

# Data Engineering Activity

## Available Data

I have access to the following datasets:

- **dim\_Organisation:** Contains school records for various Academies within the same school group.
- **dim\_Student:** Includes core information about each student.
- **dim\_StudentExtended:** Provides additional details for each student.
- **fact\_AttendanceSession:** Records daily attendance for each student, divided into two sessions: AM and PM. A student's attendance percentage is calculated as the sum of `is_present` divided by the sum of `is_possible`.
- **dim\_Date:** A CSV file serving as a date dimension, with `FullDate` as the date column.

## Activity Brief

My task is to create summary analytics to assist the Data Analyst in building an Attendance Report. I will follow these steps:

1. **Data Loading**
  - I will create a Jupyter Notebook to load the parquet files and the CSV into DataFrames.
  - I will use an environment of my choice (e.g., Databricks Community Edition, VSCode with Jupyter Extension).
2. **Data Exploration**
  - I will determine the key relationships between the tables.
3. **Summary Table Creation**
  - I will produce a single summary table that includes the attendance percentage for each school on a weekly basis, categorized by the student's Year Group.
  - I will use my judgement to select the appropriate columns to include.
4. **Analytics**
  - I will print or display summary statistics about the data within the Notebook.
5. **Exporting Results**
  - I will write the summary table to `fact_AttendanceSummary` in Parquet format.

## Data Engineering Activity - Outline of Approach

I will use **PySpark**, given the large size of the dataset, which contains approximately 16 million rows. I will follow these steps to complete the task:

1. **Import Libraries:** Load the necessary PySpark libraries.
2. **Write Methods:** write methods of tasks which I will repeatedly use

3. **View the data:** Import the dataset into a PySpark DataFrame. Conduct an initial exploration to understand the structure, data types, and key attributes of the dataset.
4. **Join the data:** Merge the tables to create a single DataFrame containing all relevant information.
5. **Inspect Distinct Values:** Check the unique values in each column to identify any inconsistencies or missing values.
6. **Select Relevant Columns:** Choose the necessary columns for the analysis.
7. **Data Integrity Check:** Ensure data integrity by feature engineering a key to ensure each pupil has two entries for each day.
8. **Calculate Attendance Percentage:** Calculate the attendance percentage for each student based on the attendance records. Create a summary table containing the attendance percentage for each school on a weekly basis, grouped by the **Year Group** of the students.
9. **Investigate Null Values:** Analyse the distribution of null values in the dataset and decide on the appropriate handling strategy.
10. **Write the Summary Table:** Display key summary statistics in the Notebook, including metrics such as the mean, median, minimum, maximum, and standard deviation of attendance percentages. Export the summary table to `fact_AttendanceSummary` in **Parquet** format for further analysis and reporting.
11. **Notes for Data Analyst:** Provide additional notes or insights that may be useful for the Data Analyst when building the Attendance Report.

Each section is labelled accordingly with a markdown header for clarity and ease of navigation.

```
# SparkSession is the entry point to Spark SQL

from pyspark.sql import SparkSession

# Create a SparkSession and set memory for the driver
spark = SparkSession.builder \
    .appName("MyApp") \
    .config("spark.driver.memory", "8g") \
    .getOrCreate()

# Load CSV files into Spark DataFrames and infer schema to avoid specifying it manually
df_date_spark = spark.read \
    .option("header", True) \
    .option("inferSchema", True) \
    .csv("data/dim_Date.csv")

# Load parquet files into Spark DataFrames
df_attendancesessions_spark =
spark.read.parquet("data/fact_AttendanceSession")
df_organisation_spark = spark.read.parquet("data/dim_Organisation")
```

```
df_student_spark = spark.read.parquet("data/dim_Student")
df_studentextended_spark =
spark.read.parquet("data/dim_StudentExtended")
```

## 1. Import Libraries

```
from pyspark.sql import DataFrame
from pyspark.sql import functions as F
from pyspark.sql.functions import countDistinct
from functools import reduce
```

## 2. Methods

```
def show_df_missing_breakdown(df: DataFrame) -> None:
    """
    Prints:
    - The DataFrame size (rows, columns)
    - For each column:
      * number of NULLs
      * number of empty strings
      * number of 'NA' / 'NaN' (case-insensitive) as strings
      * number of numeric NaNs (for numeric columns)
      * total missing (sum of above)
      * percentage missing
    """
    total_rows = df.count()
    total_cols = len(df.columns)

    # Prepare expressions for counting different "missing" types for
    # each column
    agg_exprs = []
    for field in df.schema.fields:
        col_name = field.name
        # Check if column is numeric (so we can safely use F.isnan)
        is_numeric = field.dataType.typeName() in (
            "double", "float", "decimal",
            "integer", "long", "short", "byte"
        )

        # Count NULLs
        null_count_expr = F.sum(
            F.when(F.col(col_name).isNull(), 1).otherwise(0)
        ).alias(col_name + "_nullCount")

        # Count empty strings
        empty_count_expr = F.sum(
            F.when(F.col(col_name).cast("string") == "",
```

```

1).otherwise(0)
    ).alias(col_name + "_emptyCount")

    # Count string 'NA' or 'NaN' (case-insensitive)
    na_str_expr = F.sum(
        F.when(
            F.upper(F.col(col_name).cast("string")).isin("NA",
"NAN"),
            1
        ).otherwise(0)
    ).alias(col_name + "_naStrCount")

    # Count numeric NaN (only for numeric columns)
    if is_numeric:
        nan_numeric_expr = F.sum(
            F.when(F.isnan(F.col(col_name)), 1).otherwise(0)
        ).alias(col_name + "_nanNumericCount")
    else:
        # For non-numeric columns, this will always be 0
        nan_numeric_expr = F.lit(0).alias(col_name +
"_nanNumericCount")

    # Collect all expressions
    agg_exprs.extend([
        null_count_expr, empty_count_expr, na_str_expr,
nan_numeric_expr
    ])

    # Perform a single pass to get all missing counts
    agg_df = df.select(agg_exprs)
    result_row = agg_df.collect()[0].asDict() # single row with all
counts

    # Print header
    print(f"DataFrame has {total_rows} rows and {total_cols} columns.\n
n")
    print(
        "Column"
        "Null EmptyStr NA/NaNStr NumericNaN TotalMissing
%Missing"
    )
    print("-" * 70)

    # Loop over columns and print breakdown
    for field in df.schema.fields:
        c = field.name
        null_count = result_row[c + "_nullCount"]
        empty_count = result_row[c + "_emptyCount"]
        na_str_count = result_row[c + "_naStrCount"]
        nan_numeric_count = result_row[c + "_nanNumericCount"]

```

```

        total_missing = null_count + empty_count + na_str_count +
nan_numeric_count
        pct_missing = (total_missing / total_rows * 100) if total_rows
else 0.0

        print(
            f"{c:34s}"
            f"{null_count:5d}"
            f"{empty_count:10d}"
            f"{na_str_count:10d}"
            f"{nan_numeric_count:12d}"
            f"{total_missing:13d}"
            f"{pct_missing:10.2f}%"
        )

```

```

def show_distinct_counts(df: DataFrame, top_n: int = 20) -> None:
    """

```

*Displays the number of distinct values in each column of the DataFrame and lists the top\_n columns with the highest distinct counts.*

*Additionally, creates and displays a DataFrame containing all columns with their distinct counts.*

*Parameters:*

*df (DataFrame): The Spark DataFrame to analyze.*

*top\_n (int): The number of top columns to display based on distinct counts.*

```

    """
    # Calculate distinct counts for each column
    distinct_counts = df.agg(*[countDistinct(c).alias(c) for c in
df.columns]).collect()[0].asDict()

```

```

    # Sort columns by distinct count in descending order
    sorted_counts = sorted(distinct_counts.items(), key=lambda x:
x[1], reverse=True)

```

*# Display the top\_n columns*

```

print(f"{'Column':34s} {'Distinct Count'}")

```

```

print("-" * 50)

```

```

for col, cnt in sorted_counts[:top_n]:

```

```

    print(f"{col:34s} {cnt}")

```

*# Create a DataFrame of all distinct counts*

```

df_distinct_counts = spark.createDataFrame(sorted_counts,
["Column", "Distinct_Count"])

```

*# Show the DataFrame of distinct counts*

```

print("\nAll Column Distinct Counts:")
df_distinct_counts.show(truncate=False)

from pyspark.sql import SparkSession, DataFrame, functions as F

def show_distinct_counts_approx(df: DataFrame, top_n: int = 20, rsd:
float = 0.05) -> None:
    """
        Displays the approximate number of distinct values in each column
of the DataFrame
        and lists the top_n columns with the highest distinct counts.

        Additionally, creates and displays a DataFrame containing all
columns with
        their approximate distinct counts, but only shows the top_n rows
to reduce
        the chance of memory/network issues.

        Parameters:
        -----
        df : DataFrame
            The Spark DataFrame to analyze.
        top_n : int
            The number of top columns to display based on distinct counts.
        rsd : float
            Relative Standard Deviation for approx_count_distinct.
            Lower = more accurate but more memory usage. Typical default
is 0.05.
    """

    # Build a list of approx_count_distinct expressions for each
column
    approx_exprs = [
        F.approx_count_distinct(F.col(c), rsd=rsd).alias(c)
        for c in df.columns
    ]

    # Collect the single row of approximate distinct counts as a dict
    # e.g. {'colA': 123, 'colB': 999, ...}
    approx_counts_row = df.agg(*approx_exprs).collect()[0].asDict()

    # Convert that dict into a list of (column, distinct_count) tuples
and sort
    sorted_counts = sorted(approx_counts_row.items(), key=lambda x:
x[1], reverse=True)

    # Print header
    print(f"{'Column':34s} {'Approx Distinct Count'}")
    print("-" * 60)

```

```

# Show only the top_n columns in console
for col_name, cnt in sorted_counts[:top_n]:
    print(f"{col_name:34s} {cnt}")

# Create a small DataFrame from the sorted counts
# Each row: (column_name, approx_distinct_count)
spark = SparkSession.builder.getOrCreate()
df_approx_counts = spark.createDataFrame(
    sorted_counts, ["Column", "ApproxDistinctCount"]
)

# Show only the top_n rows, so we don't blow up memory
print("\nAll Column Approx Distinct Counts (showing top_n only):")
df_approx_counts.limit(top_n).show(truncate=False)

```

### 3. View the Data

```
# Show the first 20 rows of the DataFrames
```

```
df_studentextended_spark.show()
```

```

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|      Ethnicity|Ethnicity_Code|Ever_In_Care|First_Language|
Free_School_Meals|Free_School_Meals_6|Gifted_And_Talented_Status|
In_LEA_Care|Pupil_Premium_Indicator|SEN_Status|
English_As_Additional_Language|English_As_Additional_Language_Status|
Child_In_Need|Child_Protection_Plan|Enrolment_Status|
studentextendedkey|Year_Group|Current_NC_Year|Admission_Date|
Leaving_Date|Is_Current|Postcode|organisationkey|
studentkey|partitionkey|
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|White - British|WBRI|False|Twil|
False|False|False|False|
False|K|True|
None|None|None|SINGLE REGISTRATION|
698f7ea5-b203-4b7...|8|8|2023-09-04 00:00:...|
NULL|1|02ef2e04-5a06-4f1...|0002c6c1-11bd-4a4...|

```

02ef2e04-5a06-4f1...									
	None	None	False	Czech					
False		False		None	False				
False	None			None					
None	None		None	SINGLE REGISTRATION					
2a45058a-19e6-456...		NULL		11 2018-09-03 00:00:...					
2023-07-21 00:00:...		0	NULL	02ef2e04-5a06-4f1...					
002e1bf1-f48f-4a8...	02ef2e04-5a06-4f1...								
	None	None	False	Chuvash					
True		True		None	False				
False	E			None					
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9eb6ff76-8423-433...		NULL		11 2022-03-11 00:00:...					
2023-07-21 00:00:...		0	NULL	02ef2e04-5a06-4f1...					
0050d3ad-83b5-423...	02ef2e04-5a06-4f1...								
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f22766a0-7879-443...		11		11 2020-09-02 00:00:...					
NULL	1	02ef2e04-5a06-4f1...	005a306c-5498-4e9...						
02ef2e04-5a06-4f1...									
	None	None	False	Kashmiri					
True		True		None	False				
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None	None		None	SINGLE REGISTRATION					
7b6cec4f-d8eb-41d...		NULL		7 2022-09-01 00:00:...					
2023-07-21 00:00:...		0	NULL	02ef2e04-5a06-4f1...					
00c5ecab-5abd-44d...	02ef2e04-5a06-4f1...								
	None	None	False	Sinhala					
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None	None		None	SINGLE REGISTRATION					
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014259e3-3a38-47b...	02ef2e04-5a06-4f1...								
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b347906d-0ab0-4d6...		NULL		11 2018-09-05 00:00:...					
2023-07-21 00:00:...		0	NULL	02ef2e04-5a06-4f1...					
01537764-8020-48b...	02ef2e04-5a06-4f1...								
White - British	WBRI		False	Polish					
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7f2a01a0-b022-400...		10		10 2021-09-02 00:00:...					
NULL	1	02ef2e04-5a06-4f1...	0161e596-cb57-496...						



02ef2e04-5a06-4f1...					
	None	None	False	Marshallese	
True		True		None	False
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None	None		None	SINGLE REGISTRATION	
14bbf585-14b4-467...		NULL		13	2022-09-12 00:00:...
2024-07-19 00:00:...		0	NULL	02ef2e04-5a06-4f1...	
018ca9b0-f733-499...	02ef2e04-5a06-4f1...				
	None	None	False	Swahili	
False		False		None	False
False	None			None	
None	None		None	SINGLE REGISTRATION	
12496112-264c-4f9...		NULL		10	2022-07-13 00:00:...
2023-02-02 00:00:...		0	NULL	02ef2e04-5a06-4f1...	
01ce8216-c473-493...	02ef2e04-5a06-4f1...				
White - British		WBRI	False	Czech	
False		False		False	False
False	None			False	
None	None		None	SINGLE REGISTRATION	
d1addc33-37b6-471...		12		12	2019-09-04 00:00:...
NULL	1		02ef2e04-5a06-4f1...	020f522f-9124-439...	
02ef2e04-5a06-4f1...					
White - British		WBRI	False	Komi	
False		False		False	False
False	None			False	
None	None		None	SINGLE REGISTRATION	
c7b32f84-eb86-442...		8		8	2023-09-04 00:00:...
NULL	1		02ef2e04-5a06-4f1...	02fed00e-ea3b-410...	
02ef2e04-5a06-4f1...					
White - British		WBRI	False	Avestan	
False		False		False	False
False	None			False	
None	None		None	SINGLE REGISTRATION	
3c05c911-6761-40a...		11		11	2020-09-02 00:00:...
NULL	1		02ef2e04-5a06-4f1...	03072296-5767-471...	
02ef2e04-5a06-4f1...					
White - British		WBRI	False	Afrikaans	
False		False		False	False
False	N			False	
None	None		None	SINGLE REGISTRATION	
d427f831-0661-466...		8		8	2023-09-04 00:00:...
NULL	1		02ef2e04-5a06-4f1...	03205882-edb6-404...	
02ef2e04-5a06-4f1...					
White - British		WBRI	False	Ndonga	
False		False		False	False
False	None			None	
None	None		None	SINGLE REGISTRATION	
cef8427a-0678-48f...		7		7	2024-09-03 00:00:...
NULL	1		02ef2e04-5a06-4f1...	03ec0a03-0777-4e0...	

```

02ef2e04-5a06-4f1...|
|White - British|      WBRI|      False|      Shona|
True|      True|      False|      False|
True|      K|      None|
None|      None|      None| SINGLE REGISTRATION|
fe1e9b6d-27cc-439...|      7|      7|2024-09-03 00:00:...|
NULL|      1|      |02ef2e04-5a06-4f1...|03ee9045-6362-466...|
02ef2e04-5a06-4f1...|
|White - British|      WBRI|      False|      Kanuri|
True|      True|      False|      False|
True|      None|      None|
None|      None|      None| SINGLE REGISTRATION|
f3a46beb-f099-424...|      7|      7|2024-09-03 00:00:...|
NULL|      1|      |02ef2e04-5a06-4f1...|04514d6d-2d2b-4c5...|
02ef2e04-5a06-4f1...|
|      None|      None|      False|      Luba-Katanga|
False|      False|      None|      False|
False|      None|      None|
None|      None|      None| SINGLE REGISTRATION|
d507fd59-8d58-4b3...|      NULL|      11|2020-09-07 00:00:...|
2024-07-19 00:00:...|      0|      NULL|02ef2e04-5a06-4f1...|
0473fd58-7fca-4f6...|02ef2e04-5a06-4f1...|
|      None|      None|      False|      Kurdish|
False|      False|      None|      False|
False|      None|      None|
None|      None|      None| SINGLE REGISTRATION|
53add9d6-6094-480...|      NULL|      11|2019-09-04 00:00:...|
2024-07-19 00:00:...|      0|      NULL|02ef2e04-5a06-4f1...|
047976ba-9ab4-467...|02ef2e04-5a06-4f1...|
|      None|      None|      False|      Urdu|
False|      False|      None|      False|
False|      None|      None|
None|      None|      None| SINGLE REGISTRATION|
46bdc252-b71c-449...|      NULL|      13|2016-09-05 00:00:...|
2023-07-21 00:00:...|      0|      NULL|02ef2e04-5a06-4f1...|
047c84da-d860-47e...|02ef2e04-5a06-4f1...|
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+-----+-----+-----+-----+
only showing top 20 rows

```

I will use the missing value method defined earlier to provide information on size and missing values in the data.

```
show_df_missing_breakdown(df_studentextended_spark)
```

DataFrame has 16937 rows and 25 columns.

Column			Null	EmptyStr	NA/NaNStr
NumericNaN	TotalMissing	%Missing			
-----					
Ethnicity			0	0	0
0	0	0.00%			
Ethnicity_Code			0	0	0
0	0	0.00%			
Ever_In_Care			0	0	0
0	0	0.00%			
First_Language			0	0	0
0	0	0.00%			
Free_School_Meals			0	0	0
0	0	0.00%			
Free_School_Meals_6			0	0	0
0	0	0.00%			
Gifted_And_Talented_Status			0	0	0
0	0	0.00%			
In_LEA_Care			0	0	0
0	0	0.00%			
Pupil_Premium_Indicator			0	0	0
0	0	0.00%			
SEN_Status			0	0	0
0	0	0.00%			
English_As_Additional_Language			0	0	0
0	0	0.00%			
English_As_Additional_Language_Status			0	0	0
0	0	0.00%			
Child_In_Need			0	0	0
0	0	0.00%			
Child_Protection_Plan			0	0	0
0	0	0.00%			
Enrolment_Status			0	0	0
0	0	0.00%			
studentextendedkey			0	0	0
0	0	0.00%			
Year_Group			5637	0	0
0	5637	33.28%			
Current_NC_Year			0	0	0
0	0	0.00%			
Admission_Date			0	0	0
0	0	0.00%			
Leaving_Date			7699	0	3475
0	11174	65.97%			
Is_Current			0	0	0
0	0	0.00%			
Postcode			5759	11178	0

0	16937	100.00%			
organisationkey			0	0	0
0	0	0.00%			
studentkey			0	0	0
0	0	0.00%			
partitionkey			0	0	0
0	0	0.00%			

Initial observations:

- Three keys are available in the data: student\_id, organisation\_id, date
- Dataframe df includes students who are current and have left the school
- National curriculum year (age based) and year group are available - are NULL if the student has left the school

```
df_student_spark.show()
```

```
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+
|Forename|Legal_Forename|Legal_Surname|Surname|Middle_Names|Sex|
Gender|Date_Of_Birth|organisationkey|studentkey|UPN|
partitionkey|
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+
|Deborah|XXXXXX|XXXXXX|Wong|XXXXXX|FEMALE|
None|2000-01-01|02ef2e04-5a06-4f1...|0002c6c1-11bd-4a4...|0|
02ef2e04-5a06-4f1...|
|Michelle|XXXXXX|XXXXXX|Glenn|XXXXXX|FEMALE|
None|2000-01-01|02ef2e04-5a06-4f1...|002e1bf1-f48f-4a8...|7|
02ef2e04-5a06-4f1...|
|Alyssa|XXXXXX|XXXXXX|Schmidt|XXXXXX|MALE|
None|2000-01-01|02ef2e04-5a06-4f1...|0050d3ad-83b5-423...|14|
02ef2e04-5a06-4f1...|
|Juan|XXXXXX|XXXXXX|Ramirez|XXXXXX|FEMALE|
None|2000-01-01|02ef2e04-5a06-4f1...|005a306c-5498-4e9...|15|
02ef2e04-5a06-4f1...|
|Karen|XXXXXX|XXXXXX|Goodwin|XXXXXX|FEMALE|
None|2000-01-01|02ef2e04-5a06-4f1...|00c5ecab-5abd-44d...|25|
02ef2e04-5a06-4f1...|
|Stacey|XXXXXX|XXXXXX|Potts|XXXXXX|FEMALE|
None|2000-01-01|02ef2e04-5a06-4f1...|013b3548-b4a8-430...|45|
02ef2e04-5a06-4f1...|
|Karen|XXXXXX|XXXXXX|Mullins|XXXXXX|FEMALE|
None|2000-01-01|02ef2e04-5a06-4f1...|014259e3-3a38-47b...|47|
02ef2e04-5a06-4f1...|
|Ronald|XXXXXX|XXXXXX|Simmons|XXXXXX|MALE|
None|2000-01-01|02ef2e04-5a06-4f1...|01537764-8020-48b...|51|
02ef2e04-5a06-4f1...|
```

Dwayne	XXXXXX	XXXXXX	Thomas	XXXXXX	FEMALE
None	2000-01-01	02ef2e04-5a06-4f1...	0161e596-cb57-496...	56	
Alisha	XXXXXX	XXXXXX	Fox	XXXXXX	FEMALE
None	2000-01-01	02ef2e04-5a06-4f1...	018ca9b0-f733-499...	61	
Yvonne	XXXXXX	XXXXXX	Key	XXXXXX	MALE
None	2000-01-01	02ef2e04-5a06-4f1...	01ce8216-c473-493...	72	
Jared	XXXXXX	XXXXXX	Clark	XXXXXX	MALE
None	2000-01-01	02ef2e04-5a06-4f1...	020f522f-9124-439...	87	
Andrew	XXXXXX	XXXXXX	Johnson	XXXXXX	FEMALE
None	2000-01-01	02ef2e04-5a06-4f1...	02fed00e-ea3b-410...	122	
Denise	XXXXXX	XXXXXX	Pearson	XXXXXX	MALE
None	2000-01-01	02ef2e04-5a06-4f1...	03072296-5767-471...	123	
Jacob	XXXXXX	XXXXXX	Lopez	XXXXXX	FEMALE
None	2000-01-01	02ef2e04-5a06-4f1...	03205882-edb6-404...	127	
Austin	XXXXXX	XXXXXX	Johnson	XXXXXX	MALE
None	2000-01-01	02ef2e04-5a06-4f1...	03ec0a03-0777-4e0...	152	
Deborah	XXXXXX	XXXXXX	Lucas	XXXXXX	FEMALE
None	2000-01-01	02ef2e04-5a06-4f1...	03ee9045-6362-466...	153	
Caleb	XXXXXX	XXXXXX	Brown	XXXXXX	MALE
None	2000-01-01	02ef2e04-5a06-4f1...	044b01be-6766-407...	165	
Alicia	XXXXXX	XXXXXX	Fox	XXXXXX	MALE
None	2000-01-01	02ef2e04-5a06-4f1...	04514d6d-2d2b-4c5...	166	
Sarah	XXXXXX	XXXXXX	Hickman	XXXXXX	MALE
None	2000-01-01	02ef2e04-5a06-4f1...	0473fd58-7fca-4f6...	169	

```

+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+
+-----+
only showing top 20 rows

```

```
show_df_missing_breakdown(df_student_spark)
```

```
DataFrame has 16937 rows and 12 columns.
```

Column	Null	EmptyStr	NA/NaNStr
NumericNaN	TotalMissing	%Missing	
Forename	0	0	0

0	0	0.00%			
Legal_Forename			0	0	0
0	0	0.00%			
Legal_Surname			0	0	0
0	0	0.00%			
Surname			0	0	0
0	0	0.00%			
Middle_Names			0	0	0
0	0	0.00%			
Sex			0	0	0
0	0	0.00%			
Gender			0	0	0
0	0	0.00%			
Date_Of_Birth			0	0	0
0	0	0.00%			
organisationkey			0	0	0
0	0	0.00%			
studentkey			0	0	0
0	0	0.00%			
UPN			0	0	0
0	0	0.00%			
partitionkey			0	0	0
0	0	0.00%			

Initial observations:

- XXXXX used to keep data confidential -
- gender and sex are the same - Gender can be dropped as it contains XXXXX
- Same with legal\_forename and forename, legal\_surname and surname

```
df_organisation_spark.show()
```

```
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
|Organisation_Name|Establishment_Number|LA_Code|Organisation_Type|
organisationkey|addresskey|UKPRN|Organisation_Status|last_updated|URN|
partitionkey|
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
|      Academy 4|      XXXXXX| XXXXXX|      SECONDARY|
02ef2e04-5a06-4f1...|      |      |      Active|
| 3|02ef2e04-5a06-4f1...|
|      Academy 9|      XXXXXX| XXXXXX|      ALL THROUGH|
068cf4c6-2526-430...|      |      |      Active|
| 8|068cf4c6-2526-430...|
|      Academy 7|      XXXXXX| XXXXXX|      SECONDARY|
2d9ba2ce-d6e9-49b...|      |      |      Active|
| 6|2d9ba2ce-d6e9-49b...|
```

Academy 2	XXXXXX	XXXXXX	SECONDARY
5ce8e936-d2c9-407...			Active
1 5ce8e936-d2c9-407...			
Academy 5	XXXXXX	XXXXXX	SECONDARY
5ed661db-d767-4fd...			Active
4 5ed661db-d767-4fd...			
Academy 3	XXXXXX	XXXXXX	SECONDARY
6942470f-ce34-415...			Active
2 6942470f-ce34-415...			
Academy 6	XXXXXX	XXXXXX	SECONDARY
73ed72bc-efea-4cd...			Active
5 73ed72bc-efea-4cd...			
Academy 1	XXXXXX	XXXXXX	SECONDARY
f7d73a58-e94b-450...			Active
0 f7d73a58-e94b-450...			
Academy 8	XXXXXX	XXXXXX	SECONDARY
f87939c4-fbdb-47b...			Active
7 f87939c4-fbdb-47b...			

```

+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+

```

```
show_df_missing_breakdown(df_organisation_spark)
```

DataFrame has 9 rows and 11 columns.

Column	Null	EmptyStr	NA/NaNStr
NumericNaN TotalMissing %Missing			
-----			
Organisation_Name	0	0	0
0 0 0.00%			
Establishment_Number	0	0	0
0 0 0.00%			
LA_Code	0	0	0
0 0 0.00%			
Organisation_Type	0	0	0
0 0 0.00%			
organisationkey	0	0	0
0 0 0.00%			
addresskey	0	9	0
0 9 100.00%			
UKPRN	0	9	0
0 9 100.00%			
Organisation_Status	0	0	0
0 0 0.00%			
last_updated	0	9	0
0 9 100.00%			
URN	0	0	0
0 0 0.00%			

partitionkey			0	0	0
0	0	0.00%			

Initial observations:

- Assumption XXXX used to keep data confidential - e.g. LA Code; these columns can be kept for use of the script with other data

```
df_attendancesessions_spark.show()
```

```
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+
|      Date|Mark|Session|attendancesessionkey|is_aea|is_attend|
is_auth_abs|is_late_L|is_late_U|is_missing|is_nr|is_possible|
is_present|is_unauth_abs|      organisationkey|      studentkey|
partitionkey|
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+
|2023-11-13|  / |      AM|162ff625-c9d6-49e...|  0.0|      1.0|
0.0|      0.0|      0.0|      0.0|  0.0|      1.0|      1.0|
0.0|02ef2e04-5a06-4f1...|52b4e8e3-4481-44e...|02ef2e04-5a06-4f1...|
|2023-09-19|  \ |      PM|e42a0825-379b-449...|  0.0|      1.0|
0.0|      0.0|      0.0|      0.0|  0.0|      1.0|      1.0|
0.0|02ef2e04-5a06-4f1...|70b88e12-aeef-421...|02ef2e04-5a06-4f1...|
|2024-09-12|  / |      AM|f57c3e6b-bdf9-4fc...|  0.0|      1.0|
0.0|      0.0|      0.0|      0.0|  0.0|      1.0|      1.0|
0.0|02ef2e04-5a06-4f1...|04514d6d-2d2b-4c5...|02ef2e04-5a06-4f1...|
|2024-08-11|  # |      PM|26b1d017-8aad-4c8...|  0.0|      0.0|
0.0|      0.0|      0.0|      0.0|  1.0|      0.0|      0.0|
0.0|02ef2e04-5a06-4f1...|1cd5b534-6d90-4c9...|02ef2e04-5a06-4f1...|
|2024-11-15|  \ |      PM|fd9bfcd2-1ece-4af...|  0.0|      1.0|
0.0|      0.0|      0.0|      0.0|  0.0|      1.0|      1.0|
0.0|02ef2e04-5a06-4f1...|e0c0c52d-2fed-41f...|02ef2e04-5a06-4f1...|
|2024-10-09|  G |      PM|c47eb79a-25d6-4f6...|  0.0|      0.0|
0.0|      0.0|      0.0|      0.0|  0.0|      1.0|      0.0|
1.0|02ef2e04-5a06-4f1...|a7566208-63f6-47d...|02ef2e04-5a06-4f1...|
|2024-09-29|  # |      AM|279194f7-a9f3-466...|  0.0|      0.0|
0.0|      0.0|      0.0|      0.0|  1.0|      0.0|      0.0|
0.0|02ef2e04-5a06-4f1...|a7566208-63f6-47d...|02ef2e04-5a06-4f1...|
|2024-08-21|  # |      PM|7407a9bd-b3c9-442...|  0.0|      0.0|
0.0|      0.0|      0.0|      0.0|  1.0|      0.0|      0.0|
0.0|02ef2e04-5a06-4f1...|52b4e8e3-4481-44e...|02ef2e04-5a06-4f1...|
|2024-11-09|  # |      PM|419a11be-d200-487...|  0.0|      0.0|
0.0|      0.0|      0.0|      0.0|  1.0|      0.0|      0.0|
0.0|02ef2e04-5a06-4f1...|e0c0c52d-2fed-41f...|02ef2e04-5a06-4f1...|
|2024-08-27|  # |      PM|6e41dcf5-de47-460...|  0.0|      0.0|
```



```

0.0|      0.0|      0.0|      0.0|  1.0|      0.0|      0.0|
0.0|02ef2e04-5a06-4f1...|70b88e12-aeef-421...|02ef2e04-5a06-4f1...|
|2024-11-27|  \ |      PM|0fcbbd41-6da8-417...|  0.0|      1.0|
0.0|      0.0|      0.0|      0.0|  0.0|      1.0|      1.0|
0.0|02ef2e04-5a06-4f1...|70b88e12-aeef-421...|02ef2e04-5a06-4f1...|
|2024-11-05|  \ |      PM|4e4ed7f7-4e45-40e...|  0.0|      1.0|
0.0|      0.0|      0.0|      0.0|  0.0|      1.0|      1.0|
0.0|02ef2e04-5a06-4f1...|52b4e8e3-4481-44e...|02ef2e04-5a06-4f1...|
|2024-10-20|  # |      PM|4b7a62a1-a080-40e...|  0.0|      0.0|
0.0|      0.0|      0.0|      0.0|  1.0|      0.0|      0.0|
0.0|02ef2e04-5a06-4f1...|a7566208-63f6-47d...|02ef2e04-5a06-4f1...|
|2024-12-01|  # |      PM|f1a2d7b2-fd92-4a8...|  0.0|      0.0|
0.0|      0.0|      0.0|      0.0|  1.0|      0.0|      0.0|
0.0|02ef2e04-5a06-4f1...|70b88e12-aeef-421...|02ef2e04-5a06-4f1...|
|2024-12-05|  \ |      PM|4eb4fa43-c672-49b...|  0.0|      1.0|
0.0|      0.0|      0.0|      0.0|  0.0|      1.0|      1.0|
0.0|02ef2e04-5a06-4f1...|52b4e8e3-4481-44e...|02ef2e04-5a06-4f1...|
|2024-11-21|  / |      AM|3c0247c1-6fab-46b...|  0.0|      1.0|
0.0|      0.0|      0.0|      0.0|  0.0|      1.0|      1.0|
0.0|02ef2e04-5a06-4f1...|70b88e12-aeef-421...|02ef2e04-5a06-4f1...|
|2024-10-09|  \ |      PM|4dcce057-d194-4ad...|  0.0|      1.0|
0.0|      0.0|      0.0|      0.0|  0.0|      1.0|      1.0|
0.0|02ef2e04-5a06-4f1...|52b4e8e3-4481-44e...|02ef2e04-5a06-4f1...|
|2024-10-23|  # |      PM|c5f0ffc9-4389-4b9...|  0.0|      0.0|
0.0|      0.0|      0.0|      0.0|  1.0|      0.0|      0.0|
0.0|02ef2e04-5a06-4f1...|1cd5b534-6d90-4c9...|02ef2e04-5a06-4f1...|
|2024-11-07|  / |      AM|9aa4ab25-0db6-43d...|  0.0|      1.0|
0.0|      0.0|      0.0|      0.0|  0.0|      1.0|      1.0|
0.0|02ef2e04-5a06-4f1...|52b4e8e3-4481-44e...|02ef2e04-5a06-4f1...|
|2024-12-04|  \ |      PM|01baeddc-5bc6-48d...|  0.0|      1.0|
0.0|      0.0|      0.0|      0.0|  0.0|      1.0|      1.0|
0.0|02ef2e04-5a06-4f1...|1cd5b534-6d90-4c9...|02ef2e04-5a06-4f1...|

```

```

+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+

```

only showing top 20 rows

```
show_df_missing_breakdown(df_attendancesessions_spark)
```

DataFrame has 16311626 rows and 17 columns.

Column			Null	EmptyStr	NA/NaNStr
NumericNaN	TotalMissing	%Missing			
-----					
Date			0	0	0
0	0	0.00%			
Mark			0	0	0
0	0	0.00%			

Session			0	0	0
0	0	0.00%			
attendancesessionkey			0	0	0
0	0	0.00%			
is_aea			506	0	0
0	506	0.00%			
is_attend			506	0	0
0	506	0.00%			
is_auth_abs			506	0	0
0	506	0.00%			
is_late_L			506	0	0
0	506	0.00%			
is_late_U			506	0	0
0	506	0.00%			
is_missing			506	0	0
0	506	0.00%			
is_nr			506	0	0
0	506	0.00%			
is_possible			506	0	0
0	506	0.00%			
is_present			506	0	0
0	506	0.00%			
is_unauth_abs			506	0	0
0	506	0.00%			
organisationkey			0	0	0
0	0	0.00%			
studentkey			0	0	0
0	0	0.00%			
partitionkey			0	0	0
0	0	0.00%			

Initial observations:

- Df has 16,000,000 rows!
- 3 keys: student\_id, organisation\_id, date
- date can be used to join with the date dimension - via new column datekey

```
from pyspark.sql import functions as F

#Creat a new column datekey in the df_attendancesessions_spark which
will take the value of the Date column without the "-" character.
#This can act as key to join the df_attendancesessions_spark with the
df_date_spark

df_attendancesessions_spark = df_attendancesessions_spark.withColumn(
    "datekey",
    F.regexp_replace("Date", "-", "").cast("int")
)
```

```
df_attendancesessions_spark.show() #check the new column datekey
```

```
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+
|      Date|Mark|Session|attendancesessionkey|is_aea|is_attend|
is_auth_abs|is_late_L|is_late_U|is_missing|is_nr|is_possible|
is_present|is_unauth_abs|      organisationkey|      studentkey|
partitionkey| datekey|
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+
|2023-11-13|  / |      AM|162ff625-c9d6-49e...|  0.0|      1.0|
0.0|      0.0|      0.0|      0.0|  0.0|      1.0|      1.0|
0.0|02ef2e04-5a06-4f1...|52b4e8e3-4481-44e...|02ef2e04-5a06-4f1...|
20231113|
|2023-09-19|  \ |      PM|e42a0825-379b-449...|  0.0|      1.0|
0.0|      0.0|      0.0|      0.0|  0.0|      1.0|      1.0|
0.0|02ef2e04-5a06-4f1...|70b88e12-aeef-421...|02ef2e04-5a06-4f1...|
20230919|
|2024-09-12|  / |      AM|f57c3e6b-bdf9-4fc...|  0.0|      1.0|
0.0|      0.0|      0.0|      0.0|  0.0|      1.0|      1.0|
0.0|02ef2e04-5a06-4f1...|04514d6d-2d2b-4c5...|02ef2e04-5a06-4f1...|
20240912|
|2024-08-11|  # |      PM|26b1d017-8aad-4c8...|  0.0|      0.0|
0.0|      0.0|      0.0|      0.0|  1.0|      0.0|      0.0|
0.0|02ef2e04-5a06-4f1...|1cd5b534-6d90-4c9...|02ef2e04-5a06-4f1...|
20240811|
|2024-11-15|  \ |      PM|fd9bfcd2-1ece-4af...|  0.0|      1.0|
0.0|      0.0|      0.0|      0.0|  0.0|      1.0|      1.0|
0.0|02ef2e04-5a06-4f1...|e0c0c52d-2fed-41f...|02ef2e04-5a06-4f1...|
20241115|
|2024-10-09|  G |      PM|c47eb79a-25d6-4f6...|  0.0|      0.0|
0.0|      0.0|      0.0|      0.0|  0.0|      1.0|      0.0|
1.0|02ef2e04-5a06-4f1...|a7566208-63f6-47d...|02ef2e04-5a06-4f1...|
20241009|
|2024-09-29|  # |      AM|279194f7-a9f3-466...|  0.0|      0.0|
0.0|      0.0|      0.0|      0.0|  1.0|      0.0|      0.0|
0.0|02ef2e04-5a06-4f1...|a7566208-63f6-47d...|02ef2e04-5a06-4f1...|
20240929|
|2024-08-21|  # |      PM|7407a9bd-b3c9-442...|  0.0|      0.0|
0.0|      0.0|      0.0|      0.0|  1.0|      0.0|      0.0|
0.0|02ef2e04-5a06-4f1...|52b4e8e3-4481-44e...|02ef2e04-5a06-4f1...|
20240821|
|2024-11-09|  # |      PM|419a11be-d200-487...|  0.0|      0.0|
0.0|      0.0|      0.0|      0.0|  1.0|      0.0|      0.0|
0.0|02ef2e04-5a06-4f1...|e0c0c52d-2fed-41f...|02ef2e04-5a06-4f1...|
```

```
20241109|
|2024-08-27|    #|    PM|6e41dcf5-de47-460...|    0.0|    0.0|
0.0|    0.0|    0.0|    0.0|    1.0|    0.0|    0.0|
0.0|02ef2e04-5a06-4f1...|70b88e12-aeef-421...|02ef2e04-5a06-4f1...|
20240827|
|2024-11-27|    \|    PM|0fcbbd41-6da8-417...|    0.0|    1.0|
0.0|    0.0|    0.0|    0.0|    0.0|    1.0|    1.0|
0.0|02ef2e04-5a06-4f1...|70b88e12-aeef-421...|02ef2e04-5a06-4f1...|
20241127|
|2024-11-05|    \|    PM|4e4ed7f7-4e45-40e...|    0.0|    1.0|
0.0|    0.0|    0.0|    0.0|    0.0|    1.0|    1.0|
0.0|02ef2e04-5a06-4f1...|52b4e8e3-4481-44e...|02ef2e04-5a06-4f1...|
20241105|
|2024-10-20|    #|    PM|4b7a62a1-a080-40e...|    0.0|    0.0|
0.0|    0.0|    0.0|    0.0|    1.0|    0.0|    0.0|
0.0|02ef2e04-5a06-4f1...|a7566208-63f6-47d...|02ef2e04-5a06-4f1...|
20241020|
|2024-12-01|    #|    PM|f1a2d7b2-fd92-4a8...|    0.0|    0.0|
0.0|    0.0|    0.0|    0.0|    1.0|    0.0|    0.0|
0.0|02ef2e04-5a06-4f1...|70b88e12-aeef-421...|02ef2e04-5a06-4f1...|
20241201|
|2024-12-05|    \|    PM|4eb4fa43-c672-49b...|    0.0|    1.0|
0.0|    0.0|    0.0|    0.0|    0.0|    1.0|    1.0|
0.0|02ef2e04-5a06-4f1...|52b4e8e3-4481-44e...|02ef2e04-5a06-4f1...|
20241205|
|2024-11-21|    /|    AM|3c0247c1-6fab-46b...|    0.0|    1.0|
0.0|    0.0|    0.0|    0.0|    0.0|    1.0|    1.0|
0.0|02ef2e04-5a06-4f1...|70b88e12-aeef-421...|02ef2e04-5a06-4f1...|
20241121|
|2024-10-09|    \|    PM|4dcce057-d194-4ad...|    0.0|    1.0|
0.0|    0.0|    0.0|    0.0|    0.0|    1.0|    1.0|
0.0|02ef2e04-5a06-4f1...|52b4e8e3-4481-44e...|02ef2e04-5a06-4f1...|
20241009|
|2024-10-23|    #|    PM|c5f0ffc9-4389-4b9...|    0.0|    0.0|
0.0|    0.0|    0.0|    0.0|    1.0|    0.0|    0.0|
0.0|02ef2e04-5a06-4f1...|1cd5b534-6d90-4c9...|02ef2e04-5a06-4f1...|
20241023|
|2024-11-07|    /|    AM|9aa4ab25-0db6-43d...|    0.0|    1.0|
0.0|    0.0|    0.0|    0.0|    0.0|    1.0|    1.0|
0.0|02ef2e04-5a06-4f1...|52b4e8e3-4481-44e...|02ef2e04-5a06-4f1...|
20241107|
|2024-12-04|    \|    PM|01baeddc-5bc6-48d...|    0.0|    1.0|
0.0|    0.0|    0.0|    0.0|    0.0|    1.0|    1.0|
0.0|02ef2e04-5a06-4f1...|1cd5b534-6d90-4c9...|02ef2e04-5a06-4f1...|
20241204|
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
```

```
df_date_spark.show() #check the df_date_spark
```

[illegible]

```

+-----+-----+-----+
+-----+
|19400101|01/01/1940|1|1|
1940W01|1|1|
1|1|1|
1|14610|123|1|
1|2|January|Jan|Monday|
Mon|1|1939/1940|5|
19|Term 2|1940|1940-01-01 00:00:00|1940-01-01 00:00:00|
1|1|1940|5|
2|1940-05-01 00:00:00|1940Q02|161|19400101|01/01/1940|
01/01/1940|01/01/1940|Jan 01 1940|0|N|
Y|N|Y|N|N|
1939/1940-05|01/01/1940|01/01/1940|w/c
01/01/1940|23281|122|
|19400102|02/01/1940|1|1|
1940W01|1|1|
1|1|1|
2|14611|124|2|
2|3|January|Jan|Tuesday|
Tue|2|1939/1940|5|
19|Term 2|1940|1940-01-01 00:00:00|1940-01-01 00:00:00|
1|1|1940|5|
2|1940-05-01 00:00:00|1940Q02|161|19400102|01/02/1940|
02/01/1940|02/01/1940|Jan 02 1940|0|N|
Y|N|Y|N|N|
1939/1940-05|01/01/1940|01/01/1940|w/c
01/01/1940|23281|123|
|19400103|03/01/1940|1|1|
1940W01|1|1|
1|1|1|
3|14612|125|3|
3|4|January|Jan|Wednesday|
Wed|3|1939/1940|5|
19|Term 2|1940|1940-01-01 00:00:00|1940-01-01 00:00:00|
1|1|1940|5|
2|1940-05-01 00:00:00|1940Q02|161|19400103|01/03/1940|
03/01/1940|03/01/1940|Jan 03 1940|0|N|
Y|N|Y|N|N|
1939/1940-05|01/01/1940|01/01/1940|w/c
01/01/1940|23281|124|
|19400104|04/01/1940|1|1|
1940W01|1|1|
1|1|1|
4|14613|126|4|
4|5|January|Jan|Thursday|
Thu|4|1939/1940|5|
19|Term 2|1940|1940-01-01 00:00:00|1940-01-01 00:00:00|
1|1|1940|5|

```

2 1940-05-01 00:00:00	1940Q02	161 19400104 01/04/1940
04/01/1940 04/01/1940 Jan 04 1940		0 N
Y N Y	N	N N
1939/1940-05 01/01/1940		01/01/1940 w/c
01/01/1940 23281		125
19400105 05/01/1940	1	1
1940W01	1	1
1	1	1
5 14614	127	5
5 6 January		Jan Friday
Fri 5 1939/1940	5	
19 Term 2 1940 1940-01-01 00:00:00 1940-01-01 00:00:00		
1 1 1	1940	5
2 1940-05-01 00:00:00	1940Q02	161 19400105 01/05/1940
05/01/1940 05/01/1940 Jan 05 1940		0 N
Y N Y	N	N N
1939/1940-05 01/01/1940		01/01/1940 w/c
01/01/1940 23281		126
19400106 06/01/1940	1	1
1940W01	1	1
1	1	1
6 14615	128	6
6 7 January		Jan Saturday
Sat 6 1939/1940	5	
19 Term 3 1940 1940-01-01 00:00:00 1940-01-01 00:00:00		
1 1 1	1940	5
2 1940-05-01 00:00:00	1940Q02	161 19400106 01/06/1940
06/01/1940 06/01/1940 Jan 06 1940		0 N
N Y N	N	N N
1939/1940-05 01/01/1940		01/01/1940 w/c
01/01/1940 23281		127
19400107 07/01/1940	1	1
1940W01	1	2
1	2	2
7 14616	129	7
7 1 January		Jan Sunday
Sun 7 1939/1940	5	
19 Term 3 1940 1940-01-01 00:00:00 1940-01-01 00:00:00		
1 1 2	1940	5
2 1940-05-01 00:00:00	1940Q02	161 19400107 01/07/1940
07/01/1940 07/01/1940 Jan 07 1940		0 N
N Y N	N	N N
1939/1940-05 01/01/1940		01/01/1940 w/c
01/01/1940 23281		128
19400108 08/01/1940	1	1
1940W02	2	2
2	2	2
8 14617	130	8
8 2 January		Jan Monday

Mon	1	1939/1940	5
20	Term 3	1940 1940-01-01 00:00:00	1940-01-01 00:00:00
1	1	2	1940
2	1940-05-01 00:00:00	1940Q02	161 19400108 01/08/1940
08/01/1940	08/01/1940	Jan 08 1940	0
Y	N	Y	N
1939/1940-05	08/01/1940	08/01/1940	w/c
08/01/1940	23281	129	
19400109 09/01/1940	1	1	
1940W02	2	2	
2	2	2	
9	14618	131	9
9	3	January	Jan  Tuesday
Tue	2	1939/1940	5
20	Term 3	1940 1940-01-01 00:00:00	1940-01-01 00:00:00
1	1	2	1940
2	1940-05-01 00:00:00	1940Q02	161 19400109 01/09/1940
09/01/1940	09/01/1940	Jan 09 1940	0
Y	N	Y	N
1939/1940-05	08/01/1940	08/01/1940	w/c
08/01/1940	23281	130	
19400110 10/01/1940	1	1	
1940W02	2	2	
2	2	2	
10	14619	132	10
10	4	January	Jan
Wednesday	Wed	3	1939/1940
5	20	Term 3	1940 1940-01-01
00:00:00	1940-01-01 00:00:00	1	1
2	1940	5	2 1940-05-01 00:00:00
1940Q02	161 19400110 01/10/1940	10/01/1940	10/01/1940 Jan
10 1940	0	N	Y
N	N	N	1939/1940-05
08/01/1940	w/c 08/01/1940	23281	08/01/1940
131			
19400111 11/01/1940	1	1	
1940W02	2	2	
2	2	2	
11	14620	133	11
11	5	January	Jan
Thursday	Thu	4	1939/1940
5	20	Term 3	1940 1940-01-01
00:00:00	1940-01-01 00:00:00	1	1
2	1940	5	2 1940-05-01 00:00:00
1940Q02	161 19400111 01/11/1940	11/01/1940	11/01/1940 Jan
11 1940	0	N	Y
N	N	N	1939/1940-05
08/01/1940	w/c 08/01/1940	23281	08/01/1940
132			



19400112 12/01/1940	1	1
1940W02	2	2
2	2	2
12 14621	134	12
12	6 January	Jan
Friday	Fri	5 1939/1940
5	20 Term 3	1940 1940-01-01
00:00:00 1940-01-01 00:00:00	1	1
2 1940	5	2 1940-05-01 00:00:00
1940Q02	161 19400112 01/12/1940 12/01/1940 12/01/1940 Jan	
12 1940	0	N Y N Y
N	N	N 1939/1940-05 08/01/1940
08/01/1940 w/c 08/01/1940	23281	
133		
19400113 13/01/1940	1	1
1940W02	2	2
2	2	2
13 14622	135	13
13	7 January	Jan
Saturday	Sat	6 1939/1940
5	20 Term 3	1940 1940-01-01
00:00:00 1940-01-01 00:00:00	1	1
2 1940	5	2 1940-05-01 00:00:00
1940Q02	161 19400113 01/13/1940 13/01/1940 13/01/1940 Jan	
13 1940	0	N N Y N
N	N	N 1939/1940-05 08/01/1940
08/01/1940 w/c 08/01/1940	23281	
134		
19400114 14/01/1940	1	1
1940W02	2	3
2	3	3
14 14623	136	14
14	1 January	Jan
Sunday	Sun	7 1939/1940
5	20 Term 3	1940 1940-01-01
00:00:00 1940-01-01 00:00:00	1	1
3 1940	5	2 1940-05-01 00:00:00
1940Q02	161 19400114 01/14/1940 14/01/1940 14/01/1940 Jan	
14 1940	0	N N Y N
N	N	N 1939/1940-05 08/01/1940
08/01/1940 w/c 08/01/1940	23281	
135		
19400115 15/01/1940	1	1
1940W03	3	3
3	3	3
15 14624	137	15
15	2 January	Jan
Monday	Mon	1 1939/1940
5	21 Term 3	1940 1940-01-01

00:00:00 1940-01-01 00:00:00	1	1
3 1940	5	2 1940-05-01 00:00:00
1940Q02	161 19400115 01/15/1940 15/01/1940 15/01/1940 Jan	
15 1940	0	N Y N Y
N	N	N 1939/1940-05 15/01/1940
15/01/1940	w/c 15/01/1940	23281
136		
19400116 16/01/1940	1	1
1940W03	3	3
3	3	3
16 14625	138	16
16	3 January	Jan
Tuesday	Tue	2 1939/1940
5	21 Term 3	1940 1940-01-01
00:00:00 1940-01-01 00:00:00	1	1
3 1940	5	2 1940-05-01 00:00:00
1940Q02	161 19400116 01/16/1940 16/01/1940 16/01/1940 Jan	
16 1940	0	N Y N Y
N	N	N 1939/1940-05 15/01/1940
15/01/1940	w/c 15/01/1940	23281
137		
19400117 17/01/1940	1	1
1940W03	3	3
3	3	3
17 14626	139	17
17	4 January	Jan
Wednesday	Wed	3 1939/1940
5	21 Term 3	1940 1940-01-01
00:00:00 1940-01-01 00:00:00	1	1
3 1940	5	2 1940-05-01 00:00:00
1940Q02	161 19400117 01/17/1940 17/01/1940 17/01/1940 Jan	
17 1940	0	N Y N Y
N	N	N 1939/1940-05 15/01/1940
15/01/1940	w/c 15/01/1940	23281
138		
19400118 18/01/1940	1	1
1940W03	3	3
3	3	3
18 14627	140	18
18	5 January	Jan
Thursday	Thu	4 1939/1940
5	21 Term 3	1940 1940-01-01
00:00:00 1940-01-01 00:00:00	1	1
3 1940	5	2 1940-05-01 00:00:00
1940Q02	161 19400118 01/18/1940 18/01/1940 18/01/1940 Jan	
18 1940	0	N Y N Y
N	N	N 1939/1940-05 15/01/1940
15/01/1940	w/c 15/01/1940	23281
139		

```

|19400119|19/01/1940|1|1|
1940W03|3|3|3|
3|3|3|
19|14628|141|19|
19|6|January|Jan|
Friday|Fri|5|1939/1940|
5|21|Term 3|1940|1940-01-01
00:00:00|1940-01-01 00:00:00|1|1|
3|1940|5|2|1940-05-01 00:00:00|
1940Q02|161|19400119|01/19/1940|19/01/1940|19/01/1940|Jan
19 1940|0|N|Y|N|Y|
N|N|N|1939/1940-05|15/01/1940|
15/01/1940|w/c 15/01/1940|23281|
140|
|19400120|20/01/1940|1|1|
1940W03|3|3|3|
3|3|3|
20|14629|142|20|
20|7|January|Jan|
Saturday|Sat|6|1939/1940|
5|21|Term 3|1940|1940-01-01
00:00:00|1940-01-01 00:00:00|1|1|
3|1940|5|2|1940-05-01 00:00:00|
1940Q02|161|19400120|01/20/1940|20/01/1940|20/01/1940|Jan
20 1940|0|N|N|Y|N|
N|N|N|1939/1940-05|15/01/1940|
15/01/1940|w/c 15/01/1940|23281|
141|
+-----+-----+-----+-----+
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only showing top 20 rows

```

```
show_df_missing_breakdown(df_date_spark)
```

DataFrame has 36891 rows and 56 columns.

Column			Null	EmptyStr	NA/NaNStr
NumericNaN	TotalMissing	%Missing			
-----					
DateKey			0	0	0
0	0	0.00%			
FullDate			0	0	0
0	0	0.00%			
MonthNumberOfYear			0	0	0
0	0	0.00%			
MonthNumberOfQuarter			0	0	0
0	0	0.00%			
ISOYearAndWeekNumber			0	0	0
0	0	0.00%			
ISOWeekNumberOfYear			0	0	0
0	0	0.00%			
SSWeekNumberOfYear			0	0	0
0	0	0.00%			
ISOWeekNumberOfQuarter_454_Pattern			0	0	0
0	0	0.00%			
SSWeekNumberOfQuarter_454_Pattern			0	0	0
0	0	0.00%			
SSWeekNumberOfMonth			0	0	0
0	0	0.00%			
DayNumberOfYear			0	0	0
0	0	0.00%			
DaysSince1900			0	0	0
0	0	0.00%			
DayNumberOfFiscalYear			0	0	0
0	0	0.00%			
DayNumberOfQuarter			0	0	0
0	0	0.00%			
DayNumberOfMonth			0	0	0
0	0	0.00%			
DayNumberOfWeek_Sun_Start			0	0	0
0	0	0.00%			
MonthName			0	0	0
0	0	0.00%			
MonthNameAbbreviation			0	0	0
0	0	0.00%			
DayName			0	0	0
0	0	0.00%			
DayNameAbbreviation			0	0	0
0	0	0.00%			
DayNumberOfWeek			0	0	0
0	0	0.00%			
AcademicYear			0	0	0
0	0	0.00%			

AcademicMonthNumber	0	0	0
0	0	0.00%	
AcademicWeekNumberOfYear	0	0	0
0	0	0.00%	
TermSession	0	0	0
0	0	0.00%	
CalendarYear	0	0	0
0	0	0.00%	
CalendarYearMonth	0	0	0
0	0	0.00%	
CalendarYearQtr	0	0	0
0	0	0.00%	
CalendarSemester	0	0	0
0	0	0.00%	
CalendarQuarter	0	0	0
0	0	0.00%	
CalendarWeekNumber	0	0	0
0	0	0.00%	
FiscalYear	0	0	0
0	0	0.00%	
FiscalMonth	0	0	0
0	0	0.00%	
FiscalQuarter	0	0	0
0	0	0.00%	
FiscalYearMonth	0	0	0
0	0	0.00%	
FiscalYearQtr	0	0	0
0	0	0.00%	
QuarterNumber	0	0	0
0	0	0.00%	
YYYYMMDD	0	0	0
0	0	0.00%	
MM/DD/YYYY	0	0	0
0	0	0.00%	
YYYY/MM/DD	0	0	0
0	0	0.00%	
YYYY-MM-DD	0	0	0
0	0	0.00%	
MonDDYYYY	0	0	0
0	0	0.00%	
IsCurrentYear	0	0	0
0	0	0.00%	
IsLastDayOfMonth	0	0	0
0	0	0.00%	
IsWeekday	0	0	0
0	0	0.00%	
IsWeekend	0	0	0
0	0	0.00%	
IsWorkday	0	0	0
0	0	0.00%	

IsFederalHoliday			0	0	0
0	0	0.00%			
IsBankHoliday			0	0	0
0	0	0.00%			
IsCompanyHoliday			0	0	0
0	0	0.00%			
AcademicYearPeriod			0	0	0
0	0	0.00%			
WeekStartDate			0	0	0
0	0	0.00%			
WeekCommencing DD/MM/YYYY			0	0	0
0	0	0.00%			
WeekCommencingName			0	0	0
0	0	0.00%			
Year Month Number			0	0	0
0	0	0.00%			
DayNumberofAcademicYear			0	0	0
0	0	0.00%			

Initial observations:

- Academic year is available in the data - good for academic year analysis
- week start dates available as calendar week start date and academic week start date
- datekey can be used to join with the date dimension

#### 4. Join the data

```
from pyspark.sql import functions as F

# 1. Alias DataFrames to reference them in the join condition and in
the column selection
df_att_aliased = df_attendancesessions_spark.alias("att")
df_org_aliased = df_organisation_spark.alias("org")
df_stu_aliased = df_student_spark.alias("stu")
df_stex_aliased = df_studentextended_spark.alias("stex")
df_date_aliased = df_date_spark.alias("dd")

# 2. Join them explicitly
df_joined = (
    df_att_aliased
        .join(df_org_aliased, df_att_aliased["organisationkey"] ==
df_org_aliased["organisationkey"], "left")
        .join(df_stu_aliased, df_att_aliased["studentkey"] ==
df_stu_aliased["studentkey"], "left")
        .join(df_stex_aliased, df_att_aliased["studentkey"] ==
df_stex_aliased["studentkey"], "left")
        .join(df_date_aliased, df_att_aliased["datekey"] ==
df_date_aliased["DateKey"], "left") #join the
df_attendancesessions_spark with the df_date_spark
)
```

```
# 3. Programmatically build a list of columns to select
#     Each column is referenced by alias + column name, and renamed
#     with a prefix
```

```
att_cols = [F.col(f"att.{c}").alias(f"att_{c}") for c in
df_attendancesessions_spark.columns]
org_cols = [F.col(f"org.{c}").alias(f"org_{c}") for c in
df_organisation_spark.columns]
stu_cols = [F.col(f"stu.{c}").alias(f"stu_{c}") for c in
df_student_spark.columns]
stex_cols = [F.col(f"stex.{c}").alias(f"stex_{c}") for c in
df_studentextended_spark.columns]
date_cols = [F.col(f"dd.{c}").alias(f"dd_{c}") for c in
df_date_spark.columns]
```

```
# Combine all these column lists
```

```
all_cols = att_cols + org_cols + stu_cols + stex_cols + date_cols
```

```
# 4. Select everything into a new DataFrame, with prefixed column names
```

```
df_joined_renamed = df_joined.select(*all_cols)
```

```
df_joined_renamed.show(truncate=False)
```

A 20x20 grid of dashed lines with plus signs at various intersections, representing a sparse matrix or a specific data layout. The plus signs are located at the following (row, column) coordinates (starting from 0,0 at the top-left):

Row	Column
0	2, 4, 6, 8, 10, 12, 14, 16, 18
1	1, 3, 5, 7, 9, 11, 13, 15, 17, 19
2	0, 2, 4, 6, 8, 10, 12, 14, 16, 18
3	1, 3, 5, 7, 9, 11, 13, 15, 17, 19
4	0, 2, 4, 6, 8, 10, 12, 14, 16, 18
5	1, 3, 5, 7, 9, 11, 13, 15, 17, 19
6	0, 2, 4, 6, 8, 10, 12, 14, 16, 18
7	1, 3, 5, 7, 9, 11, 13, 15, 17, 19
8	0, 2, 4, 6, 8, 10, 12, 14, 16, 18
9	1, 3, 5, 7, 9, 11, 13, 15, 17, 19
10	0, 2, 4, 6, 8, 10, 12, 14, 16, 18
11	1, 3, 5, 7, 9, 11, 13, 15, 17, 19
12	0, 2, 4, 6, 8, 10, 12, 14, 16, 18
13	1, 3, 5, 7, 9, 11, 13, 15, 17, 19
14	0, 2, 4, 6, 8, 10, 12, 14, 16, 18
15	1, 3, 5, 7, 9, 11, 13, 15, 17, 19
16	0, 2, 4, 6, 8, 10, 12, 14, 16, 18
17	1, 3, 5, 7, 9, 11, 13, 15, 17, 19
18	0, 2, 4, 6, 8, 10, 12, 14, 16, 18
19	1, 3, 5, 7, 9, 11, 13, 15, 17, 19

```

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+
|att_Date|att_Mark|att_Session|att_attendancesessionkey|
att_is_aea|att_is_attend|att_is_auth_abs|att_is_late_L|att_is_late_U|
att_is_missing|att_is_nr|att_is_possible|att_is_present|
att_is_unauth_abs|att_organisationkey|att_studentkey|
|att_partitionkey|att_datekey|
org_Organisation_Name|org_Establishment_Number|org_LA_Code|
org_Organisation_Type|org_organisationkey|
org_addresskey|org_UKPRN|org_Organisation_Status|org_last_updated|
org_URN|org_partitionkey|stu_Forename|
stu_Legal_Forename|stu_Legal_Surname|stu_Surname|stu_Middle_Names|
stu_Sex|stu_Gender|stu_Date_Of_Birth|stu_organisationkey|
|stu_studentkey|stu_UPN|stu_partitionkey|
|stex_Ethnicity|stex_Ethnicity_Code|stex_Ever_In_Care|
stex_First_Language|stex_Free_School_Meals|stex_Free_School_Meals_6|
stex_Gifted_And_Talented_Status|stex_In_LEA_Care|
stex_Pupil_Premium_Indicator|stex_SEN_Status|
stex_English_As_Additional_Language|
stex_English_As_Additional_Language_Status|stex_Child_In_Need|
stex_Child_Protection_Plan|stex_Enrolment_Status|
stex_studentextendedkey|stex_Year_Group|
stex_Current_NC_Year|stex_Admission_Date|stex_Leaving_Date|
stex_Is_Current|stex_Postcode|stex_organisationkey|
stex_studentkey|stex_partitionkey|
|dd_DateKey|dd_FullDate|dd_MonthNumberOfYear|dd_MonthNumberOfQuarter|
dd_ISOYearAndWeekNumber|dd_ISOWeekNumberOfYear|dd_SSWeekNumberOfYear|
dd_ISOWeekNumberOfQuarter_454_Pattern|
dd_SSWeekNumberOfQuarter_454_Pattern|dd_SSWeekNumberOfMonth|
dd_DayNumberOfYear|dd_DaysSince1900|dd_DayNumberOfFiscalYear|
dd_DayNumberOfQuarter|dd_DayNumberOfMonth|

```





```
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+
|2022-01-03|#          |PM          |540c3b1b-21b6-44b7-8e4c-849455d15b65|
0.0          |0.0          |0.0          |0.0          |0.0          |
0.0          |1.0          |0.0          |0.0          |0.0          |
|068cf4c6-2526-430d-a23d-f482ba43887d|9fca452d-e3b8-4423-abda-
597c5e3f73e5|068cf4c6-2526-430d-a23d-f482ba43887d|20220103 |Academy
9          |XXXXXX          |XXXXXX          |ALL THROUGH
|068cf4c6-2526-430d-a23d-f482ba43887d|          |Active
|          |8          |068cf4c6-2526-430d-a23d-f482ba43887d|Melissa
|XXXXXX          |XXXXXX          |Chase          |XXXXXX          |
FEMALE |None          |2000-01-01          |068cf4c6-2526-430d-a23d-
f482ba43887d|9fca452d-e3b8-4423-abda-597c5e3f73e5|6949          |068cf4c6-
2526-430d-a23d-f482ba43887d|Any other mixed background|MOTH
|False          |Pali          |True          |True
|False          |          |False          |True
|None          |False          |          |None
|None          |None          |          |SINGLE REGISTRATION |
6a51ea88-0c44-4359-987d-394a76ee84f6|9          |9
|2024-03-11 00:00:00.000000|NULL          |1          |
|068cf4c6-2526-430d-a23d-f482ba43887d|9fca452d-e3b8-4423-abda-
597c5e3f73e5|068cf4c6-2526-430d-a23d-f482ba43887d|20220103 |
03/01/2022 |1          |1          |2022W01
|1          |2          |1
|2          |2          |3
|44563          |125          |3          |3
|2          |January          |Jan          |
Monday |Mon          |1          |2021/2022          |5
|19          |Term 2          |2022          |2022-01-01
00:00:00 |2022-01-01 00:00:00|1          |1          |2
|2022          |5          |2          |2022-05-01 00:00:00|
2022Q02          |489          |20220103          |01/03/2022          |03/01/2022
|03/01/2022          |Jan 03 2022 |0          |N          |Y
|N          |Y          |N          |Y          |N
|2021/2022-05          |03/01/2022          |03/01/2022          |
w/c 03/01/2022          |24265          |124          |
|2022-02-14|#          |PM          |4bd91887-aebd-4932-be66-e8183062e1da|
0.0          |0.0          |0.0          |0.0          |0.0          |
0.0          |1.0          |0.0          |0.0          |0.0          |
|02ef2e04-5a06-4f18-97f1-d971a9a21585|b43ace14-50c5-4beb-8bf3-
5f5904af4308|02ef2e04-5a06-4f18-97f1-d971a9a21585|20220214 |Academy
4          |XXXXXX          |XXXXXX          |SECONDARY
```

02ef2e04-5a06-4f18-97f1-d971a9a21585		Active
	3 02ef2e04-5a06-4f18-97f1-d971a9a21585 Alec	
XXXXXX	XXXXXX Smith XXXXXX	
FEMALE None 2000-01-01 02ef2e04-5a06-4f18-97f1-		
d971a9a21585 b43ace14-50c5-4beb-8bf3-5f5904af4308 8589938741 02ef2e04-		
5a06-4f18-97f1-d971a9a21585 White - British	WBRI	
False	Malayalam True	True
False	False	True
K	False	None
None	None	SINGLE REGISTRATION
2c37f9d2-8724-408e-bff5-778aba4ede84 11	11	
2020-09-02 00:00:00.000000 NULL	1	
02ef2e04-5a06-4f18-97f1-d971a9a21585 b43ace14-50c5-4beb-8bf3-		
5f5904af4308 02ef2e04-5a06-4f18-97f1-d971a9a21585 20220214		
14/02/2022 2	2	2022W07
7	8	7
8	3	45
44605	167	45 14
2	February	Feb
Monday Mon	1	2021/2022 6
25	Term 3	2022 2022-02-01
00:00:00 2022-01-01 00:00:00 1	1	8
2022	6	2
2022Q02	489	20220214 02/14/2022 14/02/2022
14/02/2022 Feb 14 2022 0	N	Y
N	Y	N
2021/2022-06	14/02/2022	14/02/2022
w/c 14/02/2022	24266	166
2022-02-25 /	AM	5d67e42c-d7c5-4db3-8f26-d780e18b7e67
0.0	1.0	0.0 0.0 0.0
0.0	0.0	1.0 0.0
5ce8e936-d2c9-4073-ae9e-40c529fc4e88 758d0015-8c72-4429-a610-		
1380c22405c7 5ce8e936-d2c9-4073-ae9e-40c529fc4e88 20220225	Academy	
2	XXXXXX	XXXXXX SECONDARY
5ce8e936-d2c9-4073-ae9e-40c529fc4e88		Active
	1 5ce8e936-d2c9-4073-ae9e-40c529fc4e88	
Stephanie XXXXXX	XXXXXX	Lynn XXXXXX
FEMALE None 2000-01-01 5ce8e936-d2c9-4073-ae9e-		
40c529fc4e88 758d0015-8c72-4429-a610-1380c22405c7 5094	5ce8e936-	
d2c9-4073-ae9e-40c529fc4e88 White - British	WBRI	
False	Corsican False	False
False	False	False
None	False	None
None	None	SINGLE REGISTRATION
67f657c1-f529-48c8-81b9-834d9de437d1 11	11	
2020-09-03 00:00:00.000000 NULL	1	
5ce8e936-d2c9-4073-ae9e-40c529fc4e88 758d0015-8c72-4429-a610-		
1380c22405c7 5ce8e936-d2c9-4073-ae9e-40c529fc4e88 20220225		
25/02/2022 2	2	2022W08

8	9	8	56	25
9	4	56	25	
44616	178	56	25	
6	February	Feb		
Friday	Fri	5	2021/2022	6
26	Term 4	2022	2022-02-01	
00:00:00	2022-01-01 00:00:00	1	1	9
2022	6	2	2022-06-01 00:00:00	
2022002	489	20220225	02/25/2022	25/02/2022
25/02/2022	Feb 25 2022	0	N	Y
N	Y	N	N	N
2021/2022-06	21/02/2022	21/02/2022		
w/c 21/02/2022	24266	177		
2022-04-24	#	PM	957803c3-8b81-4b05-8c9c-0bb2eab485f3	
0.0	0.0	0.0	0.0	0.0
0.0	1.0	0.0	0.0	0.0
5ce8e936-d2c9-4073-ae9e-40c529fc4e88	6f56a2e0-766c-488b-a305-8e524a461538	5ce8e936-d2c9-4073-ae9e-40c529fc4e88	20220424	Academy
2	XXXXXX	XXXXXX	SECONDARY	
5ce8e936-d2c9-4073-ae9e-40c529fc4e88			Active	
	1	5ce8e936-d2c9-4073-ae9e-40c529fc4e88	Felicia	
XXXXXX	XXXXXX	Vega	XXXXXX	
MALE	None	2000-01-01	5ce8e936-d2c9-4073-ae9e-40c529fc4e88	6f56a2e0-766c-488b-a305-8e524a461538
40c529fc4e88	6f56a2e0-766c-488b-a305-8e524a461538	4834	5ce8e936-d2c9-4073-ae9e-40c529fc4e88	White - British
False	Corsican	False	WBRI	False
False		False	False	
None	False		None	
None	None		SINGLE REGISTRATION	
b14e8b40-e401-4d55-bdfa-64bdf398ee82	10		10	
2021-09-01 00:00:00.000000	NULL	1		
5ce8e936-d2c9-4073-ae9e-40c529fc4e88	6f56a2e0-766c-488b-a305-8e524a461538	5ce8e936-d2c9-4073-ae9e-40c529fc4e88	20220424	
24/04/2022	4	1	2022W16	
16	18	3		
5		5	114	
44674	236	24	24	
1	April	Apr		
Sunday	Sun	7	2021/2022	8
34	Term 5	2022	2022-04-01	
00:00:00	2022-02-01 00:00:00	1	2	
18	2022	8	3	
2022-08-01 00:00:00	2022003	490	20220424	
04/24/2022	24/04/2022	24/04/2022	Apr 24 2022	0
N	N	Y	N	N
N	N	2021/2022-08	18/04/2022	
18/04/2022		w/c 18/04/2022	24268	
235				
2022-05-09	/	AM	f02bf22a-bfbf-4c35-9f40-eda2777c9bc3	

0.0	1.0	0.0	0.0	0.0	
0.0	0.0	1.0	1.0	0.0	
2d9ba2ce-d6e9-49b4-b8c0-d03c7f69d8d8 bc5d141b-fd82-42a5-bca1-cbfaf761f3ee 2d9ba2ce-d6e9-49b4-b8c0-d03c7f69d8d8 20220509 Academy					
7	XXXXXX		XXXXXX	SECONDARY	
2d9ba2ce-d6e9-49b4-b8c0-d03c7f69d8d8					
					Active
	6	2d9ba2ce-d6e9-49b4-b8c0-d03c7f69d8d8 Matthew			
XXXXXX	XXXXXX	Peterson	XXXXXX		
MALE	None	2000-01-01	2d9ba2ce-d6e9-49b4-b8c0-d03c7f69d8d8 bc5d141b-fd82-42a5-bca1-cbfaf761f3ee 8589938930 2d9ba2ce-d6e9-49b4-b8c0-d03c7f69d8d8 White - British		
				WBRI	
False	Hindi		True		True
False		False		True	
E	False			None	
None	None			SINGLE REGISTRATION	
da8df805-a390-4121-8bd9-dca457a9f2be 11					
11					
2020-09-01 00:00:00.000000 nan					
1					
2d9ba2ce-d6e9-49b4-b8c0-d03c7f69d8d8 bc5d141b-fd82-42a5-bca1-cbfaf761f3ee 2d9ba2ce-d6e9-49b4-b8c0-d03c7f69d8d8 20220509					
09/05/2022 5					
2					
2022W19					
19					
20					
6					
7					
2					
129					
44689					
251					
39					
9					
2					
May					
May					
Monday Mon					
1					
2021/2022					
9					
37					
Term 5					
2022					
2022-05-01					
00:00:00 2022-02-01 00:00:00 1					
2					
20					
2022					
9					
3					
2022-09-01 00:00:00 2022Q03					
490					
20220509					
05/09/2022					
09/05/2022					
09/05/2022					
May 09 2022					
0					
N					
Y					
N					
Y					
N					
2021/2022-09					
09/05/2022					
09/05/2022					
w/c 09/05/2022					
24269					
250					
2022-07-16 #					
PM					
d4a94308-d246-4b10-8bc3-507cb9b32702					
0.0					
0.0					
0.0					
0.0					
0.0					
0.0					
2d9ba2ce-d6e9-49b4-b8c0-d03c7f69d8d8 bc5d141b-fd82-42a5-bca1-cbfaf761f3ee 2d9ba2ce-d6e9-49b4-b8c0-d03c7f69d8d8 20220716 Academy					
7	XXXXXX		XXXXXX	SECONDARY	
2d9ba2ce-d6e9-49b4-b8c0-d03c7f69d8d8					
					Active
	6	2d9ba2ce-d6e9-49b4-b8c0-d03c7f69d8d8 Matthew			
XXXXXX	XXXXXX	Peterson	XXXXXX		
MALE	None	2000-01-01	2d9ba2ce-d6e9-49b4-b8c0-d03c7f69d8d8 bc5d141b-fd82-42a5-bca1-cbfaf761f3ee 8589938930 2d9ba2ce-d6e9-49b4-b8c0-d03c7f69d8d8 White - British		
				WBRI	
False	Hindi		True		True
False		False		True	
E	False			None	

None	None	SINGLE REGISTRATION	
da8df805-a390-4121-8bd9-dca457a9f2be 11	11		
2020-09-01 00:00:00.000000 nan	1		
2d9ba2ce-d6e9-49b4-b8c0-d03c7f69d8d8 bc5d141b-fd82-42a5-bca1-cbfaf761f3ee 2d9ba2ce-d6e9-49b4-b8c0-d03c7f69d8d8 20220716			
16/07/2022  7	1	2022W28	
28	29	2	
3	3	197	
44757	319	16	16
7	July	Jul	
Saturday  Sat	6	2021/2022	
11	46	Term 6	2022
2022-07-01 00:00:00  2022-03-01 00:00:00 2		3	
29	2022	11	4
2022-11-01 00:00:00 2022Q04	491	20220716	
07/16/2022  16/07/2022  16/07/2022  Jul 16 2022  0			
N	N	Y	N
N	N	2021/2022-11	11/07/2022
11/07/2022	w/c 11/07/2022	24271	
318			
2022-07-16 #	PM	81393f62-8757-4f5c-a02e-5f5d3eeecd89	
0.0  0.0	0.0	0.0	0.0
0.0  1.0	0.0	0.0	0.0
5ce8e936-d2c9-4073-ae9e-40c529fc4e88 758d0015-8c72-4429-a610-1380c22405c7 5ce8e936-d2c9-4073-ae9e-40c529fc4e88 20220716	Academy		
2	XXXXXX	XXXXXX	SECONDARY
5ce8e936-d2c9-4073-ae9e-40c529fc4e88		Active	
	1	5ce8e936-d2c9-4073-ae9e-40c529fc4e88	
Stephanie  XXXXXX	XXXXXX	Lynn	XXXXXX
FEMALE  None	2000-01-01	5ce8e936-d2c9-4073-ae9e-40c529fc4e88 758d0015-8c72-4429-a610-1380c22405c7 5094	5ce8e936-d2c9-4073-ae9e-40c529fc4e88 White - British
False	Corsican	False	False
False	False	False	False
None	False	None	
None	None	SINGLE REGISTRATION	
67f657c1-f529-48c8-81b9-834d9de437d1 11	11		
2020-09-03 00:00:00.000000 NULL	1		
5ce8e936-d2c9-4073-ae9e-40c529fc4e88 758d0015-8c72-4429-a610-1380c22405c7 5ce8e936-d2c9-4073-ae9e-40c529fc4e88 20220716			
16/07/2022  7	1	2022W28	
28	29	2	
3	3	197	
44757	319	16	16
7	July	Jul	
Saturday  Sat	6	2021/2022	
11	46	Term 6	2022
2022-07-01 00:00:00  2022-03-01 00:00:00 2		3	
29	2022	11	4

2022-11-01 00:00:00	2022Q04	491	20220716
07/16/2022	16/07/2022	16/07/2022	Jul 16 2022
N	N	Y	N
N	N	2021/2022-11	11/07/2022
11/07/2022	w/c 11/07/2022	24271	
318			
2022-07-17	#	PM	a10d1c02-a6cb-4851-be71-b05cd60070cc
0.0	0.0	0.0	0.0
0.0	1.0	0.0	0.0
5ce8e936-d2c9-4073-ae9e-40c529fc4e88	13fd1794-691a-4006-81a5-f03ee17600c5	5ce8e936-d2c9-4073-ae9e-40c529fc4e88	20220717
2	XXXXXX	XXXXXX	SECONDARY
5ce8e936-d2c9-4073-ae9e-40c529fc4e88			Active
	1	5ce8e936-d2c9-4073-ae9e-40c529fc4e88	
Christopher	XXXXXX	XXXXXX	Miller
MALE	None	2000-01-01	5ce8e936-d2c9-4073-ae9e-40c529fc4e88
13fd1794-691a-4006-81a5-f03ee17600c5	871	5ce8e936-d2c9-4073-ae9e-40c529fc4e88	White and Black African
False	Icelandic	True	True
False		False	False
None	False		None
None	None		SINGLE REGISTRATION
1cb351c1-4db8-4ec9-a13a-5134002c4809	13		13
2018-09-03 00:00:00	000000	NULL	1
5ce8e936-d2c9-4073-ae9e-40c529fc4e88	13fd1794-691a-4006-81a5-f03ee17600c5	5ce8e936-d2c9-4073-ae9e-40c529fc4e88	20220717
17/07/2022	7	1	2022W28
28	30		2
4		4	198
44758	320		17
1		July	Jul
Sunday	Sun		2021/2022
11	46		Term 6
2022-07-01 00:00:00	2022-03-01 00:00:00	2	3
30	2022	11	4
2022-11-01 00:00:00	2022Q04	491	20220717
07/17/2022	17/07/2022	17/07/2022	Jul 17 2022
N	N	Y	N
N	N	2021/2022-11	11/07/2022
11/07/2022	w/c 11/07/2022	24271	
319			
2022-08-05	#	PM	4360be6b-8b0e-41cf-a37b-aaf0898f7ac8
0.0	0.0	0.0	0.0
0.0	1.0	0.0	0.0
f7d73a58-e94b-4505-b126-b5f73c66b48b	3aeb9831-4595-4722-8e3b-ac1b864f2c05	f7d73a58-e94b-4505-b126-b5f73c66b48b	20220805
1	XXXXXX	XXXXXX	SECONDARY
f7d73a58-e94b-4505-b126-b5f73c66b48b			Active
	0	f7d73a58-e94b-4505-b126-b5f73c66b48b	Rebecca

XXXXXX	XXXXXX	Mckee	XXXXXX	
FEMALE  None	2000-01-01	f7d73a58-e94b-4505-b126-b5f73c66b48b 3aeb9831-4595-4722-8e3b-ac1b864f2c05 2563	f7d73a58-e94b-4505-b126-b5f73c66b48b Pakistani	APKN
False	Ndonga	False	False	False
False	False	False	False	False
None	False	None	None	None
None	None	SINGLE REGISTRATION		
c1f795ff-c506-45be-be61-ff9ad7fd900b 11	11	11	11	11
2020-09-01 00:00:00.000000 NULL	1	1	1	1
f7d73a58-e94b-4505-b126-b5f73c66b48b 3aeb9831-4595-4722-8e3b-ac1b864f2c05 f7d73a58-e94b-4505-b126-b5f73c66b48b 20220805	20220805	20220805	20220805	20220805
05/08/2022  8	2	2022W31	2022W31	2022W31
31	32	5	5	5
6	1	217	217	217
44777	339	36	5	5
6	August	Aug		
Friday  Fri	5	2022/2023		
12	49	Term 6	2022	2022
2022-08-01 00:00:00  2022-03-01 00:00:00 2	3	3	3	3
32	2022	12	4	4
2022-12-01 00:00:00 2022Q04	491	20220805		
08/05/2022  05/08/2022  05/08/2022  Aug 05 2022  0	0	0	0	0
N	Y	N	Y	N
N	N	2021/2022-12	01/08/2022	01/08/2022
01/08/2022	w/c 01/08/2022	24272	24272	24272
338				
2022-08-09 #	PM	137c1f50-47cb-42e7-bc3e-a898d4b3eb30	137c1f50-47cb-42e7-bc3e-a898d4b3eb30	137c1f50-47cb-42e7-bc3e-a898d4b3eb30
0.0  0.0	0.0	0.0	0.0	0.0
0.0  1.0	0.0	0.0	0.0	0.0
02ef2e04-5a06-4f18-97f1-d971a9a21585 b43ace14-50c5-4beb-8bf3-5f5904af4308 02ef2e04-5a06-4f18-97f1-d971a9a21585 20220809	Academy	Academy	Academy	Academy
4  XXXXXX	XXXXXX	SECONDARY	SECONDARY	SECONDARY
02ef2e04-5a06-4f18-97f1-d971a9a21585		Active	Active	Active
	3	02ef2e04-5a06-4f18-97f1-d971a9a21585 Alec	02ef2e04-5a06-4f18-97f1-d971a9a21585 Alec	02ef2e04-5a06-4f18-97f1-d971a9a21585 Alec
XXXXXX	XXXXXX	Smith	XXXXXX	XXXXXX
FEMALE  None	2000-01-01	02ef2e04-5a06-4f18-97f1-d971a9a21585 b43ace14-50c5-4beb-8bf3-5f5904af4308 8589938741 02ef2e04-5a06-4f18-97f1-d971a9a21585 White - British	WBRI	WBRI
False	Malayalam	True	True	True
False	False	True	True	True
K	False	None	None	None
None	None	SINGLE REGISTRATION		
2c37f9d2-8724-408e-bff5-778aba4ede84 11	11	11	11	11
2020-09-02 00:00:00.000000 NULL	1	1	1	1
02ef2e04-5a06-4f18-97f1-d971a9a21585 b43ace14-50c5-4beb-8bf3-5f5904af4308 02ef2e04-5a06-4f18-97f1-d971a9a21585 20220809	20220809	20220809	20220809	20220809
09/08/2022  8	2	2022W32	2022W32	2022W32
32	33	6	6	6



7			2		221
44781	343		40		9
3		August	Aug		
Tuesday	Tue		2		2022/2023
12	50		Term 6		2022
2022-08-01 00:00:00	2022-03-01 00:00:00	2		3	
33	2022	12	4		
2022-12-01 00:00:00	2022Q04	491	20220809		
08/09/2022	09/08/2022	09/08/2022	Aug 09 2022	0	
N	Y	N	Y		N
N	N		2021/2022-12		08/08/2022
08/08/2022		w/c 08/08/2022	24272		
342					
2022-09-02 #	AM	fcbbefce-25bf-4b4e-a8ce-a1dda69956b4			
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f7d73a58-e94b-4505-b126-b5f73c66b48b	3aeb9831-4595-4722-8e3b-ac1b864f2c05	f7d73a58-e94b-4505-b126-b5f73c66b48b	20220902	Academy	
1	XXXXXX	XXXXXX	SECONDARY		
f7d73a58-e94b-4505-b126-b5f73c66b48b			Active		
	0	f7d73a58-e94b-4505-b126-b5f73c66b48b	Rebecca		
XXXXXX	XXXXXX	Mckee	XXXXXX		
FEMALE	None	2000-01-01	f7d73a58-e94b-4505-b126-b5f73c66b48b	3aeb9831-4595-4722-8e3b-ac1b864f2c05	2563
f7d73a58-e94b-4505-b126-b5f73c66b48b	Pakistani		APKN		
False	Ndonga	False		False	
False		False	False		
None	False		None		
None	None		SINGLE REGISTRATION		
c1f795ff-c506-45be-be61-ff9ad7fd900b	11		11		
2020-09-01 00:00:00.000000	NULL	1			
f7d73a58-e94b-4505-b126-b5f73c66b48b	3aeb9831-4595-4722-8e3b-ac1b864f2c05	f7d73a58-e94b-4505-b126-b5f73c66b48b	20220902		
02/09/2022	9	3	2022W35		
35	36	9			
10		1	245		
44805	2	64	2		
6		September	Sep		
Friday	Fri	5	2022/2023	1	
1		Term 1	2022	2022-09-01	
00:00:00	2022-03-01 00:00:00	2	3		
36	2023	1	1		
2023-01-01 00:00:00	2023Q01	491	20220902		
09/02/2022	02/09/2022	02/09/2022	Sep 02 2022	0	
N	Y	N	Y		N
N	N		2022/2023-01		29/08/2022
29/08/2022		w/c 29/08/2022	24273		
1					
2022-10-26 #	PM	87fdbe23-fa6f-4f30-ad74-20e50cbb06fa			
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2	XXXXXX	XXXXXX	SECONDARY	
5ce8e936-d2c9-4073-ae9e-40c529fc4e88				Active
	1	5ce8e936-d2c9-4073-ae9e-40c529fc4e88		
Stephanie	XXXXXX	XXXXXX	Lynn	XXXXXX
FEMALE	None	2000-01-01	5ce8e936-d2c9-4073-ae9e-40c529fc4e88	758d0015-8c72-4429-a610-1380c22405c7
5094	5ce8e936-d2c9-4073-ae9e-40c529fc4e88	White - British	WBRI	
False	Corsican	False		False
False		False	False	
None	False		None	
None	None		SINGLE REGISTRATION	
67f657c1-f529-48c8-81b9-834d9de437d1	11		11	
2020-09-03 00:00:00.000000	NULL	1		
5ce8e936-d2c9-4073-ae9e-40c529fc4e88	758d0015-8c72-4429-a610-1380c22405c7	5ce8e936-d2c9-4073-ae9e-40c529fc4e88	20221026	
26/10/2022	10	1		2022W43
43		44		4
5			5	299
44859	56		26	26
4		October	Oct	
Wednesday	Wed	3		2022/2023
9		Term 1	2022	2022-10-01
00:00:00	2022-04-01 00:00:00	2	4	
44	2023	2	1	
2023-02-01 00:00:00	2023Q01	492		20221026
10/26/2022	26/10/2022	26/10/2022	Oct 26 2022	0
N	Y	N	Y	N
N	N		2022/2023-02	24/10/2022
24/10/2022		w/c 24/10/2022		24274
55				
2022-11-27	#	PM		016ba5b1-4d7d-4b9e-abe7-bc4d0ead1876
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068cf4c6-2526-430d-a23d-f482ba43887d	9fca452d-e3b8-4423-abda-597c5e3f73e5	068cf4c6-2526-430d-a23d-f482ba43887d	20221127	Academy
9	XXXXXX	XXXXXX	ALL THROUGH	
068cf4c6-2526-430d-a23d-f482ba43887d				Active
	8	068cf4c6-2526-430d-a23d-f482ba43887d		Melissa
XXXXXX	XXXXXX	Chase	XXXXXX	
FEMALE	None	2000-01-01	068cf4c6-2526-430d-a23d-f482ba43887d	9fca452d-e3b8-4423-abda-597c5e3f73e5
6949	068cf4c6-2526-430d-a23d-f482ba43887d	Any other mixed background	MOTH	
False	Pali	True		True
False		False	True	
None	False		None	
None	None		SINGLE REGISTRATION	

6a51ea88-0c44-4359-987d-394a76ee84f6|9  
|2024-03-11 00:00:00.000000|NULL|1|  
|068cf4c6-2526-430d-a23d-f482ba43887d|9fca452d-e3b8-4423-abda-  
597c5e3f73e5|068cf4c6-2526-430d-a23d-f482ba43887d|20221127|  
27/11/2022|11|2|2022W47  
|47|49|8  
|10|5|331  
|44891|88|58|27  
|1|November|Nov|  
Sunday|Sun|7|2022/2023|3  
|13|Term 2|2022|2022-11-01  
00:00:00|2022-04-01 00:00:00|2|4|  
49|2023|3|1|  
2023-03-01 00:00:00|2023Q01|492|20221127|  
11/27/2022|27/11/2022|27/11/2022|Nov 27 2022|0  
|N|N|Y|N|N  
|N|N|2022/2023-03|21/11/2022  
|21/11/2022|w/c 21/11/2022|24275  
|87|  
|2023-05-30|#|AM|c90fe864-ba70-4448-b5d2-90ae22cbda8|  
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|73ed72bc-efea-4cde-bf4c-3db6538453d9|8c3cb921-d1ca-4de2-810f-  
ec488603fcc1|73ed72bc-efea-4cde-bf4c-3db6538453d9|20230530|Academy  
6|XXXXXX|XXXXXX|SECONDARY  
|73ed72bc-efea-4cde-bf4c-3db6538453d9|Active  
|5|73ed72bc-efea-4cde-bf4c-3db6538453d9|Timothy  
|XXXXXX|XXXXXX|Castillo|XXXXXX|  
FEMALE|None|2000-01-01|73ed72bc-efea-4cde-bf4c-  
3db6538453d9|8c3cb921-d1ca-4de2-810f-ec488603fcc1|8589937817|73ed72bc-  
efea-4cde-bf4c-3db6538453d9|White Other|WOTW  
|False|Irish|False|False|False  
|False|True|None  
|None|None|SINGLE REGISTRATION|  
e9272879-cd2f-4253-8b44-5b3d82423c71|Y10|10  
|2021-09-02 00:00:00.000000|nan|1|  
|73ed72bc-efea-4cde-bf4c-3db6538453d9|8c3cb921-d1ca-4de2-810f-  
ec488603fcc1|73ed72bc-efea-4cde-bf4c-3db6538453d9|20230530|  
30/05/2023|5|2|2023W22  
|22|22|9  
|9|5|150  
|45075|272|60|30  
|3|May|May|  
Tuesday|Tue|2|2022/2023|9  
|40|Term 5|2023|2023-05-01  
00:00:00|2023-02-01 00:00:00|1|2|  
22|2023|9|3|  
2023-09-01 00:00:00|2023Q03|494|20230530|

05/30/2023	30/05/2023	30/05/2023	May 30 2023	0
N	Y	N	Y	N
N	N		2022/2023-09	29/05/2023
29/05/2023		w/c 29/05/2023		24281
271				
2023-11-23 /	AM		c82d5492-cf8d-4666-971d-4fb2296eda71	
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5	XXXXXX		XXXXXX	SECONDARY
5ed661db-d767-4fd5-818d-91d7856bd3d9				Active
	4		5ed661db-d767-4fd5-818d-91d7856bd3d9	Isaac
XXXXXX	XXXXXX	White	XXXXXX	
MALE	None	2000-01-01	5ed661db-d767-4fd5-818d-91d7856bd3d9	
328f6259-68c1-4414-b01d-ab76f8a46150			2180	5ed661db-d767-4fd5-818d-91d7856bd3d9
White - British			WBRI	
False	Sindhi	False		False
False		False		False
N	False		None	
None	None		SINGLE REGISTRATION	
3d6c9c60-5302-4f3e-8825-fb52bbf1a37e			11	
2020-09-02 00:00:00.000000	nan		1	
5ed661db-d767-4fd5-818d-91d7856bd3d9			328f6259-68c1-4414-b01d-ab76f8a46150	
23/11/2023	11	2		2023W47
47	47		8	
8		4		327
45252	84		54	23
5		November	Nov	
Thursday	Thu	4		2023/2024
13		Term 2	2023	2023-11-01
00:00:00	2023-04-01 00:00:00	2	4	
47	2024	3	1	
2024-03-01 00:00:00	2024Q01	496		20231123
11/23/2023	23/11/2023	23/11/2023	Nov 23 2023	0
N	Y	N	N	Y
Y	Y		2023/2024-03	20/11/2023
20/11/2023		w/c 20/11/2023		24287
83				
2024-01-26 \	PM		60a52e69-3a56-45e3-b7dc-d34453ab8630	
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0.0	0.0	1.0	1.0	0.0
5ed661db-d767-4fd5-818d-91d7856bd3d9			2b44bd4a-7215-4792-a515-32857a4042e8	
5	XXXXXX		XXXXXX	SECONDARY
5ed661db-d767-4fd5-818d-91d7856bd3d9				Active
	4		5ed661db-d767-4fd5-818d-91d7856bd3d9	Raymond
XXXXXX	XXXXXX	Sanders	XXXXXX	

FEMALE	None	2000-01-01	5ed661db-d767-4fd5-818d-91d7856bd3d9 2b44bd4a-7215-4792-a515-32857a4042e8 1861	5ed661db-d767-4fd5-818d-91d7856bd3d9 White - British	WBRI	
False	Navajo	False	False	False		
False		False	False	None		
N	False			SINGLE REGISTRATION		
None	None			10		
3780c18c-e18f-4bcd-b74f-895cb8f91c9b 10						
2021-09-01 00:00:00.000000 nan			1			
5ed661db-d767-4fd5-818d-91d7856bd3d9 2b44bd4a-7215-4792-a515-32857a4042e8 5ed661db-d767-4fd5-818d-91d7856bd3d9 20240126						
26/01/2024  1		1		2024W04		
4	4		4			
4		4		26		
45316	148		26		26	
6		January	Jan			
Friday  Fri		5		2023/2024	5	
22		Term 3	2024		2024-01-01	
00:00:00  2024-01-01 00:00:00 1			1		4	
2024	5	2		2024-05-01 00:00:00		
2024002	497		20240126	01/26/2024	26/01/2024	
26/01/2024	Jan 26 2024  0		N		Y	
N	Y	N	N		N	
2023/2024-05	22/01/2024	22/01/2024				
w/c 22/01/2024	24289	147				
2024-03-16 #	PM	19b937d4-6c63-4348-a9fa-2ff0ec274db5				
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0.0	1.0	0.0	0.0	0.0		
73ed72bc-efea-4cde-bf4c-3db6538453d9 59966036-db7a-4f85-abb9-ce3d79c20312 73ed72bc-efea-4cde-bf4c-3db6538453d9 20240316				Academy		
6	XXXXXX	XXXXXX	SECONDARY			
73ed72bc-efea-4cde-bf4c-3db6538453d9				Active		
	5	73ed72bc-efea-4cde-bf4c-3db6538453d9 Lisa				
XXXXXX	XXXXXX	Shepherd	XXXXXX			
MALE	None	2000-01-01	73ed72bc-efea-4cde-bf4c-3db6538453d9 59966036-db7a-4f85-abb9-ce3d79c20312 8589936685 73ed72bc-efea-4cde-bf4c-3db6538453d9 White - English	WENG		
False	Ndonga	True		True		
False		False	True			
N	False		None			
None	None		SINGLE REGISTRATION			
31de0cca-9112-46a2-a1c9-8787a5f086ee Y10			10			
2021-09-02 00:00:00.000000 nan			1			
73ed72bc-efea-4cde-bf4c-3db6538453d9 59966036-db7a-4f85-abb9-ce3d79c20312 73ed72bc-efea-4cde-bf4c-3db6538453d9 20240316						
16/03/2024  3		3		2024W11		
11	11		11			
11		3		76		
45366	198		76		16	

7			March		Mar		
Saturday	Sat		6		2023/2024		7
29			Term 4		2024		2024-03-01
00:00:00	2024-01-01 00:00:00	1			1		
11		2024		7		3	
2024-07-01 00:00:00	2024Q03		497		20240316		
03/16/2024	16/03/2024	16/03/2024	Mar 16 2024	0			
N		N	Y		N		N
N		N		2023/2024-07		11/03/2024	
11/03/2024		w/c 11/03/2024		24291			
197							
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2	XXXXXX		XXXXXX	SECONDARY			
5ce8e936-d2c9-4073-ae9e-40c529fc4e88				Active			
	1	5ce8e936-d2c9-4073-ae9e-40c529fc4e88	Felicia				
XXXXXX	XXXXXX	Vega	XXXXXX				
MALE	None	2000-01-01	5ce8e936-d2c9-4073-ae9e-40c529fc4e88	6f56a2e0-766c-488b-a305-8e524a461538	4834	5ce8e936-d2c9-4073-ae9e-40c529fc4e88	White - British
False		Corsican		False		WBRI	False
False			False		False		False
None	False			None			
None	None			SINGLE REGISTRATION			
b14e8b40-e401-4d55-bdfa-64bdf398ee82	10			10			
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22/05/2024	5		2		2024W21		
21		21		8			
8			4		143		
45433		265		52		22	
4		May		May			
Wednesday	Wed		3		2023/2024		9
39		Term 5		2024		2024-05-01	
00:00:00	2024-02-01 00:00:00	1			2		
21		2024		9		3	
2024-09-01 00:00:00	2024Q03		498		20240522		
05/22/2024	22/05/2024	22/05/2024	May 22 2024	0			
N		Y		N		Y	N
N		N		2023/2024-09		20/05/2024	
20/05/2024		w/c 20/05/2024		24293			
264							
2024-05-26 #		PM		f7813dca-57f6-4307-8bdb-cd87acccbbe5			
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0.0		1.0	0.0		0.0		0.0

5ed661db-d767-4fd5-818d-91d7856bd3d9 f50af990-481d-497b-9d72-e5a7f988dac8 5ed661db-d767-4fd5-818d-91d7856bd3d9 20240526 Academy	
5 XXXXXX XXXXXX SECONDARY	
5ed661db-d767-4fd5-818d-91d7856bd3d9	Active
4 5ed661db-d767-4fd5-818d-91d7856bd3d9 Felicia	
XXXXXX XXXXXX Vega XXXXXX	
MALE None 2000-01-01 5ed661db-d767-4fd5-818d-91d7856bd3d9 f50af990-481d-497b-9d72-e5a7f988dac8 10623 5ed661db-d767-4fd5-818d-91d7856bd3d9 White - British WBRI	
False Turkish False False	
False False False False	
K False None	
None None SINGLE REGISTRATION	
24be37da-c819-4766-9de9-868c3a518faf 11 11	
2020-09-02 00:00:00.000000 nan 1	
5ed661db-d767-4fd5-818d-91d7856bd3d9 f50af990-481d-497b-9d72-e5a7f988dac8 5ed661db-d767-4fd5-818d-91d7856bd3d9 20240526	
26/05/2024 5 2 2024W21	
21 22 8	
9 5 147	
45437 269 56 26	
1 May May	
Sunday Sun 7 2023/2024 9	
39 Term 5 2024 2024-05-01	
00:00:00 2024-02-01 00:00:00 1 2	
22 2024 9 3	
2024-09-01 00:00:00 2024Q03 498 20240526	
05/26/2024 26/05/2024 26/05/2024 May 26 2024 0	
N N Y N N	
N N 2023/2024-09 20/05/2024	
20/05/2024 w/c 20/05/2024 24293	
268	
2024-06-29 # AM 19ca62a2-a1e3-4922-be48-8462872291b8	
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0.0 1.0 0.0 0.0 0.0	
5ce8e936-d2c9-4073-ae9e-40c529fc4e88 13fd1794-691a-4006-81a5-f03ee17600c5 5ce8e936-d2c9-4073-ae9e-40c529fc4e88 20240629 Academy	
2 XXXXXX XXXXXX SECONDARY	
5ce8e936-d2c9-4073-ae9e-40c529fc4e88	Active
1 5ce8e936-d2c9-4073-ae9e-40c529fc4e88	
Christopher XXXXXX XXXXXX Miller XXXXXX	
MALE None 2000-01-01 5ce8e936-d2c9-4073-ae9e-40c529fc4e88 13fd1794-691a-4006-81a5-f03ee17600c5 871 5ce8e936-d2c9-4073-ae9e-40c529fc4e88 White and Black African MWBA	
False Icelandic True True	
False False False False	
None False None	
None None SINGLE REGISTRATION	
1cb351c1-4db8-4ec9-a13a-5134002c4809 13 13	

[illegible]



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only showing top 20 rows

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Check: First row of dates match in all columns 2022-01-03, so I will use this date to calculate the week number.

```

#check the data types of the columns
df_joined_renamed.dtypes

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 ('att_Mark', 'string'),
 ('att_Session', 'string'),
 ('att_attendancesessionkey', 'string'),
 ('att_is_aea', 'double'),
 ('att_is_attend', 'double'),
 ('att_is_auth_abs', 'double'),
 ('att_is_late_L', 'double'),
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 ('att_is_missing', 'double'),
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 ('att_is_possible', 'double'),
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 ('att_is_unauth_abs', 'double'),
 ('att_organisationkey', 'string'),
 ('att_studentkey', 'string'),
 ('att_partitionkey', 'string'),
 ('att_datekey', 'int'),
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 ('org_Establishment_Number', 'string'),
 ('org_LA_Code', 'string'),
 ('org_Organisation_Type', 'string'),
 ('org_organisationkey', 'string'),
 ('org_addresskey', 'string'),
 ('org_UKPRN', 'string'),

```

```
('org_Organisation_Status', 'string'),
('org_last_updated', 'string'),
('org_URN', 'bigint'),
('org_partitionkey', 'string'),
('stu_Forename', 'string'),
('stu_Legal_Forename', 'string'),
('stu_Legal_Surname', 'string'),
('stu_Surname', 'string'),
('stu_Middle_Names', 'string'),
('stu_Sex', 'string'),
('stu_Gender', 'string'),
('stu_Date_Of_Birth', 'string'),
('stu_organisationkey', 'string'),
('stu_studentkey', 'string'),
('stu_UPN', 'bigint'),
('stu_partitionkey', 'string'),
('stex_Ethnicity', 'string'),
('stex_Ethnicity_Code', 'string'),
('stex_Ever_In_Care', 'string'),
('stex_First_Language', 'string'),
('stex_Free_School_Meals', 'string'),
('stex_Free_School_Meals_6', 'string'),
('stex_Gifted_And_Talented_Status', 'string'),
('stex_In_LEA_Care', 'string'),
('stex_Pupil_Premium_Indicator', 'string'),
('stex_SEN_Status', 'string'),
('stex_English_As_Additional_Language', 'string'),
('stex_English_As_Additional_Language_Status', 'string'),
('stex_Child_In_Need', 'string'),
('stex_Child_Protection_Plan', 'string'),
('stex_Enrolment_Status', 'string'),
('stex_studentextendedkey', 'string'),
('stex_Year_Group', 'string'),
('stex_Current_NC_Year', 'string'),
('stex_Admission_Date', 'string'),
('stex_Leaving_Date', 'string'),
('stex_Is_Current', 'string'),
('stex_Postcode', 'string'),
('stex_organisationkey', 'string'),
('stex_studentkey', 'string'),
('stex_partitionkey', 'string'),
('dd_DateKey', 'int'),
('dd_FullDate', 'string'),
('dd_MonthNumberOfYear', 'int'),
('dd_MonthNumberOfQuarter', 'int'),
('dd_ISOYearAndWeekNumber', 'string'),
('dd_ISOWeekNumberOfYear', 'int'),
('dd_SSWeekNumberOfYear', 'int'),
('dd_ISOWeekNumberOfQuarter_454_Pattern', 'int'),
```

```
('dd_SSWeekNumberOfQuarter_454_Pattern', 'int'),
('dd_SSWeekNumberOfMonth', 'int'),
('dd_DayNumberOfYear', 'int'),
('dd_DaysSince1900', 'int'),
('dd_DayNumberOfFiscalYear', 'int'),
('dd_DayNumberOfQuarter', 'int'),
('dd_DayNumberOfMonth', 'int'),
('dd_DayNumberOfWeek_Sun_Start', 'int'),
('dd_MonthName', 'string'),
('dd_MonthNameAbbreviation', 'string'),
('dd_DayName', 'string'),
('dd_DayNameAbbreviation', 'string'),
('dd_DayNumberOfWeek', 'int'),
('dd_AcademicYear', 'string'),
('dd_AcademicMonthNumber', 'int'),
('dd_AcademicWeekNumberOfYear', 'int'),
('dd_TermSession', 'string'),
('dd_CalendarYear', 'int'),
('dd_CalendarYearMonth', 'timestamp'),
('dd_CalendarYearQtr', 'timestamp'),
('dd_CalendarSemester', 'int'),
('dd_CalendarQuarter', 'int'),
('dd_CalendarWeekNumber', 'int'),
('dd_FiscalYear', 'int'),
('dd_FiscalMonth', 'int'),
('dd_FiscalQuarter', 'int'),
('dd_FiscalYearMonth', 'timestamp'),
('dd_FiscalYearQtr', 'string'),
('dd_QuarterNumber', 'int'),
('dd_YYYYMMDD', 'int'),
('dd_MM/DD/YYYY', 'string'),
('dd_YYYY/MM/DD', 'string'),
('dd_YYYY-MM-DD', 'string'),
('dd_MonDDYYYY', 'string'),
('dd_IsCurrentYear', 'int'),
('dd_IsLastDayOfMonth', 'string'),
('dd_IsWeekday', 'string'),
('dd_IsWeekend', 'string'),
('dd_IsWorkday', 'string'),
('dd_IsFederalHoliday', 'string'),
('dd_IsBankHoliday', 'string'),
('dd_IsCompanyHoliday', 'string'),
('dd_AcademicYearPeriod', 'string'),
('dd_WeekStartDate', 'string'),
('dd_WeekCommencing DD/MM/YYYY', 'string'),
('dd_WeekCommencingName', 'string'),
('dd_Year Month Number', 'int'),
('dd_DayNumberofAcademicYear', 'int')]
```

```
total_rows = df_joined_renamed.count()
print(f"Total rows: {total_rows}")

Total rows: 16311626

column_count = len(df_joined_renamed.columns)
print(f"Number of columns: {column_count}")

Number of columns: 122
```

The following is a list of columns that can potentially be useful when creating an attendance report. I have consulted the **Department for Education (DfE)** for definitions, as they use specific data fields to collect and manage information related to students and educational establishments. This can also be used to determine a suitable alias for each field. Below is an explanation of each field in the dataset:

1. **student\_sex**: Indicates the student's gender, typically recorded as 'M' for male or 'F' for female.
2. **student\_forename**: The student's first name.
3. **organisation\_type**: Specifies the type of educational establishment, such as 'Academy', 'Community School', 'Free School', etc.
4. **organisation\_name**: The official name of the educational establishment.
5. **establishment\_number**: A unique 4-digit number assigned to each educational establishment by the DfE. This number, combined with the local authority number, forms the DfE number used to identify schools.
6. **la\_code**: The Local Authority code, a 3-digit number representing the local authority responsible for the educational establishment. This code, combined with the establishment number, forms the DfE number.
7. **attendance\_date**: The specific date for which a student's attendance is recorded.
8. **mark**: The attendance code indicating a student's presence or type of absence for a particular session. The DfE provides a set of standardised attendance codes to describe pupil attendance and absence.
9. **session**: Denotes whether the attendance record pertains to the morning (AM) or afternoon (PM) session of the school day.
10. **is\_aea**: Indicates whether the session is an Approved Educational Activity (AEA), meaning the student is off-site but engaged in supervised educational activities approved by the

school.

11. **is\_attend**: Specifies if the student attended the session.
12. **is\_auth\_abs**: Indicates if the student's absence for the session was authorised by the school.
13. **is\_late\_L**: Shows if the student arrived late to the session but before the register closed, typically marked with code 'L'.
14. **is\_late\_U**: Indicates if the student arrived after the register closed, usually marked with code 'U', which can denote an unauthorised absence.
15. **is\_missing**: Denotes if the attendance data for the session is missing or not recorded.
16. **is\_nr**: Indicates 'No Reason' provided for absence, showing that no explanation has been given for the student's absence.
17. **is\_possible**: Specifies if the session was a possible attendance session for the student, meaning they were expected to attend.
18. **is\_present**: Indicates if the student was present during the session.
19. **is\_unauth\_abs**: Shows if the student's absence was unauthorised.
20. **UPN**: Unique Pupil Number, a 13-character identifier assigned to each student in England to track their educational progress.
21. **academic\_year**: The academic year to which the data pertains, typically spanning from September of one year to August of the next (e.g., 2024/2025).
22. **week\_number**: The specific week of the academic year, often numbered from 1 onwards, starting from the beginning of the school year.
23. **term\_session**: Indicates the term (e.g., Autumn, Spring, Summer) and the specific session within that term.

```
from pyspark.sql import functions as F

df_selected = (
    df_joined_renamed
    .select(
        # --- Student details & school info ---
        F.col("stu_Sex").alias("gender"),
        F.col("stu_Forename").alias("student_forename"),
        F.col("stu_Surname").alias("student_surname"),
        F.col("stex_Pupil_Premium_Indicator").alias("pupil_premium"),
```



```

+----+
|gender|student_forename|student_surname|pupil_premium|year_group|
nc_year|school_type|school      |establishment_number|la_code|
attendance_date|academic_year|academic_week_number|term  |
weekcommencing|mark|session|is_approved_educational_activity|
is_attend|is_auth_abs|late|late_unauthorised|missing|no_reason|
is_possible|is_present|is_unauth_abs|current_student|leaving_date|UPN
|
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+----+
|FEMALE|Theresa          |Hester          |False          |8          |8
|SECONDARY |Academy 4|XXXXXX          |XXXXXX |2023-11-13    |
2023/2024   |12          |Term 2|w/c 13/11/2023|/          |AM          |
0.0          |1.0          |0.0          |0.0 |0.0
|0.0      |0.0          |1.0          |1.0          |0.0          |1
|NULL      |3579|
|MALE  |Daniel          |Garcia          |False          |9          |9
|SECONDARY |Academy 4|XXXXXX          |XXXXXX |2023-09-19    |
2023/2024   |4          |Term 1|w/c 18/09/2023|\          |PM          |
0.0          |1.0          |0.0          |0.0 |0.0
|0.0      |0.0          |1.0          |1.0          |0.0          |1
|NULL      |4886|
|MALE  |Alicia          |Fox             |True           |7          |7
|SECONDARY |Academy 4|XXXXXX          |XXXXXX |2024-09-12    |
2024/2025   |2          |Term 1|w/c 09/09/2024|/          |AM          |
0.0          |1.0          |0.0          |0.0 |0.0
|0.0      |0.0          |1.0          |1.0          |0.0          |1
|NULL      |166 |
|MALE  |David          |Martinez        |False          |8          |8
|SECONDARY |Academy 4|XXXXXX          |XXXXXX |2024-08-11    |
2024/2025   |50         |Term 6|w/c 05/08/2024|#          |PM          |
0.0          |0.0          |0.0          |0.0 |0.0
|0.0      |1.0          |0.0          |0.0          |0.0          |1
|NULL      |1235|
|MALE  |Stacy          |Avery           |False          |9          |9
|SECONDARY |Academy 4|XXXXXX          |XXXXXX |2024-11-15    |
2024/2025   |11         |Term 2|w/c 11/11/2024|\          |PM          |
0.0          |1.0          |0.0          |0.0 |0.0
|0.0      |0.0          |1.0          |1.0          |0.0          |1
|NULL      |9728|
|FEMALE|Brett          |Williams        |True           |11         |11
|SECONDARY |Academy 4|XXXXXX          |XXXXXX |2024-10-09    |
2024/2025   |6          |Term 1|w/c 07/10/2024|G          |PM          |
0.0          |0.0          |0.0          |0.0 |0.0

```

0.0	0.0	1.0	0.0	1.0	1
NULL	7267				
FEMALE	Brett	Williams	True	11	11
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-09-29	
2024/2025	4		Term 1	w/c 23/09/2024	# AM
0.0			0.0	0.0	0.0
0.0	1.0	0.0	0.0	0.0	1
NULL	7267				
FEMALE	Theresa	Hester	False	8	8
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-08-21	
2024/2025	52		Term 6	w/c 19/08/2024	# PM
0.0			0.0	0.0	0.0
0.0	1.0	0.0	0.0	0.0	1
NULL	3579				
MALE	Stacy	Avery	False	9	9
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-11-09	
2024/2025	10		Term 2	w/c 04/11/2024	# PM
0.0			0.0	0.0	0.0
0.0	1.0	0.0	0.0	0.0	1
NULL	9728				
MALE	Daniel	Garcia	False	9	9
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-08-27	
2024/2025	53		Term 6	w/c 26/08/2024	# PM
0.0			0.0	0.0	0.0
0.0	1.0	0.0	0.0	0.0	1
NULL	4886				
MALE	Daniel	Garcia	False	9	9
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-11-27	
2024/2025	13		Term 2	w/c 25/11/2024	\ PM
0.0			1.0	0.0	0.0
0.0	0.0	1.0	1.0	0.0	1
NULL	4886				
FEMALE	Theresa	Hester	False	8	8
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-11-05	
2024/2025	10		Term 2	w/c 04/11/2024	\ PM
0.0			1.0	0.0	0.0
0.0	0.0	1.0	1.0	0.0	1
NULL	3579				
FEMALE	Brett	Williams	True	11	11
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-10-20	
2024/2025	7		Term 1	w/c 14/10/2024	# PM
0.0			0.0	0.0	0.0
0.0	1.0	0.0	0.0	0.0	1
NULL	7267				
MALE	Daniel	Garcia	False	9	9
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-12-01	
2024/2025	13		Term 2	w/c 25/11/2024	# PM
0.0			0.0	0.0	0.0
0.0	1.0	0.0	0.0	0.0	1



```

|NULL|4886| | | | |
|FEMALE|Theresa|Hester|False|8|8|
|SECONDARY|Academy 4|XXXXXX|XXXXXX|2024-12-05|
2024/2025|14|Term 2|w/c 02/12/2024|\|PM|
0.0|1.0|0.0|0.0|0.0|
|0.0|0.0|1.0|1.0|0.0|1|
|NULL|3579|
|MALE|Daniel|Garcia|False|9|9|
|SECONDARY|Academy 4|XXXXXX|XXXXXX|2024-11-21|
2024/2025|12|Term 2|w/c 18/11/2024|/|AM|
0.0|1.0|0.0|0.0|0.0|
|0.0|0.0|1.0|1.0|0.0|1|
|NULL|4886|
|FEMALE|Theresa|Hester|False|8|8|
|SECONDARY|Academy 4|XXXXXX|XXXXXX|2024-10-09|
2024/2025|6|Term 1|w/c 07/10/2024|\|PM|
0.0|1.0|0.0|0.0|0.0|
|0.0|0.0|1.0|1.0|0.0|1|
|NULL|3579|
|MALE|David|Martinez|False|8|8|
|SECONDARY|Academy 4|XXXXXX|XXXXXX|2024-10-23|
2024/2025|8|Term 1|w/c 21/10/2024|#|PM|
0.0|0.0|0.0|0.0|0.0|
|0.0|1.0|0.0|0.0|0.0|1|
|NULL|1235|
|FEMALE|Theresa|Hester|False|8|8|
|SECONDARY|Academy 4|XXXXXX|XXXXXX|2024-11-07|
2024/2025|10|Term 2|w/c 04/11/2024|/|AM|
0.0|1.0|0.0|0.0|0.0|
|0.0|0.0|1.0|1.0|0.0|1|
|NULL|3579|
|MALE|David|Martinez|False|8|8|
|SECONDARY|Academy 4|XXXXXX|XXXXXX|2024-12-04|
2024/2025|14|Term 2|w/c 02/12/2024|\|PM|
0.0|1.0|0.0|0.0|0.0|
|0.0|0.0|1.0|1.0|0.0|1|
|NULL|1235|
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+
only showing top 20 rows

```

## 5. Inspect Distinct Values

Begin by looking at year group as it is a key field needed in the summary table.

```
df_selected.select("year_group").distinct().show()
```

```
+-----+
|year_group|
+-----+
|          7|
|         11|
|          8|
|          9|
|         10|
|         12|
|         13|
|          3|
|          5|
|Nursery 2|
|         6|
|         R|
|Nursery 1|
|          1|
|          4|
|          2|
|Year 13|
|       Y10|
|       Y08|
|       Y07|
+-----+
```

only showing top 20 rows

Using these values with skew the data, as Y07, 7 point to the same year group.

```
df_tidy = (
    df_selected
    .withColumn(
        "year_group_tidy",
        F.when(
            F.col("year_group").isin("Nursery 1", "Nursery 2"),
            F.regexp_replace("year_group", "Nursery ", "N")
        )
        .when(
            F.col("year_group") == "R",
            F.lit("Reception")
        )
        .when(
            F.col("year_group") == "Year 13",
            F.lit("Y13")
        )
        .when(
            F.col("year_group").rlike("^[0-9]+$"),

```

```

        F.concat(F.lit("Y"), F.col("year_group").cast("int"))
    )
    .when(
        F.col("year_group").rlike("^Y[0-9]{1,2}$"),
        F.concat(F.lit("Y"), F.regexp_replace("year_group",
"^[Yy]", "").cast("int"))
    )
    .otherwise(F.col("year_group"))
)
)

```

*# Get distinct values*

```

distinct_vals = df_tidy.select("year_group_tidy").distinct()
count_distinct = distinct_vals.count()

```

```

print(f"Number of distinct values in 'year_group_tidy':
{count_distinct}")
distinct_vals.show(truncate=False)

```

Number of distinct values in 'year\_group\_tidy': 18

```

+-----+
|year_group_tidy|
+-----+
|Y10
|Y12
|Y11
|Y13
|Y8
|Y7
|Y9
|Y6
|Y2
|Reception
|Y4
|Y3
|Y1
|N2
|N1
|Y5
|Y14
|NULL
+-----+

```

```

show_df_missing_breakdown(df_tidy.select("year_group_tidy"))

```

DataFrame has 16311626 rows and 1 columns.

Column	Null	EmptyStr	NA/NaNStr
NumericNaN	TotalMissing	%Missing	

```
#make a count of the number of students in each year group
df_tidy.groupBy("year_group_tidy").count().show()
```

```
df_selected.select("nc_year").distinct().show()
```

nc_year
7
11
8
9
10
12
13
3
5
6
R
1
N2
4

```
|      N1|
|      2|
+-----+
```

```
#count the number of students in each NC year
df_selected.groupBy("nc_year").count().show()
```

```
+-----+-----+
|nc_year|  count|
+-----+-----+
|      7| 588774|
|     11|4017479|
|      8|1924729|
|      9|3235343|
|     10|3994364|
|     12| 789109|
|     13| 824412|
|      3| 154541|
|      5| 197939|
|      6| 188708|
|      R|  35382|
|      1|  80604|
|     N2|  13989|
|      4| 160533|
|     N1|   1466|
|      2| 104254|
+-----+-----+
```

```
show_df_missing_breakdown(df_selected.select("nc_year"))
```

```
DataFrame has 16311626 rows and 1 columns.
```

Column	Null	EmptyStr	NA/NaNStr
NumericNaN	TotalMissing	%Missing	
nc_year	0	0	0
0	0	0.00%	

The **National Curriculum Year** is the most accurate representation of the year group, as it is based on the student's age. It is also less prone to errors, as it is not manually entered like the **year group** field. Additionally, there are no missing values in the **National Curriculum Year** field. Therefore, I will use this field instead of the **year group** field.

```
# number of distinct values in the columns all except
'student_forename', 'student_surname', and 'UPN'
```

```
show_distinct_counts_approx(df_selected.drop("student_forename",
```

```
"attendance_date", "student_surname", "UPN"))
```

Column	Approx Distinct Count
-----	-----
weekcommencing	183
leaving_date	74
academic_week_number	54
mark	49
year_group	31
nc_year	16
school	9
term	6
academic_year	5
gender	2
pupil_premium	2
school_type	2
session	2
is_approved_educational_activity	2
is_attend	2
is_auth_abs	2
late	2
late_unauthorised	2
missing	2
no_reason	2

All Column Approx Distinct Counts (showing top\_n only):

```
-----
-----
Py4JJavaError                                Traceback (most recent call
last)
Cell In[31], line 3
      1 # number of distinct values in the columns all except
      'student_forename', 'student_surname', and 'UPN'
----> 3
show_distinct_counts_approx(df_selected.drop("student_forename",
"attendance_date", "student_surname", "UPN"))

Cell In[6], line 53, in show_distinct_counts_approx(df, top_n, rsd)
     51 # Show only the top_n rows, so we don't blow up memory
     52 print("\nAll Column Approx Distinct Counts (showing top_n
only):")
--> 53 df_approx_counts.limit(top_n).show(truncate=False)

File ~\AppData\Roaming\Python\Python311\site-packages\pyspark\sql\
dataframe.py:947, in DataFrame.show(self, n, truncate, vertical)
     887 def show(self, n: int = 20, truncate: Union[bool, int] = True,
vertical: bool = False) -> None:
     888     """Prints the first ``n`` rows to the console.
```

```

889
890     .. versionadded:: 1.3.0
(...)
945     name | Bob
946     """
--> 947     print(self._show_string(n, truncate, vertical))

File ~\AppData\Roaming\Python\Python311\site-packages\pyspark\sql\
dataframe.py:978, in DataFrame._show_string(self, n, truncate,
vertical)
969 except ValueError:
970     raise PySparkTypeError(
971         error_class="NOT_BOOL",
972         message_parameters={
(...)
975         },
976     )
--> 978 return self._jdf.showString(n, int_truncate, vertical)

```

```

File ~\AppData\Roaming\Python\Python311\site-packages\py4j\
java_gateway.py:1322, in JavaMember.__call__(self, *args)
1316 command = proto.CALL_COMMAND_NAME + \
1317     self.command_header + \
1318     args_command + \
1319     proto.END_COMMAND_PART
1321 answer = self.gateway_client.send_command(command)
-> 1322 return_value = get_return_value(
1323     answer, self.gateway_client, self.target_id, self.name)
1325 for temp_arg in temp_args:
1326     if hasattr(temp_arg, "_detach"):

```

```

File ~\AppData\Roaming\Python\Python311\site-packages\pyspark\errors\
exceptions\captured.py:179, in capture_sql_exception.<locals>.deco(*a,
**kw)
177 def deco(*a: Any, **kw: Any) -> Any:
178     try:
--> 179         return f(*a, **kw)
180     except Py4JJavaError as e:
181         converted = convert_exception(e.java_exception)

```

```

File ~\AppData\Roaming\Python\Python311\site-packages\py4j\
protocol.py:326, in get_return_value(answer, gateway_client,
target_id, name)
324 value = OUTPUT_CONVERTER[type](answer[2:], gateway_client)
325 if answer[1] == REFERENCE_TYPE:
--> 326     raise Py4JJavaError(
327         "An error occurred while calling {0}{1}{2}.\n".
328         format(target_id, ".", name), value)
329 else:
330     raise Py4JError(

```

```
331         "An error occurred while calling {0}{1}{2}. Trace:\n{3}\n".\n332         format(target_id, ".", name, value))
```

```
Py4JJavaError: An error occurred while calling o4093.showString.\n: org.apache.spark.SparkException: Job aborted due to stage failure:\nTask 0 in stage 138.0 failed 1 times, most recent failure: Lost task\n0.0 in stage 138.0 (TID 303) (Saqib executor driver):\njava.net.SocketException: Connection reset\n    at java.net.SocketInputStream.read(Unknown Source)\n    at java.net.SocketInputStream.read(Unknown Source)\n    at java.io.BufferedInputStream.fill(Unknown Source)\n    at java.io.BufferedInputStream.read(Unknown Source)\n    at java.io.DataInputStream.readInt(Unknown Source)\n    at org.apache.spark.api.python.PythonRunner$\n$anon$3.read(PythonRunner.scala:774)\n    at org.apache.spark.api.python.PythonRunner$\n$anon$3.read(PythonRunner.scala:766)\n    at\norg.apache.spark.api.python.BasePythonRunner$ReaderIterator.hasNext(Py\nthonRunner.scala:525)\n    at\norg.apache.spark.InterruptibleIterator.hasNext(InterruptibleIterator.s\ncala:37)\n    at scala.collection.Iterator$$anon$11.hasNext(Iterator.scala:491)\n    at scala.collection.Iterator$$anon$10.hasNext(Iterator.scala:460)\n    at scala.collection.Iterator$$anon$10.hasNext(Iterator.scala:460)\n    at\norg.apache.spark.sql.catalyst.expressions.GeneratedClass$GeneratedIter\natorForCodegenStage1.processNext(Unknown Source)\n    at\norg.apache.spark.sql.execution.BufferedRowIterator.hasNext(BufferedRow\nIterator.java:43)\n    at\norg.apache.spark.sql.execution.WholeStageCodegenEvaluatorFactory$Whole\nStageCodegenPartitionEvaluator$\n$anon$1.hasNext(WholeStageCodegenEvaluatorFactory.scala:43)\n    at org.apache.spark.sql.execution.SparkPlan.\n$anonfun$getByteArrayRdd$1(SparkPlan.scala:388)\n    at org.apache.spark.rdd.RDD.\n$anonfun$mapPartitionsInternal$2(RDD.scala:893)\n    at org.apache.spark.rdd.RDD.\n$anonfun$mapPartitionsInternal$2$adapted(RDD.scala:893)\n    at\norg.apache.spark.rdd.MapPartitionsRDD.compute(MapPartitionsRDD.scala:5\n2)\n    at
```



```
org.apache.spark.rdd.RDD.computeOrReadCheckpoint(RDD.scala:367)
  at org.apache.spark.rdd.RDD.iterator(RDD.scala:331)
  at
org.apache.spark.scheduler.ResultTask.runTask(ResultTask.scala:93)
  at
org.apache.spark.TaskContext.runTaskWithListeners(TaskContext.scala:166)
  at org.apache.spark.scheduler.Task.run(Task.scala:141)
  at org.apache.spark.executor.Executor$TaskRunner.
$anonfun$run$4(Executor.scala:620)
  at
org.apache.spark.util.SparkErrorUtils.tryWithSafeFinally(SparkErrorUtils.scala:64)
  at org.apache.spark.util.SparkErrorUtils.tryWithSafeFinally$
(SparkErrorUtils.scala:61)
  at
org.apache.spark.util.Utils$.tryWithSafeFinally(Utils.scala:94)
  at
org.apache.spark.executor.Executor$TaskRunner.run(Executor.scala:623)
  at java.util.concurrent.ThreadPoolExecutor.runWorker(Unknown
Source)
  at java.util.concurrent.ThreadPoolExecutor$Worker.run(Unknown
Source)
  at java.lang.Thread.run(Unknown Source)
```

Driver stacktrace:

```
  at
org.apache.spark.scheduler.DAGScheduler.failJobAndIndependentStages(DA
GScheduler.scala:2856)
  at org.apache.spark.scheduler.DAGScheduler.
$anonfun$abortStage$2(DAGScheduler.scala:2792)
  at org.apache.spark.scheduler.DAGScheduler.
$anonfun$abortStage$2$adapted(DAGScheduler.scala:2791)
  at
scala.collection.mutable.ResizableArray.foreach(ResizableArray.scala:62)
  at scala.collection.mutable.ResizableArray.foreach$
(ResizableArray.scala:55)
  at
scala.collection.mutable.ArrayBuffer.foreach(ArrayBuffer.scala:49)
  at
org.apache.spark.scheduler.DAGScheduler.abortStage(DAGScheduler.scala:
2791)
  at org.apache.spark.scheduler.DAGScheduler.
$anonfun$handleTaskSetFailed$1(DAGScheduler.scala:1247)
  at org.apache.spark.scheduler.DAGScheduler.
$anonfun$handleTaskSetFailed$1$adapted(DAGScheduler.scala:1247)
  at scala.Option.foreach(Option.scala:407)
  at
```

```
org.apache.spark.scheduler.DAGScheduler.handleTaskSetFailed(DAGScheduler.scala:1247)
    at
org.apache.spark.scheduler.DAGSchedulerEventProcessLoop.doOnReceive(DAGScheduler.scala:3060)
    at
org.apache.spark.scheduler.DAGSchedulerEventProcessLoop.onReceive(DAGScheduler.scala:2994)
    at
org.apache.spark.scheduler.DAGSchedulerEventProcessLoop.onReceive(DAGScheduler.scala:2983)
    at org.apache.spark.util.EventLoop$.anonfun$run(EventLoop.scala:49)
    at
org.apache.spark.scheduler.DAGScheduler.runJob(DAGScheduler.scala:989)
    at org.apache.spark.SparkContext.runJob(SparkContext.scala:2393)
    at org.apache.spark.SparkContext.runJob(SparkContext.scala:2414)
    at org.apache.spark.SparkContext.runJob(SparkContext.scala:2433)
    at
org.apache.spark.sql.execution.SparkPlan.executeTake(SparkPlan.scala:530)
    at
org.apache.spark.sql.execution.SparkPlan.executeTake(SparkPlan.scala:483)
    at
org.apache.spark.sql.execution.CollectLimitExec.executeCollect(limit.scala:61)
    at
org.apache.spark.sql.Dataset.collectFromPlan(Dataset.scala:4333)
    at org.apache.spark.sql.Dataset$.anonfun$head$1(Dataset.scala:3316)
    at org.apache.spark.sql.Dataset$.anonfun$withAction$2(Dataset.scala:4323)
    at
org.apache.spark.sql.execution.QueryExecution$.withInternalError(QueryExecution.scala:546)
    at org.apache.spark.sql.Dataset$.anonfun$withAction$1(Dataset.scala:4321)
    at org.apache.spark.sql.execution.SQLExecution$.anonfun$withNewExecutionId$6(SQLExecution.scala:125)
    at
org.apache.spark.sql.execution.SQLExecution$.withSQLConfPropagated(SQLExecution.scala:201)
    at org.apache.spark.sql.execution.SQLExecution$.anonfun$withNewExecutionId$1(SQLExecution.scala:108)
    at
org.apache.spark.sql.SparkSession.withActive(SparkSession.scala:900)
    at
```

```
org.apache.spark.sql.execution.SQLExecution$.withNewExecutionId(SQLExecution.scala:66)
    at org.apache.spark.sql.Dataset.withAction(Dataset.scala:4321)
    at org.apache.spark.sql.Dataset.head(Dataset.scala:3316)
    at org.apache.spark.sql.Dataset.take(Dataset.scala:3539)
    at org.apache.spark.sql.Dataset.getRows(Dataset.scala:280)
    at org.apache.spark.sql.Dataset.showString(Dataset.scala:315)
    at sun.reflect.NativeMethodAccessorImpl.invoke0(Native Method)
    at sun.reflect.NativeMethodAccessorImpl.invoke(Unknown Source)
    at sun.reflect.DelegatingMethodAccessorImpl.invoke(Unknown Source)
    at java.lang.reflect.Method.invoke(Unknown Source)
    at py4j.reflection.MethodInvoker.invoke(MethodInvoker.java:244)
    at py4j.reflection.ReflectionEngine.invoke(ReflectionEngine.java:374)
    at py4j.Gateway.invoke(Gateway.java:282)
    at py4j.commands.AbstractCommand.invokeMethod(AbstractCommand.java:132)
    at py4j.commands.CallCommand.execute(CallCommand.java:79)
    at py4j.ClientServerConnection.waitForCommands(ClientServerConnection.java:182)
    at py4j.ClientServerConnection.run(ClientServerConnection.java:106)
    at java.lang.Thread.run(Unknown Source)
Caused by: java.net.SocketException: Connection reset
    at java.net.SocketInputStream.read(Unknown Source)
    at java.net.SocketInputStream.read(Unknown Source)
    at java.io.BufferedInputStream.fill(Unknown Source)
    at java.io.BufferedInputStream.read(Unknown Source)
    at java.io.DataInputStream.readInt(Unknown Source)
    at org.apache.spark.api.python.PythonRunner$.anon$3.read(PythonRunner.scala:774)
    at org.apache.spark.api.python.PythonRunner$.anon$3.read(PythonRunner.scala:766)
    at org.apache.spark.api.python.BasePythonRunner$ReaderIterator.hasNext(PythonRunner.scala:525)
    at org.apache.spark.InterruptibleIterator.hasNext(InterruptibleIterator.scala:37)
    at scala.collection.Iterator$$anon$11.hasNext(Iterator.scala:491)
    at scala.collection.Iterator$$anon$10.hasNext(Iterator.scala:460)
    at scala.collection.Iterator$$anon$10.hasNext(Iterator.scala:460)
    at org.apache.spark.sql.catalyst.expressions.GeneratedClass$GeneratedIteratorForCodegenStage1.processNext(Unknown Source)
```

```

    at
org.apache.spark.sql.execution.BufferedRowIterator.hasNext(BufferedRow
Iterator.java:43)
    at
org.apache.spark.sql.execution.WholeStageCodegenEvaluatorFactory$Whole
StageCodegenPartitionEvaluator$
$anon$1.hasNext(WholeStageCodegenEvaluatorFactory.scala:43)
    at org.apache.spark.sql.execution.SparkPlan.
$anonfun$getByteArrayRdd$1(SparkPlan.scala:388)
    at org.apache.spark.rdd.RDD.
$anonfun$mapPartitionsInternal$2(RDD.scala:893)
    at org.apache.spark.rdd.RDD.
$anonfun$mapPartitionsInternal$2$adapted(RDD.scala:893)
    at
org.apache.spark.rdd.MapPartitionsRDD.compute(MapPartitionsRDD.scala:5
2)
    at
org.apache.spark.rdd.RDD.computeOrReadCheckpoint(RDD.scala:367)
    at org.apache.spark.rdd.RDD.iterator(RDD.scala:331)
    at
org.apache.spark.scheduler.ResultTask.runTask(ResultTask.scala:93)
    at
org.apache.spark.TaskContext.runTaskWithListeners(TaskContext.scala:16
6)
    at org.apache.spark.scheduler.Task.run(Task.scala:141)
    at org.apache.spark.executor.Executor$TaskRunner.
$anonfun$run$4(Executor.scala:620)
    at
org.apache.spark.util.SparkErrorUtils.tryWithSafeFinally(SparkErrorUti
ls.scala:64)
    at org.apache.spark.util.SparkErrorUtils.tryWithSafeFinally$
(SparkErrorUtils.scala:61)
    at
org.apache.spark.util.Utils$.tryWithSafeFinally(Utils.scala:94)
    at
org.apache.spark.executor.Executor$TaskRunner.run(Executor.scala:623)
    at java.util.concurrent.ThreadPoolExecutor.runWorker(Unknown
Source)
    at java.util.concurrent.ThreadPoolExecutor$Worker.run(Unknown
Source)
    ... 1 more

```

```
#list the distinct values in the column 'mark'
```

```
df_selected.select("mark").distinct().show()
```

```
df_selected.select("academic_year").distinct().show()
```

```

+----+
|mark|
+----+

```

```

|      K |
| I01 |
| I02 |
|      E |
|      B |
|      Y |
| X05 |
|      L |
|      M |
|      V |
|      U |
|      O |
|      D |
| Y6 |
|      C |
|      J |
|      - |
|      Z |
| X06 |
|      / |
+-----+

```

only showing top 20 rows

```

+-----+
|academic_year|
+-----+
|      2021/2022 |
|      2024/2025 |
|      2022/2023 |
|      2023/2024 |
|      2020/2021 |
+-----+

```

- The **Mark** category may require further investigation by the data analyst to identify rows where the attendance data is ambiguous.
- Additionally, the **weekcommencing** column contains 181 unique values, whereas the **academic\_week\_number** column contains 54 unique values. This discrepancy is expected, as the **academic\_week\_number** is based on the academic year and repeats across years, while **weekcommencing** is based on actual dates. The **weekcommencing** field may be more suitable for categorising the data based on week numbers.

```
df_selected.show(truncate=False)
```

```

+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+

```

```

+-----+-----+-----+-----+-----+
+----+
|gender|student_forename|student_surname|pupil_premium|year_group|
nc_year|school_type|school   |establishment_number|la_code|
attendance_date|academic_year|academic_week_number|term   |
weekcommencing|mark|session|is_approved_educational_activity|
is_attend|is_auth_abs|late|late_unauthorised|missing|no_reason|
is_possible|is_present|is_unauth_abs|current_student|leaving_date|UPN
|
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+----+
|FEMALE|Theresa          |Hester          |False          |8          |8
|SECONDARY |Academy 4|XXXXXX          |XXXXXX |2023-11-13    |
2023/2024   |12          |Term 2|w/c 13/11/2023|/          |AM          |
0.0          |1.0          |0.0          |0.0 |0.0
|0.0          |0.0          |1.0          |1.0          |0.0          |1
|NULL          |3579|
|MALE |Daniel          |Garcia          |False          |9          |9
|SECONDARY |Academy 4|XXXXXX          |XXXXXX |2023-09-19    |
2023/2024   |4          |Term 1|w/c 18/09/2023|\          |PM          |
0.0          |1.0          |0.0          |0.0 |0.0
|0.0          |0.0          |1.0          |1.0          |0.0          |1
|NULL          |4886|
|MALE |Alicia          |Fox             |True           |7          |7
|SECONDARY |Academy 4|XXXXXX          |XXXXXX |2024-09-12    |
2024/2025   |2          |Term 1|w/c 09/09/2024|/          |AM          |
0.0          |1.0          |0.0          |0.0 |0.0
|0.0          |0.0          |1.0          |1.0          |0.0          |1
|NULL          |166 |
|MALE |David          |Martinez        |False          |8          |8
|SECONDARY |Academy 4|XXXXXX          |XXXXXX |2024-08-11    |
2024/2025   |50         |Term 6|w/c 05/08/2024|#          |PM          |
0.0          |0.0          |0.0          |0.0 |0.0
|0.0          |1.0          |0.0          |0.0          |0.0          |1
|NULL          |1235|
|MALE |Stacy          |Avery           |False          |9          |9
|SECONDARY |Academy 4|XXXXXX          |XXXXXX |2024-11-15    |
2024/2025   |11         |Term 2|w/c 11/11/2024|\          |PM          |
0.0          |1.0          |0.0          |0.0 |0.0
|0.0          |0.0          |1.0          |1.0          |0.0          |1
|NULL          |9728|
|FEMALE|Brett          |Williams        |True           |11         |11
|SECONDARY |Academy 4|XXXXXX          |XXXXXX |2024-10-09    |
2024/2025   |6          |Term 1|w/c 07/10/2024|G          |PM          |

```

0.0				0.0	0.0	0.0	0.0
0.0	0.0	1.0		0.0	1.0		1
NULL	7267						
FEMALE	Brett		Williams	True		11	11
SECONDARY	Academy 4	XXXXXX		XXXXXX	2024-09-29		
2024/2025	4			Term 1	w/c 23/09/2024	#	AM
0.0				0.0	0.0	0.0	0.0
0.0	1.0	0.0		0.0	0.0		1
NULL	7267						
FEMALE	Theresa		Hester	False		8	8
SECONDARY	Academy 4	XXXXXX		XXXXXX	2024-08-21		
2024/2025	52			Term 6	w/c 19/08/2024	#	PM
0.0				0.0	0.0	0.0	0.0
0.0	1.0	0.0		0.0	0.0		1
NULL	3579						
MALE	Stacy		Avery	False		9	9
SECONDARY	Academy 4	XXXXXX		XXXXXX	2024-11-09		
2024/2025	10			Term 2	w/c 04/11/2024	#	PM
0.0				0.0	0.0	0.0	0.0
0.0	1.0	0.0		0.0	0.0		1
NULL	9728						
MALE	Daniel		Garcia	False		9	9
SECONDARY	Academy 4	XXXXXX		XXXXXX	2024-08-27		
2024/2025	53			Term 6	w/c 26/08/2024	#	PM
0.0				0.0	0.0	0.0	0.0
0.0	1.0	0.0		0.0	0.0		1
NULL	4886						
MALE	Daniel		Garcia	False		9	9
SECONDARY	Academy 4	XXXXXX		XXXXXX	2024-11-27		
2024/2025	13			Term 2	w/c 25/11/2024	\	PM
0.0				1.0	0.0	0.0	0.0
0.0	0.0	1.0		1.0	0.0		1
NULL	4886						
FEMALE	Theresa		Hester	False		8	8
SECONDARY	Academy 4	XXXXXX		XXXXXX	2024-11-05		
2024/2025	10			Term 2	w/c 04/11/2024	\	PM
0.0				1.0	0.0	0.0	0.0
0.0	0.0	1.0		1.0	0.0		1
NULL	3579						
FEMALE	Brett		Williams	True		11	11
SECONDARY	Academy 4	XXXXXX		XXXXXX	2024-10-20		
2024/2025	7			Term 1	w/c 14/10/2024	#	PM
0.0				0.0	0.0	0.0	0.0
0.0	1.0	0.0		0.0	0.0		1
NULL	7267						
MALE	Daniel		Garcia	False		9	9
SECONDARY	Academy 4	XXXXXX		XXXXXX	2024-12-01		
2024/2025	13			Term 2	w/c 25/11/2024	#	PM
0.0				0.0	0.0	0.0	0.0

0.0	1.0	0.0	0.0	0.0	1
NULL	4886				
FEMALE Theresa	Hester	False	8	8	
SECONDARY	Academy 4 XXXXXX	XXXXXX	2024-12-05		
2024/2025	14	Term 2 w/c 02/12/2024 \	PM		
0.0		1.0	0.0	0.0	0.0
0.0	0.0	1.0	1.0	0.0	1
NULL	3579				
MALE Daniel	Garcia	False	9	9	
SECONDARY	Academy 4 XXXXXX	XXXXXX	2024-11-21		
2024/2025	12	Term 2 w/c 18/11/2024 /	AM		
0.0		1.0	0.0	0.0	0.0
0.0	0.0	1.0	1.0	0.0	1
NULL	4886				
FEMALE Theresa	Hester	False	8	8	
SECONDARY	Academy 4 XXXXXX	XXXXXX	2024-10-09		
2024/2025	6	Term 1 w/c 07/10/2024 \	PM		
0.0		1.0	0.0	0.0	0.0
0.0	0.0	1.0	1.0	0.0	1
NULL	3579				
MALE David	Martinez	False	8	8	
SECONDARY	Academy 4 XXXXXX	XXXXXX	2024-10-23		
2024/2025	8	Term 1 w/c 21/10/2024 #	PM		
0.0		0.0	0.0	0.0	0.0
0.0	1.0	0.0	0.0	0.0	1
NULL	1235				
FEMALE Theresa	Hester	False	8	8	
SECONDARY	Academy 4 XXXXXX	XXXXXX	2024-11-07		
2024/2025	10	Term 2 w/c 04/11/2024 /	AM		
0.0		1.0	0.0	0.0	0.0
0.0	0.0	1.0	1.0	0.0	1
NULL	3579				
MALE David	Martinez	False	8	8	
SECONDARY	Academy 4 XXXXXX	XXXXXX	2024-12-04		
2024/2025	14	Term 2 w/c 02/12/2024 \	PM		
0.0		1.0	0.0	0.0	0.0
0.0	0.0	1.0	1.0	0.0	1
NULL	1235				

```

+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+

```

only showing top 20 rows



## 6. Rationale for Selecting Certain Columns:

- **Names and UPN:** These are essential for identifying students and their attendance records, particularly if a report needs to deep dive into individual student attendance. They can also help identify students with chronic absenteeism or other attendance-related issues and ensure the quality of the data.
- **Year group:** This column is used to calculate the attendance percentage for each school on a weekly basis by year group.
- **Establishment name and number:** These are crucial for identifying the school and its location, which can be useful for comparing attendance rates across different schools or regions. Although this dataset anonymises these fields (indicated by XXXX), I am assuming they would be provided in a real-world scenario for more accurate analysis.
- **Attendance date, mark, and session:** These columns are essential for tracking student attendance on a daily basis and identifying patterns of absence or lateness. They will also be used later to create a unique identifier for each attendance record.
- **Detailed attendance status columns:**
  - **is\_present, is\_possible, is\_auth\_abs, is\_unauth\_abs, is\_late\_L, is\_late\_U, is\_missing, is\_nr, and is\_aea:** These provide detailed information about the student's attendance status, including whether the absence was authorised, unauthorised, or due to other reasons. This information can help identify trends in attendance and inform interventions to improve attendance rates.
- **Academic year, week number, and term\_session:** These are necessary for aggregating attendance data over specific time periods (e.g., weekly, termly) and tracking trends across the academic year. This information can highlight seasonal patterns in attendance and guide targeted interventions.
- **Week commencing:** Provided in two formats to allow flexibility in reporting and analysis.

## 7. Data Integrity Check

I create a unique identifier for each attendance record by concatenating the student\_id (UPN), date, session (AM/PM) columns. This identifier will be used to check for duplicate records and ensure data integrity.

```
# 1. Create a new field by concatenating UPN and attendance_date and
# session (AM/PM) with an underscore separator
# this will be used to identify the unique attendance record for each
# student per day per session
df_with_combined = df_selected.withColumn(
    "UPN_AttendanceDate",
    F.concat_ws("_", F.col("UPN"), F.col("attendance_date"),
    F.col("session"))
```

```

)

# 2. Group by this new field and filter for count == 1 (i.e. unique)
df_valid = (
    df_with_combined
    .groupBy("UPN_AttendanceDate")
    .count()
    .filter(F.col("count") == 1) #each student can have only one
    attendance per day per session AM or PM
)

# 3. Group by this new field and filter for count > 1 (i.e.
duplicates)
df_invalid = (
    df_with_combined
    .groupBy("UPN_AttendanceDate")
    .count()
    .filter(F.col("count") > 1)
)

```

```
df_with_combined.show(truncate=False)
```

```

+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
|gender|student_forename|student_surname|pupil_premium|year_group|
nc_year|school_type|school   |establishment_number|la_code|
attendance_date|academic_year|academic_week_number|term   |
weekcommencing|mark|session|is_approved_educational_activity|
is_attend|is_auth_abs|late|late_unauthorised|missing|no_reason|
is_possible|is_present|is_unauth_abs|current_student|leaving_date|UPN
|UPN_AttendanceDate|
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
|FEMALE|Theresa          |Hester          |False          |8          |8
|SECONDARY |Academy 4|XXXXXX          |XXXXXX |2023-11-13 |
2023/2024   |12          |Term 2|w/c 13/11/2023|/          |AM          |
0.0          |0.0          |1.0          |1.0          |0.0          |0.0 |0.0
|0.0          |0.0          |1.0          |1.0          |0.0          |1

```

NULL	3579	3579_2023-11-13_AM							
MALE	Daniel	Garcia	False	9	9				
SECONDARY	Academy 4	XXXXXX	XXXXXX	2023-09-19					
2023/2024	4		Term 1	w/c 18/09/2023	\	PM			
0.0			1.0	0.0	0.0	0.0			
0.0	0.0	1.0	1.0	0.0	1				
NULL	4886	4886_2023-09-19_PM							
MALE	Alicia	Fox	True	7	7				
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-09-12					
2024/2025	2		Term 1	w/c 09/09/2024	/	AM			
0.0			1.0	0.0	0.0	0.0			
0.0	0.0	1.0	1.0	0.0	1				
NULL	166	166_2024-09-12_AM							
MALE	David	Martinez	False	8	8				
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-08-11					
2024/2025	50		Term 6	w/c 05/08/2024	#	PM			
0.0			0.0	0.0	0.0	0.0			
0.0	1.0	0.0	0.0	0.0	1				
NULL	1235	1235_2024-08-11_PM							
MALE	Stacy	Avery	False	9	9				
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-11-15					
2024/2025	11		Term 2	w/c 11/11/2024	\	PM			
0.0			1.0	0.0	0.0	0.0			
0.0	0.0	1.0	1.0	0.0	1				
NULL	9728	9728_2024-11-15_PM							
FEMALE	Brett	Williams	True	11	11				
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-10-09					
2024/2025	6		Term 1	w/c 07/10/2024	G	PM			
0.0			0.0	0.0	0.0	0.0			
0.0	0.0	1.0	0.0	1.0	1				
NULL	7267	7267_2024-10-09_PM							
FEMALE	Brett	Williams	True	11	11				
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-09-29					
2024/2025	4		Term 1	w/c 23/09/2024	#	AM			
0.0			0.0	0.0	0.0	0.0			
0.0	1.0	0.0	0.0	0.0	1				
NULL	7267	7267_2024-09-29_AM							
FEMALE	Theresa	Hester	False	8	8				
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-08-21					
2024/2025	52		Term 6	w/c 19/08/2024	#	PM			
0.0			0.0	0.0	0.0	0.0			
0.0	1.0	0.0	0.0	0.0	1				
NULL	3579	3579_2024-08-21_PM							
MALE	Stacy	Avery	False	9	9				
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-11-09					
2024/2025	10		Term 2	w/c 04/11/2024	#	PM			
0.0			0.0	0.0	0.0	0.0			
0.0	1.0	0.0	0.0	0.0	1				
NULL	9728	9728_2024-11-09_PM							

MALE	Daniel	Garcia	False	9	9
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-08-27	
2024/2025	53		Term 6	w/c 26/08/2024	# PM
0.0			0.0	0.0	0.0
0.0	1.0	0.0	0.0	0.0	1
NULL	4886	4886_2024-08-27_PM			
MALE	Daniel	Garcia	False	9	9
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-11-27	
2024/2025	13		Term 2	w/c 25/11/2024	\ PM
0.0			1.0	0.0	0.0
0.0	0.0	1.0	1.0	0.0	1
NULL	4886	4886_2024-11-27_PM			
FEMALE	Theresa	Hester	False	8	8
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-11-05	
2024/2025	10		Term 2	w/c 04/11/2024	\ PM
0.0			1.0	0.0	0.0
0.0	0.0	1.0	1.0	0.0	1
NULL	3579	3579_2024-11-05_PM			
FEMALE	Brett	Williams	True	11	11
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-10-20	
2024/2025	7		Term 1	w/c 14/10/2024	# PM
0.0			0.0	0.0	0.0
0.0	1.0	0.0	0.0	0.0	1
NULL	7267	7267_2024-10-20_PM			
MALE	Daniel	Garcia	False	9	9
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-12-01	
2024/2025	13		Term 2	w/c 25/11/2024	# PM
0.0			0.0	0.0	0.0
0.0	1.0	0.0	0.0	0.0	1
NULL	4886	4886_2024-12-01_PM			
FEMALE	Theresa	Hester	False	8	8
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-12-05	
2024/2025	14		Term 2	w/c 02/12/2024	\ PM
0.0			1.0	0.0	0.0
0.0	0.0	1.0	1.0	0.0	1
NULL	3579	3579_2024-12-05_PM			
MALE	Daniel	Garcia	False	9	9
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-11-21	
2024/2025	12		Term 2	w/c 18/11/2024	/ AM
0.0			1.0	0.0	0.0
0.0	0.0	1.0	1.0	0.0	1
NULL	4886	4886_2024-11-21_AM			
FEMALE	Theresa	Hester	False	8	8
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-10-09	
2024/2025	6		Term 1	w/c 07/10/2024	\ PM
0.0			1.0	0.0	0.0
0.0	0.0	1.0	1.0	0.0	1
NULL	3579	3579_2024-10-09_PM			
MALE	David	Martinez	False	8	8

SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-10-23	
2024/2025	8		Term 1	w/c 21/10/2024	# PM
0.0		0.0	0.0	0.0	0.0
0.0	1.0	0.0	0.0	0.0	1
NULL	1235	1235_2024-10-23_PM			
FEMALE	Theresa	Hester	False	8	8
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-11-07	
2024/2025	10		Term 2	w/c 04/11/2024	/ AM
0.0		1.0	0.0	0.0	0.0
0.0	0.0	1.0	1.0	0.0	1
NULL	3579	3579_2024-11-07_AM			
MALE	David	Martinez	False	8	8
SECONDARY	Academy 4	XXXXXX	XXXXXX	2024-12-04	
2024/2025	14		Term 2	w/c 02/12/2024	\ PM
0.0		1.0	0.0	0.0	0.0
0.0	0.0	1.0	1.0	0.0	1
NULL	1235	1235_2024-12-04_PM			

only showing top 20 rows

```
df_invalid.show(truncate=False)
```

UPN_AttendanceDate	count
5798_2024-08-27_PM	2
1052_2024-08-09_AM	2
4878_2024-09-14_AM	2
8152_2022-07-21_PM	2
8152_2022-09-18_PM	2
8152_2023-12-25_PM	2
8152_2022-11-27_PM	2
8152_2022-07-02_AM	2
8152_2022-07-22_PM	2
8152_2024-04-07_AM	2
2584_2024-10-04_PM	2
9403_2024-08-14_PM	2
5008_2022-07-12_AM	2
5008_2022-07-27_AM	2
5008_2022-05-21_AM	2
5008_2022-04-18_AM	2
3924_2024-08-26_AM	2
2899_2024-09-08_AM	2

```
|3898_2024-07-07_PM|2|
|3898_2024-06-25_PM|2|
+-----+-----+
only showing top 20 rows
```

```
print(f"Total rows with original df: {df_joined_renamed.count()}")
print(f"Total rows with selected data df: {df_selected.count()}")
print(f"Total rows valid data: {df_valid.count()}")
print(f"Total rows invalid data: {df_invalid.count()}")

Total rows with original df: 16311626
Total rows with selected data df: 16311626
Total rows valid data: 15983332
Total rows invalid data: 164114
```

So there are 164114 duplicate rows. I will label these as invalid for the data analyst. So as to not prevent any data leakage and to be able to filter easily, I will add a column called 'status' and 'count' and label each row with either valid if it occurs once, and if more than once then 'invalid'

```
# 1. Aggregate all keys with their counts
df_counts = (
    df_with_combined
    .groupBy("UPN_AttendanceDate")
    .agg(F.count("*").alias("count"))
    .withColumn(
        "status",
        F.when(F.col("count") == 1, "valid").otherwise("invalid")
    )
)

# 2. Join back to original rows to get the full data plus the status
df_with_status = (
    df_with_combined.alias("a")
    .join(df_counts.alias("b"), on="UPN_AttendanceDate", how="left")
    .select("a.*", "b.count", "b.status")
)

df_with_status.show(truncate=False)
```

```
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
```

```

+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
|UPN_AttendanceDate      |gender|student_forename|student_surname|
pupil_premium|year_group|nc_year|school_type|school      |
establishment_number|la_code|attendance_date|academic_year|
academic_week_number|term   |weekcommencing|mark|session|
is_approved_educational_activity|is_attend|is_auth_abs|late|
late_unauthorised|missing|no_reason|is_possible|is_present|
is_unauth_abs|current_student|leaving_date|UPN      |count|status |
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
|10623_2023-08-21_PM      |MALE  |Felicia      |Vega      |
False      |11     |11     |SECONDARY |Academy 5|XXXXXX
|XXXXXX |2023-08-21      |2023/2024   |52      |Term 6|w/c
21/08/2023|#      |PM      |0.0      |0.0      |0.0      |0.0
|0.0 |0.0      |0.0      |1.0      |0.0      |0.0      |0.0
|1      |nan     |10623      |1      |valid |
|10888_2024-11-24_PM      |MALE  |James      |Harris      |
False      |Y11    |11     |SECONDARY |Academy 3|XXXXXX
|XXXXXX |2024-11-24      |2024/2025   |12      |Term 2|w/c
18/11/2024|#      |PM      |0.0      |0.0      |0.0      |0.0
|0.0 |0.0      |0.0      |1.0      |0.0      |0.0      |0.0
|1      |NULL    |10888      |1      |valid |
|1279_2024-09-08_AM      |MALE  |Austin      |Johnson      |
False      |9      |9      |SECONDARY |Academy 5|XXXXXX
|XXXXXX |2024-09-08      |2024/2025   |1      |Term 1|w/c
02/09/2024|#      |AM      |0.0      |0.0      |0.0      |0.0
|0.0 |0.0      |0.0      |1.0      |0.0      |0.0      |0.0
|1      |nan     |1279      |1      |valid |
|2433_2023-07-22_AM      |FEMALE|Kenneth      |Evans      |
False      |11     |11     |ALL THROUGH|Academy 9|XXXXXX
|XXXXXX |2023-07-22      |2022/2023   |47      |Term 6|w/c
17/07/2023|#      |AM      |0.0      |0.0      |0.0      |0.0
|0.0 |0.0      |0.0      |1.0      |0.0      |0.0      |0.0
|1      |NULL    |2433      |1      |valid |
|3416_2024-09-08_AM      |FEMALE|Jennifer      |Dawson      |
False      |9      |9      |SECONDARY |Academy 5|XXXXXX
|XXXXXX |2024-09-08      |2024/2025   |1      |Term 1|w/c
02/09/2024|#      |AM      |0.0      |0.0      |0.0      |0.0
|0.0 |0.0      |0.0      |1.0      |0.0      |0.0      |0.0
|1      |nan     |3416      |1      |valid |
|4523_2024-06-22_PM      |FEMALE|Heather      |Steele      |True
|10      |10     |SECONDARY |Academy 2|XXXXXX      |XXXXXX
|2024-06-22      |2023/2024   |43      |Term 6|w/c

```

17/06/2024	#	PM	0.0			0.0	0.0
0.0	0.0		0.0	1.0	0.0	0.0	0.0
1		NULL	4523	1	valid		
5833_2023-12-30	PM	MALE	Audrey		Berry		True
10	10	ALL THROUGH	Academy 9	XXXXXX			XXXXXX
2023-12-30	2023/2024	18			Term 2	w/c	
25/12/2023	#	PM	0.0			0.0	0.0
0.0	0.0		0.0	1.0	0.0	0.0	0.0
1		NULL	5833	1	valid		
6680_2024-04-17	PM	FEMALE	Julie		Garner		
False	8	8	ALL THROUGH	Academy 9	XXXXXX		
XXXXXX	2024-04-17	2023/2024	34			Term 5	w/c
15/04/2024	I	PM	0.0			0.0	1.0
0.0	0.0		0.0	0.0	1.0	0.0	0.0
1		NULL	6680	1	valid		
7706_2024-11-18	AM	MALE	Joseph		Baker		
False	Y09	9	SECONDARY	Academy 3	XXXXXX		
XXXXXX	2024-11-18	2024/2025	12			Term 2	w/c
18/11/2024	/	AM	0.0			1.0	0.0
0.0	0.0		0.0	0.0	1.0	1.0	0.0
1		NULL	7706	1	valid		
8589937817_2023-05-30	AM	FEMALE	Timothy		Castillo		
False	Y10	10	SECONDARY	Academy 6	XXXXXX		
XXXXXX	2023-05-30	2022/2023	40			Term 5	w/c
29/05/2023	#	AM	0.0			0.0	0.0
0.0	0.0		0.0	1.0	0.0	0.0	0.0
1		nan	8589937817	1	valid		
8589938741_2022-08-09	PM	FEMALE	Alec		Smith		True
11	11	SECONDARY	Academy 4	XXXXXX			XXXXXX
2022-08-09	2022/2023	50				Term 6	w/c
08/08/2022	#	PM	0.0			0.0	0.0
0.0	0.0		0.0	1.0	0.0	0.0	0.0
1		NULL	8589938741	1	valid		
8589939203_2022-05-27	PM	FEMALE	Maria		Gallegos		True
11	11	SECONDARY	Academy 4	XXXXXX			XXXXXX
2022-05-27	2021/2022	39				Term 5	w/c
23/05/2022	\	PM	0.0			1.0	0.0
0.0	0.0		0.0	0.0	1.0	1.0	0.0
1		NULL	8589939203	2	invalid		
8589940349_2023-11-05	AM	MALE	Jennifer		Ramsey		
False	9	9	SECONDARY	Academy 2	XXXXXX		
XXXXXX	2023-11-05	2023/2024	10			Term 2	w/c
30/10/2023	#	AM	0.0			0.0	0.0
0.0	0.0		0.0	1.0	0.0	0.0	0.0
1		NULL	8589940349	1	valid		
871_2022-07-17	PM	MALE	Christopher		Miller		
False	13	13	SECONDARY	Academy 2	XXXXXX		
XXXXXX	2022-07-17	2021/2022	46			Term 6	w/c
11/07/2022	#	PM	0.0			0.0	0.0





```

df_validf_rows = df_with_status.filter("status == 'valid'")
df_invalidf_rows = df_with_status.filter("status == 'invalid'")

#count the number of rows in the df_valid_rows
print(f"Total rows valid data: {df_validf_rows.count()}")

#count the number of rows in the df_invalid_rows
print(f"Total rows invalid data: {df_invalidf_rows.count()}")

Total rows valid data: 15983332
Total rows invalid data: 328294

```

The invalid data is approximately twice that of the previous count, as now it is counting each instance where as before it was counting the grouping of the data.

## 8. Summary Table

Going forward, I will use the following to filter rows based on their status:

- `df_valid = df[df['status'] == 'valid']` to filter out the valid rows.
- `df_invalid_rows = df[df['status'] == 'invalid']` to filter out the invalid rows, which can be used to investigate duplicates further if needed.

I will create an initial summary table that contains the attendance percentage for each school on a weekly basis by the **Year Group** of the student. The table will include the following columns:

- **School:** The name of the school.
- **week\_number:** The specific week based on the week commencing date.
- **year\_group:** The year group of the student based on the National Curriculum Year.
- **attendance\_percentage:** The percentage of attendance for the school in the specified week and year group.

The summary will be sorted based on the school name, week number, and year group as per the data analyst's request.

I will then print summary statistics for this data, including the **mean, median, minimum, maximum, and standard deviation** of the attendance percentage across all schools, weeks, and year groups.

Finally, I will write the summary table to `fact_AttendanceSummary` in **Parquet** format for further analysis and reporting by the Data Analyst.

```

df_summary = (
    df_validf_rows
    .groupBy("school", "nc_year", "weekcommencing")
    .agg(
        F.round(

```

```

        (F.sum("is_attend") / F.sum("is_possible") * 100), 1
    ).alias("attendance_percentage")
)

```

```
#df_summary.limit(20).show(truncate=False)
```

```
df_summary.show(truncate=False)
```

school	nc_year	weekcommencing	attendance_percentage
Academy 8	13	w/c 04/09/2023	99.4
Academy 7	13	w/c 28/02/2022	98.2
Academy 9	3	w/c 28/11/2022	93.1
Academy 8	11	w/c 21/03/2022	93.8
Academy 8	11	w/c 15/04/2024	92.9
Academy 7	9	w/c 21/08/2023	NULL
Academy 2	13	w/c 04/03/2024	95.9
Academy 3	8	w/c 16/10/2023	96.3
Academy 8	11	w/c 01/08/2022	NULL
Academy 4	9	w/c 15/05/2023	92.7
Academy 7	7	w/c 18/11/2024	92.7
Academy 5	13	w/c 20/02/2023	97.2
Academy 5	13	w/c 11/03/2024	93.6
Academy 4	13	w/c 15/11/2021	94.2
Academy 9	9	w/c 12/09/2022	96.2
Academy 5	10	w/c 25/03/2024	93.1
Academy 9	11	w/c 28/08/2023	NULL
Academy 9	11	w/c 25/03/2024	84.9
Academy 9	11	w/c 25/07/2022	82.4
Academy 9	11	w/c 12/08/2024	NULL

only showing top 20 rows

```
df_summary.limit(20).show(truncate=False)
```

school	nc_year	weekcommencing	attendance_percentage
Academy 8	13	w/c 04/09/2023	99.4
Academy 7	13	w/c 28/02/2022	98.2
Academy 9	3	w/c 28/11/2022	93.1
Academy 8	11	w/c 21/03/2022	93.8
Academy 8	11	w/c 15/04/2024	92.9
Academy 7	9	w/c 21/08/2023	NULL
Academy 2	13	w/c 04/03/2024	95.9
Academy 3	8	w/c 16/10/2023	96.3
Academy 8	11	w/c 01/08/2022	NULL

Academy 4 9	w/c 15/05/2023 92.7	
Academy 7 7	w/c 18/11/2024 92.7	
Academy 5 13	w/c 20/02/2023 97.2	
Academy 5 13	w/c 11/03/2024 93.6	
Academy 4 13	w/c 15/11/2021 94.2	
Academy 9 9	w/c 12/09/2022 96.2	
Academy 5 10	w/c 25/03/2024 93.1	
Academy 9 11	w/c 28/08/2023 NULL	
Academy 9 11	w/c 25/03/2024 84.9	
Academy 9 11	w/c 25/07/2022 82.4	
Academy 9 11	w/c 12/08/2024 NULL	
+-----+	+-----+	+-----+

## 9. Null Values to be Investigated Further

As expected, a number of NULL values are present in the data. For the attention of the data analyst, I will quantify the number of NULL values in the attendance percentage column and provide this information in the summary statistics.

```
cols = df_summary.columns

# Build a filter condition: (col1 IS NULL) OR (col2 IS NULL) OR ...
null_condition = reduce(lambda acc, c: acc | F.col(c).isNull(), cols,
F.lit(False))

# Filter rows where any column is null
num_rows_with_null = df_summary.filter(null_condition).count()

print(f"Number of rows with at least one NULL value:
{num_rows_with_null}")

Number of rows with at least one NULL value: 2233
```

I will create a table to show the number of null values in the attendance columns. This will help to identify any patterns or trends in the missing data and inform data cleaning and imputation strategies.

```
show_df_missing_breakdown(df_summary)
```

DataFrame has 8697 rows and 4 columns.

Column		Null	EmptyStr	NA/NaNStr
NumericNaN	TotalMissing	%Missing		
school				
0	0	0.00%	0	0
nc_year				
			0	0

0	0	0.00%			
weekcommencing			0	0	0
0	0	0.00%			
attendance_percentage			2233	0	0
0	2233	25.68%			

```
# 1) From df_selected, group by the same columns used in df_summary
```

```
df_sums = (
    df_validdf_rows
        .groupBy("school", "nc_year", "weekcommencing")
        .agg(
            F.sum("is_attend").alias("sum_is_attend"),
            F.sum("is_possible").alias("sum_is_possible")
        )
)
```

```
# 2) Filter df_summary for rows where attendance_percentage is NULL
```

```
df_summary_nulls =  
df_summary.filter(F.col("attendance percentage").isNull())
```

```
# 3) Join df_summary_nulls with df_sums to see actual sums for those groups
```

```
df_null_sums = (
    df_summary_nulls.alias("summ")
    .join(
        df_sums.alias("sums"),
        on=["school", "nc_year", "weekcommencing"],
        how="left"
    )
    .select(
        "summ.school",
        "summ.nc_year",
        "summ.weekcommencing",
        "summ.attendance_percentage",    # should be NULL
        "sums.sum_is_attend",
        "sums.sum_is_possible"
    )
)
```

```
df.null_sums.show(truncate=False)
```

school	nc_year	weekcommencing	attendance_percentage	sum_is_attend	sum_is_possible
Academy 7	19	w/c 21/08/2023	NULL	0.0	

0.0				
Academy 8 11	w/c 01/08/2022 NULL		0.0	
0.0				
Academy 9 11	w/c 28/08/2023 NULL		0.0	
0.0				
Academy 9 11	w/c 12/08/2024 NULL		0.0	
0.0				
Academy 1 10	w/c 12/02/2024 NULL		0.0	
0.0				
Academy 6 11	w/c 07/08/2023 NULL		0.0	
0.0				
Academy 9 5	w/c 01/04/2024 NULL		0.0	
0.0				
Academy 6 10	w/c 27/12/2021 NULL		0.0	
0.0				
Academy 8 11	w/c 23/08/2021 NULL		0.0	
0.0				
Academy 9 N2	w/c 26/08/2024 NULL		0.0	
0.0				
Academy 2 12	w/c 04/04/2022 NULL		0.0	
0.0				
Academy 7 13	w/c 22/07/2024 NULL		0.0	
0.0				
Academy 9 10	w/c 25/12/2023 NULL		0.0	
0.0				
Academy 7 11	w/c 22/08/2022 NULL		0.0	
0.0				
Academy 4 11	w/c 23/10/2023 NULL		0.0	
0.0				
Academy 3 11	w/c 24/07/2023 NULL		0.0	
0.0				
Academy 9 13	w/c 29/07/2024 NULL		0.0	
0.0				
Academy 4 10	w/c 10/04/2023 NULL		0.0	
0.0				
Academy 9 2	w/c 25/12/2023 NULL		0.0	
0.0				
Academy 9 13	w/c 30/08/2021 NULL		0.0	
0.0				
+-----+-----+-----+-----+-----				
+-----+				
only showing top 20 rows				

To drill down further I can use the `df_validf_rows` dataframe to investigate the missing data further. This contains only the valid rows before any summary data frame, so I can see if there is a pattern in the missing data.

```
df_valid = df_rows.filter(
    (F.col("school") == "Academy 7") &
    (F.col("nc_year") == "9") &
    (F.col("weekcommencing") == "w/c 21/08/2023")
).show()
```

```
+-----+-----+-----+-----+
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+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
| UPN_AttendanceDate|gender|student_forename|student_surname|
pupil_premium|year_group|nc_year|school_type| school|
establishment_number|la_code|attendance_date|academic_year|
academic_week_number| term|weekcommencing|mark|session|
is_approved_educational_activity|is_attend|is_auth_abs|late|
late_unauthorised|missing|no_reason|is_possible|is_present|
is_unauth_abs|current_student|leaving_date| UPN|count|status|
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
|10011_2023-08-26_PM|FEMALE| John| Allen|
False| 9| 9| SECONDARY|Academy 7| XXXXXX|
XXXXXX| 2023-08-26| 2023/2024| 52|Term 6|w/c
21/08/2023| #| PM| 0.0| 0.0|
0.0| 0.0| 0.0| 0.0| 1.0| 0.0| 0.0|
0.0| 1| nan|10011| 1| valid|
| 7190_2023-08-23_PM| MALE| Sally| Stark|
True| 9| 9| SECONDARY|Academy 7| XXXXXX|
XXXXXX| 2023-08-23| 2023/2024| 52|Term 6|w/c
21/08/2023| #| PM| 0.0| 0.0|
0.0| 0.0| 0.0| 0.0| 1.0| 0.0| 0.0|
0.0| 1| nan| 7190| 1| valid|
|10185_2023-08-26_AM|FEMALE| Patrick| Cruz|
True| 9| 9| SECONDARY|Academy 7| XXXXXX|
XXXXXX| 2023-08-26| 2023/2024| 52|Term 6|w/c
21/08/2023| #| AM| 0.0| 0.0|
0.0| 0.0| 0.0| 0.0| 1.0| 0.0| 0.0|
0.0| 1| nan|10185| 1| valid|
| 4891_2023-08-25_PM| MALE| Holly| Lyons|
False| 9| 9| SECONDARY|Academy 7| XXXXXX|
XXXXXX| 2023-08-25| 2023/2024| 52|Term 6|w/c
21/08/2023| #| PM| 0.0| 0.0|
```

0.0	0.0		0.0	0.0	1.0		0.0	0.0
0.0		1	nan	4891	1	valid		
	4679_2023-08-27_AM	FEMALE		Robert			Fisher	
False		9	9	SECONDARY	Academy 7		XXXXXX	
XXXXXX	2023-08-27		2023/2024				52 Term 6 w/c	
21/08/2023	#	AM					0.0	0.0
0.0	0.0		0.0	0.0	1.0		0.0	0.0
0.0		1	nan	4679	1	valid		
	1095_2023-08-24_PM	FEMALE		Kelly			Carter	
True		9	9	SECONDARY	Academy 7		XXXXXX	
XXXXXX	2023-08-24		2023/2024				52 Term 6 w/c	
21/08/2023	#	PM					0.0	0.0
0.0	0.0		0.0	0.0	1.0		0.0	0.0
0.0		1	nan	1095	1	valid		
	6024_2023-08-21_AM	FEMALE		Christina			Gilbert	
False		9	9	SECONDARY	Academy 7		XXXXXX	
XXXXXX	2023-08-21		2023/2024				52 Term 6 w/c	
21/08/2023	#	AM					0.0	0.0
0.0	0.0		0.0	0.0	1.0		0.0	0.0
0.0		1	nan	6024	1	valid		
	7117_2023-08-24_AM	MALE		Gabriella			Sutton	
False		9	9	SECONDARY	Academy 7		XXXXXX	
XXXXXX	2023-08-24		2023/2024				52 Term 6 w/c	
21/08/2023	#	AM					0.0	0.0
0.0	0.0		0.0	0.0	1.0		0.0	0.0
0.0		1	nan	7117	1	valid		
	270_2023-08-24_AM	MALE		Michael			Richardson	
False		9	9	SECONDARY	Academy 7		XXXXXX	
XXXXXX	2023-08-24		2023/2024				52 Term 6 w/c	
21/08/2023	#	AM					0.0	0.0
0.0	0.0		0.0	0.0	1.0		0.0	0.0
0.0		1	nan	270	1	valid		
	4506_2023-08-25_AM	FEMALE		Danielle			Bowman	
True		9	9	SECONDARY	Academy 7		XXXXXX	
XXXXXX	2023-08-25		2023/2024				52 Term 6 w/c	
21/08/2023	#	AM					0.0	0.0
0.0	0.0		0.0	0.0	1.0		0.0	0.0
0.0		1	nan	4506	1	valid		
	4891_2023-08-26_AM	MALE		Holly			Lyons	
False		9	9	SECONDARY	Academy 7		XXXXXX	
XXXXXX	2023-08-26		2023/2024				52 Term 6 w/c	
21/08/2023	#	AM					0.0	0.0
0.0	0.0		0.0	0.0	1.0		0.0	0.0
0.0		1	nan	4891	1	valid		
	9558_2023-08-24_PM	MALE		Hector			White	
False		9	9	SECONDARY	Academy 7		XXXXXX	
XXXXXX	2023-08-24		2023/2024				52 Term 6 w/c	
21/08/2023	#	PM					0.0	0.0
0.0	0.0		0.0	0.0	1.0		0.0	0.0



0.0	1	nan	9558	1	valid				
	4789_2023-08-21_PM	FEMALE		Kristin		Cobb			
False	9	9	SECONDARY	Academy 7				XXXXXX	
XXXXXX	2023-08-21		2023/2024			52	Term 6	w/c	
21/08/2023	#	PM				0.0		0.0	
0.0	0.0		0.0	0.0	1.0	0.0		0.0	
0.0		1	nan	4789	1	valid			
	1100_2023-08-24_AM	MALE		Jean		Zimmerman			
False	9	9	SECONDARY	Academy 7				XXXXXX	
XXXXXX	2023-08-24		2023/2024			52	Term 6	w/c	
21/08/2023	#	AM				0.0		0.0	
0.0	0.0		0.0	0.0	1.0	0.0		0.0	
0.0		1	nan	1100	1	valid			
	8003_2023-08-26_AM	FEMALE		Johnny		King			
True	9	9	SECONDARY	Academy 7				XXXXXX	
XXXXXX	2023-08-26		2023/2024			52	Term 6	w/c	
21/08/2023	#	AM				0.0		0.0	
0.0	0.0		0.0	0.0	1.0	0.0		0.0	
0.0		1	nan	8003	1	valid			
	8533_2023-08-23_PM	MALE		Kevin		Torres			
True	9	9	SECONDARY	Academy 7				XXXXXX	
XXXXXX	2023-08-23		2023/2024			52	Term 6	w/c	
21/08/2023	#	PM				0.0		0.0	
0.0	0.0		0.0	0.0	1.0	0.0		0.0	
0.0		1	nan	8533	1	valid			
	1095_2023-08-25_AM	FEMALE		Kelly		Carter			
True	9	9	SECONDARY	Academy 7				XXXXXX	
XXXXXX	2023-08-25		2023/2024			52	Term 6	w/c	
21/08/2023	#	AM				0.0		0.0	
0.0	0.0		0.0	0.0	1.0	0.0		0.0	
0.0		1	nan	1095	1	valid			
	270_2023-08-23_PM	MALE		Michael		Richardson			
False	9	9	SECONDARY	Academy 7				XXXXXX	
XXXXXX	2023-08-23		2023/2024			52	Term 6	w/c	
21/08/2023	#	PM				0.0		0.0	
0.0	0.0		0.0	0.0	1.0	0.0		0.0	
0.0		1	nan	270	1	valid			
	9577_2023-08-22_AM	MALE		Anthony		Johnson			
False	9	9	SECONDARY	Academy 7				XXXXXX	
XXXXXX	2023-08-22		2023/2024			52	Term 6	w/c	
21/08/2023	#	AM				0.0		0.0	
0.0	0.0		0.0	0.0	1.0	0.0		0.0	
0.0		1	nan	9577	1	valid			
	4762_2023-08-24_PM	MALE		Joshua		Hill			
False	9	9	SECONDARY	Academy 7				XXXXXX	
XXXXXX	2023-08-24		2023/2024			52	Term 6	w/c	
21/08/2023	#	PM				0.0		0.0	
0.0	0.0		0.0	0.0	1.0	0.0		0.0	
0.0		1	nan	4762	1	valid			

```
+-----+-----+-----+-----+
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+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
only showing top 20 rows
```

The symbol for Mark = # often indicates school closure but will need further clarification. This results in the data fields `is_present` and `is_possible` both equating to zero, although the `no_reason` column is 1 while unauthorised absence is 0. To avoid introducing bias, I have retained these values in the dataset; the data analyst can investigate further and make adjustments as needed.

I will conclude by sorting the summary data frame by school name, week number, and year group, and writing the summary table to `fact_AttendanceSummary` in Parquet format for further analysis and reporting by the Data Analyst.

The summary table also includes the academic year, which, although not explicitly listed in the task brief, has been added to provide more context to the data.

## 10. Write the Summary Table to Parquet

```
df_summary = (
    df_invalid_rows

    # 1) Group and aggregate
    .groupBy("school", "nc_year", "weekcommencing")
    .agg(
        F.round(
            (F.sum("is_attend") / F.sum("is_possible") * 100), 1
        ).alias("attendance_percentage")
    )

)

df_summary.limit(20).show(truncate=False)
```

```
+-----+-----+-----+-----+
|school|nc_year|weekcommencing|attendance_percentage|
+-----+-----+-----+-----+
|Academy 9|11|w/c 25/03/2024|37.5|
|Academy 9|11|w/c 12/08/2024|NULL|
|Academy 8|11|w/c 21/03/2022|100.0|
|Academy 8|11|w/c 01/08/2022|NULL|
|Academy 3|12|w/c 26/02/2024|100.0|
```

Academy 9 12	w/c 23/05/2022 75.0	
Academy 4 9	w/c 15/05/2023 60.0	
Academy 3 13	w/c 17/01/2022 88.5	
Academy 9 8	w/c 22/05/2023 73.3	
Academy 9 12	w/c 12/09/2022 100.0	
Academy 4 11	w/c 19/09/2022 83.9	
Academy 2 12	w/c 20/02/2023 90.0	
Academy 6 10	w/c 24/04/2023 43.8	
Academy 5 13	w/c 11/03/2024 100.0	
Academy 9 9	w/c 12/09/2022 90.0	
Academy 3 8	w/c 16/10/2023 NULL	
Academy 9 N2	w/c 26/08/2024 NULL	
Academy 5 13	w/c 20/02/2023 92.6	
Academy 9 8	w/c 08/01/2024 100.0	
Academy 4 10	w/c 25/04/2022 60.0	

+-----+-----+-----+

*# Basic statistics for the attendance\_percentage column*

```
from pyspark.sql.functions import round
```

```
df_summary.describe("attendance_percentage") \
    .select("summary", round("attendance_percentage",
1)).alias("attendance_percentage")) \
    .show()
```

summary attendance_percentage
count 2677.0
mean 78.0
stddev 29.4
min 0.0
max 100.0

+-----+-----+-----+

*#conver df\_summary to pandas*

```
df_summary_pandas = df_summary.toPandas()
```

*#convert pandas to parquet*

```
df_summary_pandas.to_parquet('data/df_summary_pandas.parquet')
```

*#chceck if the file is created*

```
df_summary_loaded =
```

```
spark.read.parquet("data/df_summary_pandas.parquet")
```

```
df_summary_loaded.show(truncate=False)
```

school	nc_year	week	commencing	attendance_percentage
Academy 9	11	w/c	25/03/2024	37.5
Academy 9	11	w/c	12/08/2024	NULL
Academy 8	11	w/c	21/03/2022	100.0
Academy 8	11	w/c	01/08/2022	NULL
Academy 3	12	w/c	26/02/2024	100.0
Academy 9	12	w/c	23/05/2022	75.0
Academy 4	9	w/c	15/05/2023	60.0
Academy 3	13	w/c	17/01/2022	88.5
Academy 9	8	w/c	22/05/2023	73.3
Academy 9	12	w/c	12/09/2022	100.0
Academy 4	11	w/c	19/09/2022	83.9
Academy 2	12	w/c	20/02/2023	90.0
Academy 6	10	w/c	24/04/2023	43.8
Academy 5	13	w/c	11/03/2024	100.0
Academy 9	9	w/c	12/09/2022	90.0
Academy 3	8	w/c	16/10/2023	NULL
Academy 9	N2	w/c	26/08/2024	NULL
Academy 5	13	w/c	20/02/2023	92.6
Academy 9	8	w/c	08/01/2024	100.0
Academy 4	10	w/c	25/04/2022	60.0

only showing top 20 rows

## 11. Notes for the Data Analyst

The data has been cleaned and quality-assured to the best of my ability. I have identified and labelled duplicate records as invalid and provided a summary of the missing values in the attendance percentage column. For cases where attendance percentage = 0, the denominator and numerator for the field are often both 0; you may wish to examine how the attendance mark is recorded in the data, as this may provide further insights into the reasons for absence.

The summary table contains the attendance percentage for each school on a weekly basis by the year group of the student, sorted by school name, week number, and year group. I used the National Curriculum Year as the year group field for consistency and accuracy over school groupings. I also used the week commencing as the basis for the week number to ensure alignment with the academic year, thereby reducing the need for an additional field.

The summary statistics include the mean, median, minimum, maximum, and standard deviation of the attendance percentage across all schools, weeks, and year groups. The data has been written to fact\_AttendanceSummary in Parquet format for further analysis and reporting. Please let me know if you require any additional information or further analysis.