

Air Pollution: A Global Crisis Demanding Action

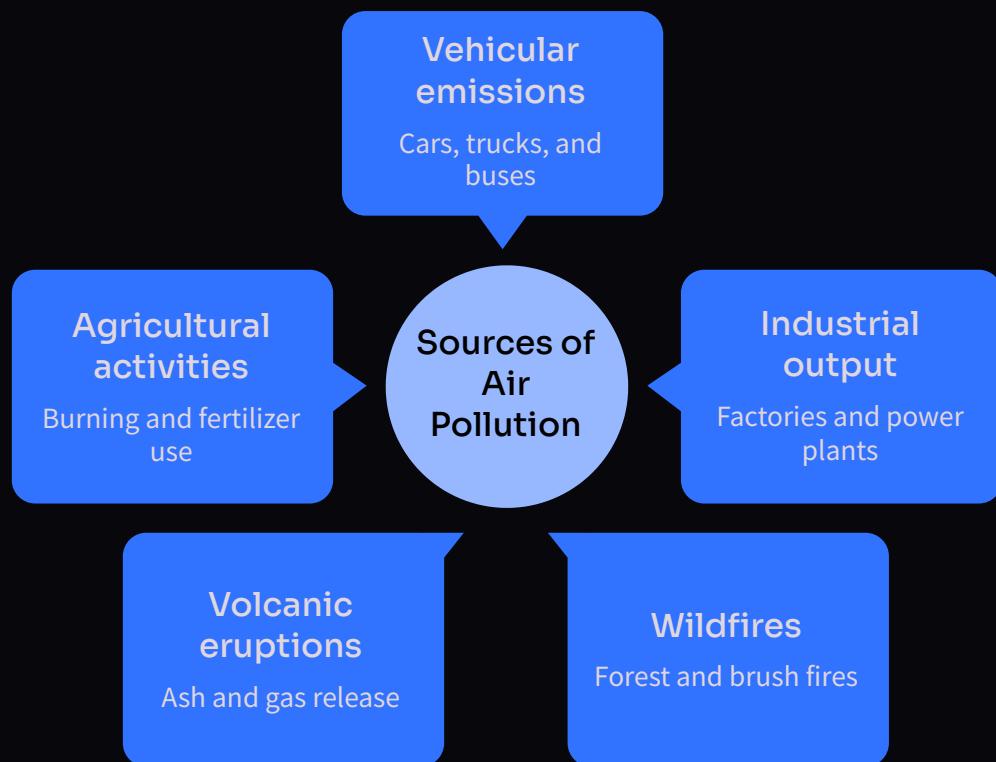
R by Rawan Ahmed

[Full project on GitHub](#)



What Is Air Pollution?

Air pollution is the introduction of harmful substances such as tiny particulates, toxic gases, and biological molecules into Earth's atmosphere.



The Deadly Scale: 8.1 Million Premature Deaths in 2021

8.1M

Lives Lost

Air pollution caused more deaths than HIV/AIDS and malaria combined in 2021.

700K+

Children Affected

Over 700,000 children under five died due to exposure to polluted air.

99%

Global Exposure

Virtually the entire global population breathes air exceeding WHO pollution limits.



Key Pollutants to Know

Particulate Matter (PM_{2.5} & PM₁₀)

Tiny particles that can penetrate deep into the lungs and bloodstream, causing severe health issues.

Nitrogen Oxides (NO_x) & Sulfur Dioxide (SO₂)

Primarily from fossil fuel combustion, these lead to smog formation and acid rain, harming ecosystems.

Carbon Monoxide (CO)

An odorless, toxic gas produced by incomplete combustion, highly dangerous in enclosed spaces.

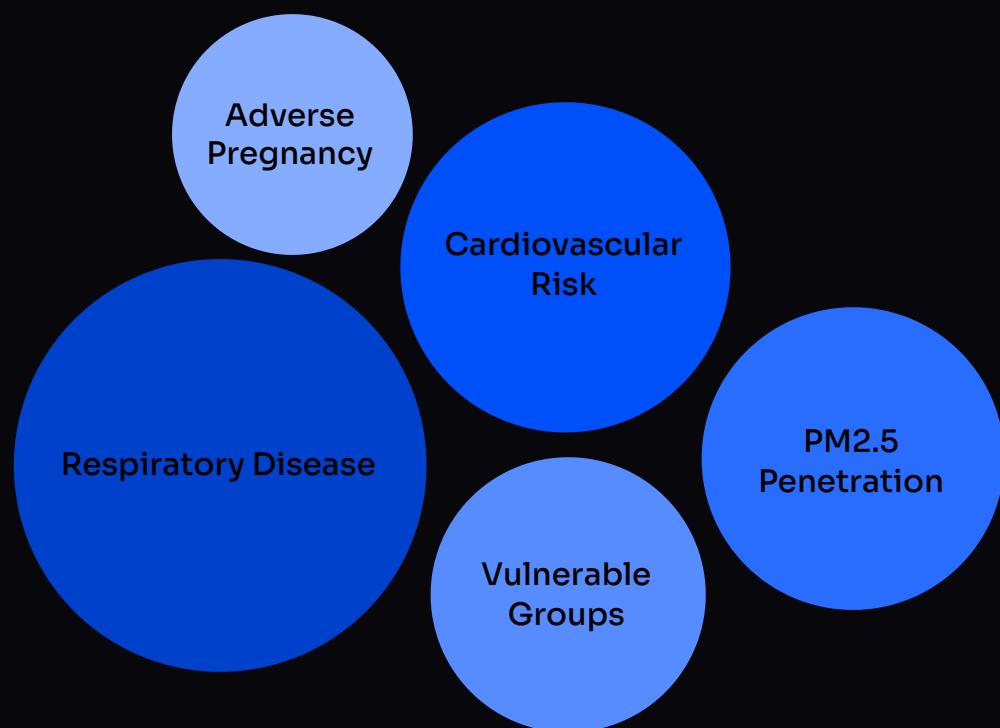
Volatile Organic Compounds (VOCs)

Carcinogenic chemicals emitted from industrial processes and common household products.

Health Impacts: More Than Just a Cough

Air pollution's effects extend far beyond respiratory issues. It is strongly linked to serious conditions such as pneumonia, asthma, lung cancer, and increases the risk of heart disease and stroke.

Fine particulate matter (**PM2.5**) is particularly dangerous as it can enter the bloodstream, affecting multiple organs and body systems. Children, the elderly, and individuals with pre-existing conditions are the most vulnerable to long-term damage, facing chronic illnesses and even premature death. It also contributes to adverse pregnancy outcomes like stillbirths.



Dataset overview:

Data Gathering By: Rawan Ahmed



Air Quality Dataset summary:

Source of Data: WHO (world health organization).

- This data set provides a snapshot of air quality metrics and demographic information.
 - Geographical Focus.
 - Time Reference.
 - Air Pollutant Concentration.
-



Health Dataset summary:

Source of Data: IHME (Global Burden of disease).

- This data set appears to contain global or multi-country health statistics.
- Primary Focus (Measure).
- Geographical Scope.
- Observed Values.

Data Cleaning:

Data Cleaning by: Rawan Ahmed and Ibrahim Abdelmohsen

AirQuality Data:

```
AirQuality.info()

✓ 0.0s

<class 'pandas.core.frame.DataFrame'>
Index: 40095 entries, 0 to 40097
Data columns (total 18 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   who_region       40095 non-null   object  
 1   iso3             40095 non-null   object  
 2   country_name     40095 non-null   object  
 3   city              40095 non-null   object  
 4   year              40095 non-null   int64  
 5   version           40095 non-null   object  
 6   pm10_concentration 40095 non-null   float64 
 7   pm25_concentration 40095 non-null   float64 
 8   no2_concentration 40095 non-null   float64 
 9   pm10_tempcov      40095 non-null   float64 
 10  pm25_tempcov      40095 non-null   float64 
 11  no2_tempcov       40095 non-null   float64 
 12  type_of_stations  40095 non-null   object  
 13  number_of_stations 40095 non-null   int64  
 14  population         40095 non-null   float64 
 15  population_source  40095 non-null   object  
 16  latitude            40095 non-null   float64 
 17  longitude           40095 non-null   float64 
dtypes: float64(9), int64(2), object(7)
memory usage: 5.8+ MB
```

```
AirQuality.isnull().sum()

58]
..   who_region          0
    iso3                 0
    country_name         0
    city                 0
    year                 0
    version               0
    pm10_concentration   0
    pm25_concentration   0
    no2_concentration    0
    pm10_tempcov         0
    pm25_tempcov         0
    no2_tempcov          0
    type_of_stations     0
    number_of_stations   0
    population            0
    population_source     0
    latitude               0
    longitude              0
dtype: int64
```

Data Cleaning:

Health Data

```
#detect duplicates:  
health_data.duplicated()
```

```
...      0  
0    False  
1    False  
2    False  
3    False  
4    False  
...  
9823  False  
9824  False  
9825  False  
9826  False  
9827  False  
9828 rows × 1 columns
```

dtype: bool

all the id columns are unnecessary because their names are existing in the data

```
health_data= health_data.drop(columns=[  
    'measure_id',  
    'location_id',  
    'sex_id',  
    'age_id',  
    'cause_id',  
    'metric_id'  
])
```

Change Names of some Columns to match other data sets:

```
health_data = health_data.rename(columns={  
    'location_name': 'country',  
    'measure_name': 'measure',  
    'cause_name': 'disease',  
    'val': 'value'  
})
```

SQL Analysis (Using SQL Server)

Analysis by: Ghalia Hamed and Mohamed Sayed

“After cleaning and preparing the dataset, we moved to the SQL analysis phase using SQL Server to answer key questions about air pollution trends and health impact.”

▪ Top 5 Worst Air-Quality Countries (Annually)

We identified the top 5 countries with the highest PM2.5 concentrations each year.

This helps highlight the most polluted regions annually and shows how rankings change over time.

```
--4)Top 5 worst air-quality countries annually
WITH aggregated AS (
    SELECT
        country_name,
        year,
        AVG(pm25_concentration) AS avg_pm25
    FROM AirQuality_cleaned2
    GROUP BY country_name, year
),
ranked AS (
    SELECT
        country_name,
        year,
        avg_pm25,
        DENSE_RANK() OVER (PARTITION BY year ORDER BY avg_pm25 DESC) AS pollution_rank
    FROM aggregated
)
SELECT *
FROM ranked
WHERE pollution_rank <= 5;
```

SQL Analysis (Using SQL Server)

- Did deaths rise or fall?

This SQL query calculates the **Year-over-Year (YoY) change in total deaths** for each country using a Common Table Expression (CTE) to aggregate the data first. The **LAG window function** is then applied to retrieve the total death figure from the preceding year for comparison.

```
--6)Did deaths rise or fall?
WITH deaths_agg AS (
    SELECT
        country,
        year,
        SUM(value) AS total_deaths
    FROM [cleaned_health_data-1]
    GROUP BY country, year
)
SELECT
    country,
    year,
    total_deaths,
    LAG(total_deaths) OVER (PARTITION BY country ORDER BY year) AS prev_year_deaths,
    total_deaths - LAG(total_deaths) OVER (PARTITION BY country ORDER BY year) AS yoy_change_deaths
FROM deaths_agg
ORDER BY country, year;
```

SQL Analysis (Using SQL Server)

▪Top 3 Pollution-Related Mortality Leaders (Per Region)

We identified the three countries with the highest pollution-related death rates in each WHO region. This helps highlight the main hotspots within every region and shows where the health impact of air pollution is most severe.

```
--12)Which age group has the highest average death count across all countries and years.  
SELECT  
    age,  
    AVG(value) AS avg_deaths,  
    SUM(value) AS total_deaths  
FROM [cleaned_health_data-1]  
GROUP BY age  
ORDER BY avg_deaths DESC;
```

SQL Analysis (Using SQL Server)

▪Which countries have the highest long-term pollution risk score?

This SQL query identifies countries with the **highest long-term pollution risk** by calculating a yearly risk score (PM2.5 concentration times population). It then averages these yearly scores to determine the overall **long-term risk** for each country and ranks them in descending order.

```
--15)Which countries have the highest long-term pollution risk score (PM2.5 * population).
WITH yearly AS (
    SELECT
        country_name,
        year,
        AVG(pm25_concentration) AS avg_pm25,
        AVG(population) AS population
    FROM AirQuality_cleaned2
    GROUP BY country_name, year
),
risk AS (
    SELECT
        country_name,
        year,
        (avg_pm25 * population) AS risk_score
    FROM yearly
)
SELECT
    country_name,
    AVG(risk_score) AS long_term_risk
FROM risk
GROUP BY country_name
ORDER BY long_term_risk DESC;
```

Dashboard & Visualization Insights (Using Power Bi)

Created By: Abdallah Ahmed

Dashboard Overview

After completing the data cleaning and SQL analysis, we built an interactive dashboard to visualize the key insights from the air-quality dataset.

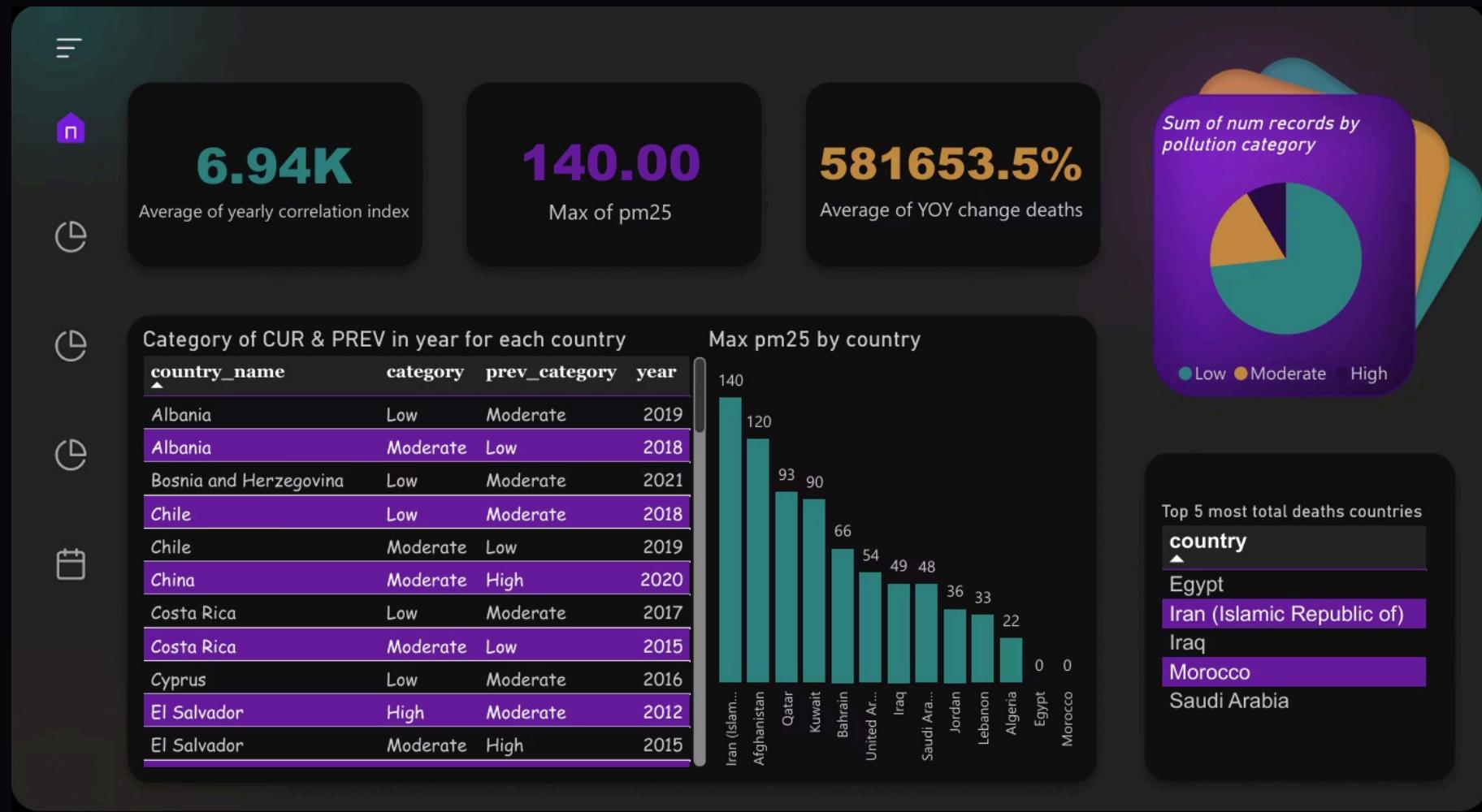
The dashboard helps us clearly understand pollution patterns, category distribution, and health impacts across countries.

Dashboard Purpose

The dashboard transforms raw numbers into easy-to-understand visuals, allowing us to:

- Identify high-risk countries
- Monitor pollution trends
- Detect category shifts over time
- Link pollution with health impact

Dashboard Details



Key Metrics Cards

Max PM2.5 by Country

Pollution Category Distribution

Top 5 Countries with Highest Total Deaths

Category Change per Country

Dashboard Details



Key Health and Environmental Metrics

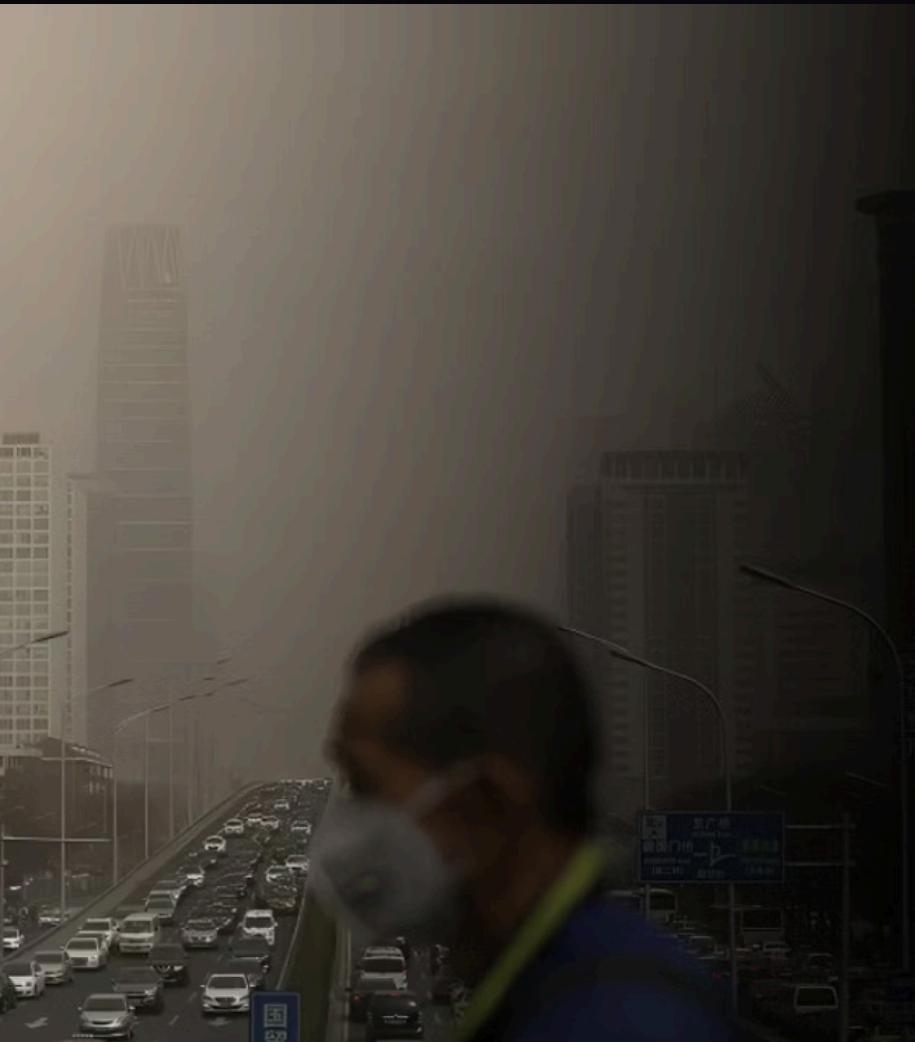
Country Contribution to Improvement

Mortality by Age Group

PM2.5 or Risk Index Trend

Focus on Environmental Risk

Dashboard Details

A large, semi-transparent background image shows a man wearing a dark face mask, walking through a hazy, smog-filled city street. In the background, there are several modern skyscrapers and a multi-lane highway with cars. The overall atmosphere is one of air pollution.

Air Pollution & Health Impact Analysis

This report presents the design and development of an interactive **Power BI dashboard** that analyzes global air pollution trends and their relationship with health outcomes. Using cleaned by **Python** datasets for PM_{2.5} concentration and mortality statistics, we built analytical **SQL** queries and visualizations that explore:

- Year-to-year changes in pollution for each country
- Long-term pollution trends using rolling averages
- Death statistics and trends across countries
- Correlation indicators between PM_{2.5} and mortality
- Country rankings for pollution and health impact
- Risk classifications and percentile distributions
- Population-weighted pollution risk assessment
- Identification of high-risk countries and improvement patterns

The dashboard offers a comprehensive, data-driven understanding of how air quality changes over time and how it relates to public health.

THANK YOU....:)

Course Title: Data Analyst Specialist
Instructor: Eng. Menna Selim

←

Key Findings



Air pollution levels vary significantly across countries and years



PM2.5 is the most critical pollutant and reaches extremely high values in certain regions



Several countries showed year-to-year improvement, while others experienced continuous decline.

3

The 3-year rolling trend helped reveal long-term patterns beyond annual fluctuations.



Pollution-related deaths remain highest in a small group of countries, emphasizing the health risk.

Tools Used:



Python

Versatile scripting
and data processing



Jupyter

Interactive
notebooks for
analysis



SQL Server

Reliable relational
database engine



Power BI

Visualize insights
and dashboards

