

A decorative pattern of green squares and rectangles of varying shades, arranged in a grid-like fashion, located on the right side of the slide.

# HarvestMatch

Mona Khosla, Tanvi Pabbathi, Siri Nellutla, Hugo Leung  
CIS 5500 Final Presentation

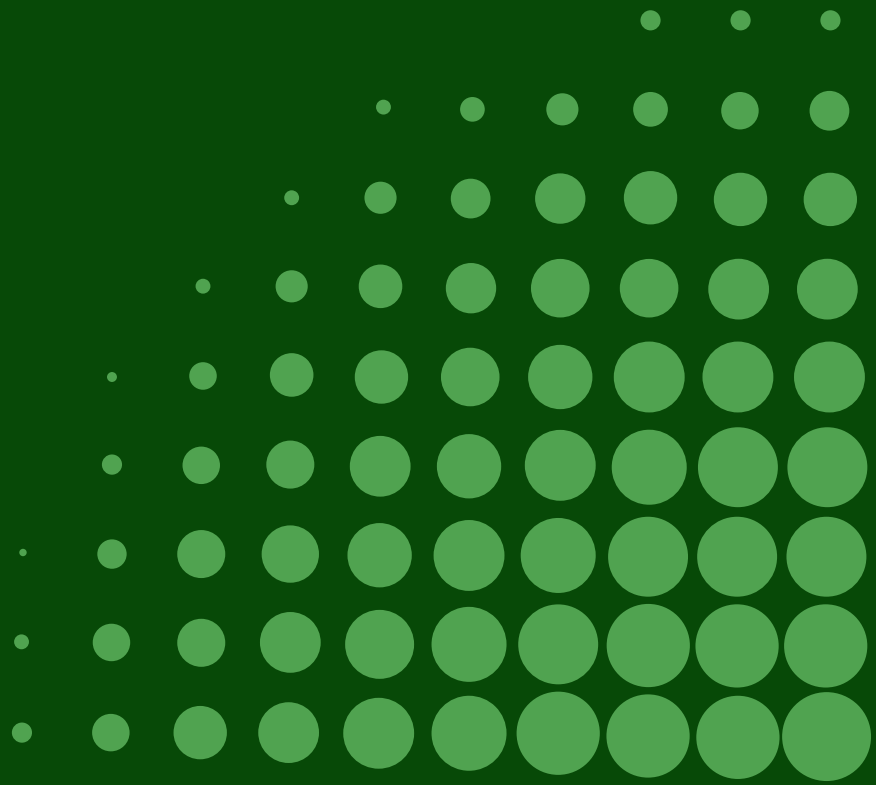
# Presentation overview

- Motivation
- Datasets
- Relation Schema
- Demo
- Queries + Performance
- Challenges



# Motivation

Farmers, gardeners, and agricultural planners often lack accessible, region-specific crop guidance that accounts for complex environmental variables.





What crops grow best in my state?

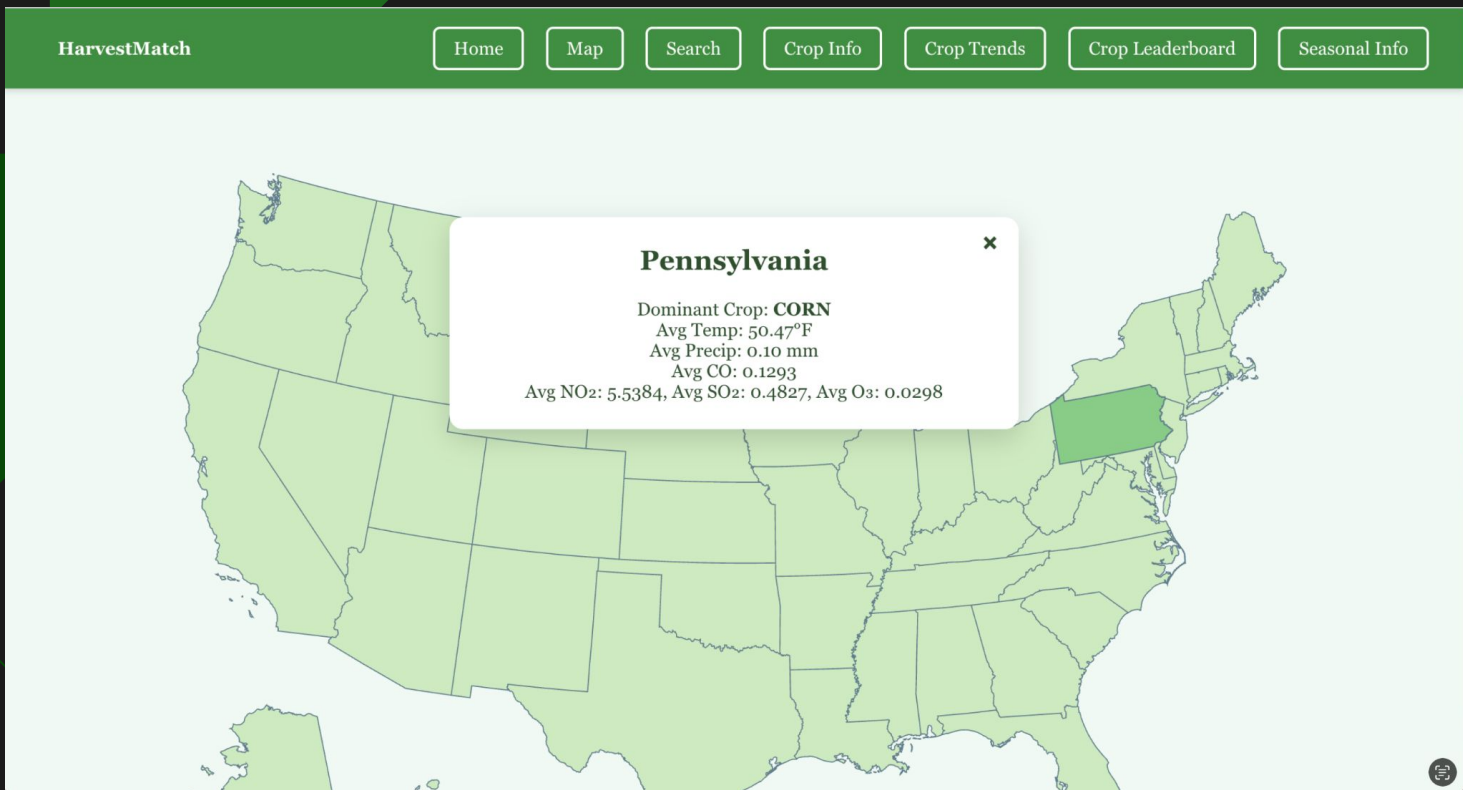
What crop is most resilient to climate change?

Which crops thrive under current pollution,  
temperature, or precipitation?

When should I plant a specific crop?

Questions you can ask  
HarvestMatch

# Snapshot



# Datasets



## Pollution dataset

Tracks daily air quality metrics ( $O_3$ ,  $CO$ ,  $SO_2$ ,  $NO_2$ ) by state and city. It helps identify how pollution levels affect crop performance and resilience.



## Crops dataset

Includes historical crop yields (kg/acre) by state, crop type, and month. This dataset enables comparison of crop productivity across time and regions.



## Temperature dataset

Captures monthly average temperatures per state with geographic coordinates. The temperature dataset supports climate-based crop suitability analysis.

## Weather dataset

Captures historical weather events across regions in the US. Each event includes location, severity, precipitation, and timing data.



We built a web app that empowers users to explore crop suitability using three interlinked datasets across climate, environment, and agricultural yield.

### Pollution Data Pre-processing

- Cleaned missing AQI and mean values
- Mapped pollution readings to seasons
- Dropped columns irrelevant to our application
- Removed null values

### Crop Data Pre-processing

- Normalized crop yield units to kg/acre
- Month to Season Conversion for ease of joins on other relations
- Removed null values

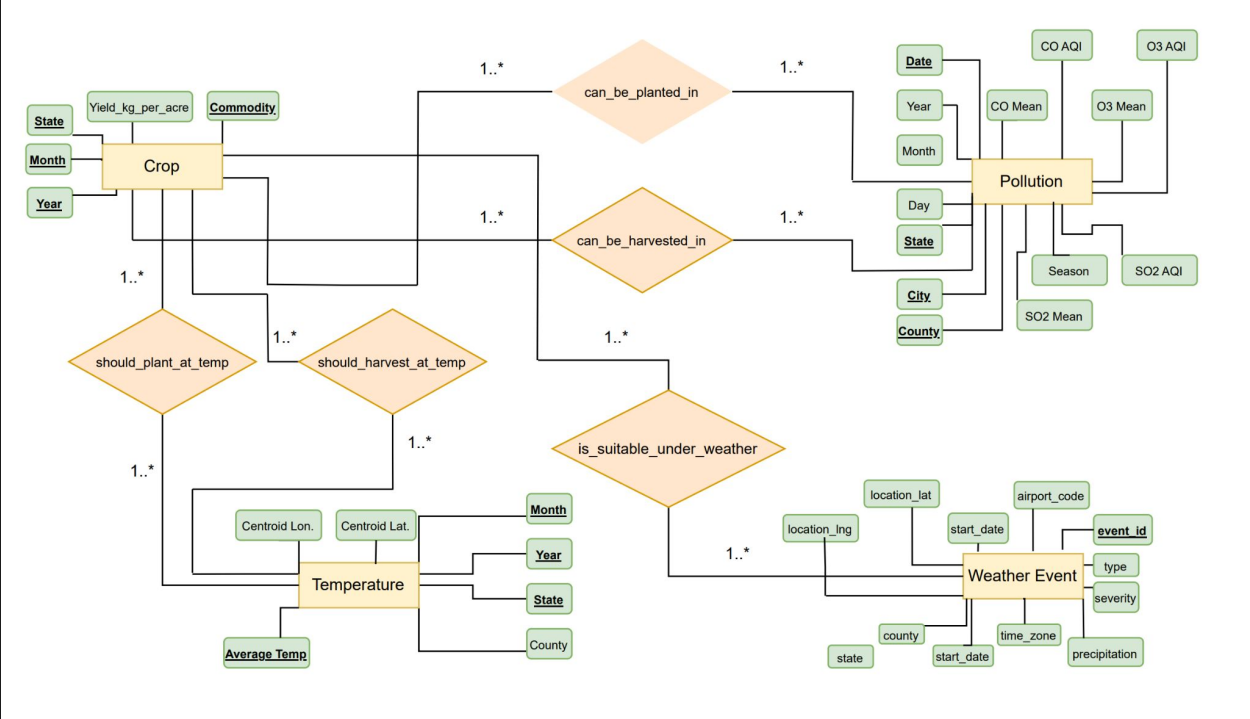
### Temperature Data Pre-processing

- Mapped months to their respective seasons in order to join on the variable with other relations
- Removed null values

### Weather Data Pre-processing

- Converted state abbreviations to names in order to perform joins on the other relations
- Assigned seasons based on month
- Dropped irrelevant columns
- Removed null values

# Entity Relationship Diagram (3NF)







Demo!

# Complex Query Example 1: Most to Least Climate Resilient Crops (35s)

```
WITH yearly_precip AS (  
  SELECT  
    EXTRACT(YEAR FROM start_date)::int AS year, UPPER(state) AS state, AVG(precipitation) AS avg_precip  
    FROM weather_events  
    WHERE start_date BETWEEN '2016-01-01' AND '2022-12-31'  
    GROUP BY EXTRACT(YEAR FROM start_date), UPPER(state)  
, crop_env AS (  
  SELECT  
    c.crop, c.yield_kg_per_acre, (p."CO Mean" + p."NO2 Mean" + p."SO2 Mean" + p."O3 Mean") AS pollution, t.average_temp, y.avg_precip  
    FROM crop_data c  
    JOIN pollution_data p ON c.year = p."Year" AND UPPER(c.state) = UPPER(p."State")  
    JOIN temperature_data t ON c.year = t.year AND UPPER(c.state) = UPPER(t.state)  
    JOIN yearly_precip y ON c.year = y.year AND UPPER(c.state) = y.state  
    WHERE c.year BETWEEN 2016 AND 2022  
, classified AS (  
  SELECT  
    crop, yield_kg_per_acre,  
    CASE  
      WHEN pollution < 15 OR pollution > 35 THEN 1 ELSE 0  
    END +  
    CASE  
      WHEN average_temp < 15 OR average_temp > 25 THEN 1 ELSE 0  
    END +  
    CASE  
      WHEN avg_precip < 400 OR avg_precip > 900 THEN 1 ELSE 0  
    END AS extreme_score  
    FROM crop_env  
, crop_resilience AS (  
  SELECT  
    crop, AVG(yield_kg_per_acre) FILTER (WHERE extreme_score >= 2) AS avg_yield_in_extremes  
    FROM classified  
    GROUP BY crop  
    HAVING COUNT(*) FILTER (WHERE extreme_score >= 2) > 1  
) SELECT  
  crop,  
  ROUND(avg_yield_in_extremes::numeric, 2) AS avg_yield_in_extremes  
  FROM crop_resilience  
  ORDER BY avg_yield_in_extremes DESC;
```

## Complex Query Example 2: Avg Crop Yield Based on Avg Pollution, Precipitation, and Temperature (27s)

```
WITH crop_yearly AS (  
  SELECT year,UPPER(state) AS state, AVG(yield_kg_per_acre) AS avg_yield  
  FROM crop_data  
  WHERE year BETWEEN 2016 AND 2021  
  GROUP BY year, UPPER(state)  
) , pollution_yearly AS (  
  SELECT  
    "Year" AS year, UPPER("State") AS state, AVG("CO Mean") AS avg_co, AVG("NO2 Mean") AS avg_no2, AVG("SO2 Mean") AS avg_so2, AVG("O3 Mean") AS avg_o3  
  FROM pollution_data  
  WHERE "Year" BETWEEN 2016 AND 2021  
  GROUP BY "Year", UPPER("State")  
) , precip_yearly AS (  
  SELECT  
    EXTRACT(YEAR FROM start_date)::int AS year, UPPER(state) AS state, AVG(precipitation) AS avg_precipitation  
  FROM weather_events  
  WHERE EXTRACT(YEAR FROM start_date)::int BETWEEN 2016 AND 2021  
  GROUP BY EXTRACT(YEAR FROM start_date), UPPER(state)  
) , temperature_yearly AS (  
  SELECT year, UPPER(state) AS state, AVG(average_temp) AS avg_temp  
  FROM temperature_data  
  WHERE year BETWEEN 2016 AND 2021  
  GROUP BY year, UPPER(state)  
)SELECT c.year, c.state, ROUND(c.avg_yield::numeric, 2) AS avg_yield, ROUND(p.avg_co::numeric, 4) AS avg_co, ROUND(p.avg_no2::numeric, 4) AS avg_no2,  
  ROUND(p.avg_so2::numeric, 4) AS avg_so2, ROUND(p.avg_o3::numeric, 4) AS avg_o3, ROUND(w.avg_precipitation::numeric, 2) AS avg_precipitation, ROUND(t.avg_temp::numeric, 2)  
  AS avg_temp  
FROM crop_yearly c  
LEFT JOIN pollution_yearly p ON c.year = p.year AND c.state = p.state  
LEFT JOIN precip_yearly w ON c.year = w.year AND c.state = w.state  
LEFT JOIN temperature_yearly t ON c.year = t.year AND c.state = t.state  
ORDER BY c.state, c.year;
```



# Performance

Query	Initial Execution Time	Optimized Execution Time
Get historical averages by state and season	12s 86 ms	107 ms
Get avg crop yield based on avg pollution, precipitation, and temperature	26s 593 ms	176 ms
Get best conditions to grow each crop	17s 643ms	170ms
Get a ranking from most to least climate resilient crops	36s 802 ms	321 ms
Best crop to plant based on precipitation	38s 657ms	570ms

# Technical Challenges

## Massive Datasets

Imported datasets with hundreds of thousands of rows, faced issues with schema design, indexing, and query performance, required data cleaning and normalization before ingestion

## Routing to Interactive Map

Managing React Router across multiple interactive components, ensuring state persistence and debugging navigation between map and search pages, Required careful UI-state and query coordination.

## Query Optimization

Even with optimized views, high join cardinality (~600K rows) from pollution data required aggregation to ~3K rows to achieve sub-1s query speed.



# Thank you!

# Image Citations

<https://www.nrdc.org/stories/air-pollution-everything-you-need-know>

<http://rodaleinstitute.org/why-organic/organic-farming-practices/crop-rotations/>

<https://www.google.com/url?sa=i&url=https%3A%2F%2Fwww.twinkl.nl%2Fteaching-wiki%2Fweather&psig=AOvVaw0ePDdZrXcYW9goqooepBPv&ust=1746884594623000&source=images&cd=vfe&opi=89978449&ved=0CBcQjhxqFwoTCMjX47DClo0DFQAAAAAdAAAAABAE>