700+ Free Datasets in Python

that you can use today...







🤴 700+ Free Datasets in Python

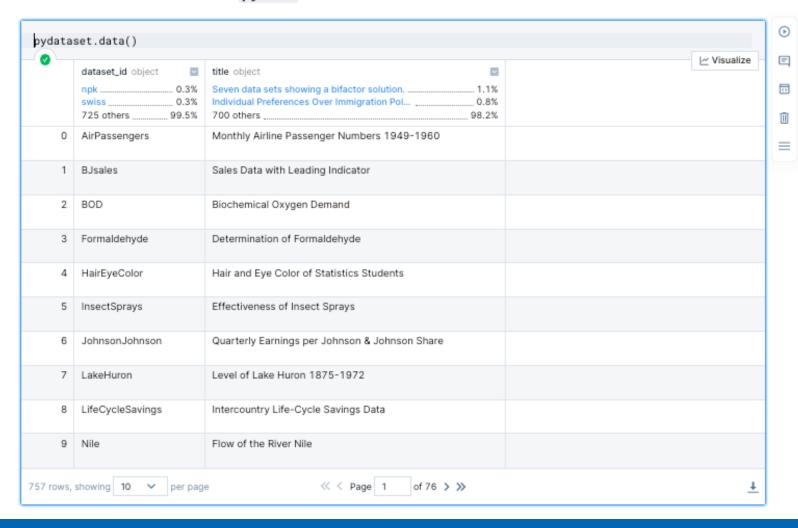
Use the package pydataset to get instant access!

import sys
!{sys.executable} --m pip install pydataset

A quick analyses to show you the power of pydataset:

import pydataset import pandas as pd

All the datasets available via pydata





All the datasets with Month or Quarter in Title

	dataset_id object	title object	
	AirPassengers 10% JohnsonJohnson 10% 8 others 80%	Monthly Airline Passenger Numbers 1949-1960 10 Quarterly Earnings per Johnson & Johnson Share 10 8 others 8	0%
0	AirPassengers	Monthly Airline Passenger Numbers 1949-1960	
6	JohnsonJohnson	Quarterly Earnings per Johnson & Johnson Share	
17	UKgas	UK Quarterly Gas Consumption	
30	austres	Quarterly Time Series of the Number of Australian Residents	
49	nottem	Average Monthly Temperatures at Nottingham, 1920-1939	
53	presidents	Quarterly Approval Ratings of US Presidents	
61	sunspot.month	Monthly Sunspot Data, from 1749 to "Present"	
63	sunspots	Monthly Sunspot Numbers, 1749-1983	
71	acme	Monthly Excess Returns	
468	deaths	Monthly Deaths from Lung Diseases in the UK	



All the datasets with Month or Quarter in Title

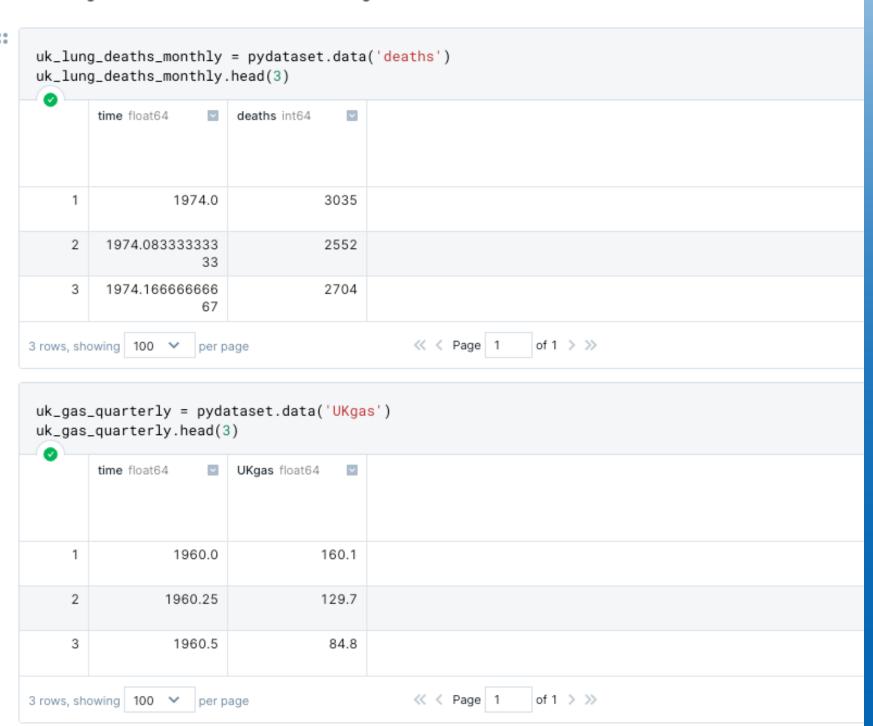
possible_data = pydataset.data() possible_data[possible_data['title'].str.contains('Quarter|Month')] dataset_id object title object ¥ AirPassengers 10% Monthly Airline Passenger Numbers 1949-1960 10% JohnsonJohnson 10% Quarterly Earnings per Johnson & Johnson Share 10% 8 others 80% 8 others 80% Monthly Airline Passenger Numbers 1949-1960 AirPassengers JohnsonJohnson Quarterly Earnings per Johnson & Johnson Share UK Quarterly Gas Consumption UKgas Quarterly Time Series of the Number of Australian Residents 30 austres Average Monthly Temperatures at Nottingham, 1920-1939 49 nottem Quarterly Approval Ratings of US Presidents 53 presidents sunspot.month Monthly Sunspot Data, from 1749 to "Present" Monthly Sunspot Numbers, 1749-1983 63 sunspots Monthly Excess Returns acme Monthly Deaths from Lung Diseases in the UK 468 deaths 10 rows, showing 100 V << < Page 1</p> of 1 > >> per page



We found our time series data!

Question:

Are lung deaths in the UK correlated with UK gas utilization?





Prepare the data...

```
# turn month to quarters
 uk_lung_deaths_monthly['is_quarter'] = uk_lung_deaths_monthly['time'] % 0.25
 uk_lung_deaths_quarterly = uk_lung_deaths_monthly[
     uk_lung_deaths_monthly['is_quarter']==0
 uk_lung_deaths_quarterly = uk_lung_deaths_quarterly[['time', 'deaths']]
 uk_lung_deaths_quarterly.head()
         time float64
                          deaths int64
     1
                  1974.0
                                      3035
     4
                 1974.25
                                      2554
     7
                  1974.5
                                      1721
    10
                 1974.75
                                      2074
    13
                  1975.0
                                      2933
5 rows, showing 10
                                                 << < Page 1</p>
                                                                 of 1 > >>
                      per page
```



```
# get unique times
uk_lung_deaths_time = uk_lung_deaths_quarterly['time'].tolist()
uk_gas_time = uk_gas_quarterly['time'].tolist()
combined_time = uk_lung_deaths_time + uk_gas_time
unique_time = set(combined_time)
unique_time_df = pd.DataFrame(unique_time, columns=['time'])
# get max and min times
min_time = max(
    min(uk_lung_deaths_time),
    min(uk_gas_time)
max_time = min(
    max(uk_lung_deaths_time),
    max(uk_gas_time)
# left join data
join_df = unique_time_df \
    .merge(uk_lung_deaths_quarterly, on='time', how='left') \
    .merge(uk_gas_quarterly, on='time', how='left')
# filter and organize data
join_filter_df = join_df[
    (join_df['time'] >= min_time) &
    (join_df['time'] <= max_time)
final_df = join_filter_df \
    .sort_values('time', ascending=True) \
    .set_index('time')
```



Plot the data...

```
final_df.plot.line(subplots=True)
array([<AxesSubplot:xlabel='time'>, <AxesSubplot:xlabel='time'>],
      dtype=object)
 3000
 2500
                                                                 deaths
 2000
 1500
  800
              UKgas
  600
  400
  200
        1974
                  1975
                             1976
                                       1977
                                                  1978
                                                            1979
                                                                       1980
                                      time
```

Calculate the correlation...

```
final_df['deaths'].corr(final_df['UKgas'])

0.6094685074564089
```





Final Thoughts:

Though this is a very simple analysis, it shows you how easy test data can become available to you via pydataset.

How can you use this in your data workflow?

- Mock data for unit tests
- Simple data to test out unfamiliar functions
 - Quick data for demos

Link to the Deepnote project with code in the comments!