DATA INGESTION PIPELINE

TRANSFORMING AIR QUALITY DATA

AUTHOR: CARL BEBLI

PRODUCT OBJECTIVE



Objective

To build an end-to-end data ingestion pipeline that fetches, processes, and stores Air quality data for further analysis and reporting



Focus

Automating the data collection from
Air Quality API, processing it, and
storing it in a PostgreSQL database using
Docker and Airflow.

ROLE

Project Planning

Defined the project scope, objectives and timeline

Documentation

Created detailed
documentation and a
comprehensive README
file

Development

Implemented the data ingestion pipeline, developed Airflow DAGs, and configured Docker container

Testing and Debugging

Ensured data accuracy, reliability, and handled any technical issues.

PROJECT OVERVIEW

airflow: Contains Airflow DAGs and configuration files

data_fetch: Scripts for fetching air quality data from an API

database: Contains SQL scripts for database schema creation

visualize_air_quality_data.py: Python script for data visualization.

Various visualizations: Box plot, histogram, line plot, pie chart, and seaborn visualization

README.md: Documentation of the project.

TECHNICAL CHALLENGES

DOCKER CONFIGURATION



Setting up and configuring
Docker containers for Airflow
and PostgreSQL

ERROR HANDLING



Implementing robust error handling and logging within the data pipeline.

DATA
TRANSFORMATION



Implementing robust error handling and logging within the data pipeline.



SOLUTION APPROACH



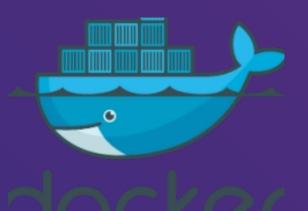
DATA INGESTION



ORCHESTRATION



CONTAINERIZATION



DATA STORAGE



Used Python to fetch weather data from Air Quality API Employed Apache Airflow to schedule and manage the ETL pipeline.

Utilized Docker to containerize the application, ensuring consistent environments.

Stored processed data in a
PostgreSQL database, ensuring
it's readily accessible for
analysis.

RESULT AND IMPACT



AUTOMATED PIPELINE



SCALABILITY



DATA
ACCESSIBILITY



EFFICIENCY

Successfully automated the data ingestion and processing pipeline

Built a scalable solution that can handle varying data loads

Provided easy access to processed weather data for analytics and reporting

Reduced manual intervention, allowing for continuous data updates and improved data accuracy

LESSONS LEARNED

· Docker Mastery: Gained indepth understanding of Docker and container orchestration.

* API Integration: Learned best practices for integrating and handling third-party APIs.

· Error Handling: Developed skills in implementing robust error handling and logging mechanisms.

* Workflow Management: Enhanced my ability to design and manage data workflows using Airflow.

ADAPTABILITY

1. Designed the pipeline to accommodate additional data sources with minimal changes

2. Built a solution that can scale horizontally to handle increased data volumes and complexity

TECHNICAL TOOLS

PYTHON



Used for data fetching and processing scripts



Managed and scheduled ETL workflows

DOCKER



Containerized the application for consistency and scalability

POSTGRESQL



Stored processed data for analysis and reporting.

AIR QUALITY API



Source of data

SUMMARY

Demonstrated strong abilities in handling complex data engineering challenges.

* Applied innovative solutions to create a robust and scalable data ingestion pipeline.

• Equipped to handle evolving data sources and project requirements.



Want to make a presentation like this one?

Start with a fully customizable template, create a beautiful deck in minutes, then easily share it with anyone.

Create a presentation (It's free)