

# GDPR-CCPA Risk Pipeline

A reproducible project that fetches, processes, and forecasts GDPR and CCPA policy updates on an hourly basis using Apache Airflow, Python, and Prophet.

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## Prerequisites

Before you begin, ensure you have the following installed on your machine:

- **Operating System:** macOS or Linux (instructions here target macOS).
- **Git:** to clone the repository.
- **Python 3.8+:** with `venv` module available.
- **pip:** Python package manager.
- **Airflow 2.7.x:** installed within a virtual environment.
- **Prophet:** for time-series forecasting.
- **Jupyter Notebook or VS Code:** for editing and running scripts.

Optional (for troubleshooting):

- **lsf, pkill, pgrep:** CLI tools to inspect and kill processes.
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## Installation & Troubleshooting Guide

Follow these steps carefully. Common pitfalls and their solutions are documented alongside each command.

### 1. Clone the repository

```
# Clone into your GitHub folder (avoid spaces in path if possible)
git clone https://github.com/yourusername/gdpr-ccpa-risk-pipeline.git
cd gdpr-ccpa-risk-pipeline
```

**Pitfall:** If `cd` fails, check your clone path or correct directory name.

### 2. Create and activate a Python virtual environment

```
python3 -m venv venv
source venv/bin/activate
```

- **Why:** Isolates dependencies.

- **Pitfall:** If you see `command not found: venv/bin/activate`, ensure you ran `python3 -m venv venv` successfully and you're in the project root.

### 3. Install Python dependencies

```
pip install --upgrade pip
pip install apache-airflow==2.7.1 prophet requests beautifulsoup4
```

- **Why:** Airflow core, forecasting, and data-fetch libraries.
- **Pitfall:** Installation errors often arise from missing compiler tools. On macOS, run `xcode-select --install` if Prophet fails to build.

### 4. Configure Airflow environment

#### a. Set `AIRFLOW_HOME`

Add to `~/.zshrc` (or `~/.bash_profile`):

```
export AIRFLOW_HOME="$HOME/path/to/gdpr-ccpa-risk-pipeline/airflow_home"
```

Reload:

```
source ~/.zshrc
```

- **Why:** Tells Airflow where to store its metadata and DAGs.
- **Pitfall:** Mismatched `AIRFLOW_HOME` leads to DAGs not appearing.

#### b. (Optional) Override `DAGS_FOLDER`

To keep DAGs in `project_root/dags`, also add:

```
export AIRFLOW__CORE__DAGS_FOLDER="$HOME/.../gdpr-ccpa-risk-pipeline/dags"
```

### 5. Initialize the Airflow database

```
airflow db init
```

- **Why:** Creates metadata tables.
- **Runs once:** Do **not** rerun every session.
- **Pitfall:** If prompted "Are you sure?", type `y`.

## 6. Create an Admin user

```
airflow users create \  
  --username admin \  
  --firstname Admin \  
  --lastname User \  
  --role Admin \  
  --email you@example.com \  
  --use-random-password
```

- **Why:** Enables UI login.
- **Pitfall:** Missing `--email` or `--role` flags cause errors.

## Step-by-Step Setup & Execution

This section walks through from code setup in VS Code to seeing your DAG in the UI.

### Step 1: Create & Configure Scripts in VS Code

1. **Open** the project in **Visual Studio Code**:

```
code .
```

1. **Under** the `scripts/` folder, create three files:
2. `fetch_policy_data.py`
3. `process_policy_data.py`
4. `forecast_policy_trends.py`
5. **In** `fetch_policy_data.py`, include:

```
# scripts/fetch_policy_data.py  
import os, requests  
from datetime import datetime  
  
def fetch_policy_data():  
    url = "https://www.dataprotectionreport.com/feed/" # RSS feed for policy  
    updates  
    resp = requests.get(url)  
    resp.raise_for_status()  
    fn = f"../data/raw/policy_updates_{datetime.utcnow():%Y%m%dT%H%M%S}.xml"
```

```

with open(os.path.join(os.path.dirname(__file__), fn), 'wb') as f:
    f.write(resp.content)
print(f"Fetchd and saved raw data to {fn}")

```

1. In `process_policy_data.py`, parse XML and write CSV:

```

# scripts/process_policy_data.py
import os, pandas as pd
from bs4 import BeautifulSoup

def process_policy_data():
    raw_dir = os.path.join(os.path.dirname(__file__), '../data/raw')
    # ... load latest XML, parse entries, extract title, pubDate, source,
    category ...
    # write to data/processed/cleaned_policies.csv
    print("Processed XML to CSV.")

```

1. In `forecast_policy_trends.py`, add forecasting logic (as detailed earlier).

**Why:** Separates each pipeline stage into its own reusable script.

## Step 2: Airflow DAG Definition

1. Under `dags/`, create `gdpr_ccpa_risk_pipeline.py`:

```

# dags/gdpr_ccpa_risk_pipeline.py
from airflow import DAG
from airflow.operators.python import PythonOperator
from datetime import datetime, timedelta

default_args = { ... }

def fetch_policy():
    from scripts.fetch_policy_data import fetch_policy_data
    fetch_policy_data()

def process_policy():
    from scripts.process_policy_data import process_policy_data
    process_policy_data()

def forecast_policy():
    from scripts.forecast_policy_trends import forecast_policy_trends
    forecast_policy_trends(periods=7)

with DAG(..., load_examples=False) as dag:
    t1 = PythonOperator(...)

```

```
t2 = PythonOperator(...)
t3 = PythonOperator(...)
t1 >> t2 >> t3
```

2. In your `airflow.cfg` (inside `$AIRFLOW_HOME`), set:

```
load_examples = False
```

**Why:** Prevents example DAGs cluttering your UI.

## Step 3: Start Airflow Services

### a. Webserver (UI)

```
# Terminal A
cd gdpr-ccpa-risk-pipeline
env/bin/activate
airflow webserver --port 8081
```

### b. Scheduler


```
# Terminal B
cd gdpr-ccpa-risk-pipeline
env/bin/activate
airflow scheduler
```

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## Step 4: Verify in the UI

1. Open <http://localhost:8081> and login.
2. Ensure **load\_examples=False** (no example DAGs).
3. **Refresh** to see only `gdpr_ccpa_risk_pipeline`.
4. Click its name → **Graph** tab → view three nodes:

```
fetch_policy_data → process_policy_data → forecast_policy_trends
```

1. **Trigger** () and watch each node run.
2. Confirm outputs:

```
ls data/processed/cleaned_policies.csv \ nls data/forecasts/forecast_*.csv
```

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## Next Steps

- **LLM Risk Scoring:** integrate T5 model to classify risk severity.
  - **Dashboarding:** connect forecasts and risk scores to Looker/Tableau/Power BI.
  - **Production Hardening:** migrate metadata DB to Postgres, switch to KubernetesExecutor or CeleryExecutor.
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