

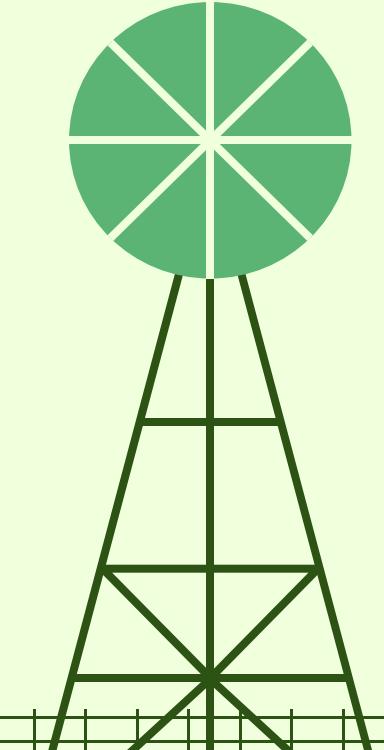


# SMART FARMING SYSTEM

Where Innovation Cultivates Growth



AGRISENSE



# Our team



**Yasmeen Abdallah**



**Yasmine Abdelgawad**



**Fareess Farrag**



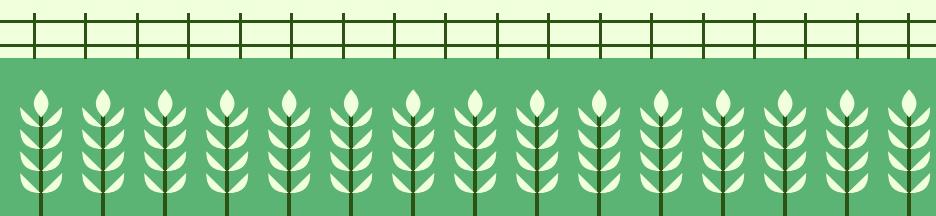
**Mohamed Sherif**

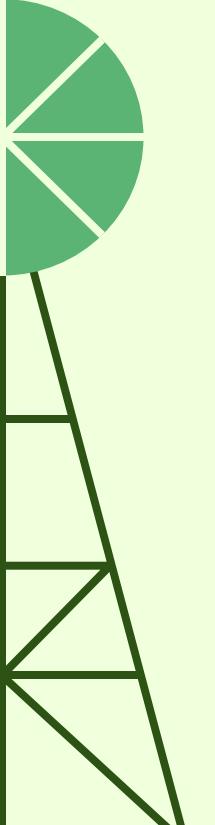


**Omar Mostafa**



**Omar Ahmed**





# Table of contents

**01**

Overview

**02**

Data Simulation &  
Real-Time  
Streaming

**03**

Setting up the  
pipeline

**04**

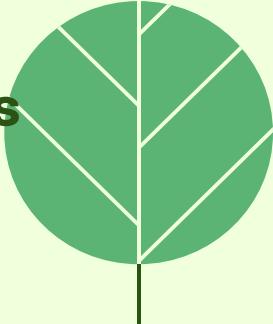
ETL

**05**

Analysis

**06**

ML models





01

# Overview

Problem statement and solution



# The problem

- “Farmers lose up to 50% of irrigation water due to poor monitoring and delayed decisions.”
- Farmers can’t monitor soil & climate continuously
- Irrigation decisions are delayed or inaccurate
- Data is scattered, not analyzed, not actionable



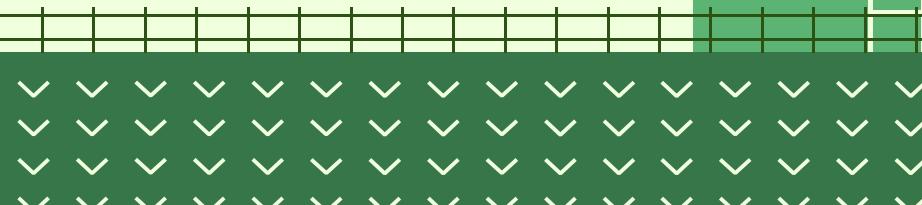
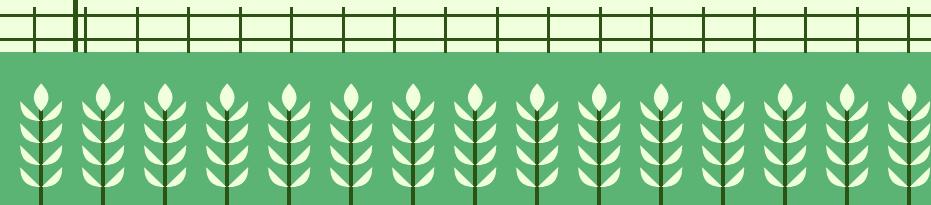
# Our solution

## AGRISENSE:

- provides a smarter, data-driven way to monitor farms
- Simulated farms using live data, ready for real production deployment.
- Gives real-time insights, alerts, and trend analysis
- Supports better irrigation and soil-health decisions



**AGRISENSE**

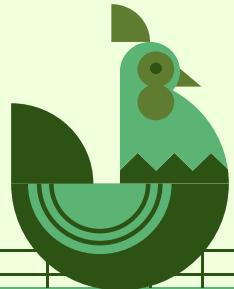




# Data Simulation & Real-Time Streaming

# 02

- Real farms normally use IoT sensors to capture soil & climate conditions.
- For this project, we created a simulation to test the system end-to-end.
- We use two data flows:
  - Historical Data: full 2024 simulation (minute-by-minute)
  - Live Stream: new reading generated every 3 seconds



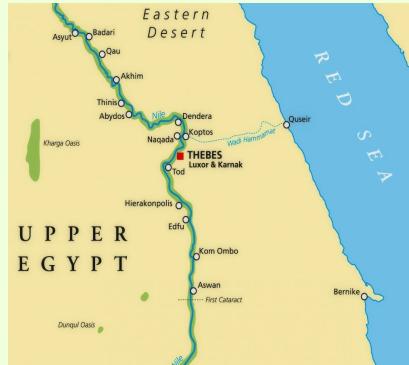
# Simulated Farms & Regions

# 10 farms created with different climate and crop conditions



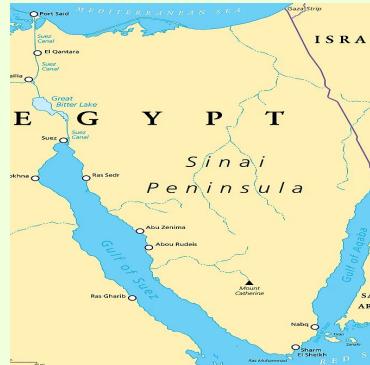
# Nile Delta

## humid & moderate climate



# Upper Egypt

## Hot & Dry



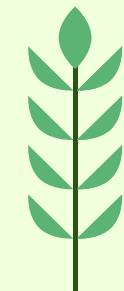
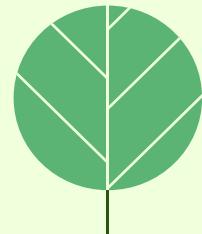
Sinai

# Cooler with Microclimate

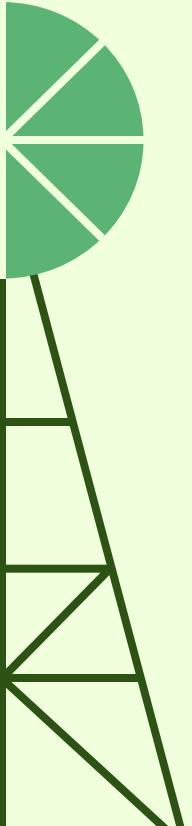


# Sensor Variables & How They Affect Each Other

Sunlight	Temp ↑	Moisture ↓	
Rain	Moisture ↑	Humidity ↑	pH ↓
Temp	Humidity ↓		
Moisture	pH ↑		



# Live Data Generation + Kafka Producer

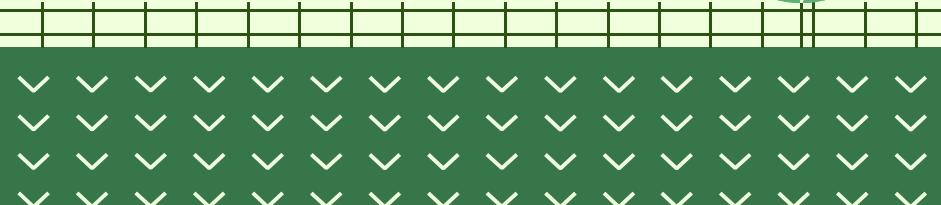
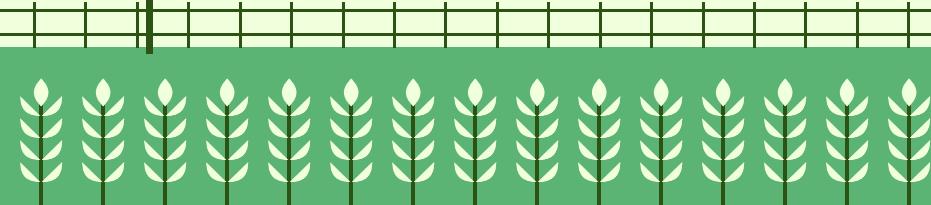
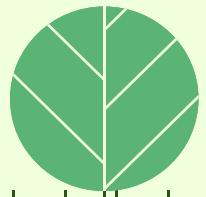
- 
- Every ~3 seconds, we generate a new sensor reading  
Calculated using:

- time of day
- farm region
- current weather patterns
- last moisture/pH state

1. Reading is turned into JSON
  2. Sent instantly to Kafka topic smart\_farming\_data
  3. Spark consumes the stream in real time
- 

# 03

## Setting up the pipeline



# End-to-End Data Pipeline

Our pipeline processes real-time agricultural sensor data through four key stages:

- **Data Ingestion** → Apache Kafka
- **Stream Processing** → PySpark Structured Streaming
- **Dual Storage** → PostgreSQL + HDFS
- **Analytics & ML** → Streamlit Dashboard + Predictive Models

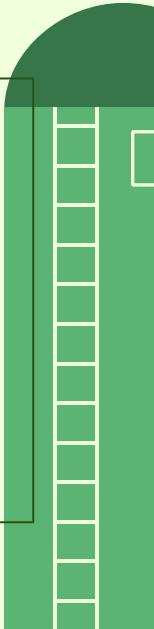


# Apache Kafka - Real-Time Data Ingestion

Distributed streaming platform for real-time data pipelines

## Our Implementation

- Topic: **smart\_farming\_data**
- Producer: Python script reads sensor data and publishes to Kafka Message
- Format: JSON with sensor readings (soil moisture, temperature, pH, rainfall, humidity, sunlight, pesticide usage)
- Frequency: Continuous streaming.

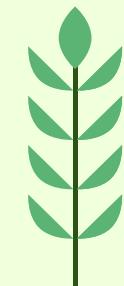


# PySpark - Connecting Kafka to Storage

Acts as the middle layer that reads from Kafka and writes to both PostgreSQL and HDFS simultaneously

## Our Implementation

- Consume from Kafka: Continuously reads messages from **smart\_farming\_data** topic
- Parse & Transform: Converts JSON to structured format
- Dual Write Operation: Sends processed data to two destinations in parallel:  
PostgreSQL → For real-time queries HDFS → For historical storage





# PostgreSQL - Real-Time Operational Database

Stores latest sensor readings for immediate access and real-time dashboards

	sensor_id [PK] uuid	timestamp timestamp with time zone	soil_moisture double precision	soil_ph double precision	temperature double precision	rainfall double precision	humidity double precision	sunlight_intensity double precision	pesticide_usage_ml double precision	farm_id character varying (50)	region character varying (100)	crop_type character varying (100)
1	b07270-3c4e-4b91-bf91-014ae3df4b7	2025-11-26 22:10:07.006285+00	40.09	6.43	20.43	0	55.03	0	0	farm_1	NileDelta	Wheat
2	11ee3ff9-a8cb-4b63-886c-9097a8c2d48a	2025-11-26 22:09:00.662812+00	40.06	6.36	22.36	0	51.54	0	0	farm_2	NileDelta	Rice
3	18235bee-d115-40f4-af65-d69cd82be18a	2025-11-26 22:10:09.681076+00	35.11	6.74	19.77	0	56.82	0	0	farm_3	NileDelta	Onion
4	f2942b4c-ab44-47e7-8d09-9c3cecc404ba	2025-11-26 22:10:10.71585+00	39.81	6.13	23.2	0	54.77	0	0	farm_4	UpperEgypt	Tomato
5	fc08fb5-d0f9-45a3-bdc4-a24ede2e5381	2025-11-26 22:10:11.763671+00	36.46	6.28	24.73	0	50.87	0	0	farm_5	UpperEgypt	Dates
6	1065f96-5531-41e1-ba26-b2e3dbdd96d	2025-11-26 22:10:12.78212+00	36.76	6.36	25.4	0	50.84	0	0	farm_6	UpperEgypt	Peanuts
7	3cb435d0-105c-49ec-8896-da49928e30...	2025-11-26 22:10:13.811801+00	30.96	6.67	23.39	0	51.78	0	0	farm_7	Sinai	Corn
8	51e44d30-6ee2-4474-8131-5fc9b19a1d...	2025-11-26 22:10:14.841818+00	37.74	6.3	19.16	0	56.14	0	0	farm_8	Sinai	Olive
9	fcd65ab-b08e4-4411-90a1-26e82512d8...	2025-11-26 22:10:15.851954+00	36.53	6.08	20.63	0	54.17	0	0	farm_9	Sinai	Barley
10	55c48152-9092-4bd0-9525-d9c1f595c7...	2025-11-26 22:10:16.863276+00	39.64	6.8	21.52	0	52.84	0	0	farm_10	Sinai	Potato
11	66fc7034-0927-4e37-b0b6-6f5c18cea424	2025-11-26 22:10:17.917938+00	40.03	6.44	18.93	0	57.59	0	0	farm_1	NileDelta	Wheat
12	4348160d-ebbd-44e8-add8-8ef3951f382	2025-11-26 22:10:18.927769+00	39.79	6.38	20.11	0	54.08	0	0	farm_2	NileDelta	Rice
13	bfe4d53b-a878-468d-a425-38b5035a3d...	2025-11-26 22:10:19.944158+00	35.21	6.73	18.16	0	56.84	0	0	farm_3	NileDelta	Onion
14	ece81923-b952-4fc0-8e04-ca84cf1ff5	2025-11-26 22:10:20.956343+00	39.66	6.13	20.54	0	57.06	0	0	farm_4	UpperEgypt	Tomato
15	5aedf060-e837-4e8e-a147-4ed152be87...	2025-11-26 22:10:21.977674+00	36.41	6.29	21.76	0	55.39	0	0	farm_5	UpperEgypt	Dates
16	e9c6e8e1-8f8f-4939-9909-2d84e227d333	2025-11-26 22:10:22.997787+00	36.66	6.36	22.13	0	50.45	0	0	farm_6	UpperEgypt	Peanuts
17	c6404e22-9d3-4f19-8fdb-a1878e461ed7	2025-11-26 22:10:24.0761+00	30.63	6.66	20.88	0	53.73	0	0	farm_7	Sinai	Corn
18	3a01428e-189d-4899-a55-108f4665t298	2025-11-26 22:10:25.015775+00	37.9	6.32	18.11	0	54.13	0	0	farm_8	Sinai	Olive
19	a7d1c495-6ace-449c-b55-37724f1436...	2025-11-26 22:10:26.033794+00	71.52	5.99	18.94	58.94	73.08	0	0	farm_9	Sinai	Barley
20	21682945-0859-41e2-bf86-d80972bdd1...	2025-11-26 22:10:27.057828+00	39.55	6.82	19.69	0	56.17	0	0	farm_10	Sinai	Potato
21	cf63b86-56ae-4dde-91e8-a318679214f	2025-11-26 22:10:28.144578+00	39.98	6.44	17.67	0	57.36	0	0	farm_1	NileDelta	Wheat





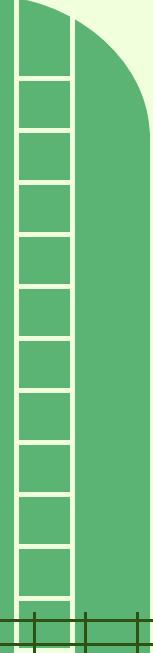
# Hadoop HDFS - Batch Data

Stores all historical sensor data for batch analytics, machine learning, and data warehousing

## Why HDFS?

- Scalable: Handles petabytes of data across distributed nodes
- Cost-Effective: Cheap storage for historical data
- Fault Tolerant: Data replicated across multiple nodes
- Analytics Ready: Optimized for large-scale batch processing

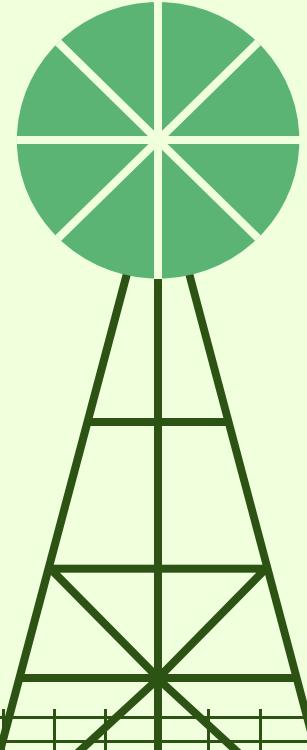




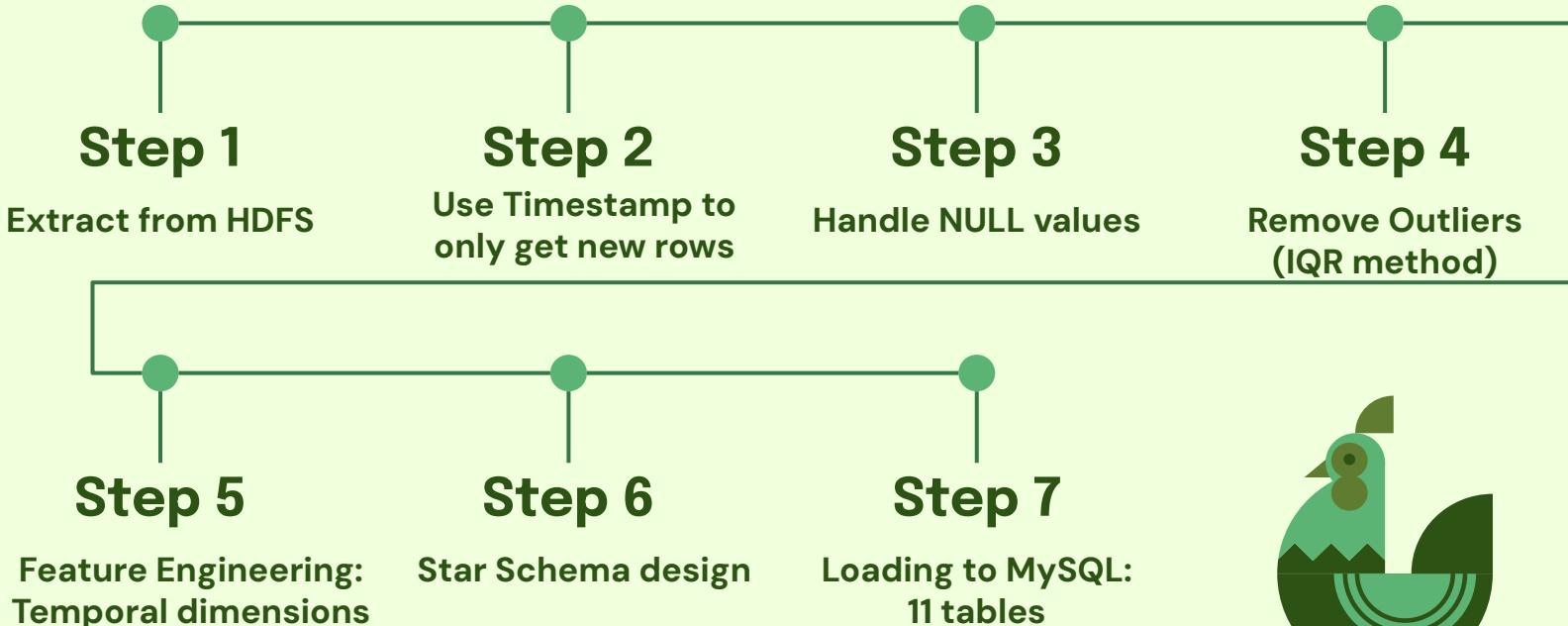
04

# Smart Farming ETL Pipeline

Incremental Data  
Processing with PySpark



# ETL Execution Flow



# Our ETL Logic with pyspark

## Incremental Loading:

- Stored last processed timestamp
- Only processed new rows using timestamp and checkpoints
- Prevented duplication
- Exits if no new data is found

## Missing Value Handling:

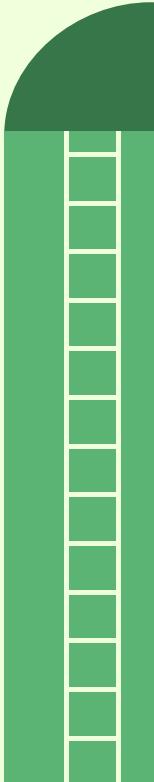
- Filled numeric NULLs using **mean imputation** for soil\_moisture, soil\_pH, temperature, humidity, sunlight\_intensity

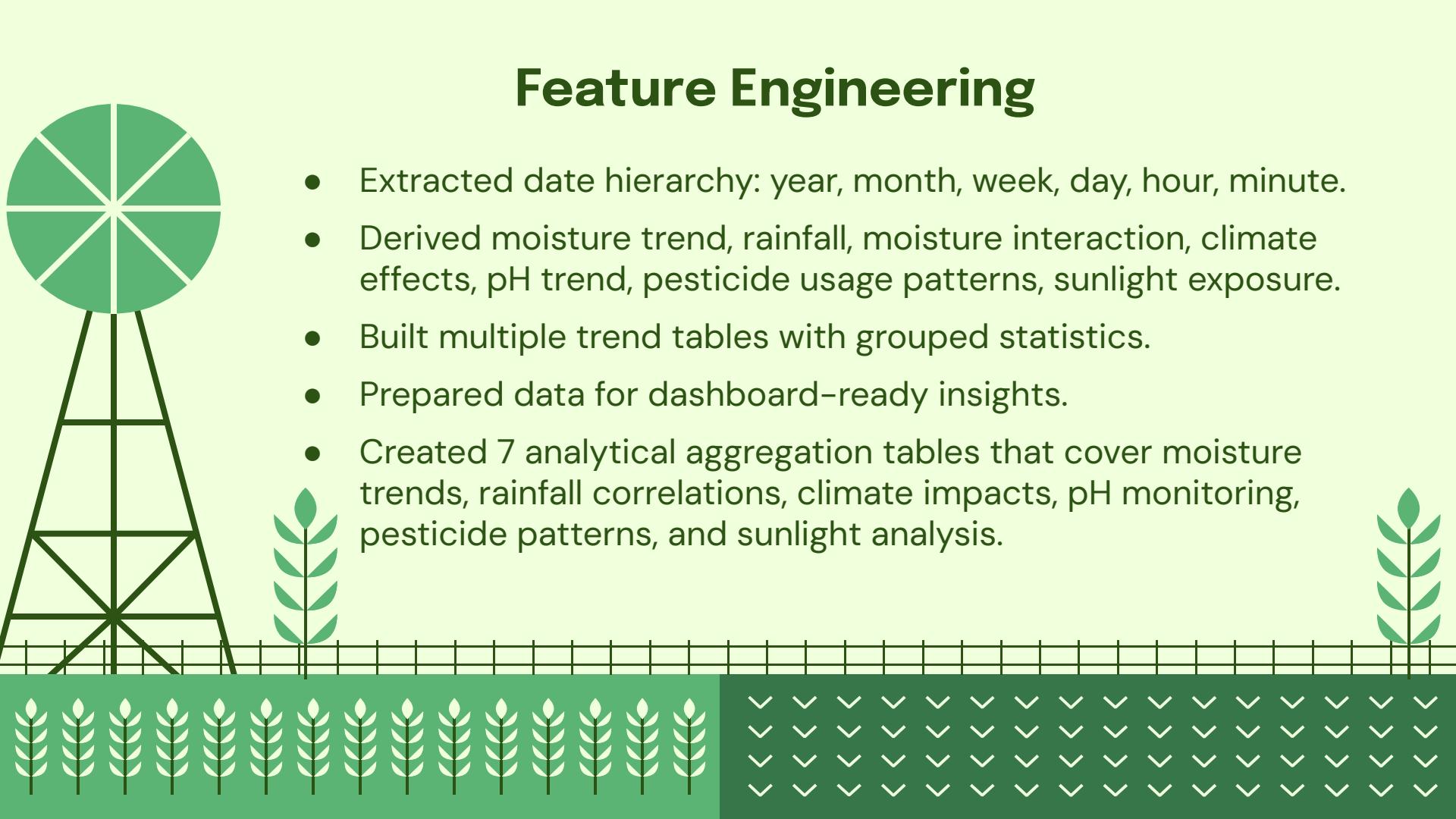
## Outlier Detection & Removal:

- IQR method on numerical columns
- Removed noisy/physically impossible sensor readings

## Text Cleaning:

- Trimmed whitespace in region & crop names





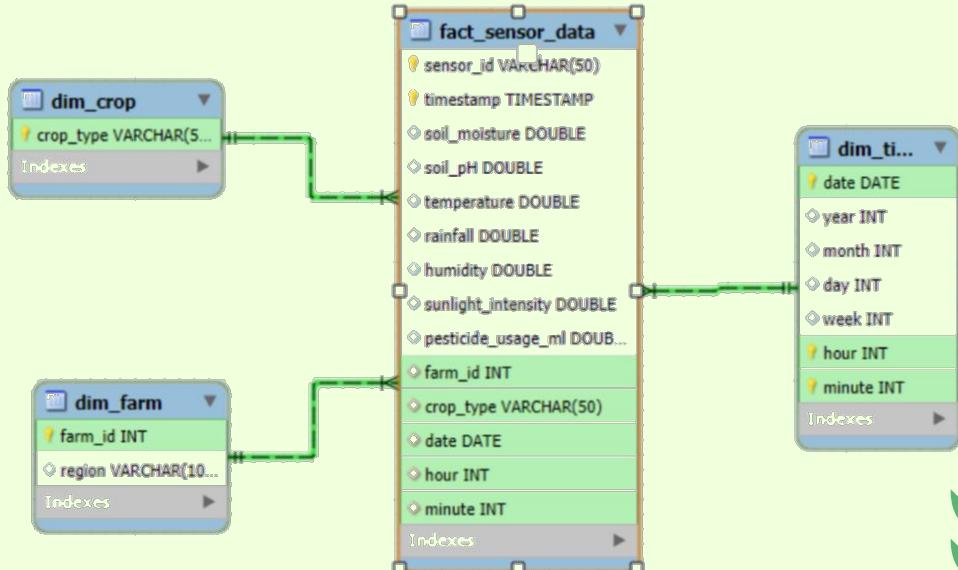
# Feature Engineering

- Extracted date hierarchy: year, month, week, day, hour, minute.
- Derived moisture trend, rainfall, moisture interaction, climate effects, pH trend, pesticide usage patterns, sunlight exposure.
- Built multiple trend tables with grouped statistics.
- Prepared data for dashboard-ready insights.
- Created 7 analytical aggregation tables that cover moisture trends, rainfall correlations, climate impacts, pH monitoring, pesticide patterns, and sunlight analysis.

# Star Schema Topology

## Data Warehouse Model

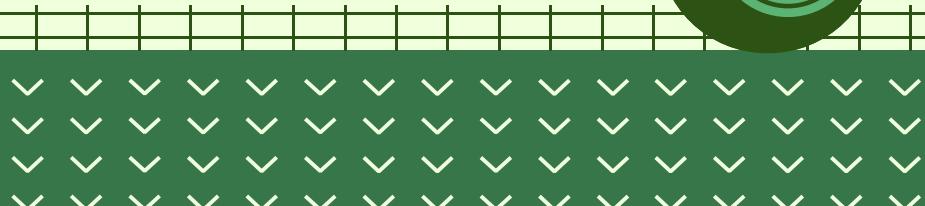
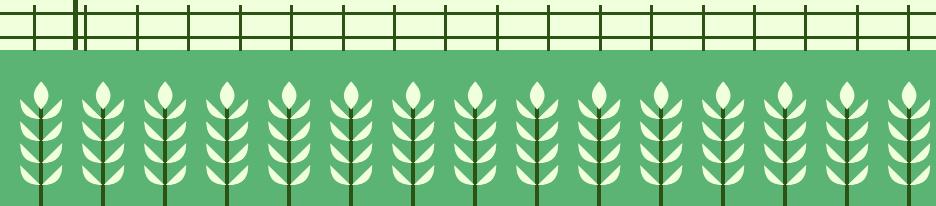
- 4 core tables:
  - 3 dimensions & 1 fact
- 7 analytical aggregations
- 11 total tables loaded incrementally to MySQL data warehouse with optimized column types and append mode



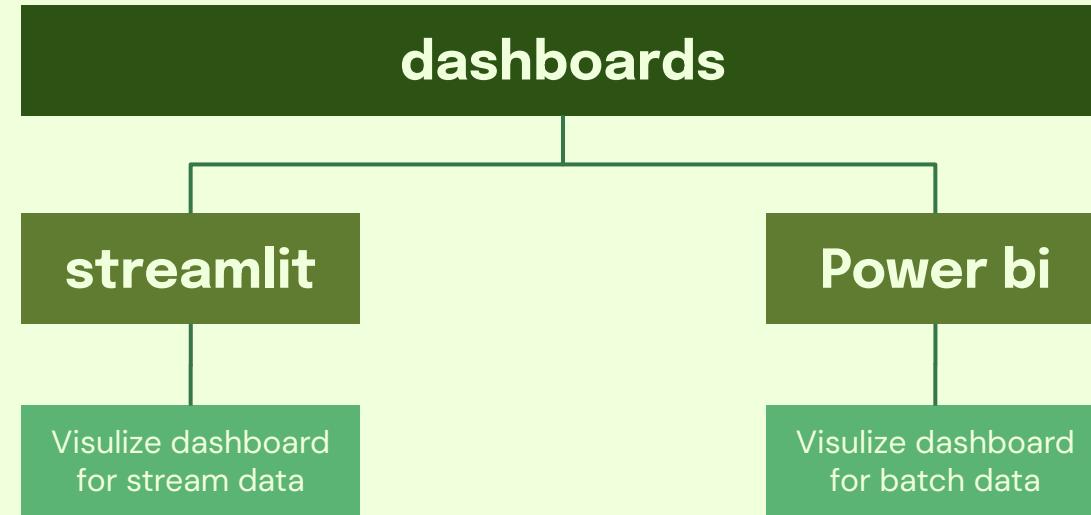
# 05

# Analysis

Batch and Streaming dashboards



# Analysis

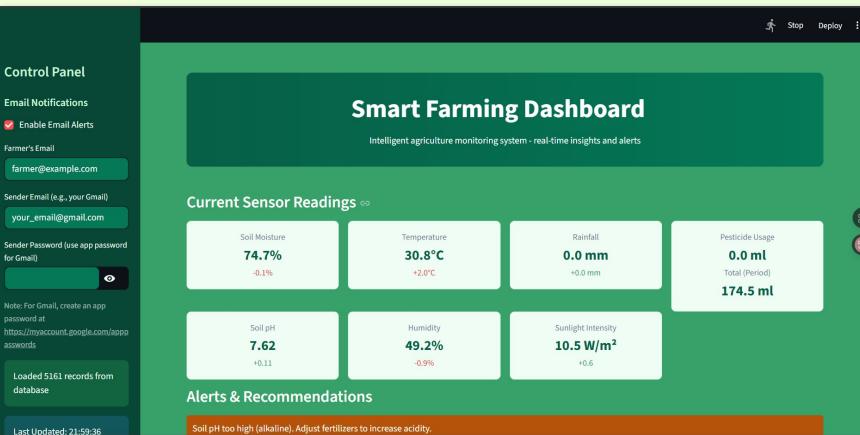


# Streamlit

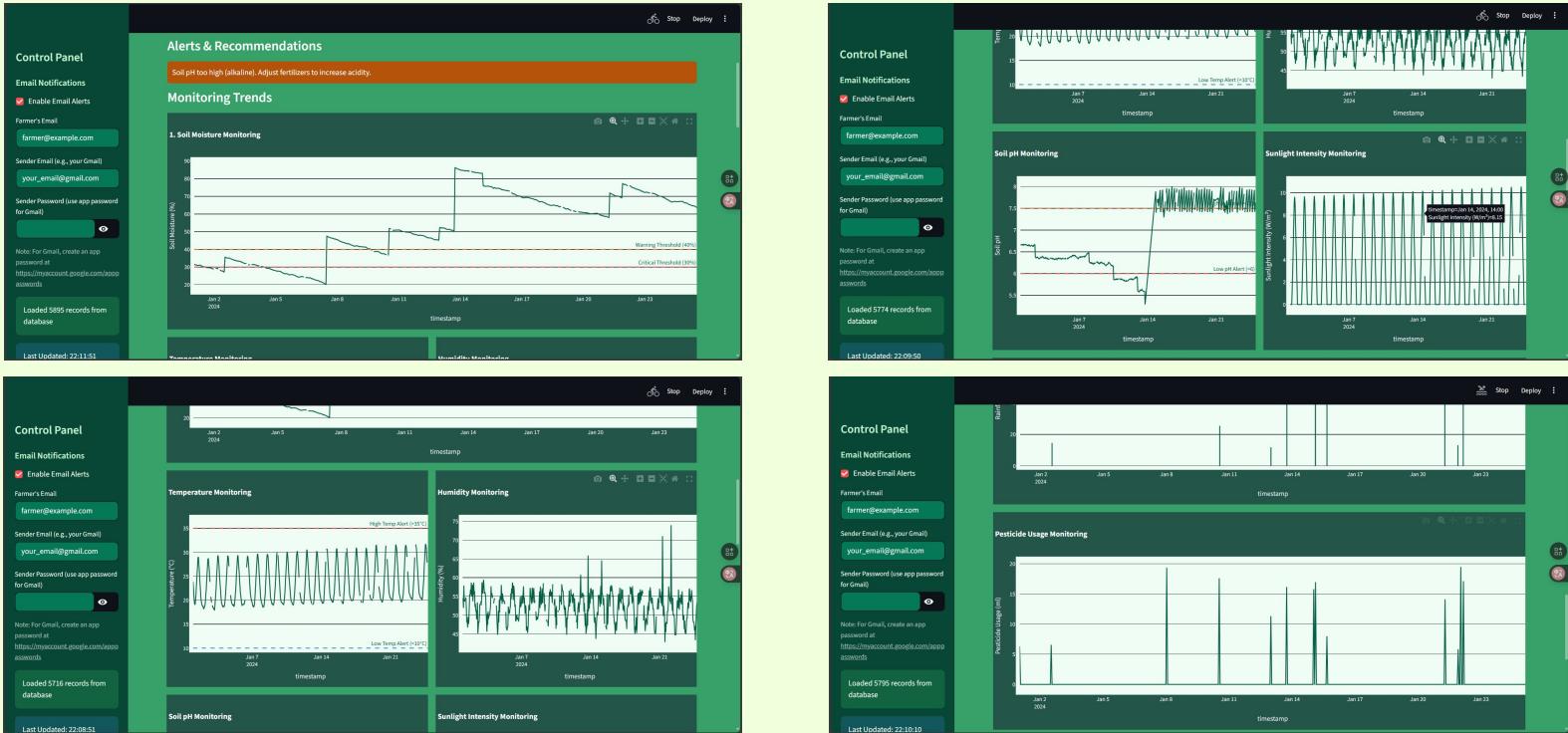
provides real-time monitoring, alerts, and interactive visualizations of sensor data.

## Dashboard Features

- Real-time sensor readings (soil moisture, temperature, pH, etc.)
- Interactive trend charts with threshold alerts
- Critical/Warning alerts based on sensor thresholds
- Email notifications for alerts (configurable in sidebar)
- Farm & Region filtering
- Historical data table for the last
- Auto-refresh every 1 second

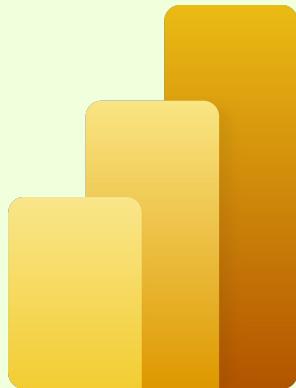


# Charts



# Power BI

Dashboard provides real-time insights into farm operations, soil health, pesticide & chemical usage, moisture levels, and climate trends for farms in Egypt (Nile Delta, Sinai, Upper Egypt, and surrounding regions).



# Overveiw

- Shows overall farm conditions: soil moisture, temperature, rainfall, pesticides.
- Map shows farm regions in Egypt (Nile Delta, Sinai, Upper Egypt).
- The main line chart shows the daily changes in rainfall, humidity, and temperature, giving insights into climate patterns and their effect on crops.
- Gauges show key ranges for sunlight, temperature, and moisture.
- The main line chart shows the daily



# Soil health

- This dashboard focuses on overall **soil health and fertility**.
- pH vs. moisture chart shows how soil conditions change.
- Compares moisture levels for each **crop type**.
- Donut chart shows moisture distribution across farms.



# Pesticide & Chemical Usage

- Displays pesticide usage by **farm and week**.
- Highlights total pesticide consumption.
- Shows trends between **rainfall and pesticide use**.
- Compares pesticide usage across **different crop types**.



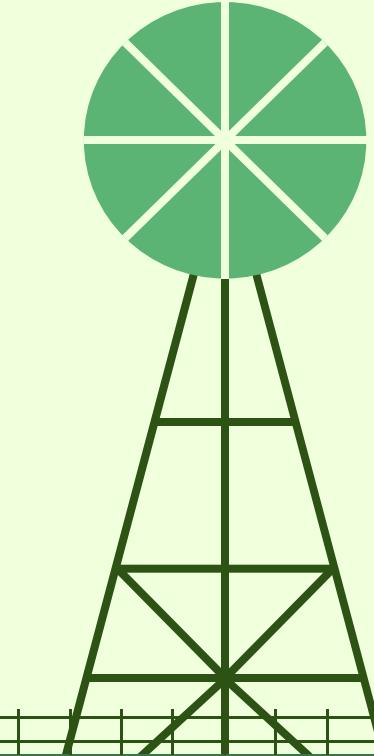
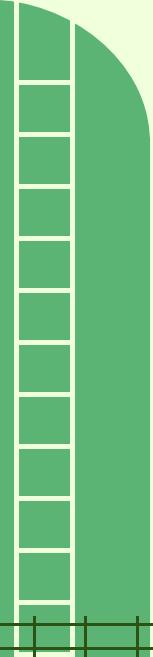
# Climate Trends Dashboard

- This dashboard shows overall climate patterns including **total rainfall, average humidity, and average temperature.**
- The gauges highlight the **minimum, maximum, and average** levels for both humidity and temperature.
- The line chart shows how **temperature, humidity, and rainfall change across weeks.**
- The bar chart displays the **rainfall distribution per week and month**, helping compare rainfall trends .



06

# ML Models

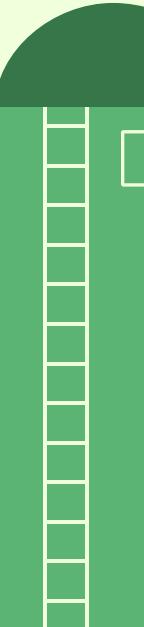


# AI-Powered Predictive Agriculture

## Model 1: Irrigation Prediction

An intelligent machine learning system that predicts irrigation needs for agricultural fields based on real-time environmental sensor data and crop characteristics.

- **Algorithm:** XGBoost Classifier
- **Accuracy:** 95%
- **Input:** Soil moisture, temperature, humidity, rainfall, sunlight, pH
- **Output:** Irrigate (1) or Not (0)
- **Purpose:** Optimize water usage and prevent over/under watering



# ML Training & Deployment Pipeline

01

## Deployment of Irrigation Prediction System

API: Flask

02



### Irrigation Prediction Result

#### Input Parameters:

**Soil Moisture:** 24.0%

**Temperature:** 34.0°C

**Humidity:** 30.0%

**Rainfall:** 0.0 mm

**Sunlight Intensity:** 900.0 W/m<sup>2</sup>

**Soil pH:** 6.2

**Irrigation Needed**

[Make Another Prediction](#)



### Irrigation Prediction System

Enter the environmental parameters to predict if irrigation is needed:

**Soil Moisture (%):**

24

**Temperature (°C):**

34

**Humidity (%):**

30

**Rainfall (mm):**

0

**Sunlight Intensity (W/m<sup>2</sup>):**

900

**Soil pH:**

6.2

**Predict Irrigation Need**

**Fill with Sample Data**



# AI-Powered Predictive Agriculture

## Model 2: Soil Health Index

A machine learning regression model that predicts soil health status based on environmental conditions, agricultural practices, and soil characteristics.

The model outputs a continuous Soil Health Index (0-1 scale).

- **Algorithm:** Random Forest Regressor
- **Accuracy:**  $R^2 = 0.9987$
- **Input:** Soil pH, moisture, climate, pesticide usage, crop type, region
- **Output:** Health score 0.0-1.0  
(🔴 Poor →🟡 Moderate →🟢 Healthy)



# Thank YOU

Special Thanks to Eng. Mohamed Hammed for his invaluable guidance, support, and mentorship throughout this project

