

General Sir John Kotelawala Defense University Faculty of Management, Social Sciences, and Humanities Department of Language

BSc in Applied Data Science Communication

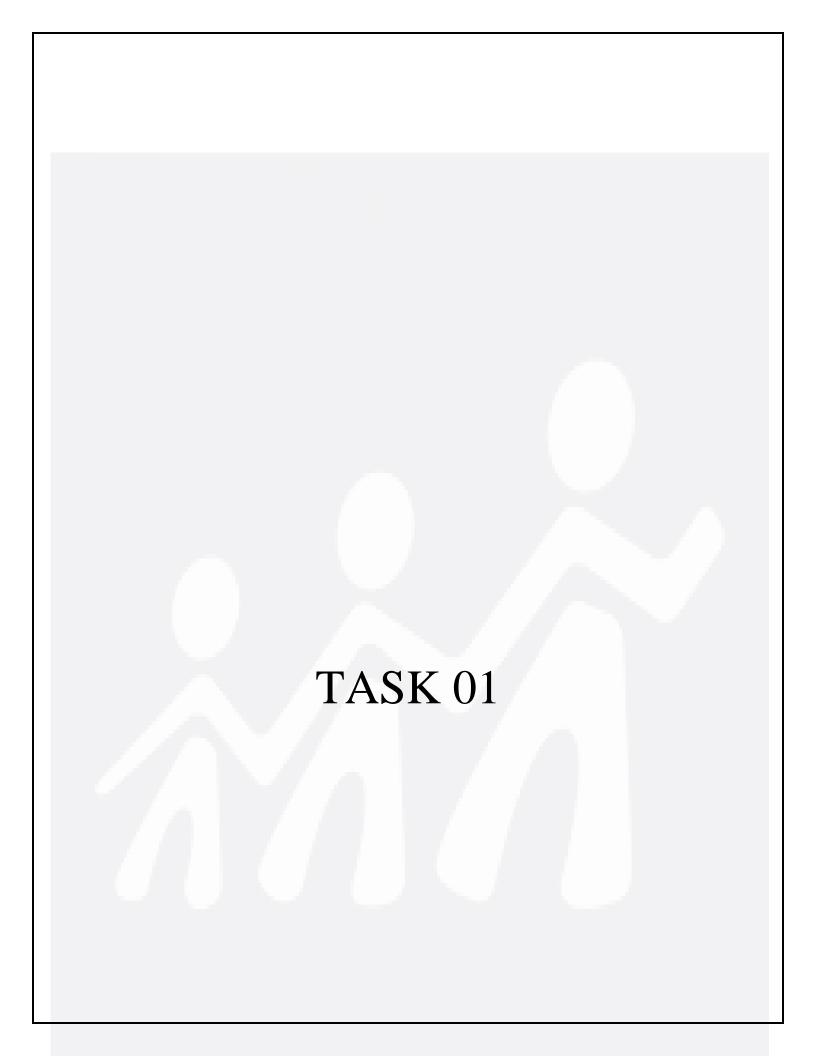
Advanced SQL and Cloud Databases
Year 2 Semester 2
Group Assignment 02
13.10.2024

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1.1 Introduction

A long-term initiative called Child Well-Being Monitor was created to assess and examine childhood poverty in lower-income nations. The project makes use of Power BI, SQL Server, and Report Builder to extract valuable insights about the well-being of children from data and present it visually. We highlight the Young Lives study, a ground-breaking research project that has been monitoring 12,000 children's lives in Ethiopia, India, Peru, and Vietnam since 2001. Enhancing knowledge of childhood poverty and its impact on health, education, and general well-being is the goal of this project, which is being conducted by the Department of International Development at the University of Oxford and funded by the Department for International Development.

We will examine statistics from Ethiopia and India in this paper, with an emphasis on four main areas: social well-being, health, education, and economic standing. We want to employ Power BI Report Builder for report development and SQL Server for data administration in order to produce comprehensive insights and visualizations that emphasize the main variables influencing child poverty. With the use of these technologies, we will be able to create a user-friendly reporting system that will guide the development of intervention plans and legislation to lessen child poverty in emerging nations.

With the use of SQL Server and Power BI, we will demonstrate our expertise in database administration, data analysis, and report creation, providing useful information on child poverty in the two chosen nations.

1.2 Downloading the dataset

We downloaded the dataset from the following link:

https://beta.ukdataservice.ac.uk/datacatalogue/series?id=2000060#!/access

We started by creating a username and registering for the UK Data Service. We downloaded the "International Study of Child Poverty: Rounds 1-5 Constructed Files, 2002-2016" dataset after obtaining access. We were able to utilize the login and password we had created upon registration to log in once we gained access. The dataset was then successfully uploaded to our account after this. We also finished the dataset download and submission process by submitting the abstract for our investigation.

1.3 Exploring the dataset

'Constructed files' are supplied to assist researchers utilizing the Young Lives data in addition to the raw information. These created files contain aggregated subsets of characteristics from various rounds of the Young Lives child and household surveys, which were carried out between 2002 and 2016. These data files are an updated version of previously archived data that encompass several rounds and guarantee uniformity in the round-to-round definitions of the variables. To keep things uniform, certain variables have been changed but the majority have remained the same.

The CSV format was created for these files using a particular procedure. Every file with tabs was initially opened with a text editor such as Notepad and saved in text format. The text files were imported into different Excel worksheets and then saved as CSV files by using Microsoft Excel's features. The tabular data from the datasets may now be more easily accessed and used in CSV format thanks to this conversion process.

Key details about the sort of data contained in each dataset are provided by the name, description, and number of rows and columns.

| File name | Description | No of Rows |
|-----------------------------------|---|-------------|
| all_countries_math_irt_scores.csv | Includes children's math proficiency | 32,632 rows |
| | ratings from several different nations. | |
| | Data is grouped by group, survey round, | |
| | and nation. | |
| | | |
| | | |

| all_countries_ppvt_irt_scores.csv | Includes results from the Peabody Picture Vocabulary Test (PPVT), which measures verbal and vocabulary skills. Scores from multiple survey rounds, countries, and linguistic groups are included in the dataset. | 34,010 rows |
|--------------------------------------|--|-------------|
| all_countries_reading_irt_scores.csv | Gives children's reading proficiency scores across national boundaries using information from many survey rounds and linguistic groupings. | 17451 rows |
| et_in_pe_sibling_ppvt_irt_scores.csv | Focuses on the PPVT results of siblings from Peru, Ethiopia, and India, emphasizing the differences in the siblings' ability and language scores. | 6164 rows |
| ethiopia_constructed.csv | A large dataset with precise socioeconomic, educational, and household demographic information from Ethiopia. This contains details about the family history, composition of the home, and child development. | 14995 rows |
| india_constructed.csv | This dataset, which is similar to the Ethiopia built file, focuses on household and child characteristics in India. It contains information on test results, household assets, education, and literacy. | 15097 rows |
| india_sibling_math_irt_scores.csv | Includes Indian siblings' math test results, enabling study of performance and educational variations between siblings. | 3023 rows |
| peru_constructed.csv | Comprehensive socioeconomic and child development data from Peru that includes test results, household variables, and education. | 13830 rows |
| vietnam_constructed.csv | Rich dataset on Vietnamese families' circumstances and children's learning | 15,000 rows |

| | outcomes, with an emphasis on home and school features. | |
|--|---|--|
| | | |

In order to give researchers context for the Young Lives data and to reflect the pertinent questions posed to households and children in subsequent survey rounds, additional variables were added (Azubuike & Briones, 2016). These variables are arranged in the created files into four primary categories: panel information, child characteristics, household characteristics, and identity and location factors. Among the categories are

- Identification and locating variables
- Panel details
- Child characteristics
- Household characteristics

For our case study in the report, we have selected the **india_constructed** and **vietnam_constructed** tables.

1.4 Importing the dataset and implementation of SQL

Once we downloaded the dataset, we saved it as a csv file format. Then we created a database called "Young lifes" in SQL server studio.

-- Creating Data Base

CREATE DATABASE Younglifes
GO

Data Base



Subsequently, we utilized the ALTER TABEL statement to incorporate a new column named 'Country' into the current database, designating it as VARCHAR(50). Next, we changed this new column so that for any entry in which the country field included either an empty string or NULL, 'Vietnam' was assigned as the country name. Ultimately, we obtained the most recent national data to confirm the modifications.

```
task_1_org.sql - L...LU4BDVA\ASUS (62)) $\frac{1}{2}$ \times \tim
```

** Although instructions were to utilize Power BI's data transformation tools to clean the dataset, we encountered technical challenges in establishing a seamless connection between the SQL database and Power BI. As a result, we opted to perform the data cleaning process directly using SQL**

We discovered inconsistent and missing data in several columns when we cleaned the dataset. We resolved this issue by using SQL Server Management Studio rather than doing the preprocessing in Excel. To ensure data consistency throughout the table, we used the COALESCE function to replace NULL or empty entries in each column with 'NA'.

```
--- Clean dataset for vietnam

UPDATE [dbo].[vietnam_constructed]

SET

country = COALESCE(NULLIF(country, ''), 'NA'),
    childid = COALESCE(NULLIF(childid, ''), 'NA'),
    chesc = COALESCE(NULLIF(childid, ''), 'NA'),
    bwght = COALESCE(NULLIF(childid, ''), 'NA'),
    chesc = COALESCE(NULLIF(childid, ''), 'NA'),
    cheight = COALESCE(NULLIF(cheight, ''), 'NA'),
    chheight = COALESCE(NULLIF(cheight, ''), 'NA'),
    chieght = COALESCE(NULLIF(cheight, ''), 'NA'),
    chieght = COALESCE(NULLIF(cheight, ''), 'NA'),
    chodesce = COALESCE(NULLIF(cheight, ''), 'NA'),
    foodsec = COALESCE(NULLIF(cheight, ''), 'NA'),
    commid = COALESCE(NULLIF(commid, ''), 'NA'),
    clustid = COALESCE(NULLIF(commid, ''), 'NA'),
    chilmes = COALESCE(NULLIF(clustid, ''), 'NA'),
    chilmes = COALESCE(NULLIF(clustid, ''), 'NA'),
    chilmes = COALESCE(NULLIF(chilmess, ''), 'NA'),
    shfam1 = COALESCE(NULLIF(shfam1, ''), 'NA'),
    shfam2 = COALESCE(NULLIF(shfam1, ''), 'NA'),
    shfam3 = COALESCE(NULLIF(shfam2, ''), 'NA'),
    shfam5 = COALESCE(NULLIF(shfam3, ''), 'NA'),
    shfam6 = COALESCE(NULLIF(shfam3, ''), 'NA'),
    shfam7 = COALESCE(NULLIF(shfam3, ''), 'NA'),
    shfam8 = COALESCE(NULLIF(shfam3, ''), 'NA'),
    shfam1 = COALESCE(NULLIF(shfam1, ''), 'NA'),
    shfam1 = COALESCE(NULLIF(shfam3, ''),
```

```
--- Clean dataset for vietnam

UPDATE [dbo].[vietnam_constructed]

SET

country = COALESCE(NULLIF(country, ''), 'NA'),
    childid = COALESCE(NULLIF(childid, ''), 'NA'),
    chsex = COALESCE(NULLIF(childid, ''), 'NA'),
    bught = COALESCE(NULLIF(bught, ''), 'NA'),
    agenon = COALESCE(NULLIF(cheeight, ''), 'NA'),
    chweight = COALESCE(NULLIF(cheeight, ''), 'NA'),
    chheight = COALESCE(NULLIF(cheeight, ''), 'NA'),
    bmi = COALESCE(NULLIF(cheight, ''), 'NA'),
    bmi = COALESCE(NULLIF(cheight, ''), 'NA'),
    zhfa = COALESCE(NULLIF(cheight, ''), 'NA'),
    choolesce = COALESCE(NULLIF(cheight, ''), 'NA'),
    region = COALESCE(NULLIF(cheight, ''), 'NA'),
    commid = COALESCE(NULLIF(counid, ''), 'NA'),
    cutstid = COALESCE(NULLIF(counid, ''), 'NA'),
    chillness = COALESCE(NULLIF(counid, ''), 'NA'),
    chillness = COALESCE(NULLIF(chilness, ''), 'NA'),
    chillness = COALESCE(NULLIF(chilness, ''), 'NA'),
    shfam1 = COALESCE(NULLIF(chilness, ''), 'NA'),
    shfam2 = COALESCE(NULLIF(shfam2, ''), 'NA'),
    shfam3 = COALESCE(NULLIF(shfam3, ''), 'NA'),
    shfam6 = COALESCE(NULLIF(shfam4, ''), 'NA'),
    shfam7 = COALESCE(NULLIF(shfam1, ''), 'NA'),
    shfam8 = COALESCE(NULLIF(shfam1, ''), 'NA'),
    shfam8 = COALESCE(NULLIF(shfam1, ''), 'NA'),
    shfam8 = COALESCE(NULLIF(shfam1, ''), 'NA'),
    shfam1 = COALESCE(NULLIF(shfam1, ''), 'NA'),
```

Delete Null Values from tables.

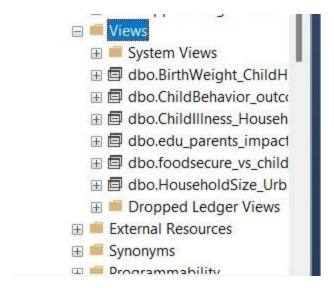
```
-- Clean dataset for the Ethiopia table using COALESCE for null/empty values
 JPDATE [dbo].[ethiopia_constructed]
       country = COALESCE(NULLIE(country.
       childid = COALESCE(NULLIF(childid, ''), 'NA'),
      Chisex = COALESCE(NULLIF(cheex, ''), 'NA'),
agemon = COALESCE(NULLIF(agemon, ''), 'NA'),
chweight = COALESCE(NULLIF(chweight, ''), 'NA'),
bmi = COALESCE(NULLIF(chheight, ''), 'NA'),
bmi = COALESCE(NULLIF(bmi, ''), 'NA'),
2wfa = COALESCE(NULLIF(zwfa, ''), 'NA'),
2wfa = COALESCE(NULLIF(zwfa, ''), 'NA'),
       zhfa = COALESCE(NULLIF(zhfa, ''), 'NA'),
foodsec = COALESCE(NULLIF(region, ''), 'NA'),
region = COALESCE(NULLIF(region, ''), 'NA'),
commid = COALESCE(NULLIF(commid, ''), 'NA'),
clustid = COALESCE(NULLIF(commid, ''), 'NA'),
chilness = COALESCE(NULLIF(chilness, ''), 'NA'),
chinjury = COALESCE(NULLIF(chinjury, ''), 'NA'),
yc = COALESCE(NULLIF(chinjury, ''), 'NA'),
deceased = COALESCE(NULLIF(deceased, ''), 'NA'),
       shfam1 = COALESCE(NULLIF(shfam1, ''), 'NA'),
shfam2 = COALESCE(NULLIF(shfam2, ''), 'NA'),
        shfam3 = COALESCE(NULLIF(shfam3,
       shfam4 = COALESCE(NULLIF(shfam4.
                                                                                       ), 'NA'),
       shfam5 = COALESCE(NULLIF(shfam5,
shfam6 = COALESCE(NULLIF(shfam6,
       shfam7 = COALESCE(NULLIF(shfam7,
       shfam8 = COALESCE(NULLIF(shfam8,
       shfam9 = COALESCE(NULLIF(shfam9, '
shfam10 = COALESCE(NULLIF(shfam10,
       shfam11 = COALESCE(NULLIF(shfam11,
shfam12 = COALESCE(NULLIF(shfam12,
       shfam13 = COALESCE(NULLIF(shfam13,
shfam14 = COALESCE(NULLIF(shfam14,
       shfam14 = COALESCE(NULLIF(Shfam18, ''), 'NA')
shfam18 = COALESCE(NULLIF(shfam18, ''), 'NA'),
'NA'),
                                                                                                 'NA').
       nnsize = COALESCE(NULLIF(nnsize, ''), 'NA'),
caredu = COALESCE(NULLIF(caredu, ''), 'NA'),
hschool = COALESCE(NULLIF(hschool, ''), 'NA'),
chalcohol = COALESCE(NULLIF(chalcohol, ''), 'NA')
```

```
country IS NULL OR country = '' OR
childid IS NULL OR childid = " OR
chsex IS NULL OR chsex = " OR
agemon IS NULL OR agemon = '' OR
chweight IS NULL OR chweight = ** OR
chheight IS NULL OR chheight = *** OR
bmi IS NULL OR bmi = "
zwfa IS NULL OR zwfa = " OR
zhfa IS NULL OR zhfa = " OR
foodsec IS NULL OR foodsec = '' OR region IS NULL OR region = '' OR
commid IS NULL OR commid = ' OR
commid IS NULL OR commid = '' OR
clustid IS NULL OR clustid = '' OR
chilness IS NULL OR chillness = '' OR
chinjury IS NULL OR chinjury = '' OR
yc IS NULL OR yc = '' OR
deceased IS NULL OR deceased = " OR
shfam1 IS NULL OR shfam1 = '' OR
shfam2 IS NULL OR shfam2 = '' OR
shfam3 IS NULL OR shfam3 = " OR
shfam4 IS NULL OR shfam4 = "
shfam5 IS NULL OR shfam5 =
shfam6 IS NULL OR shfam6 = " OR
shfam7 IS NULL OR shfam7 =
shfam8 IS NULL OR shfam8
shfam9 IS NULL OR shfam9 = "
                                       OR
shfam10 IS NULL OR shfam10 = '' OR
shfam11 IS NULL OR shfam11 = " OR
shfam12 IS NULL OR shfam12 = '' OR
shfam13 IS NULL OR shfam13 = '' OR
shfam14 IS NULL OR shfam14 = '' OR
shfam18 IS NULL OR shfam18 = '' OR
hhsize IS NULL OR hhsize = '' OR caredu IS NULL OR caredu = '' OR hschool IS NULL OR hschool = '' OR
chalcohol IS NULL OR chalcohol = "";
```

Six fresh concepts pertaining to the Young Life initiative were produced by the filtering that was done the following day. Every concept was explored by choosing the appropriate, pertinent columns.

- The link between food security and child health
- Household size differences between urban and rural areas
- Impact of birth weight on child health
- Exploring the impact of household economic strain on child health
- The Influence of Parental Education on Child Development
- Impacting of Alcohol and Smoking

In this way, we have developed six perspectives based on the six previously indicated themes.



1.5 Implementation in Microsoft Report Builder

Create an Embedded Data Source in Report Builder

1. Open Report Builder

Open the Report Builder software on your desktop.

The New Report or Dataset dialog box will show up when it opens.

2. Launch a New Report

Make sure that New Report is chosen in the dialog box's left pane.

Click on Table or Matrix Wizard in the right pane. This will launch a fresh report creation interface.

3. Select a Dataset

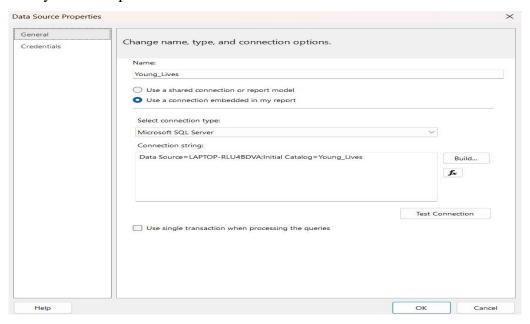
Select the Create a dataset option on the Choose a dataset screen.

To continue, click Next. This will take you to the screen where you can select a connection to a data source.

4. Establish a New Source of Data

Select New. In order to configure a new data source for the report, this will open the Data Source Properties dialog box.

Identify and Set Up



5. Construct a Connection String:

To generate a new connection string, click Build in the Connection string area.

6. Indicate the SQL Server Instance:

Enter the name of your SQL Server instance in the Server Name field or choose it from the drop-down menu.

7. Choose the Database:

Choose the required database (in this case, "younglifes") from the drop-down option.

8. Check the Link:

To make sure the connection to the data source is operating correctly, click Test Connection.

If everything goes well, the message "Connection created successfully" will appear.

9. Verify Connection:

To save the connection, click OK.

After selecting your new data source, you will be returned to the Select a connection to a data source screen. To continue, click Next.

TO CREATE A QUERY

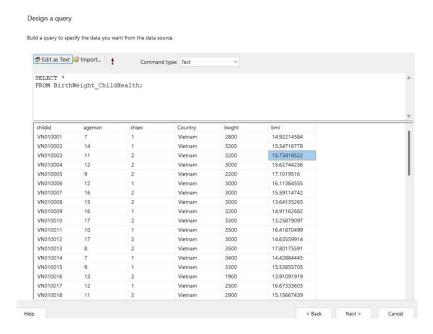
1. Design a Query:

On the Design a query page, the relational query designer is open by default. For this tutorial, switch to the text-based query designer.

2. Switch to Text-based Query:

Click Edit as Text to open the text-based query designer, displaying a query pane and a results pane.

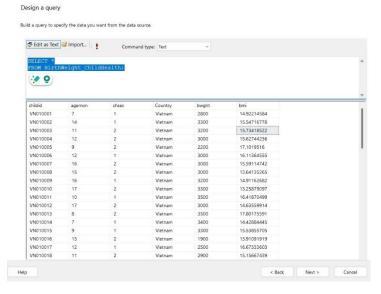
3. Input SQL Query:



4. Run the Query:

On the query designer toolbar, click Run (!) to execute the query.

The results pane will display the data fetched from the BirthWeight_ChildHealth table.

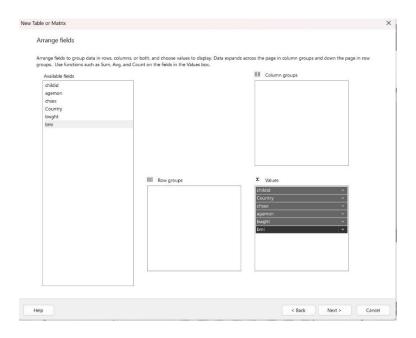


5. Proceed to the Next Step:

Click Next to move on to the grouping and layout phase.

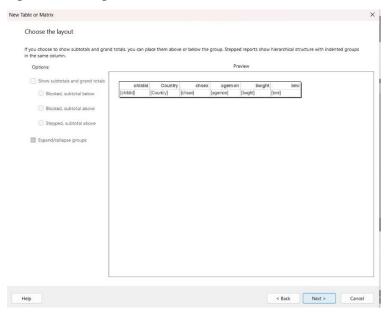
6. Organize Data into Values:

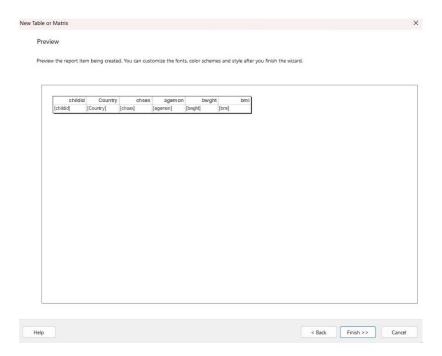
Drag all the columns into the Values section of the wizard to include them in your report.



7. Continue to Preview:

Click Next to proceed. Click Next again to preview the table structure, and then click Finish to complete the setup.

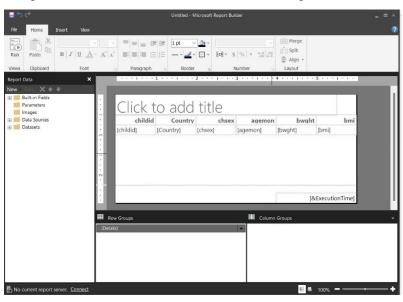




Finalizing and Previewing the Report

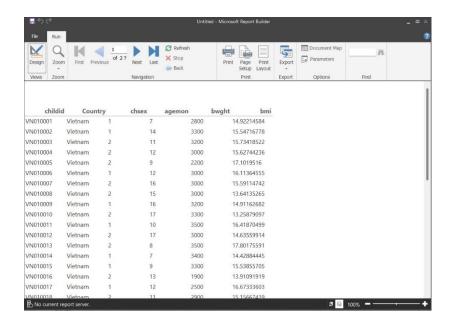
8. Add the Table to Design Surface:

After finishing the wizard, the table is added to the design surface of the report.



9. Preview the Report:

Click Run to preview your report. This allows you to see how the data will be displayed in the final report.



Creating reports using Microsoft Report Builder

❖ Report 1: The link between food security and child health

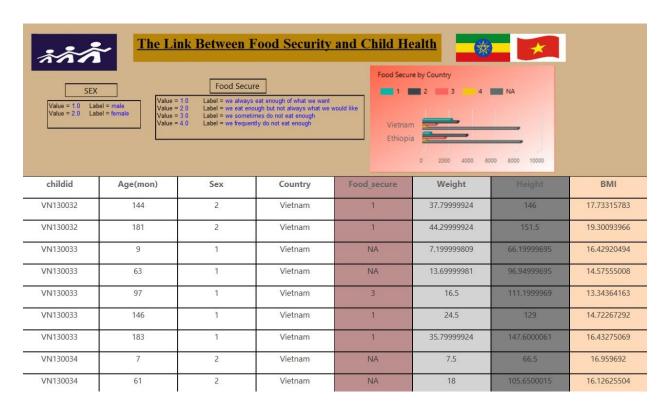
In Vietnam and Ethiopia, food security has a significant impact on the health of children. Children from households experiencing food insecurity are more likely to suffer from malnutrition in Ethiopia, where food shortages and drought are prevalent. Stunted growth compromised immune systems, and heightened susceptibility to infections are frequently the results of this. Long-term cognitive and physical problems may arise from inadequate nutrition throughout critical developmental phases.

Food insecurity is a serious problem in Vietnam as well, especially in rural regions. Low-income children might not have as much access to a variety of nutrient-dense foods, which could result

in vitamin and mineral deficits. Because undernourished children frequently find it difficult to focus in class, this has a detrimental impact on both academic performance and physical health. Improving the general health and development of children in these nations requires ensuring food security.



Final Report

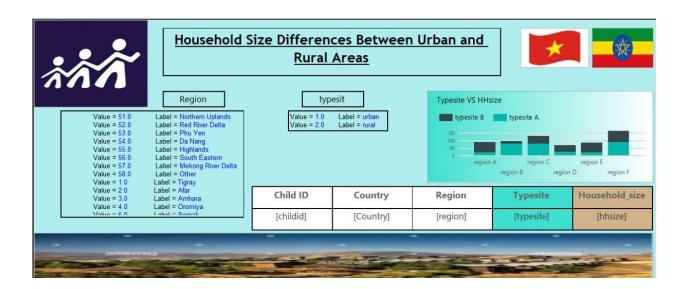


Report 2: Household size differences between urban and rural areas

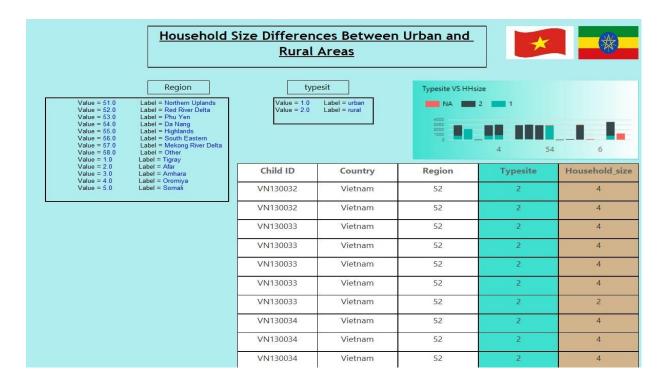
Ethiopian and Vietnamese urban-rural household size disparities are a reflection of a number of social, economic, and cultural issues. Households are typically larger in rural areas of both countries, frequently as a result of the traditional emphasis on agriculture. In many communities, having more children is viewed as advantageous for labor-intensive job, and more family members participate in farming activities. Additionally, increased birth rates in remote areas may result from less access to family planning and healthcare services.

On the other hand, household sizes are often lower in urban areas like Vietnam and Ethiopia. Reduced birth rates, a move toward nuclear families, and easier access to healthcare and education are all linked to urbanization. Economic considerations also come into play because larger families are discouraged by the higher cost of metropolitan life. Additionally, urban families frequently place a higher value on work and education than on agriculture, which results in different dynamics inside the home.

In conclusion, the disparity between rural areas' reliance on agriculture and urban areas' access to healthcare, education, and economic possibilities is what drives the variations in household sizes between Ethiopia and Vietnam.



Final Report



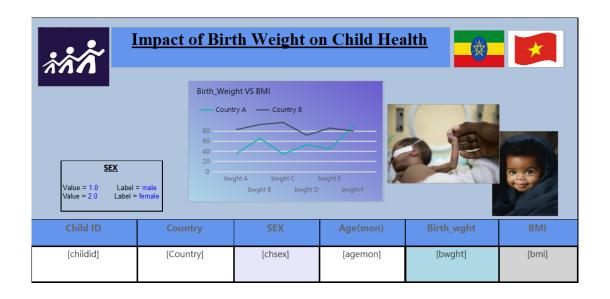
Report 3: Impact of birth weight on child health

Given that low birth weight is a key predictor of future health issues, birth weight has a substantial impact on children's health in Ethiopia and Vietnam. Low birth weight babies are

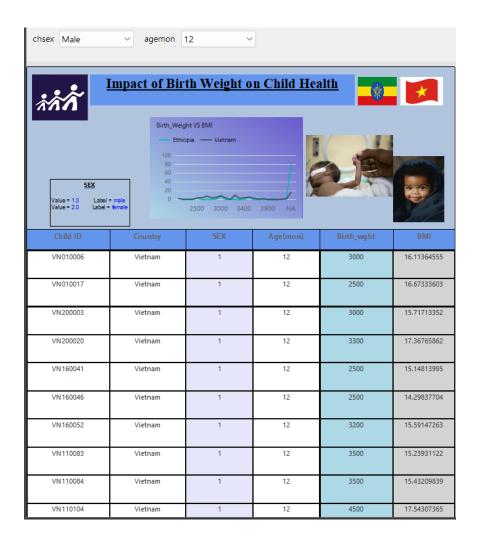
more prone to suffer from hunger, poor growth, and developmental problems in both nations. Poor maternal nutrition, restricted access to healthcare, and socioeconomic challenges that impact prenatal care are some of the contributing causes.

Low birth weight is frequently associated with high infant death rates in Ethiopia as well as long-term health problems like delayed cognitive development and infection susceptibility. Similar to this, low birth weight babies in Vietnam might have trouble developing to their full potential both intellectually and physically, which could affect their academic performance and future productivity.

Improving maternal nutrition and healthcare is essential to resolving these problems in both nations.



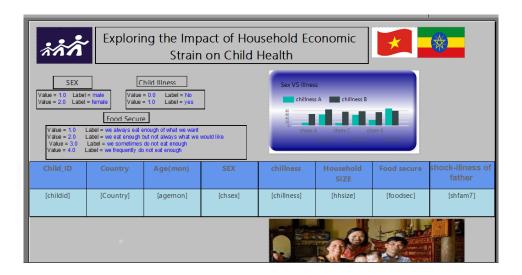
Final Report



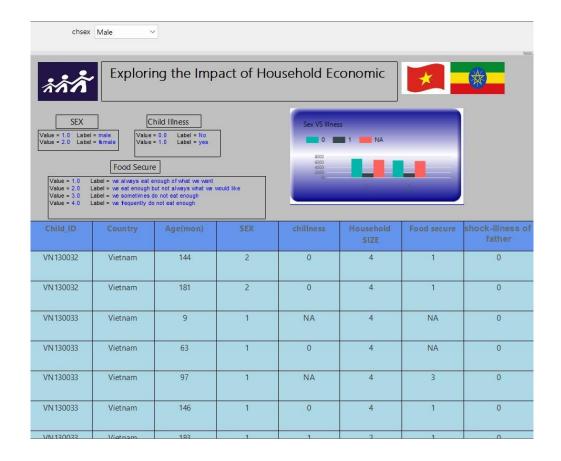
❖ Report 4: Exploring the impact of household economic strain on child health

In Ethiopia and Vietnam, household economic stress has a major impact on children's health. Children who live in families who are struggling financially are more likely to suffer from malnutrition and delayed growth. Children's immune systems can be weakened by inadequate nutrition, leaving them more vulnerable to illnesses and infections. Access to healthcare services is also limited by a lack of financial means. Families may postpone or forego medical treatment in rural areas, where financial difficulty is more common, which could impact health outcomes.

Furthermore, unstable living circumstances and a lack of support can have a detrimental impact on a child's mental health by producing stress and anxiety.



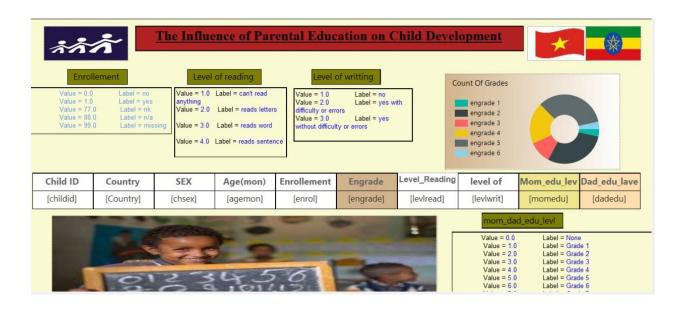
Final Report



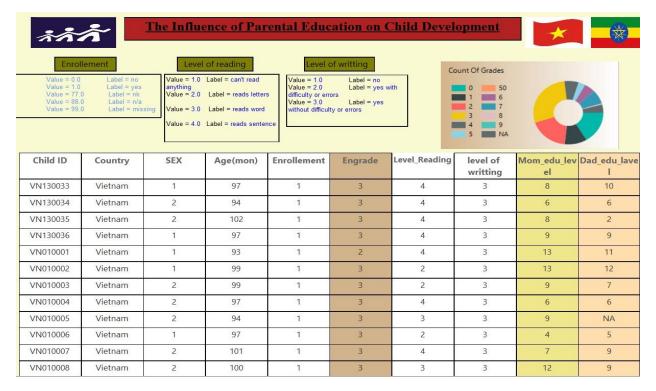
Report 5: The Influence of Parental Education on Child Development

In Vietnam and Ethiopia, parental education has a big impact on how children develop. Higher educated parents in Ethiopia typically support their children's learning more effectively at home and through their knowledge of the educational system. Parents with higher levels of education are more likely to read to their kids or encourage school attendance, two activities that foster cognitive growth and improve academic performance.

Parental education has a similar impact on children's development in Vietnam. Formally educated parents frequently place a high value on their kids' education, promoting self-discipline and academic achievement. Additionally, their children's overall physical and mental development is aided by their increased access to resources and health and nutrition information. By encouraging healthier, more nurturing surroundings, raising parental education can greatly enhance children's prospects for the future in both nations.



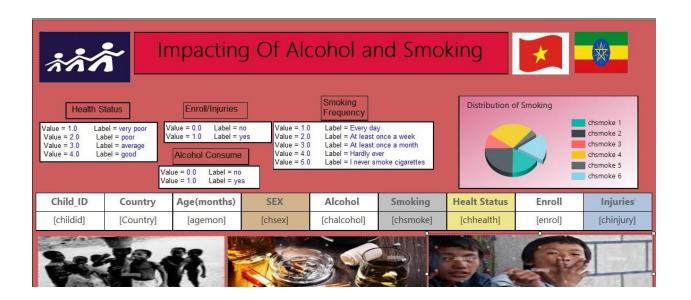
Final Report



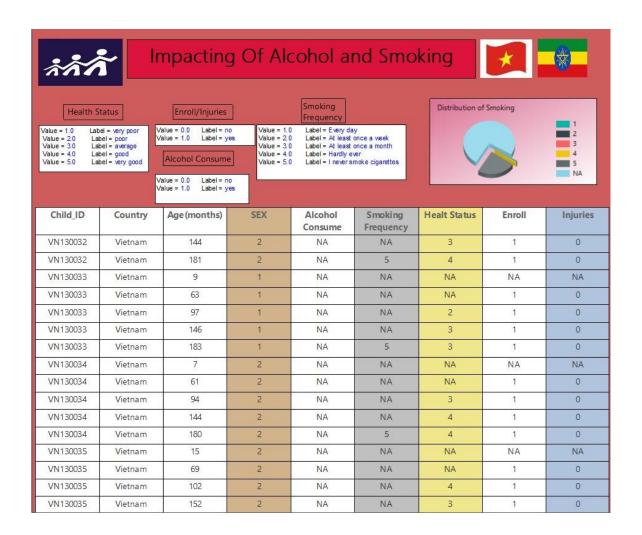
❖ Report 6: Impacting of Alcohol and Smoking

In Ethiopia and Vietnam, alcohol and tobacco use have a substantial negative influence on children's health, especially when poverty is present. In both nations, caregivers' smoking and alcohol use can have a detrimental impact on home income by taking funds away from necessities like healthcare, education, and nutrition. This can impede development and worsen poverty among children.

High alcohol use in Ethiopia is associated with socioeconomic stress, which deteriorates parental health and reduces child care. Similar to this, smoking is common in Vietnam, particularly among men. This has a direct effect on the welfare of children and their prospects for the future since it can result in health issues and decreased financial stability.



Final report



1.6 Conclusion

To sum up, this project examined a number of important variables influencing children's health and development in diverse settings, with an emphasis on Ethiopia and Vietnam. The well-being of children is impacted by the inequalities in household sizes between urban and rural settings, which reflect discrepancies in resource allocation and access to essential services. Low birth weight frequently results in long-term health issues, making it an important predictor of a child's health. Household financial distress exacerbates these problems by restricting access to nutrition and medical treatment. Food security is important since a child's physical and mental development are directly impacted by poor nutrition. Since educated parents are better able to support their children's development and education, parental education has also emerged as a critical factor. Last but not least, the negative impacts of smoking and drinking, especially in low-income households, worsen health and financial issues and further impair children's general wellbeing.

Targeted strategies that take into account the social and economic determinants of health are necessary to address these interrelated problems. Significant progress can be made in improving child health outcomes and general development in these areas by concentrating on strengthening food security, healthcare access, and education while lowering harmful habits like smoking and alcohol use.



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TASK 02

02. ANALYSIS AND VISUALIZATION OF EPC DATA IN MANCHESTER (2013-2023) USING POWER BI

1 Introduction

The Greater Manchester Domestic Energy Performance Analyzer is a large-scale project that evaluates and visualizes the energy efficiency of buildings across all of Manchester's local authorities using Energy Performance Certificate (EPC) data from 2013 to 2023. This assignment highlights the significance of data-driven insights in understanding energy use patterns and improving local sustainability activities. Using Power BI, the goal is to transform unstructured data into intelligent, interactive visualizations that identify trends in energy efficiency, identify areas for improvement, and provide useful guidance to individuals in the energy sector. This study will serve as a roadmap for future energy efficiency projects, reflecting historical progress in energy performance and assisting lawmakers, real estate developers, and homeowners in making educated decisions. Manchester's energy sustainability will rise as a result of the interactive dashboard, which will make it easier for users to explore the data and discover trends in energy performance across various property types and neighborhoods.

1.2 Exploring the dataset

Downloading the dataset

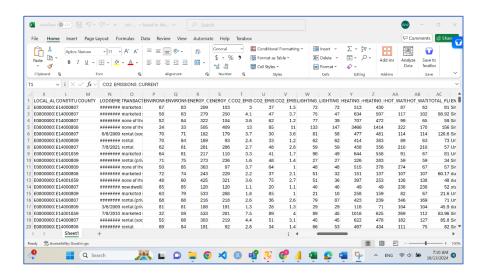
Using the provided URL, the dataset—which contains thorough information on residential Energy Performance Certificates (EPCs)was downloaded. It is important to review the dataset before beginning any analysis, making sure you understand the primary variables and have a good understanding of how it is organized. This exploratory study helps with trend identification, data quality assurance, and preparing the information for further research.

The dataset can be found at the following location:

httpshttps://epc.opendatacommunities.org/downloads/domestic#local-authority

Exploring the dataset

The EPC dataset for England and Wales provides comprehensive information on energy efficiency in residential and commercial premises. This information includes property type, insulation, heating type, emissions rating, and energy expenses. Lawmakers, tenants, homeowners, and landlords utilize this publicly available dataset to monitor their progress toward carbon reduction goals and to make well-informed decisions regarding energy efficiency. With 300,889 rows and 92 columns of property data, it provides an enormous collection that may be used to examine policy effects, energy performance trends, and opportunities for environmental improvements.

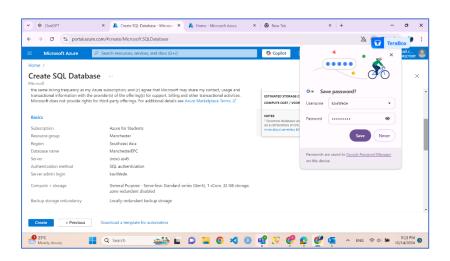


2. Database Creation

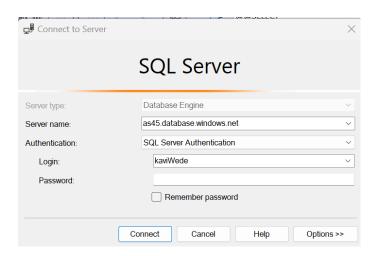
2.1 Azure cloud database

Azure MySQL Database Creation: We created a MySQL database instance in Azure to store the EPC data. This ensured reliable data storage with cloud-based scalability and security. The steps followed were:

- Azure Service Used: Azure Database for MySQL
- Database Name:ManchesterEPC
- Resource Group: Manchester
- MySQL Server Name: epc-mysql-server



• **SQL Database Connection**: The EPC data was imported into the Azure SQL Database using SQL Server Management Studio (SSMS) and Azure Data Factory. Once the database was set up, we established a secure connection to the SQL instance for querying and data management



2.2Data Cleaning and Transformation Using SQL

2.2.1 Removing Duplicates

Duplicate records can result in skewed or incorrect analyses, especially when analyzing trends in energy performance. We used SQL's ROW_NUMBER() function in conjunction with a **Common Table Expression** (**CTE**) to identify and remove duplicates based on multiple columns such as LMK_KEY, UPRN, Energy Consumption, CO2 Emissions, and other related fields.

2.2.2 Handling Missing Values

Missing values in key columns like Energy_Consumption_Current, CO2_Emissions_Current, and Current_Energy_Rating were handled by replacing them with calculated averages values.

Replace Null Values:

we replaced the null values with the average values and blank property type fields were set to "No DATA" to avoid null values in our analysis::

```
UPDATE EPC_Data

SET Energy_Consumption_Current = (SELECT AVG[Energy_Consumption_Current] FROM [dbo].[Book1])

WHERE Energy_Consumption_Current IS NULL;
```

Replace blank column with No Data

```
UPDATE [dbo].[Book1]
SET PROPERTY_TYPE = 'NO DATA'
WHERE PROPERTY_TYPE IS NULL OR PROPERTY_TYPE = '';
```

2.2.3 Filtering Data Between 2013 and 2023

To focus on recent energy performance data, we filtered the dataset to include only records between the years 2013 and 2023:

```
DELETE FROM [dbo].[Book1]
WHERE YEAR(INSPECTION_DATE) < 2013 OR YEAR(INSPECTION_DATE) > 2023;
```

2.2.4 Handling Outliers

Outliers in CO2_Emissions_Current and Energy_Consumption_Current were identified using SQL aggregate functions and manually removed based on threshold values:

Identify Outliers:

```
| SELECT * FROM [dbo].[Book1] | WHERE [CO2 Emissions Current] > 1000 OR [Energy Consumption Current] > 5000;
```

Remove Outliers:

```
DELETE FROM [dbo].[Book1]
WHERE [CO2_Emissions_Current] > 1000 OR [Energy_Consumption_Current] > 5000;
```

2.2.5 Creating Views

Create views for different aspects of the energy data, which could help simplify access to specific segments of the data, such as properties grouped by energy ratings or focusing on potential improvements.

```
CREATE VIEW View_NewEnergyRatings AS

SELECT

[POSTCODE],

[BUILDING_REFERENCE_NUMBER],

[CURRENT_ENERGY_RATING],

[POTENTIAL_ENERGY_RATING]

FROM

[dbo].[Book1]

WHERE

LODGEMENT_DATE >= '2020-01-01';
```

2.2.5. Using Common Table Expressions (CTEs)

We use CTE for perform complex queries in a more readable manner, such as calculating average current and potential energy efficiencies.

```
WITH EmissionsCTE AS (
SELECT

[LOCAL_AUTHORITY],

AVG(CAST([CURRENT_ENERGY_EFFICIENCY] AS FLOAT)) AS AVgCurrentEfficiency,

AVG(CAST([POTENTIAL_ENERGY_EFFICIENCY] AS FLOAT)) AS AVgPotentialEfficiency

FROM

[dbo].[Book1]

GROUP BY

Local_Authority
)

SELECT * FROM EmissionsCTE;
```

2.2.6. Stored Procedures

We are Creating a stored procedure to retrieve or manipulate data based on certain input parameters like energy rating or postcode.

```
@EnergyRating CHAR(1)

AS

BEGIN

[POSTCODE],

[BUILDING_REFERENCE_NUMBER],

[POTENTIAL_ENERGY_RATING],

[CURRENT_ENERGY_EFFICIENCY],

[CURRENT_ENERGY_EFFICIENCY]

FROM

[dbo].[Book1]

WHERE

[CURRENT_ENERGY_RATING] = @EnergyRating;

END;
```

2.2.7. Aggregate and Ranking Functions

We Utilize SQL's built-in functions to analyze the data, such as finding the top 10 properties with the highest potential for energy efficiency improvement.

03. Exploratory Data Analysis (EDA)

Overview

Exploratory Data Analysis (EDA) is a crucial step in understanding the structure, patterns, and relationships within the data. For this assignment, we performed EDA on the **Energy Performance Certificate (EPC)** dataset for Manchester's local authorities, covering the period from 2013 to 2023. The primary goal was to explore the data, identify trends in energy performance, CO2 emissions, and energy consumption, and uncover potential areas for improvement in energy efficiency.

Objectives of EDA

- Understand the distribution of key metrics such as energy ratings, CO2 emissions, and energy consumption.
- Detect patterns and trends in energy performance over time.
- Investigate geographical differences in energy efficiency across different local authorities.
- Identify correlations between energy ratings, CO2 emissions, and property types.
- Highlight any potential anomalies, such as outliers, that might affect the analysis.

3.2.1 Energy Ratings Distribution

The **distribution of energy ratings** was one of the first aspects we explored. We used a **bar chart** to show the proportion of properties that fall into each energy efficiency category (A-G). This allowed us to quickly see whether most properties are energy efficient (A-C ratings) or whether the majority fall into the lower ratings (D-G), which indicates a need for improvement.

Key Insights:

- The majority of properties in Manchester fall within the D and E energy rating categories, indicating a significant opportunity for energy efficiency improvements.
- Only a small percentage of properties have an A or B rating, reflecting a potential area for energy upgrades.

3.2.2 CO2 Emissions Analysis

We performed an analysis of **current and potential CO2 emissions** across different local authorities. A **line chart** was used to track the trends in CO2 emissions over time, highlighting how emissions have fluctuated across Manchester between 2013 and 2023.

In addition, a **map visualization** was created to show **geographic variations** in CO2 emissions. This allowed us to identify specific areas where properties emit higher levels of CO2, which can guide policy interventions for reducing emissions.

Key Insights:

- Certain local authorities, particularly in densely populated areas, exhibited significantly higher CO2 emissions compared to suburban and rural areas.
- CO2 emissions have gradually decreased over the years, but there is still room for improvement, especially in older properties with inefficient heating systems.

3.2.3 Energy Consumption Trends

We analyzed trends in **current energy consumption** versus **potential energy consumption** to identify opportunities for energy savings. Using a **stacked column chart**, we compared current and potential energy consumption across different property types (e.g., flats, detached houses, terraced houses).

This analysis helped highlight which property types have the highest energy consumption and the most potential for improvement if energy-saving measures are applied.

Key Insights:

- Detached houses and older buildings exhibited higher energy consumption levels, particularly in heating costs.
- There is a significant opportunity for reducing energy consumption in terraced houses and flats by upgrading insulation and heating systems.

4.2.4 Geographic Differences in Energy Efficiency

To explore geographic differences in energy efficiency, we created a **heat map** that showed the average energy efficiency scores across local authorities in Manchester. This helped us identify regions that have consistently poor energy performance, as well as those that are performing well.

Key Insights:

- Central Manchester tends to have poorer energy efficiency ratings, likely due to older buildings with outdated insulation and heating systems.
- Suburban areas generally perform better in terms of energy efficiency, with more properties achieving higher ratings (A-C).

3.3 Outlier Detection

Outliers in the dataset, such as abnormally high CO2 emissions or extremely low energy consumption, were identified and analyzed.

For instance, properties with extremely high CO2 emissions (above a certain threshold) were flagged for further investigation. These outliers could indicate errors in the data or properties with significantly outdated infrastructure that require urgent intervention.

Steps Taken to Handle Outliers:

- **Filter Outliers**: We filtered out properties with extremely high CO2 emissions or energy consumption that were far outside the normal range.
- **Replace Missing Values**: For certain outliers with missing or inconsistent data, we replaced them with average values based on the property type and location.

04. Power BI Dashboard Design

The Power BI dashboard is designed to provide an intuitive, user-friendly interface for stakeholders to explore energy performance data across Manchester's local authorities. The dashboard is divided into four distinct categories to ensure that all key aspects of the data are covered and presented in a logical flow. Each category addresses a specific area of energy performance, allowing users to dive deep into relevant data and extract meaningful insights.

Dashboard Categories:

- 1. Energy Efficiency Overview
- 2. Property Type & Energy Consumption Breakdown
- 3. CO2 Emission Analysis and Geography Insights
- 4. Risk & Opportunity Analysis

4.1 Category 1: Energy Efficiency Overview

This section provides a high-level summary of the energy efficiency ratings of properties in Manchester. It is designed to give stakeholders an immediate understanding of the overall energy performance across local authorities, highlighting areas where energy efficiency improvements can be made.

1. Key Performance Indicators (KPI Cards):



Display high-level metrics, such as the average energy efficiency score and the percentage of properties with potential for improvement..

Visual: KPI Cards

- Card 1: Average Energy Efficiency Score
- Card 2: Percentage of Properties with Potential for Improvement

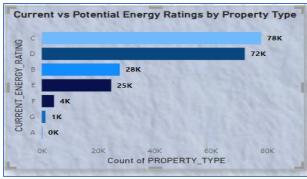
DAX Code for KPIs:

1. Average Energy Efficiency Score:

1 Avg_Energy_Efficiency_Score = AVERAGE(Book1[CURRENT_ENERGY_EFFICIENCY])

2. Percentage of Properties with Potential for Improvement

2. Bar Chart: Current vs Potential Energy Ratings by Property Type



A clustered bar chart comparing current energy ratings to potential energy ratings across different property types (House, Flat, etc.) gives a clear indication of how properties could improve. This helps you quickly see which property types have the most room for improvement

3. Table - Detailed Energy Efficiency Data

We Provides a detailed breakdown of energy efficiency data by property, including current ratings, targets, and performance against those targets.

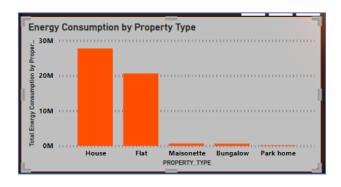


```
1 Performance Against Target =
2 IF([Avg_Energy_Efficiency_Score] > [Efficiency Target], "Above Target",
3 | IF([Avg_Energy_Efficiency_Score] < [Efficiency Target], "Below Target", "On Target"))
4
```

```
1 YOY Efficiency Change =
2 VAR CurrentYearEfficiency = CALCULATE([Avg_Energy_Efficiency_Score], 'Book1'[Year] = YEAR(TODAY()))
3 VAR LastYearEfficiency = CALCULATE([Avg_Energy_Efficiency_Score], 'Book1'[Year] = YEAR(TODAY()) - 1)
4 RETURN CurrentYearEfficiency - LastYearEfficiency
5
```

4. Energy Consumption by Property Type

The "Energy Consumption by Property Type" chart is a column chart that visually compares the total energy consumption across various property types. It specifically presents data for Houses, Flats, and Maisonettes, though only Houses and Flats show significant energy usage, nearly 30 million and just over 20 million units respectively. Maisonettes, Bungalows, and Park homes are also listed but do not display any visible energy consumption, indicating minimal or no usage in these categories. This chart effectively highlights which property types are the largest consumers of energy, aiding in targeted energy management and conservation efforts.

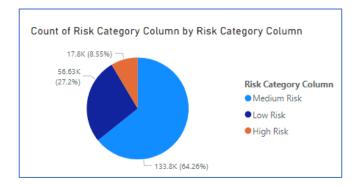


Category 2.Risk & Opportunity Analysis

1.Pie Chart - Distribution of Risk Categories

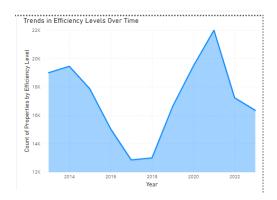
We Visualize the proportional distribution of risk based on energy efficiency levels.it immediate understanding of which risk categories contain the most properties and ability to prioritize interventions based on the size of each risk category.

DAX for Risk Category Calculation:

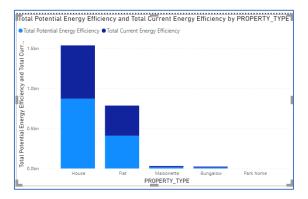


2.Stacked Area Chart - Efficiency Improvement Over Time

Monitor changes in energy efficiency levels over time across different property category.this trends in energy efficiency improvements or declines.

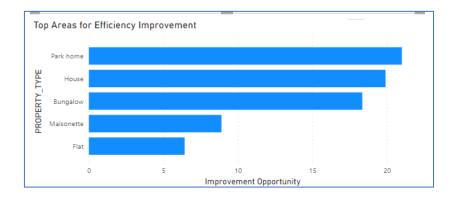


Total Potential Energy Efficiency and Total Current Energy Efficiency by PROPERTY_TYPE," is a stacked column chart that compares the total current energy efficiency against the total potential energy efficiency for various types of properties. The chart clearly shows that houses and flats are the primary focus, with houses displaying a significant gap between current and potential energy efficiency, indicating substantial room for improvement. Flats also show a notable difference but to a lesser extent, whereas Maisonettes, Bungalows, and Park homes have minimal or no data shown.



Top Areas for Efficiency Improvement," is a horizontal bar chart that ranks property types by their potential for efficiency improvement. The chart lists 'Park home' as having the highest potential, followed by 'House,' 'Bungalow,' 'Maisonette,' and 'Flat,' indicating where efforts to improve energy efficiency could be prioritized. The use of a horizontal layout makes it easy to compare across types and immediately spot which property types require the most attention

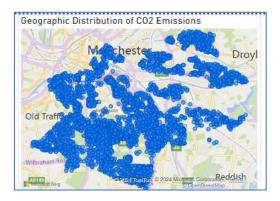
```
1 Improvement Opportunity =
2 AVERAGE('Book1'[POTENTIAL_ENERGY_EFFICIENCY]) - AVERAGE('book1'[CURRENT_ENERGY_EFFICIENCY])
3
```



Category 3.CO2 Emission Analysis and Geography Insights

1. **Geographic Distribution of CO2 Emissions**: This map highlights the distribution of CO2 emissions across the Manchester area, using blue markers to indicate emission levels in different locations. This visualization helps identify specific areas with higher emissions, aiding targeted environmental interventions.

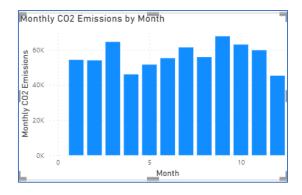
```
1 Average CO2 Emissions = AVERAGE('Book1'[CO2_EMISSIONS_CURRENT])
```



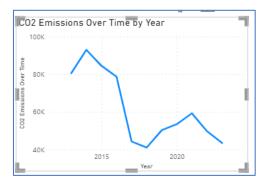
2.**Monthly CO2 Emissions by Month**: The column chart shows the fluctuations in CO2 emissions throughout the year, with each bar representing a month. This visualization reveals a trend where emissions peak around the middle of the year, possibly indicating seasonal variations in energy use or industrial activity.

```
Monthly CO2 Emissions = CALCULATE(
SUM('Book1'[CO2_EMISSIONS_CURRENT]),
ALLEXCEPT('Book1', 'Book1'[Year], 'Book1'[Month])

4 )
5
```

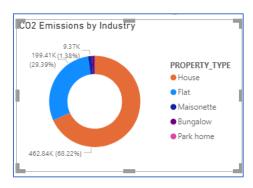


3.**CO2 Emissions Over Time by Year**: The line chart tracks the annual CO2 emissions from 2015 to beyond 2020, showing significant fluctuations over the years. The trend indicates a peak around 2015, followed by a sharp decrease and subsequent fluctuations, which could reflect changes in local industry practices, regulatory impacts, or economic factors.



4.**CO2** Emissions by Industry: The donut chart breaks down CO2 emissions by property type, indicating a predominant contribution by 'Flat' and 'House' types, which together make up the majority of emissions. This chart helps in understanding which property types are the most significant contributors to emissions, informing policy and energy efficiency strategies.





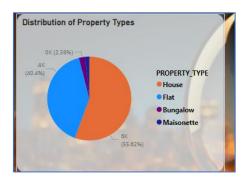
Category 4.Property Type & Energy Consumption Breakdown

1. **Energy Consumption by Property Type**: This bar chart displays the total energy consumption segmented by property type. Flats consume the most energy, significantly more than houses, with maisonettes and bungalows showing relatively minimal consumption. The visual helps identify which property types are the highest energy consumers, potentially guiding energy efficiency initiatives.

```
1 Total Energy Consumption = SUM('Book1'[ENERGY_CONSUMPTION_CURRENT])
```

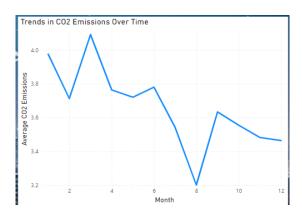


2. **Distribution of Property Types**: The pie chart shows the distribution of different property types. Houses dominate the mix, comprising 93.52% of the properties, followed by a small percentage of flats (4.45%) and an even smaller fraction of maisonettes (2.03%). This chart provides a quick view of the property landscape, useful for contextualizing energy consumption data.

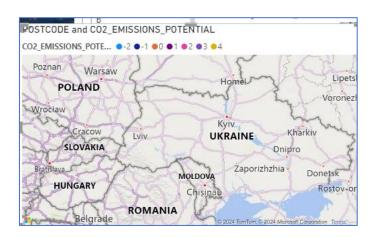


```
1 Count of Properties by Efficiency Level =
2 CALCULATE(
3 | COUNTROWS('Book1'),
4 | 'Book1'[CURRENT_ENERGY_EFFICIENCY] > 50 // Adjust thresholds as needed
5 )
```

3. **Trends in CO2 Emissions Over Time**: The line chart plots the average monthly CO2 emissions, revealing fluctuations over the course of a year. This visualization highlights the variability in emissions, which could correlate with seasonal changes, operational adjustments, or external factors affecting energy use.



4. **Map of Local Authority and Postcode**: The global map does not display specific data points but includes an indication of the local authority area (E08000003), suggesting a focus on regional analysis. This part of the dashboard could be utilized for geographically slicing the data to analyze energy and emissions data by specific areas.



5.Findings and Recommendations

Findings:

1. Property Type Energy Efficiency Variability:

• Energy consumption and efficiency levels vary significantly across different property types, with flats showing the highest energy usage

2. CO2 Emissions Insights:

Analysis of CO2 emissions reveals a pronounced seasonal variability, suggesting
higher emissions during colder months. There is also a noticeable reduction in
emissions over the years, reflecting the impact of existing energy efficiency
measures.

3. Geographical Disparities:

 The geographic distribution of CO2 emissions and energy efficiency ratings showcases notable differences between various local authorities. Some areas, particularly urban centers, show higher emissions, which could be targeted for specific policy interventions.

4. Opportunities for Improvement:

 There is a considerable potential for improvement in energy efficiency, especially in houses and park homes, as indicated by the gap between current and potential energy ratings.

Recommendations:

1. Targeted Retrofit Programs:

 Implement targeted retrofit programs for flats and houses to enhance their energy efficiency. This could include upgrading insulation, installing energy-efficient windows, and modern heating systems.

2. Seasonal Energy Efficiency Strategies:

 Develop seasonal energy efficiency strategies to address the peak in emissions during the colder months.

3. Local Authority Collaboration:

 Collaborate with local authorities to implement region-specific energy efficiency initiatives.

4. Enhanced Monitoring and Reporting:

 Strengthen monitoring and reporting mechanisms to track energy consumption and CO2 emissions more effectively.

0

Conclusion

| The analysis conducted energy consumption and for enhancing energy eareas with the highest eto improve energy efficito leverage data-driven term energy sustainability. | d CO2 emissions acros fficiency, particularly in energy consumption and ciency, reduce emission insights will be crucial | s Manchester. It high n residential properti d CO2 emissions, tar s, and move towards | nlighted significant op es. By focusing on the geted interventions ca greater sustainability | pportunities e identified in be designed . Continuing |
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