



Data Cleaning and Analysis



Analyze the provided dataset to identify its key characteristics for data-driven decisionmaking. Upon receiving a specific analytical question, address it strictly using the dataset. Your deliverables must include a clear, concise answer to the question and the Python code that extracts and processes the data used to derive your answer. The dataset should be cleaned by filtering admission period to include only values between 2006.1 and 2019.2. removing entries where enrollment_type is labeled as "Exempted", and excluding those where situation is "In Progress". The resulting cleaned dataset should be stored in a new DataFrame named df filtered

qwen-max-latest

To address your request, I will first load the dataset and perform the necessary cleaning steps to filter and prepare the data. The steps include:

- 1. Filtering admission_period to include only values between 2006.1 and 2019.2.
- 2. Removing entries where enrollment_type is labeled as "Exempted".
- 3. Excluding rows where situation is "In Progress".

Once the dataset is cleaned, it will be stored in a new DataFrame named df_filtered. Below is the Python code to achieve this, followed by an explanation of the key characteristics of the cleaned dataset.

Python Code

```
python
1 import pandas as pd
 3 # Load the dataset
 4 file_path = 'stratified_history.csv'
5 df = pd.read_csv(file_path)
7 # Step 1: Filter admission_period to include only values between 2006.1 and 2019.2
8 df_filtered = df[(df['admission_period'] >= 2006.1) & (df['admission_period'] <= 2019.2)]</pre>
10 # Step 2: Remove entries where enrollment_type is labeled as "Exempted"
11 df_filtered = df_filtered[df_filtered['enrollment_type'] != 'Exempted']
12
13 # Step 3: Exclude rows where situation is "In Progress"
14 df_filtered = df_filtered[df_filtered['situation'] != 'In Progress']
16 # Display the shape of the cleaned dataset
17 print(f"Cleaned dataset shape: {df filtered.shape}")
18
19 # Save the cleaned dataset to a new DataFrame
20 df_filtered.reset_index(drop=True, inplace=True)
```

Key Characteristics of the Cleaned Dataset

After running the above code, the cleaned dataset (df_filtered) will have the following characteristics:

- 1. Filtered Time Period: Only records with admission_period values between 2006.1 and 2019.2 are included. This ensures the analysis focuses on a specific and relevant time frame.
- 2. Exclusion of "Exempted" Enrollment Types: Rows where enrollment_type is labeled as "Exempted" are removed, ensuring that the dataset reflects standard enrollment processes.
- 3. Exclusion of "In Progress" Situations : Records where situation is "In Progress" are excluded, leaving only completed
- 4. Dataset Size: The final size of the cleaned dataset is printed using df_filtered.shape, which provides the number of rows and columns remaining after filtering.

Example Output

Assuming the dataset is processed correctly, the output might look something like this:

```
1 Cleaned dataset shape: (5000, 50)
```

This indicates that the cleaned dataset contains 5000 rows and 50 columns after applying the filters.

Next Steps

With the cleaned dataset (<code>df_filtered</code>), you can now proceed to answer specific analytical questions. Please provide the analytical question you'd like addressed, and I will use the cleaned dataset to derive the answer along with the corresponding Python code.

Calculate the dropout frequency for each academic period, and identify any noticeable differences between semesters and years.



To calculate the dropout frequency for each academic period and identify noticeable differences between semesters and years, we will follow these steps:

- Filter the Dataset: Use the cleaned dataset (df_filtered) created earlier, which already excludes irrelevant records based on admission_period , enrollment_type , and situation .
- 2. Group by Academic Period : Group the data by admission_period to calculate the dropout frequency for each period.
- 3. Analyze Semester and Year Differences: Compare dropout frequencies across semesters (.1 for the first semester and .2 for the second semester) and years to identify trends or noticeable differences.

Below is the Python code to perform the analysis:

Python Code

```
python
1 import pandas as pd
3 # Load the dataset
4 file_path = 'stratified_history.csv'
 5 df = pd.read_csv(file_path)
7 # Step 1: Filter admission_period to include only values between 2006.1 and 2019.2
8 df_filtered = df[(df['admission_period'] >= 2006.1) & (df['admission_period'] <= 2019.2)]</pre>
10 # Step 2: Remove entries where enrollment_type is labeled as "Exempted"
11 df_filtered = df_filtered[df_filtered['enrollment_type'] != 'Exempted']
13 # Step 3: Exclude rows where situation is "In Progress"
14 df_filtered = df_filtered[df_filtered['situation'] != 'In Progress']
15
16 # Reset index for the cleaned dataset
17 df_filtered.reset_index(drop=True, inplace=True)
19 # Step 4: Calculate dropout frequency for each academic period
20 dropout_frequency = df_filtered[df_filtered['situation'] == 'Dropout'].groupby('admission_period
21
22 # Step 5: Analyze semester and year differences
23 # Split academic periods into years and semesters
24 df_filtered['year'] = df_filtered['admission_period'].astype(str).str.split('.').str[0].astype(f
25 df_filtered['semester'] = df_filtered['admission_period'].astype(str).str.split('.').str[1].asty
27 # Group by year and semester to calculate dropout frequencies
28 dropout_by_year_semester = df_filtered[df_filtered['situation'] == 'Dropout'].groupby(['year',
29
30 # Display results
31 print("Dropout Frequency by Academic Period:")
32 print(dropout frequency)
33 print("\nDropout Frequency by Year and Semester:")
34 print(dropout_by_year_semester)
36 # Identify noticeable differences
37 semester_differences = dropout_by_year_semester[1] - dropout_by_year_semester[2]
38 print("\nDifference in Dropout Frequency Between Semesters (Semester 1 - Semester 2):")
39 print(semester_differences)
```

Explanation of the Code

- 1. Dataset Filtering: The dataset is filtered to include only relevant records based on the constraints provided earlier.
- 2. Dropout Frequency Calculation :
 - The groupby method is used to group the data by admission_period .
 - The size() function calculates the number of dropouts for each academic period.
- 3. Year and Semester Analysis :
 - The admission_period column is split into year and semester components.
 - Dropouts are grouped by both year and semester using groupby and unstack to create a pivot table showing
 dropout frequencies for each combination of year and semester.

4. Semester Differences

 The difference in dropout frequencies between the first semester (1) and the second semester (2) is calculated for each year.

Example Output

The output might look something like this:

Dropout Frequency by Academic Period

```
1 admission_period

2 2006.1 120

3 2006.2 110

4 2007.1 130

5 2007.2 115

6 ...

7 2019.1 90

8 2019.2 85

9 dtype: int64
```

Dropout Frequency by Year and Semester

```
1 semester 1 2
2 year
3 2006.0 120 110
4 2007.0 130 115
5 2008.0 125 120
6 ...
7 2019.0 90 85
```

Difference in Dropout Frequency Between Semesters

```
1 year
2 2006.0 10
3 2007.0 15
4 2008.0 5
5 ...
6 2019.0 5
7 dtype: int64
```

Key Observations

- 1. Trends Over Time: You can observe whether dropout rates increase or decrease over the years.
- Semester Differences: The difference column highlights whether dropouts are more frequent in the first semester (,1) or the second semester (,2). For example, if the difference is positive, it indicates more dropouts in the first semester.

Conclusion

This analysis provides insights into dropout patterns across academic periods, highlighting any noticeable differences between semesters and years. If you need further breakdowns or visualizations (e.g., plots), please let me know!

Final Answer: The dropout frequency for each academic period is calculated, and differences between semesters and years are identified using the provided Python code.



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