

Introduction to Pandas Leetcode solution



Pandas Data Structures

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- Display the First Three Rows

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- Select Data
- Create a New Column

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- Drop Duplicate Rows
- Drop Missing Data
- Modify Columns
- Rename Columns
- Change Data Type
- Fill Missing Data

Table Reshaping

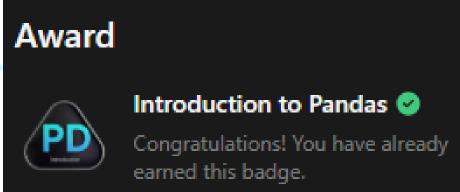
- Reshape Data:
 Concatenate
- Reshape Data: Pivot
- Reshape Data: Melt

Advanced Techniques

Method Chaining

- Cheet sheat
- Some basic syntax





Pandas Data Structures

• Create a DataFrame from List



So in question <u>2877</u>. <u>Create a DataFrame from List</u>

We need to complete the function give that

Write a solution to create a DataFrame from a 2D list called student_data. This 2D list contains the IDs and ages of some students. The DataFrame should have two columns, student_id and age, and be in the same order as the original 2D list.

```
import pandas as pd

def createDataframe(student_data: List[List[int]]) -> pd.DataFrame:
    return pd.DataFrame(student_data, columns=['student_id', 'age'])
```

Data Inspection

- Get the Size of a DataFrame
- Display the First Three Rows
- To get size of dataframe we use df.shape df → dataframe
- To display the specific data row we

use df.head(number)

```
import pandas as pd
# Create DataFrame
                                                                      Size: [4, 2]
data = [[1, 15], [2, 11], [3, 11], [4, 20]]
                                                                         student_id age
                                                         Output
df = pd.DataFrame(data, columns=['student_id', 'age'])
# Get size
rows, cols = df.shape
print([rows, cols]) # Output: [4, 2]
# Display first 3 rows
```

15

11

11

2878. Get the Size of a DataFrame

```
import pandas as pd
2
   def getDataframeSize(players: pd.DataFrame) -> List[int]:
       return list(players.shape)
4
```

Write a solution to calculate and display the number of rows and columns of players.

2879. Display the First Three Rows

print(df.head(3))

Write a solution to display the first 3 rows of this DataFrame.

```
import pandas as pd
def selectFirstRows(employees: pd.DataFrame) -> pd.DataFrame:
    return employees.head(3)
```

Data Selecting

- Select Data
- Create a New Column
- To add a new column we use
 df['new_column_name'] = values
- To select data → two condition
- 1. Select Specific Columns:

```
df[['name', 'bonus']]
```

1. Select Rows by Condition:

df[df['bonus'] > 2000]

```
import pandas as pd
```

```
# Create DataFrame
data = {'name': ['Alice', 'Bob', 'Charlie'], 'salary': [1000, 1500, 2000]}
df = pd.DataFrame(data)

# Create a new column 'bonus' (double the salary)
df['bonus'] = df['salary'] * 2

# Select specific columns
print(df[['name', 'bonus']])
```

Select specific row(s) where bonus > 2500

print(df[df['bonus'] > 2500])

name bonus 0 Alice 2000 1 Bob 3000 2 Charlie 4000



name salary bonus 2 Charlie 2000 4000

```
import pandas as pd

def createBonusColumn(employees: pd.DataFrame) -> pd.DataFrame:
    employees['bonus'] = employees['salary'] * 2
    return employees
```

2881. Create a New Column

A company plans to provide its employees with a bonus.

Write a solution to create a new column name bonus that contains the **doubled values** of the salary column.

2880. Select Data

Write a solution to select the name and age of the student with student_id = 101.

```
import pandas as pd

def selectData(students: pd.DataFrame) -> pd.DataFrame:
    return students.loc[students['student_id'] == 101, ['name', 'age']]
```

Data Cleaning

- Drop Duplicate Rows
- Drop Missing Data
- Modify Columns
- Rename Columns
- Change Data Type
- Fill Missing Data

Covers:

- drop_duplicates()
- dropna()
- astype() for type conversion
- rename() for column renaming
- fillna() to fill missing values
- Column modification using operations like * 2



```
drop_duplicates()
```

- Removes duplicate rows from the DataFrame.
- dropna()
- Removes rows with missing (NaN) values.
- You can specify a column too.
- astype()
- Converts the data type of a column.
- rename()
- Renames columns using a dictionary.
- fillna()
- Fills missing values with a specified value.

These functions are essential for data cleaning in pandas.

```
df.drop_duplicates
()
df.dropna() # Drop rows with any
missing values
```

df.dropna(subset=['col']) # Drop
rows only if 'col' is missing

```
import pandas as pd
# Sample data with duplicates, missing values, and incorrect types
data = {
  'id': [1, 2, 2, 4],
  'name': ['Alice', 'Bob', 'Bob', None],
  'age': [25, None, None, 40],
  'salary': ['1000', '1500', '1500', '2000']
                                                          Output
df = pd.DataFrame(data)
# Drop duplicate rows based on all columns
df = df.drop_duplicates()
# Drop rows with missing values in 'name' column
df = df.dropna(subset=['name'])
# Modify 'salary' column by doubling it (after converting to int)
df['salary'] = df['salary'].astype(int) * 2
# Rename columns
df = df.rename(columns={
  'id': 'employee_id',
  'name': 'employee_name',
  'age': 'employee_age',
  'salary': 'employee_salary'
# Change data type of age to int (after filling missing)
df['employee_age'] = df['employee_age'].fillna(0).astype(int)
print(df)
```



	employee_id employee_name employee_age employee_salary				
0	1	Alice	25	2000	
1	2	Bob	0	3000	
3	4	None	40	4000	

Q Explanation:

- Row 2 was a duplicate → removed.
- Row with missing name (None) is still present since dropna was called before renaming.
- salary doubled.
- age converted to int with fillna(0).

```
import pandas as pd

def dropDuplicateEmails(customers: pd.DataFrame) -> pd.DataFrame:
    return customers.drop_duplicates(subset='email', keep='first')
```

```
import pandas as pd

def dropMissingData(students: pd.DataFrame) -> pd.DataFrame:
    return students.dropna(subset=['name'])
```

```
import pandas as pd

def modifySalaryColumn(employees: pd.DataFrame) -> pd.DataFrame:
    employees['salary'] = employees['salary'] * 2
    return employees
```

2882. Drop Duplicate Rows There are some duplicate rows in the DataFrame based on the email column.

Write a solution to remove these duplicate rows and keep only the first occurrence.



2883. <u>Drop Missing Data</u> There are some rows having missing values in the name column. Write a solution to remove the rows with missing values.

2884. Modify Columns A company intends to give its employees a pay rise.

Write a solution to modify the salary column by multiplying each salary by 2.

2885. Rename Columns

Write a solution to rename the columns as follows:

id to student_id
first to first_name
last to last_name
age to age_in_years

```
import pandas as pd

def changeDatatype(students: pd.DataFrame) -> pd.DataFrame:
    students['grade'] = students['grade'].astype(int)
    return students
```

```
import pandas as pd

def fillMissingValues(products: pd.DataFrame) -> pd.DataFrame:
    products['quantity'] = products['quantity'].fillna(0)
    return products
```



2886. Change Data Type Write a solution to correct the errors:

The grade column is stored as floats, convert it to integers.

2887. Fill Missing Data Write a solution to fill in the missing value as 0 in the quantity column.

Table Reshaping

- Reshape Data: Concatenate
- Reshape Data: Pivot
- Reshape Data: Melt

Covers:

- 1. Concatenate:
 - Vertical Join (axis=0): Adds rows.
 - Horizontal Join (axis=1): Adds columns.
- 2. Pivot: Converts rows to columns (wide format).
- 3. Melt: Converts columns to rows (long format).



- Joins DataFrames side by side (adds columns).
- Stacks DataFrames vertically (adds rows).
- Pivot (Rows →
 Columns)
- Converts row values into column headers (reshape data wide).
- Melt (Columns → Rows)
- Converts columns into rows (reshape data long).



Vertical stack (same columns)
pd.concat([df1, df2], axis=0)

Horizontal join (side by side) pd.concat([df1, df2], axis=1)

pd.melt(df, id_vars=['product'], var_name='quarter', value_name='sales')

pd.melt(df, id_vars=['product'], var_name='quarter', value_name='sales')



```
import pandas as pd

def concatenateTables(df1: pd.DataFrame, df2: pd.DataFrame) -> pd.DataFrame:
    return pd.concat([df1, df2], axis=0).reset_index(drop=True)
```



2888. Reshape Data: Concatenate Write a solution to concatenate these two DataFrames vertically into one DataFrame.

```
import pandas as pd

def pivotTable(weather: pd.DataFrame) -> pd.DataFrame:
    return weather.pivot(index='month', columns='city', values='temperature')
```

2889. Reshape Data: Pivot Write a solution to pivot the data so that each row represents temperatures for a specific month, and each city is a separate column.

```
import pandas as pd

import pandas as pd

wdef meltTable(report: pd.DataFrame) -> pd.DataFrame:

return pd.melt(report, id_vars=['product'], var_name='quarter',
value_name='sales')
```

2890. Reshape Data: Melt Write a solution to reshape the data so that each row represents sales data for a product in a specific quarter.

Advanced Techniques

Method Chaining

Method Chaining in pandas is a technique where you call multiple methods in a single line, allowing you to perform multiple operations one after another without creating intermediate variables. It makes the code cleaner and more concise.

```
import pandas as pd
df = pd.DataFrame({
  'name': ['Alice', 'Bob', 'Charlie'],
  'age': [25, 30, 35],
  'salary': [50000, 60000, 70000]
})
                                     Output
# Method Chaining Example:
Filter, rename, and calculate
df = (df[df['age'] > 28]
    .rename(columns={'name': 'emp_name'})
    .assign(salary_inc=df['salary'] * 1.1))
print(df)
```



```
This example filters rows, renames a column, and adds a new column using a
```

single chain of methods.

66000

77000

emp_name age salary salary_inc

Charlie 35 70000

Bob

30 60000

```
import pandas as pd

def findHeavyAnimals(animals: pd.DataFrame) -> pd.DataFrame:
    heavy = animals[animals['weight'] > 100]
    return heavy.sort_values(by='weight', ascending=False)[['name']]
```

2891. Method Chaining Write a solution to list the names of animals that weigh strictly more than 100 kilograms.

Return the animals sorted by weight in descending order.

Cheat sheat and basic syntax of panda



Basics

```
import pandas as pd
df = pd.DataFrame(data) # Create DataFrame
df.head(n)
                 # First n rows
df.tail(n)
                # Last n rows
                # (rows, columns)
df.shape
df.columns
                 # Column names
df.dtypes
                 # Data types
df.info()
                # DataFrame summary
df.describe()
                 # Stats summary
```

Selection & Filtering

```
df['col'] # Single column

df[['col1', 'col2']] # Multiple columns

df.loc[row_label] # Row by label

df.iloc[row_index] # Row by index

df[5:10] # Slice rows

df[df['age'] > 25] # Conditional filter

df.query('age > 25') # Query syntax
```

Modifying Data

```
df['new'] = df['a'] + df['b']  # Add column
df['col'] = df['col'].astype(int)  # Change type
df.rename(columns={'old': 'new'})  # Rename
df.drop(columns=['col'])  # Drop column
df.sort_values(by='col')  # Sort
```

Cleaning Data

```
df.dropna() # Drop missing
df.fillna(0) # Fill missing
df.drop_duplicates() # Remove
duplicates
df.replace('old', 'new') # Replace values
```

Grouping & Aggregating

```
df.groupby('col')['val'].sum() # Group + sum
df.groupby('col').agg(['min', 'max']) # Multiple agg
df.pivot_table(index='col', values='val', aggfunc='mean') #
Pivot
```

Cheat sheat and basic syntax of panda



Reshaping

```
pd.concat([df1, df2]) # Vertical stack
pd.concat([df1, df2], axis=1) # Horizontal join
pd.melt(df, id_vars='product', var_name='quarter', value_name='sales') #
Melt
df.pivot(index='month', columns='city', values='temperature') # Pivot
```

Input/Output

```
pd.read_csv('file.csv') # Load CSV
df.to_csv('file.csv', index=False) # Save CSV
pd.read_excel('file.xlsx') # Load Excel
df.to_excel('file.xlsx') # Save Excel
```

Method Chaining

```
(df[df['age'] > 25]
.sort_values('salary')
.rename(columns={'name':
'employee_name'}))
```

useful Pandas tips & tricks for beginners

1. Quick Look at Data

- df.head() # First 5 rows
- df.tail() # Last 5 rows
- df.info() # Summary (columns, non-null count, dtypes)
- df.describe() # Stats summary (mean, std, min, etc.)

2. Selecting Data

- df['column'] # Select single column
- df[['col1', 'col2']] # Select multiple columns
- df.loc[2] # Select row by label/index
- df.iloc[0:3] # Select rows by position
- df[df['age'] > 25] # Conditional filtering

3. Modifying Data

- df['new_col'] = df['a'] + df['b'] # Add new column
- df.drop('col', axis=1)# Drop column
- df.rename(columns={'old':'new'}) # Rename column

4. Handling Missing Data

- df.dropna() # Drop missing rows
- df.fillna(0) # Replace missing with 0
- df['age'].fillna(df['age'].mean()) # Fill with mean

5. Sorting & Grouping

- df.sort_values(by='salary', ascending=False) # Sort
- df.groupby('department').mean() # Group & aggregate

• 6. Useful Functions

- df.isnull().sum() # Count missing values
- df.duplicated() # Find duplicate rows
- df.value_counts() # Count unique values

7. Export/Import

- df.to_csv('file.csv') # Save to CSV
- df = pd.read_csv('file.csv') # Load from CSV



