



A GenAI Pipeline for Content Generation with Apache Airflow

May 30th 2024



Agenda

- **Quick introduction:** Airflow and Astronomer
- **Challenges** of GenAI pipelines and how Airflow addresses them
- **Key Airflow features** used in the demo
 - Focus on the latest 2.9 features:
 - **Advanced Dataset Scheduling**
 - **Dynamic Task Mapping**
- **Demo:** Fine-tune GPT for an RAG pipeline for content generation
<https://github.com/astronomer/gen-ai-fine-tune-rag-use-case>



Airflow is the open standard for Workflow Management.

22M+

monthly downloads



GROWTH

46k+

Slack members



COMMUNITY

2800+

contributors



INNOVATION

1600+

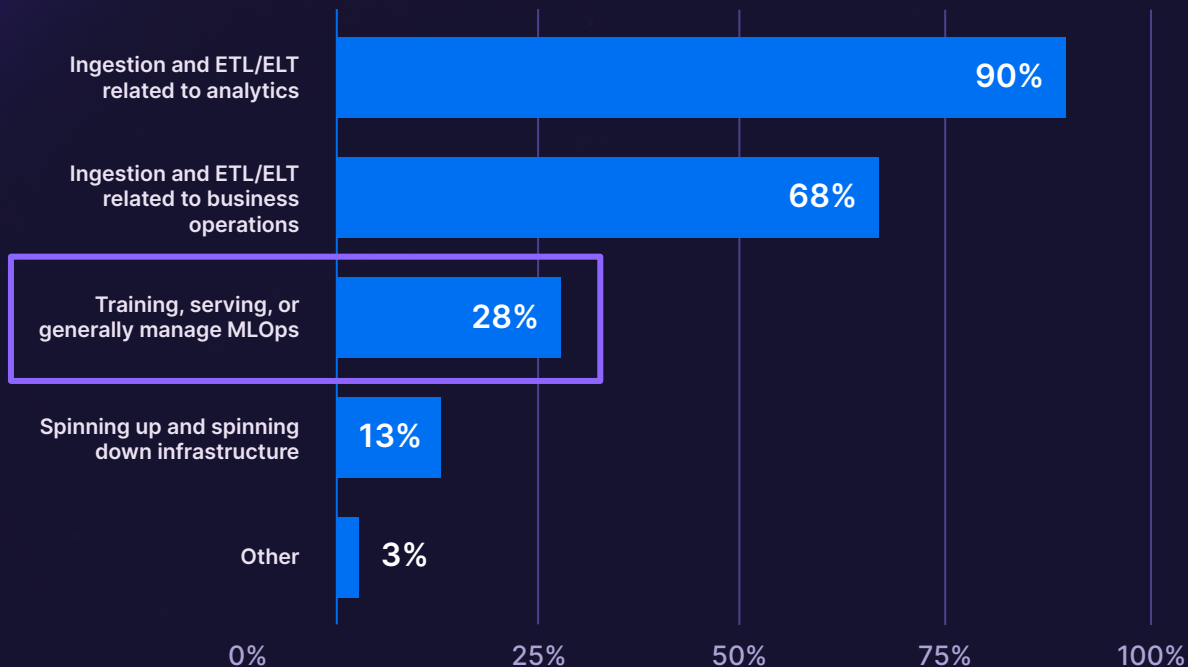
building blocks



ECOSYSTEM



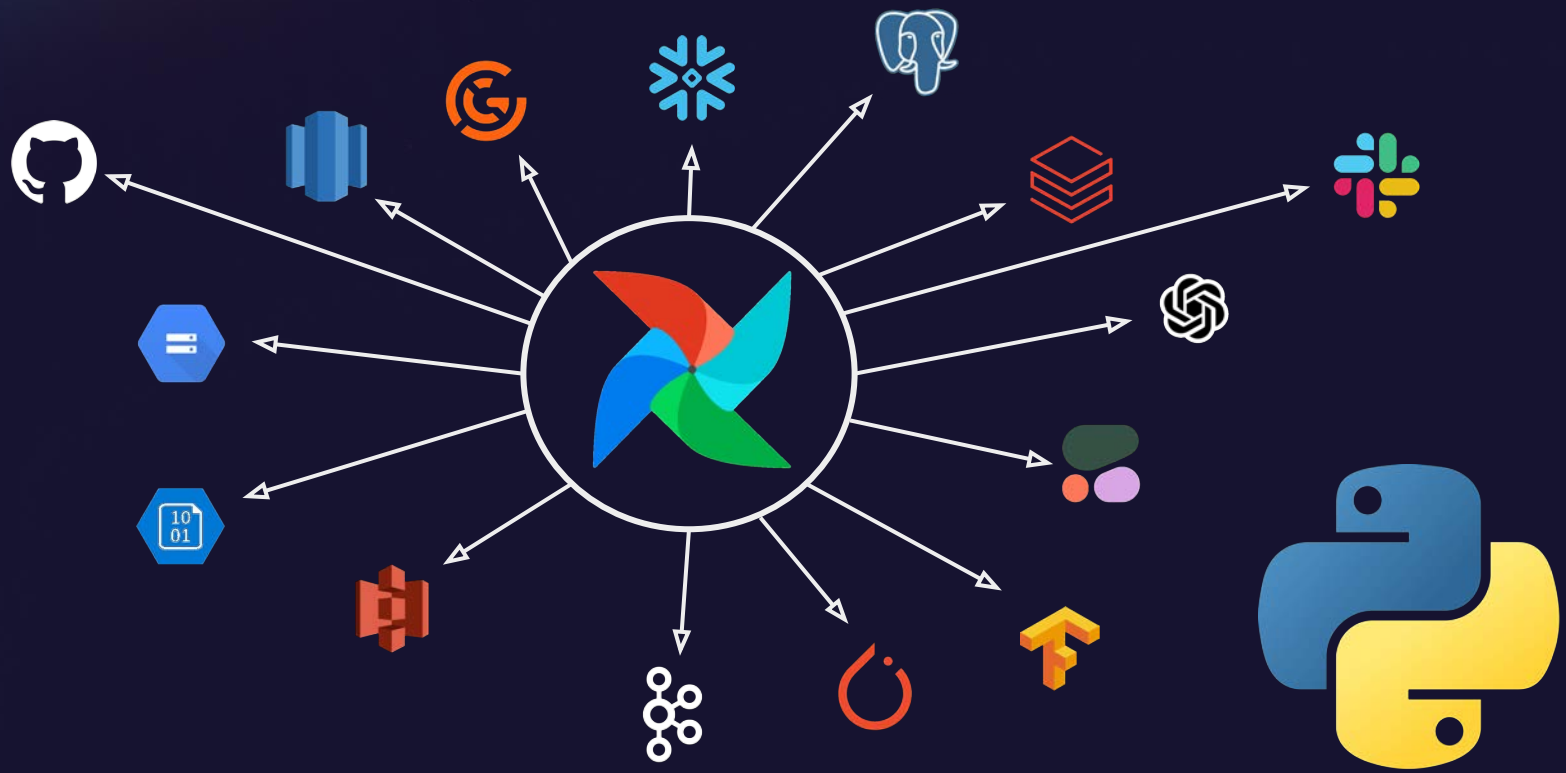
More and more people are using Airflow for ML/AI



28%

Of Airflow survey respondents in 2023 said they use Airflow for at least one ML/AI related use case.

Source: 2023 Apache Airflow Survey, n=797





ASTRONOMER

The driving force behind Apache Airflow
24x7 worldwide support | Worlds Top Airflow Experts

100%

Driving 100% of
Airflow releases

55%

Of Airflow code
contributed

18 of 25

18 of the top 25
committers on board,
8 PMC members

30K+

30K+ Airflow
students in Academy
ecosystem



Astro



Supercharge Airflow with Astronomer

Start Your 14-Day Free Trial (no credit card required)



New sign-ups receive \$300 in credits and a complimentary Airflow Fundamentals Certification Exam (normally \$150).

<https://qrco.de/bf2ICP>

Challenges when creating GenAI pipelines

The prototype works great - but production is a different beast

- API outages and rate limits
- Need to keep training data up to date
 - **Your data is what sets you apart from competitors!**
- Changing tools and APIs - new models coming out every day
- Complex pipeline structures
- Need the ability to determine which data went into training (compliance!)
- Scalability
- Reliability
- ...

Challenges when creating GenAI pipelines

The prototype works great - but production is a different beast

- API outages and rate limits -> Automatic retries
- Need to keep training data up to date -> Airflow already the standard
- Changing tools and APIs -> Airflow is tool agnostic, TaskFlow API
- Complex pipeline structures -> Datasets, dynamic task mapping, branching
- Need the ability to determine which data went into training (compliance!) -> Observability + OpenLineage integration
- Scalability -> Pluggable compute
- Reliability -> Battle tested + it is all code: CI/CD and DevOps best practices

Your data + your orchestration is what sets you apart from competitors!



Key Airflow features for GenAI

These features build a good foundation for best practice GenAI pipelines

- TaskFlow API
- Automatic retries
- Branching
- Deferrable operators
- Data-driven scheduling using Datasets
- Dynamic task mapping
- Alerts and notifications
- Setup and teardown tasks
- Backfills and reruns

TaskFlow API ⇒ Airflow decorators

The pythonic way to write Airflow DAGs

```
from airflow.operators.python import PythonOperator

def say_hi_func(name: str = "") -> str:
    return f"Hi {name}!"

say_hi_obj = PythonOperator(
    task_id="say_hi_2",
    python_callable=say_hi_func,
    op_args=["Astra"],
)
```



```
from airflow.decorators import task

@task
def say_hi_1(name: str = "") -> str:
    return f"Hi {name}!"

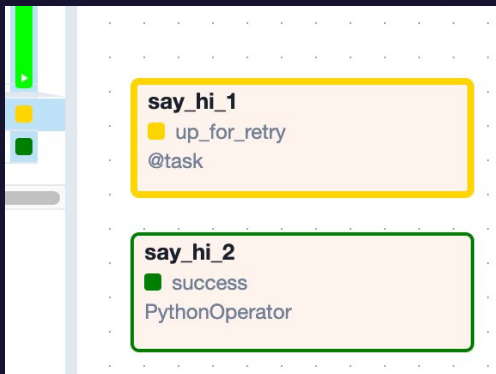
say_hi_obj = say_hi_1("Astra")
```

You can mix traditional operators and Airflow decorators!

There are many decorators: @dag, @task.kubernetes, @task.branch, @task.bash etc... see: <https://astronomer.io/docs/learn/airflow-decorators>

Automatic retries in Airflow

Protects pipelines against rate-limits and API failures



You can configure:

- Number of retries
- Delay between retries
- Exponential backoff
- Maximum delay

Ways to configure:

- Airflow config
- `default_args` in DAGs
- Individual tasks

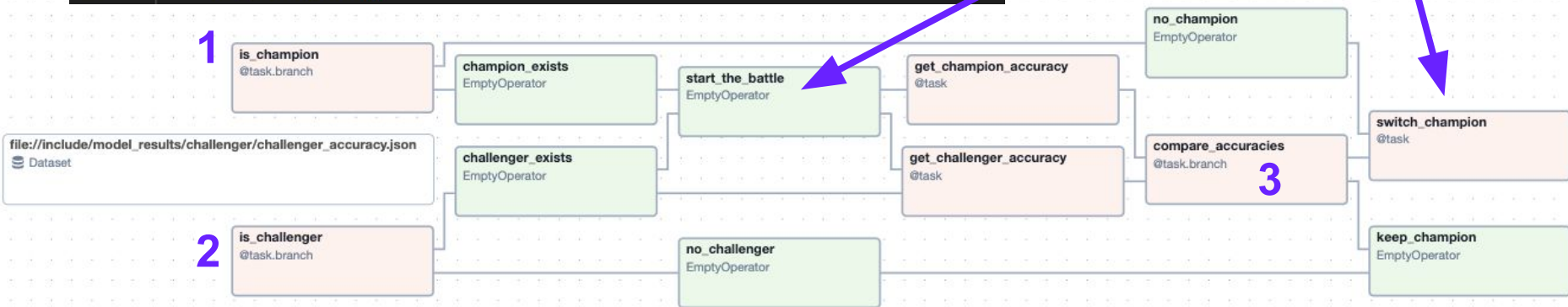
Best practice: Always set retries in production, unless a task has a reason not to

Details: <https://astronomer.io/docs/learn/rerunning-dags#automatically-retry-tasks>

Branching in Airflow

```
35 @task.branch
36 def is_champion() -> str:
37     is_champion = # logic to determine if champion exists
38     if is_champion:
39         return "champion_exists"
40     else:
41         return "no_champion"
42
43 is_champion()
```

Careful with
downstream trigger
rules!



Deferrable operators

- Deferrable operators can start async processes in the **Triggerer** component.
- Use case:
 - Waiting for a long running process to finish (e.g. model training)
 - Waiting for an event to occur in an external system (like a sensor)
- Advantage:
 - The worker slot is released = resource use optimization
- **Best practice:** Use deferrable operators whenever possible for longer tasks.



Details: <https://astronomer.io/docs/learn/deferrable-operators>



Dataset scheduling

The screenshot shows the Airflow web interface. The top navigation bar includes 'Airflow', 'DAGs', 'Cluster Activity', 'Datasets' (highlighted with a mouse cursor), 'Security', 'Browse', 'Admin', and 'Docs'. The main content area is titled 'Datasets' and includes a filter for updates in the past (All Time, 30 days, 7 days, 24 hours, 1 hour) and a search bar. Below the search bar is a table of datasets:

URI
snowflake://my_table/ Total Updates: 0
s3://bucket/data.csv Total Updates: 0

To the right of the dataset list is a DAG diagram showing a vertical flow of tasks:

```
graph TD; A[my_etl_dag] --> B[s3://bucket/data.csv]; B --> C[my_ml_dag]; C --> D[snowflake://my_table/];
```

DAG with producer task

Dataset

Consumer DAG + producer task

Next Dataset



Datasets in the Airflow 2.9 UI

Consumer DAG

The screenshot displays the Airflow 2.9 interface for a DAG run. The DAG is named 'my_ml_dag' and is in a 'Run' state, with the execution time being '2024-04-22, 14:10:22 UTC'. The interface includes tabs for 'Details', 'Graph', 'Gantt', 'Code', and 'Audit Log'. The 'Graph' tab is active, showing a task 'my_task' with a 'success' status. The task is connected to two datasets: 's3://bucket/data.csv' (highlighted with a purple box) and 'snowflake://my_table/' (highlighted with a green box). A duration bar on the left shows the task's execution time as 00:00:00.

Dataset

Producer task

Next Dataset

Advanced Dataset scheduling

Airflow 2.9 additions:

- Schedule on logical dataset expressions
 - Use AND (&) / OR (|) to create dataset logic
- Schedule on both time and datasets
 - `DatasetOrTimeSchedule` takes a `timetable` and a `dataset` argument
- REST API endpoint to update Datasets
 - Use for cross-deployment dependencies

Details: <https://astronomer.io/docs/learn/airflow-datasets>

Dynamic Task Mapping

- Create a variable number of copies of the same task based on input at runtime!
- Define parameters that stay the same (`.partial()`) and parameters that change in between task instances (`.expand()` / `.expand_kwargs()`)
- **Best practice:**
 - Use dynamic tasks when possible over dynamic DAGs
 - Customize the map index (Airflow 2.9)



Dynamic Task Mapping

Basic:

- `.partial(a=2)` → all parameters that stay the **same** for each mapped instance
- `.expand(b=[0,1])` → the parameter that changes as a **list**. Naming the kwarg is mandatory!
- `map_index_template` → customize the map index displayed in the **UI** (2.9)

Advanced:

- `.expand_kwargs({"a":1})` → map over sets of keyword arguments
- `.map(lambda x: x)` → transform the output of an upstream task before mapping over it

Details: <https://astronomer.io/docs/learn/dynamic-tasks>

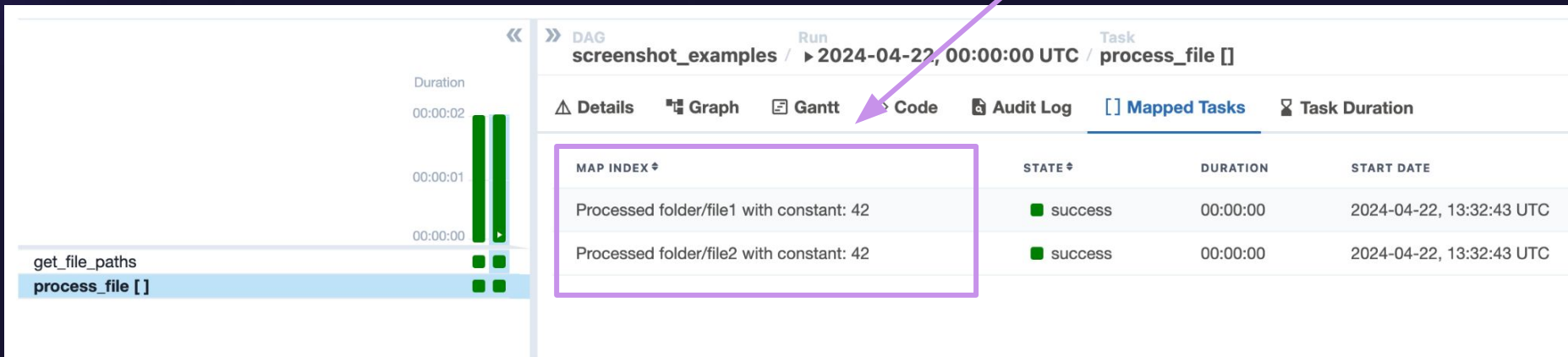
Dynamic Task Mapping - Simple example

```
44 @task
45 def get_file_paths() -> str:
46     # logic to get file paths. (potentially)
47     # results in different number of files each run
48     return ["folder/file1", "folder/file2"]
49
50 @task(map_index_template="{{ my_custom_map_index }}")
51 def process_file(constant: int, file: str) -> None:
52     # logic to process file
53
54     # create the custom map index
55     from airflow.operators.python import get_current_context
56
57     context = get_current_context()
58     context["my_custom_map_index"] = f"Processed {file} with constant: {constant}"
59
60 file_paths = get_file_paths()
61 processed_files = process_file.partial(constant=42).expand(file=file_paths)
62
```



Dynamic Task mapping custom index

Custom map index (2.9)

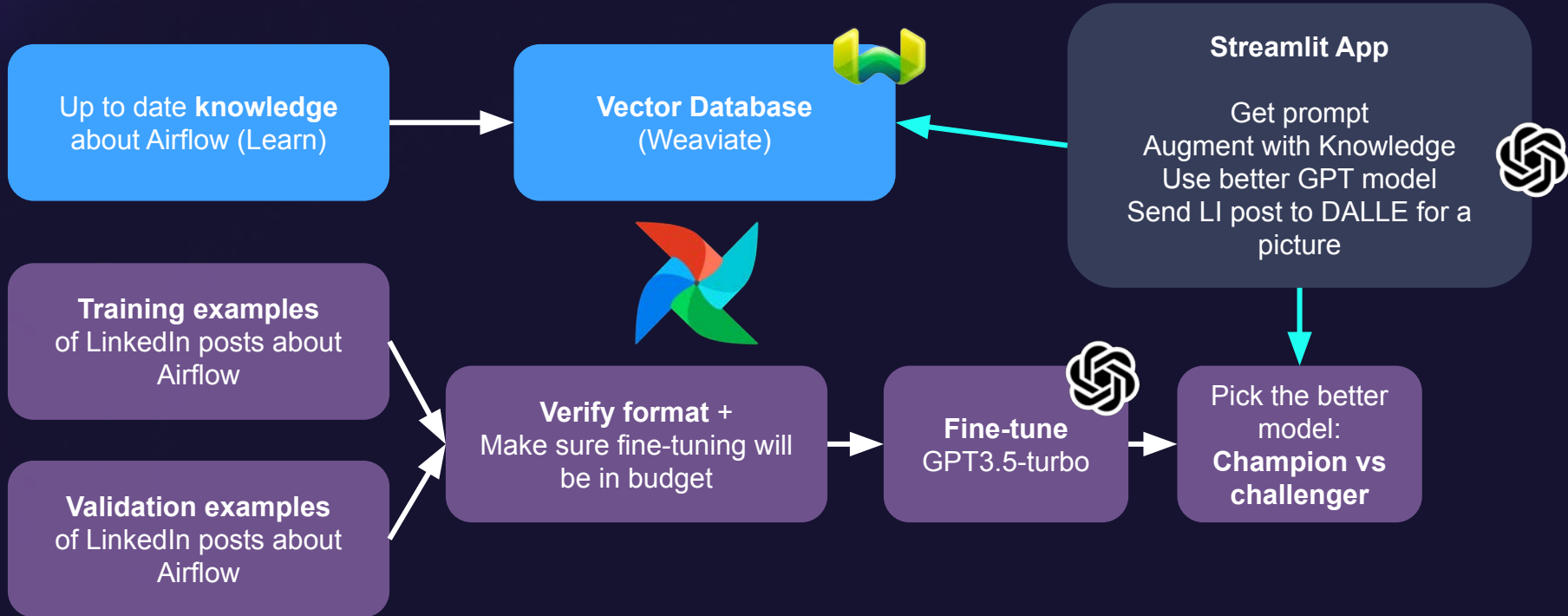


Details: <https://astronomer.io/docs/learn/dynamic-tasks>



Demo repository: Content Generation

<https://github.com/astronomer/gen-ai-fine-tune-rag-use-case>





Demo

<https://github.com/astronomer/gen-ai-fine-tune-rag-use-case>



Astro



Supercharge Airflow with Astronomer

Start Your 14-Day Free Trial
(no credit card required)



New sign-ups receive \$300 in credits *and* a complimentary Airflow Fundamentals Certification Exam (normally \$150).

<https://qrco.de/bf2ICP>



Take Home Message:
Your **data** + your **orchestration with Airflow** is
what sets you apart from competitors when
creating GenAI applications!



Appendix

Feature focus: Advanced Dataset scheduling (2.9)

Conditional Dataset Scheduling

```
5 @dag(  
6     start_date=datetime(2024, 3, 1),  
7     schedule=(  
8         (Dataset("dataset1") | Dataset("dataset2"))  
9         & (Dataset("dataset3") | Dataset("dataset4"))  
10    ), # Use () instead of [] to be able to use conditional dataset scheduling!  
11    catchup=False  
12 )  
13 def downstream2_one_in_each_group():
```

(Dataset 1 **OR** Dataset 2) **AND** (Dataset 3 **OR** Dataset 4)

Feature focus: Advanced Dataset scheduling (2.9)

Time + Dataset Scheduling

```
10 from airflow.timetables.datasets import DatasetOrTimeSchedule
11 from airflow.timetables.trigger import CronTriggerTimetable
```

```
55 @dag(
56     dag_display_name="📖 Ingest Knowledge Base",
57     start_date=datetime(2024, 4, 1),
58     schedule=DatasetOrTimeSchedule(
59         timetable=CronTriggerTimetable("0 0 * * *", timezone="UTC"),
60         datasets=reduce(
61             lambda x, y: Dataset(x) | Dataset(y), _KNOWLEDGE_BASE_DATASET_URIS
62         ),
63     ),
64     catchup=False,
```