Readme

GDPR Risk Pipeline

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

Prerequisites

- Operating System: macOS or Linux
- **Git:** to clone the repository
- Python 3.8+: with venv module
- pip: Python package manager
- Apache Airflow 2.7.x
- Prophet: for time-series forecasting
- Jupyter Notebook or VS Code: for editing scripts
- Optional: Isof, pkill, pgrep for troubleshooting

Installation & Troubleshooting Guide

Follow these steps carefully. Pitfalls and solutions are noted.

1. 1. Clone the repository

git clone https://github.com/yourusername/gdpr-ccpa-risk-pipeline.git
cd gdpr-ccpa-risk-pipeline

- Pitfall: If cd fails, check your clone path.
- 2. 2. Create & activate venv

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

- Pitfall: Ensure venv exists and you're in project root.
- 3. 3. Install dependencies

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

- Pitfall: On macOS, run xcode-select --install if build fails.
- 4. 4. Configure Airflow

```
export AIRFLOW_HOME="$HOME/.../airflow_home"
export AIRFLOW__CORE__DAGS_FOLDER="$HOME/.../dags"
```

- Pitfall: Mismatched AIRFLOW_HOME leads to missing DAGs.
- 5. 5. Initialize Airflow DB

```
airflow db init
```

- Pitfall: Type y if prompted.
- 6. 6. Create Admin user

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

• Pitfall: Include --email and --role flags.

X Step-by-Step Setup & Execution

Scripts, DAG definition, and local execution flow.

7. 1. fetch_policy_data.py snippet

```
# scripts/fetch_policy_data.py
import os, requests
from datetime import datetime

def fetch_policy_data():
    url = "https://edpb.europa.eu/news/news_en"
    resp = requests.get(url)
    resp.raise_for_status()
    ts = datetime.utcnow().strftime("%Y%m%dT%H%M%SZ")
    fn = f"data/raw/edpb_news_{ts}.json"
    with open(fn, "w") as f:
        f.write(resp.text)
    print(f"Fetched raw data to {fn}")
```

- scripts/__init__.py package marker
- scripts/process_policy_data.py cleaning & aggregation
- scripts/validate policy data.py sanity checks
- scripts/forecast_policy_trends.py forecasting logic
- 8. 2. DAG definition snippet

```
# dags/gdpr_ccpa_risk_pipeline.py
from airflow import DAG
from airflow.operators.python import PythonOperator
```

```
from datetime import datetime
from scripts.fetch policy data import fetch policy data
from scripts.process policy data import process policy data
from scripts.validate policy data import validate policy data
from scripts.forecast policy trends import forecast policy trends
with DAG("qdpr ccpa risk pipeline", schedule interval="@hourly",
catchup=False) as dag:
    t1 = PythonOperator(task id="fetch",
python callable=fetch policy data)
    t2 = PythonOperator(task id="process",
python callable=process policy data)
   t3 = PythonOperator(task id="validate",
python callable=validate policy data)
   t4 = PythonOperator(task id="forecast", python callable=lambda:
forecast policy trends(periods=7))
   t1 >> t2 >> t3 >> t4
```

Download & Inspect Input/Output

- python scripts/process_policy_data.py
- Is data/processed/cleaned_policies.csv
- head data/processed/cleaned_policies.csv
- cat data/raw/edpb news <timestamp>.json
- python scripts/forecast_policy_trends.py
- Is data/forecasts
- head data/forecasts/forecast *.csv

▼ Forecast Validation Notebook

Use validate_forecast.ipynb to load raw, processed, and forecast data; plot actual vs. forecasted trends; check confidence intervals; and validate model assumptions.

Next Steps

- Integrate LLM (T5) to classify policy update severity.
- Build dashboards in Power BI/Tableau.