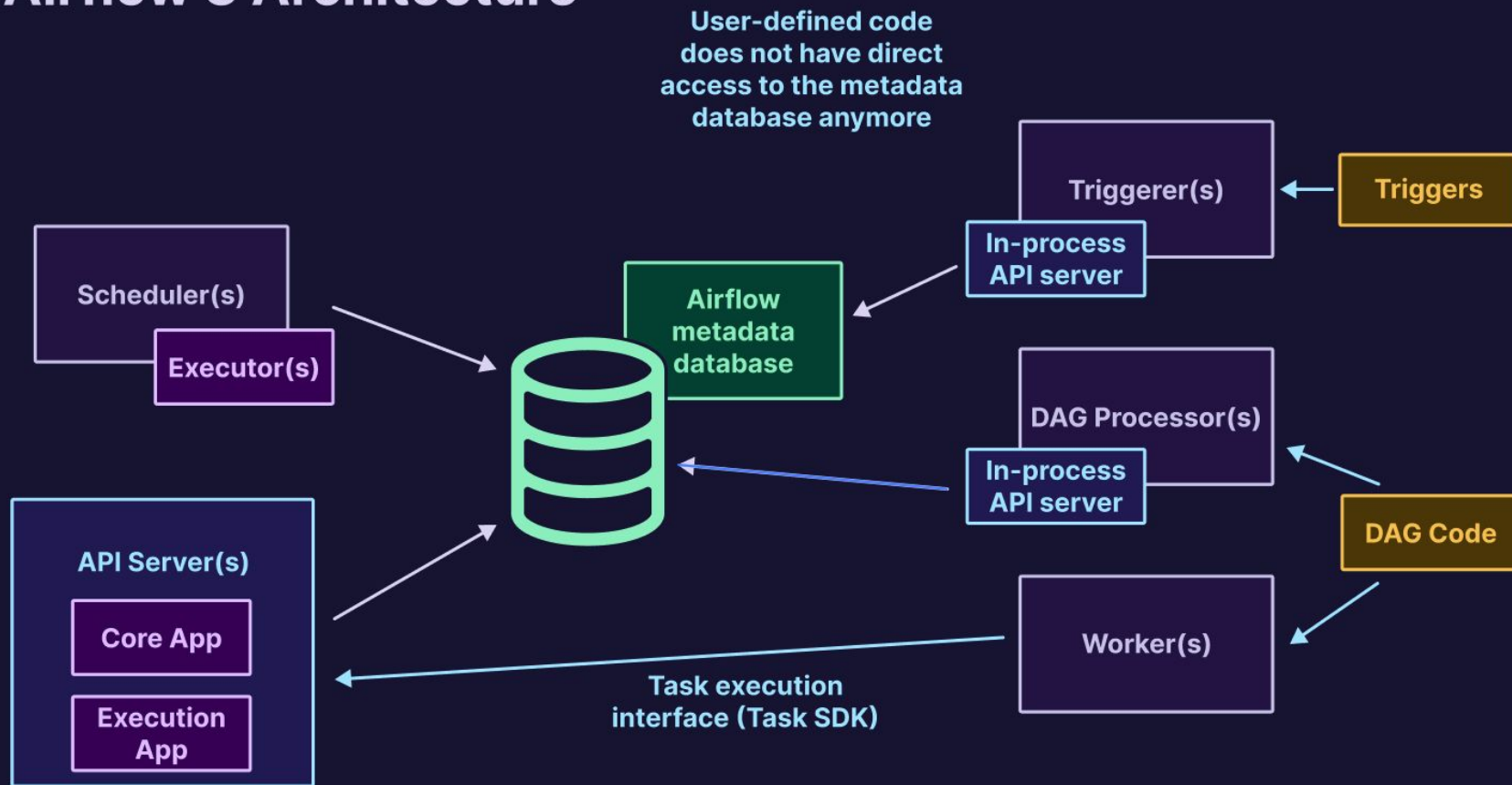


- Each task *can* create DB connections
- n tasks \Rightarrow at least n DB connections!
- DB becomes a bottleneck to scaling

The brave new world of
Airflow 3



Airflow 3 Architecture



New Terms

API Server: Airflow Server Component that provides the sole database access point for all Airflow operations for workers.

Task Execution Interface: The REST API that allows workers to communicate with Airflow.

Task SDK: The lightweight package installed on workers that enables them to talk to Airflow.

Goals

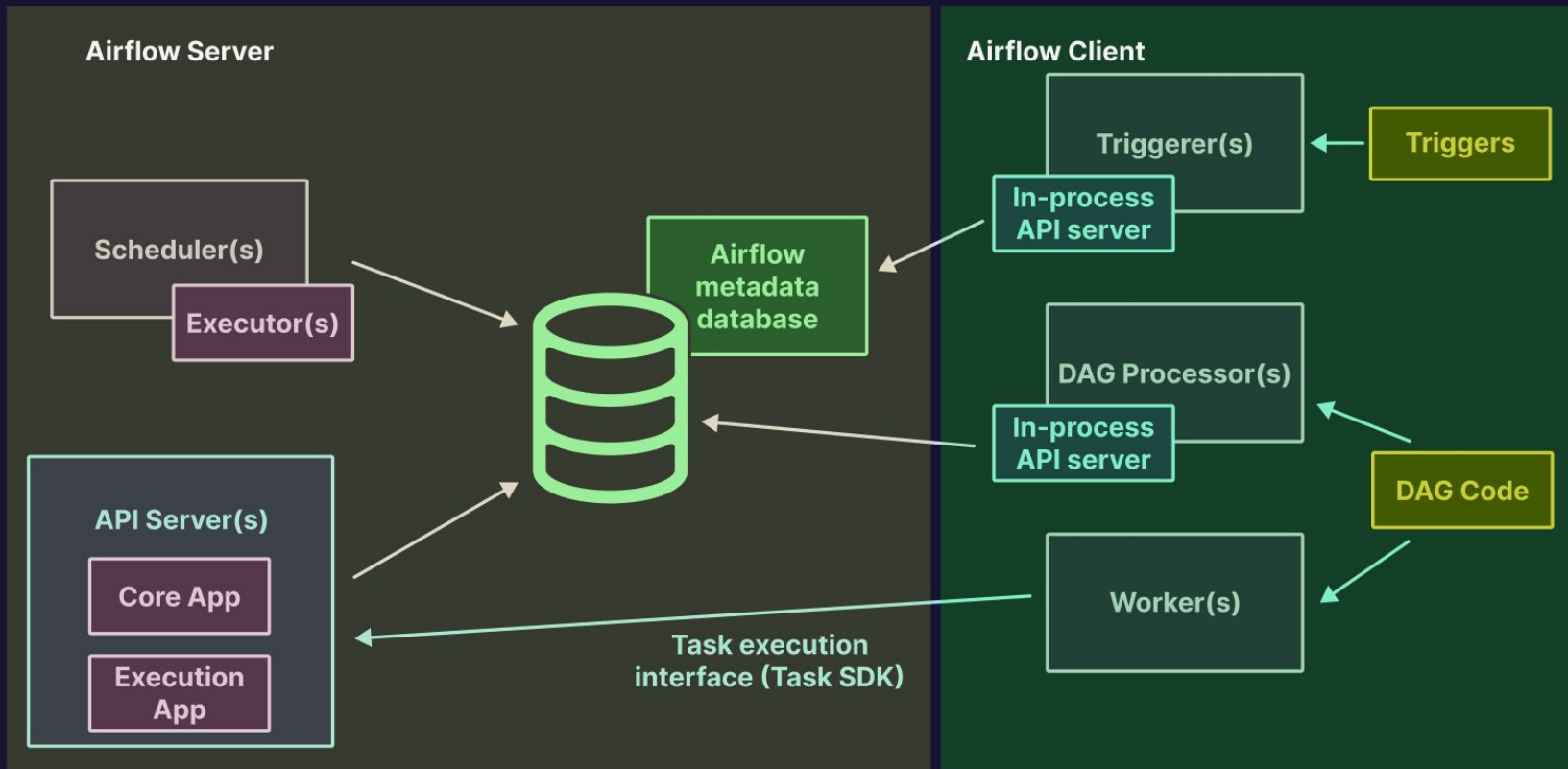
1. Forbid tasks from accessing the metadata DB
2. Workers continue to execute tasks without code change when Deployment is upgraded
3. Enable tasks in multiple languages



Goal #1: **Tasks without direct DB access**

Access everything via an API

Server/Client Split



Task SDK

Lightweight package – the thing providers and Dag authors need to use.

Well defined public python interface for Dag and Provider authors



```
from airflow import DAG
from airflow.decorators import task_group
from airflow.models import Connection, Variable
```

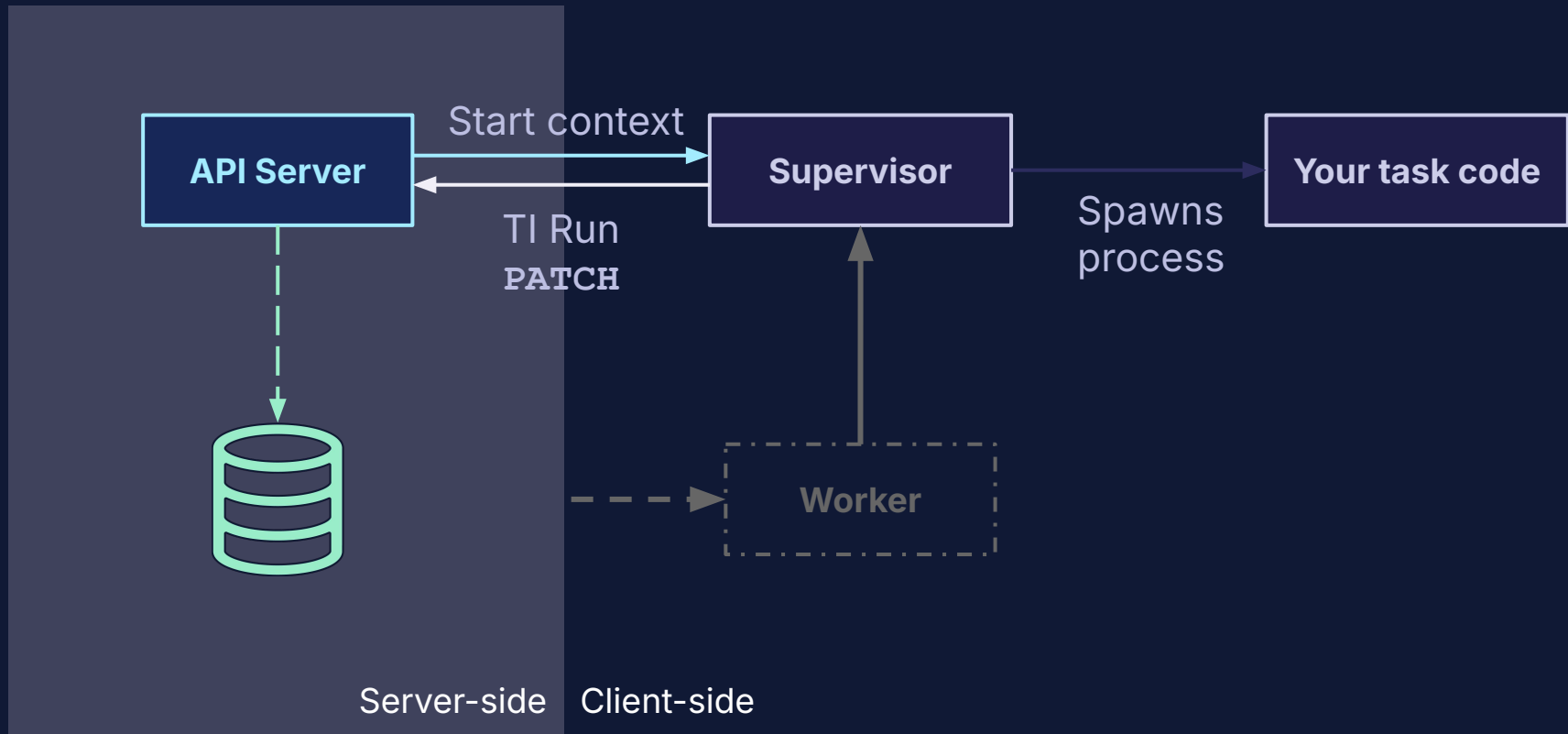
should now be



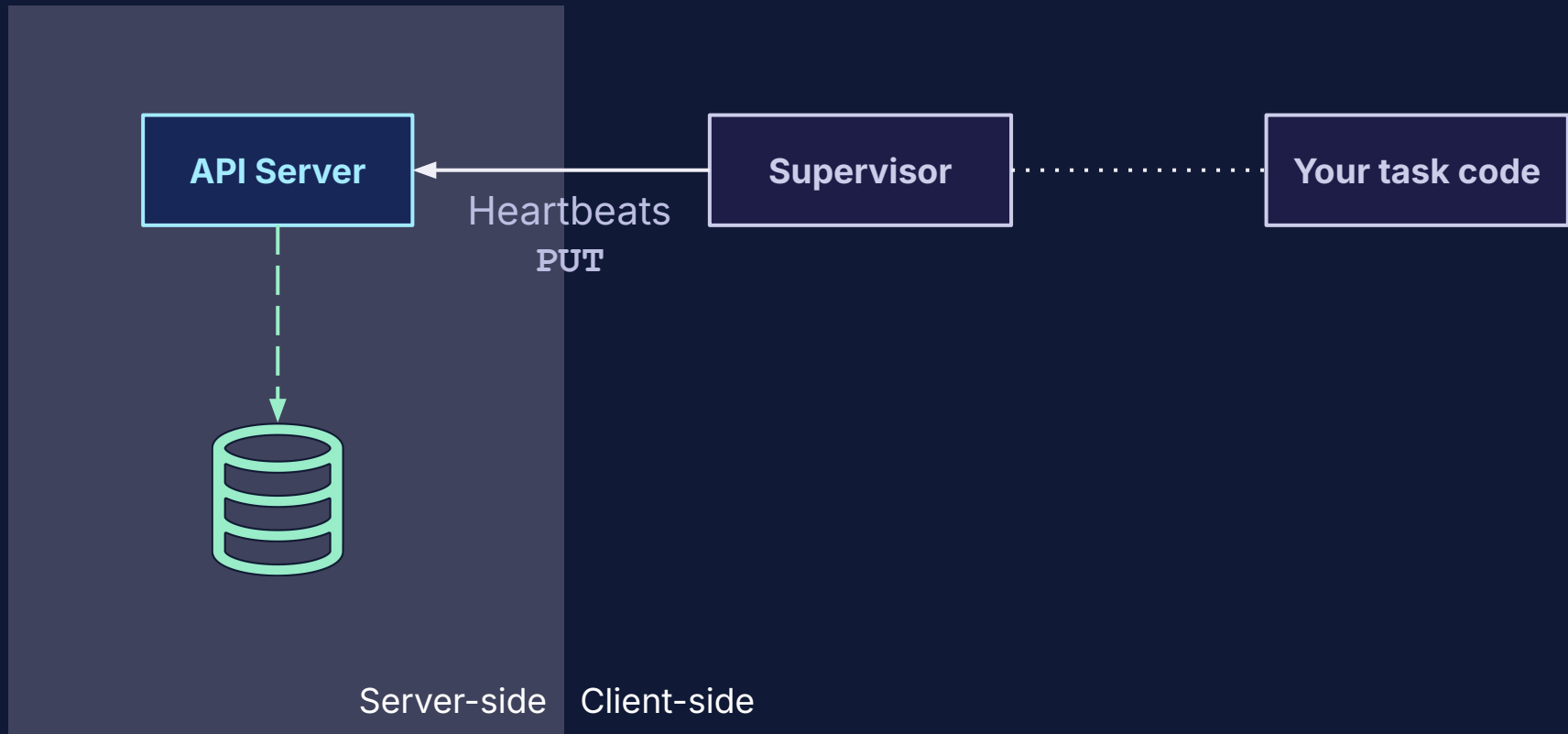
```
from airflow.sdk import Connection, DAG, Variable, task_group
```

Compat shims exist, don't worry!

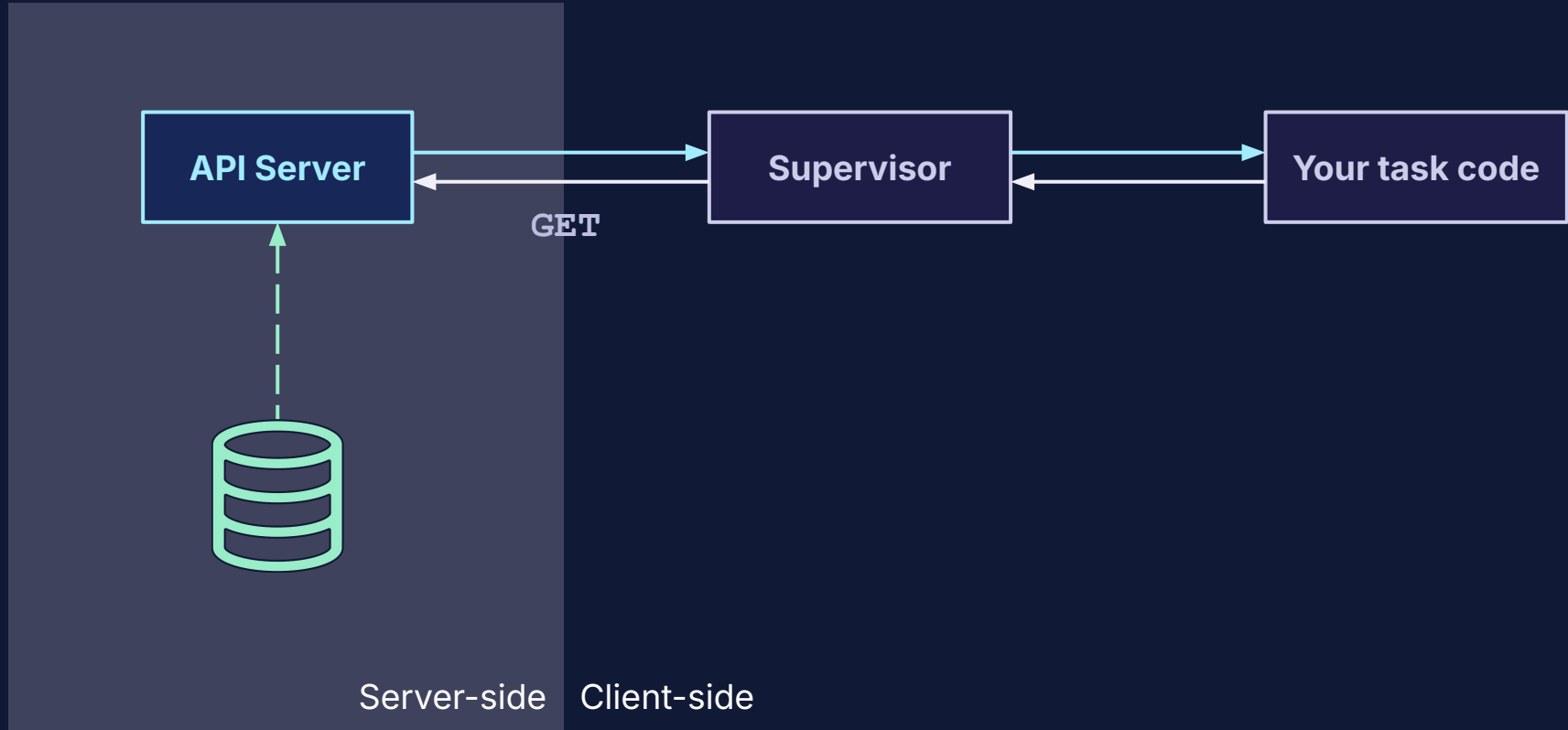
Traffic Flows: Starting a task



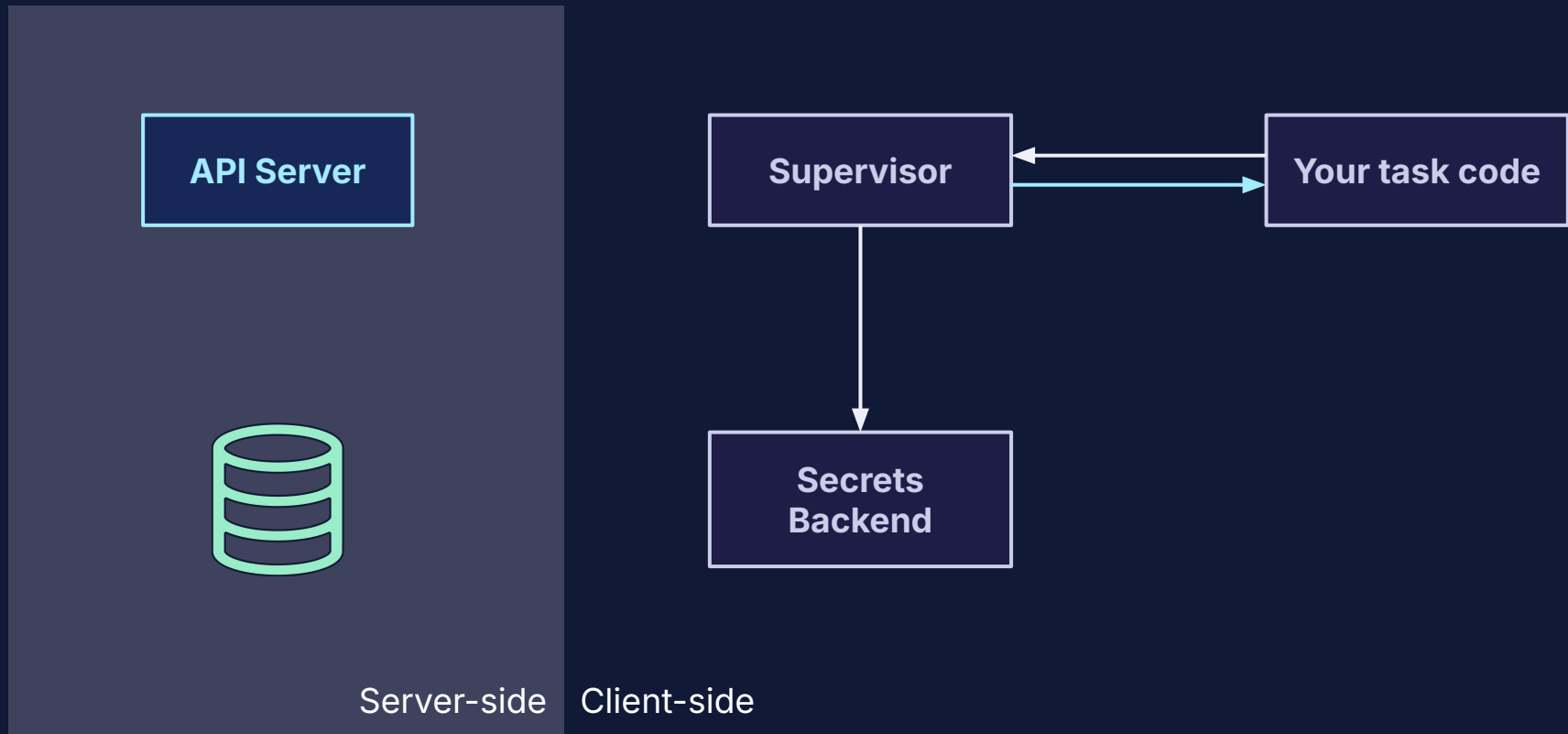
Traffic Flows: TI Heartbeat



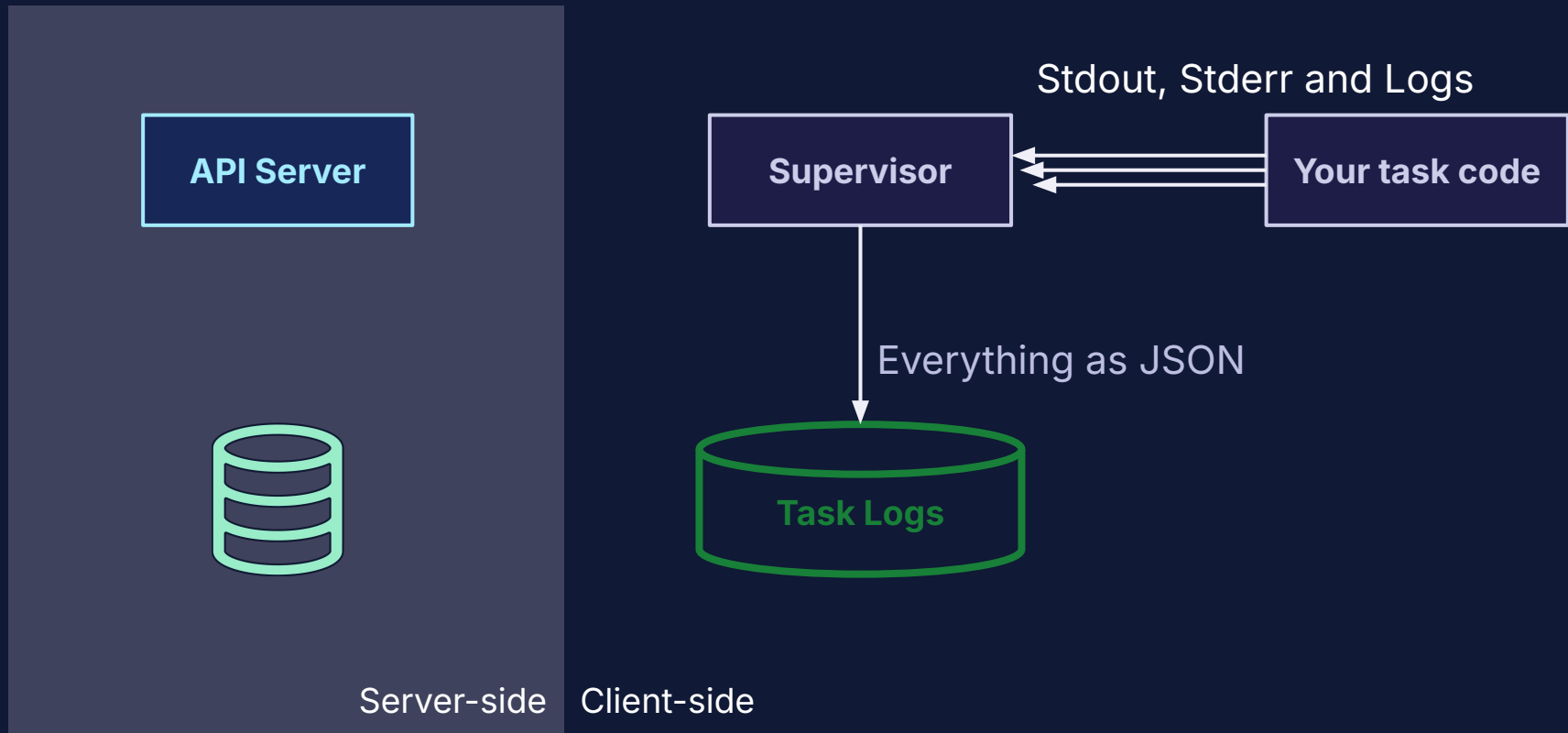
Traffic Flows: Variables/XCom/Connection



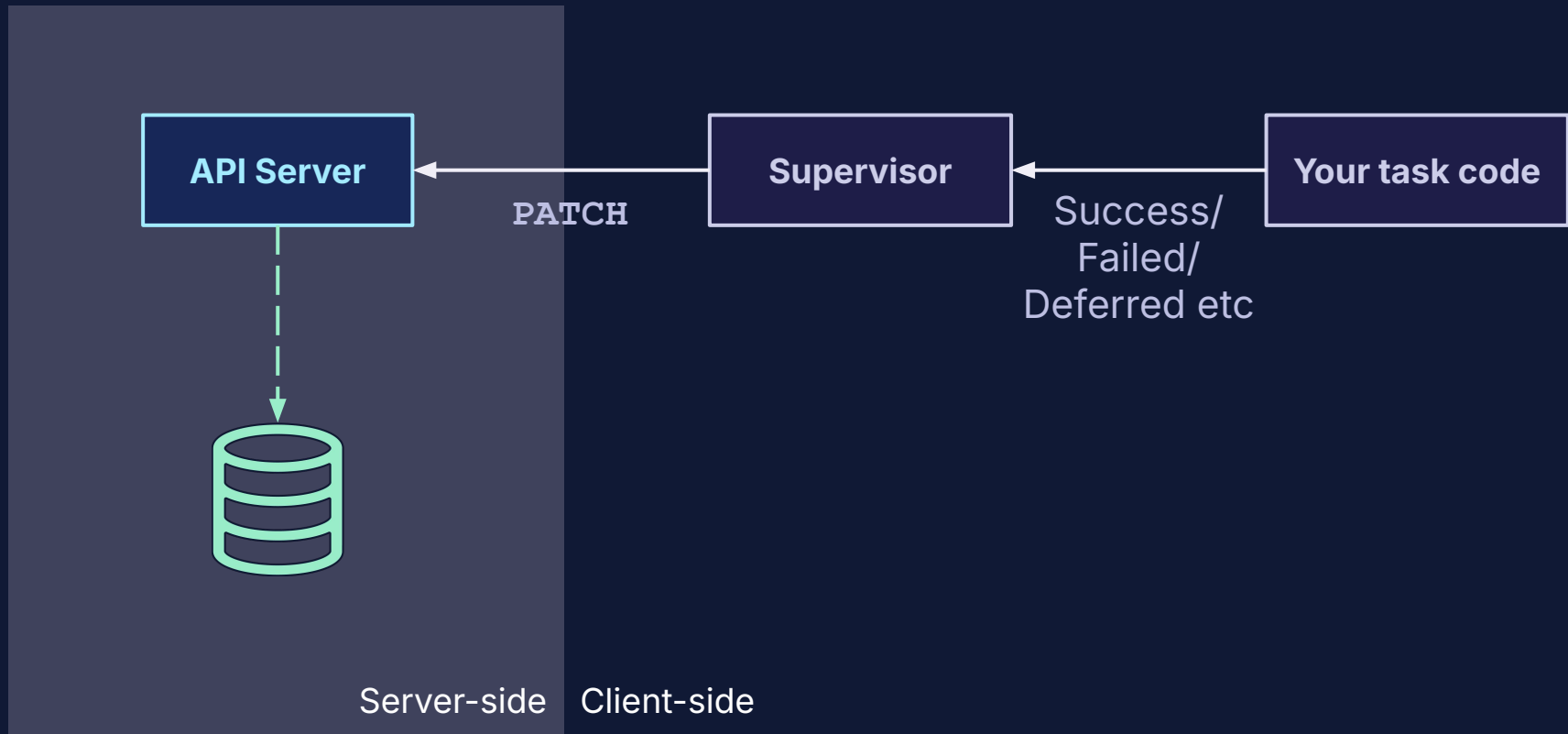
Traffic Flows: Conn w/ Secrets Backend



Traffic Flows: Output and Logs



Traffic Flows: TI Status update

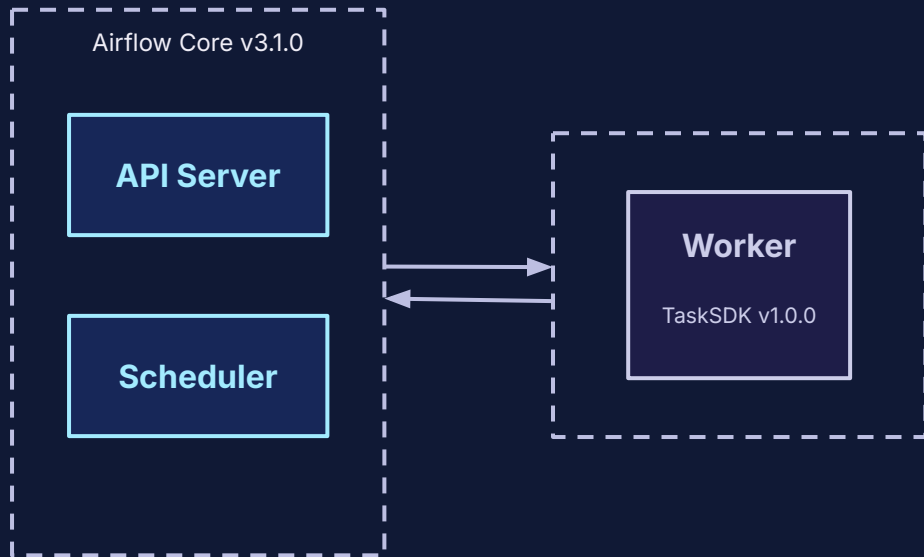


The background of the slide is a deep blue space filled with numerous small white stars and larger, colorful nebulae in shades of purple, blue, and pink. The text is overlaid on the left side of this cosmic scene.

Goal #2: **Support older Workers**

With automatic migration of HTTP
API request and responses

Why?



OLD WORKER (Task SDK v1.0.0):

"Give me the version_id and task_id for this task"

Expecting: {"task_id": "123",
"version_id": "abc", "max_tries": 3}

NEW API SERVER (v3.1.0):

"Here you go"

Response: {"task_id": "123",
"dag_version_id": "abc",
"max_tries": 3}

Common Approaches

1. Multiple Deployments (very expensive)
2. Duplicating Endpoints (not scalable)

```
GET /v1/orders → Old logic  
GET /v2/orders → New logic  
GET /v3/orders → Newest logic
```

Common Approaches

3. Schema-Only Migrations (complex transformation)

```
# Biz logic
def get_user_data():
    return {"id": 123, "name": "John", "new_field": "some_value"}

# Transform for v1 clients
def v1_transform(data):
    return {"user_id": data["id"], "name": data["name"]}

# Transform for v2 clients
def v2_transform(data):
    return {"id": data["id"], "name": data["name"]}

# Transform for v3 clients
def v3_transform(data):
    return data # No change needed - latest format
```

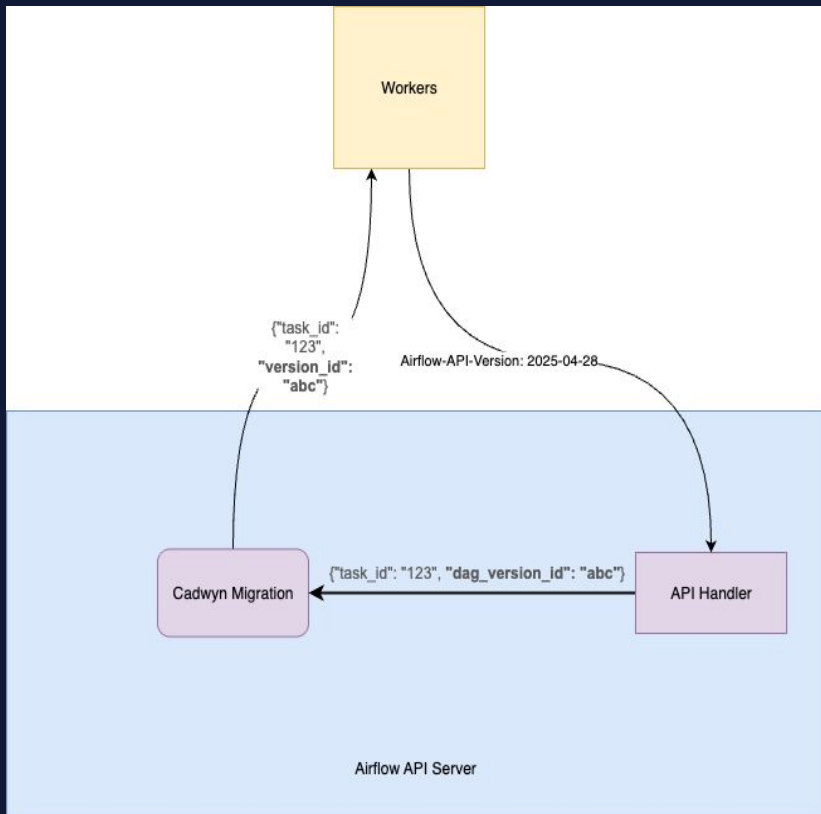

Common Approaches

4. Stripe's Method Transformation by exit "gates"

```
● ● ●  
  
# v3 → v2 gate  
def v3_to_v2(data):  
    # Remove new field added in v3  
    data.pop("new_field", None)  
    return data  
  
# v2 → v1 gate  
def v2_to_v1(data):  
    # Rename field changed in v2  
    data["user_id"] = data.pop("id")  
    return data  
  
# For v1 client: data goes through v3→v2→v1 chain
```

Scalable, Maintainable



Cadwyn Migrations



- Cadwyn: Time Travel
- "Undo" all the changes that happened between those dates

Goal:
Support all task SDK versions back to X version

Conditions for API Migrations

- Server **must** be \geq clients
 -  NEW SERVER \rightarrow OLD CLIENTS
 -  OLD SERVER \rightarrow NEW CLIENTS
- Remote Execution:
 - Server Upgrades first
 - Consumers upgrade when ready
 - Cadwyn handles the version mismatch

Goal #3: Tasks in any language

Golang, Java, *\$your_choice...*

Benefits: Golang SDK!

```
func (m *myBundle) RegisterDags(dagbag v1.Registry) error {
    dag := dagbag.AddDag("tutorial_dag")
    dag.AddTask(transform)
}

func transform(ctx context.Context, client sdk.VariableClient) error {
    val, err := client.GetVariable(ctx, "my_variable")
    if err != nil {
        return err
    }
    log.Info("Obtained variable", key, val)
    return nil
}
```

Feels *native* to Go

DAG bundles are compiled into binaries, worker loads them via hashicorp/go-plugin

Writing a new language SDK

1. Work out how ExecuteTaskWorkload will get to your workers (i.e. implement an Edge executor client)
2. Build a client for the Task Execution API OpenAPI 3.1 spec
3. Workout how you will load your Task functions (Dynamically importing? Load plugins? Precompiled in to worker?)
4. Stick it all in a `while True`
5. ???
6. Profit

Tasks as RPC?

A new way of thinking about Airflow Tasks

- Tasks in their own deployment (i.e. an Airflow worker).
Benefits from code share with main app
- Tasks as RPC: Tasks can run in an existing app deployment!
i.e. run the Airflow Task handler inside your go webserver process

One Orchestrator, Any
Language, Anywhere 