



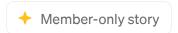




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35 Pandas Tricks to Save in Your List



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Pandas is one of the powerful dataset processing library not matter if you are a beginner in data science or an expert of python, you should use pandas to tidy, process and examine your dataset. It's a tool that's very versatile, but you can miss some of the more or **less-known tricks** that could save you time and make your code better.

In this article, I'll go over 35 Pandas tricks you should save in your list. This will make you a more efficient data scientist and these tips will make you more productive, optimize your workflow.

1. Quickly Load Data with read_csv

Loaded datasets into DataFrames using an easy to use method called read_csv() in Pandas. In addition you can pass more arguments such as usecols to read only certain columns, saving both time and memory.

```
import pandas as pd
df = pd.read_csv('data.csv', usecols=['Column1', 'Column2'])
```

2. View the First Few Rows with head()

The head() method quickly preview a dataset. They by default show you the first 5 rows.

```
df.head()
```

3. Check for Missing Values with isna()

isna() returns a DataFrame of booleans, indicating missing data before cleaning your data.

```
df.isna().sum()
```

4. Remove Duplicates with drop_duplicates()

drop_duplicates() removes duplicate rows from your dataset if your dataset has duplicate rows. This is especially helpful when you clean data to analyze.

```
df.drop_duplicates(inplace=True)
```

5. Rename Columns with rename()

Using the rename() method you can easily rename columns in case they have unclear names.

```
df.rename(columns={'old_name': 'new_name'}, inplace=True)
```

6. Convert Data Types with astype()

Your columns might sometimes be in the wrong data type. Use astype() to convert them.

```
df['Column1'] = df['Column1'].astype(float)
```

7. Handle Missing Data with fillna()

Rather than dropping rows with missing data, you can fill such rows with a specified value or method such as forward-fill or backward-fill.

```
df.fillna(method='ffill', inplace=True)
```

8. Set a Column as the Index

If you want your DataFrame to have an index, you can set a column to be index of your dataframe.

```
df.set_index('Column1', inplace=True)
```

9. Use groupby() for Aggregation

Aggregate group data based on a column you provide, allowing you to sum, mean, or count.

```
df.groupby('Category').agg({'Value': 'sum'})
```

10. Sort Values with sort_values()

To sort your DataFrame by a column <code>sort_values()</code>. You can order it ascending or descending.

```
df.sort_values(by='Column1', ascending=False)
```

11. Use apply() for Custom Functions

We use the apply() function for applying your custom function to DataFrame column or to rows.

```
df['new_column'] = df['Column1'].apply(lambda x: x * 2)
```

12. Filter Rows Based on Conditions

To filter rows of your DataFrame using conditional statements, use conditional statements.

```
df_filtered = df[df['Column1'] > 10]
```

13. Use query() for More Complex Filters

You can use query(), for better readable filtering.

```
df.query('Column1 > 10 and Column2 == "Yes"')
```

14. Concatenate DataFrames with concat()

So if you have multiple DataFrames, you can concatenate them on one axis using <code>concat()</code>.

```
df_combined = pd.concat([df1, df2], axis=0)
```

15. Merge DataFrames with merge()

Merging is like a SQL JOIN. When combining datasets based on common columns, it's incredibly useful.

```
df_merged = pd.merge(df1, df2, on='ID')
```

16. Reset Index with reset_index()

If you modify the index, or drop rows, use reset_index() to reset the index.

```
df.reset_index(drop=True, inplace=True)
```

17. Export Data to Excel with to_excel()

to_excel() is what you need to save your DataFrame to an Excel file.

```
df.to_excel('output.xlsx', index=False)
```

18. Convert to CSV with to_csv()

You can also export your DataFrame into a CSV file similarly.

```
df.to_csv('output.csv', index=False)
```

19. Quickly Get Descriptive Statistics with describe()

To quickly get an overview of the statistical summary of your DataFrame, use <code>describe()</code>.

```
df.describe()
```

20. Lambda Functions on Columns

apply() can be used, to apply a lambda function to multiple columns or rows.

```
df['new_column'] = df.apply(lambda row: row['Column1'] + row['Column2'], axis=1)
```

21. Drop Unwanted Columns with drop()

Remove unnecessary columns with <code>drop()</code>. You are able to drop one or more columns at once.

```
df.drop(columns=['Column1', 'Column2'], inplace=True)
```

22. Random sampling — use sample()

The sample() method in your DataFrame allows you to get a random sample from your dataframe.

```
df_sample = df.sample(n=5)
```

23. Extract Year, Month, or Day from DateTime Columns

If you are working with DateTime data you can extract the individual components such as year, month or day.

```
df['year'] = df['DateColumn'].dt.year
df['month'] = df['DateColumn'].dt.month
```

24. Convert Column to Categorical Type

If you don't need to do much, storing columns with few unique values can be saved using categorical type and could be fast as well.

```
df['Category'] = df['Category'].astype('category')
```

25. Find Duplicates Across Specific Contacts

duplicated() helps you check for duplicates in columns.

```
df[df.duplicated(subset=['Column1', 'Column2'])]
```

26. Use shift() for Lagging Data

On time series or sequential data shift() can create lagged columns.

```
df['lagged_column'] = df['Value'].shift(1)
```

27. Pivot Data with pivot_table()

With Create Pivot Tables, you can summarize data in a pretty similar way to what you'd do in a spreadsheet (Excel).

```
df_pivot = df.pivot_table(values='Value', index='Category', columns='Region', ag
```

28. Plot with Pandas' Built-In Plotting

You can also plot quick plots with pandas integrated with Matplotlib.

```
df['Value'].plot(kind='line')
```

29. Working with Strings on the Columns

String data can be used for applying string methods to columns directly.

```
df['name'] = df['name'].str.lower()
```

30. Make a New Column Using a Condition

With np.where() you can create a new column based on conditional logic.

```
import numpy as np
df['NewColumn'] = np.where(df['Value'] > 10, 'High', 'Low')
```

31. Using nunique() for Unique Values Count

Check how many unique values are in a column fast.

```
df['Category'].nunique()
```

32. Get Column Data Types with dtypes

Use dtypes attribute to quickly check what you have in your columns.

```
df.dtypes
```

33. Find Correlations with corr()

To correlate numerical columns you use corr().

```
df.corr()
```

34. Change Display Options

Pandas allows us to customize display of DataFrames, such as changing the max number of rows and columns shown.

```
pd.set_option('display.max_rows', 100)
```

35. Efficient Row-wise Operations

When you need to do something on a row by row basis then use <code>iterrows()</code>, otherwise go for vectorized solutions for better performance.

```
for idx, row in df.iterrows():
    print(row['Column1'])
```

Conclusion

Pandas is just plain powerful and these 35 tricks will help you doing things faster, can help make your code more readable and increase your productivity. Remember these tips and when you sit down to work with data again, you'll be more prepared to get through it more quickly.

Which of these tricks do you think is the most useful? I'd love to hear from you! So, in the comment section below, drop in any other tricks, or Pandas tips that you love and want to see in this list.

Happy coding!

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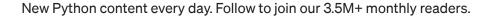
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```
4 class B(A):
       def show(self):
           print("B")
           super().show()
8 class C(A):
       def show(self):
10
           print("C")
11
           super().show()
12 class D(B, C): # Multiple inheritance
       def show(self):
```



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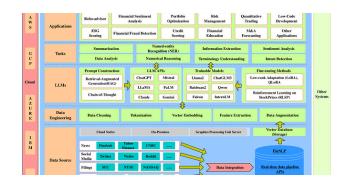


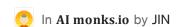
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