



Challenge task: Find other data from the UCI Machine Learning Repository. Using the previous code for reference, go explore!

```
In [21]: import requests
import zipfile
import io
import pandas as pd

# --- Step 1: Download and Extract the New Dataset ---
# URL for the Bank Marketing dataset zip file
f_zip = 'https://archive.ics.uci.edu/ml/machine-learning-databases/00222/bank-ad

print("Downloading new dataset (Bank Marketing)...")
# Send a GET request to the URL
r = requests.get(f_zip, stream=True)

print("Extracting files...")
# Create a ZipFile object from the in-memory content
bank_zip = zipfile.ZipFile(io.BytesIO(r.content))

# Extract all contents
bank_zip.extractall()
print(f"Files extracted: {bank_zip.namelist()}")
```

```
# --- Step 2: Load the CSV File into DataFrame 'df1' ---  
# The main data file in this archive is 'bank-additional/bank-additional-full.cs  
# It's a CSV file, so we use pd.read_csv()  
csv_file_path = 'bank-additional/bank-additional-full.csv'  
  
print(f"\nLoading '{csv_file_path}' into DataFrame 'df1'...")  
# Load the CSV file, noting that its separator is a semicolon ';'   
df1 = pd.read_csv(csv_file_path, sep=';')  
print("DataFrame 'df1' created successfully.")  
  
# --- Final Result ---  
# Display the head and info for the new DataFrame 'df1'  
print("\n✅ Process Complete. Here is your new DataFrame:")  
print("\n--- DataFrame df1 Head ---")  
print(df1.head())  
  
print("\n--- DataFrame df1 Info ---")  
df1.info()
```

Downloading new dataset (Bank Marketing)...

Extracting files...

Files extracted: ['bank-additional/', 'bank-additional/.DS_Store', '__MACOSX/', '__MACOSX/bank-additional/', '__MACOSX/bank-additional/.DS_Store', 'bank-additional/.Rhistory', 'bank-additional/bank-additional-full.csv', 'bank-additional/bank-additional-names.txt', 'bank-additional/bank-additional.csv', '__MACOSX/.bank-additional']

Loading 'bank-additional/bank-additional-full.csv' into DataFrame 'df1'...
DataFrame 'df1' created successfully.

✅ Process Complete. Here is your new DataFrame:

--- DataFrame df1 Head ---

	age	job	marital	education	default	housing	loan	contact \
0	56	housemaid	married	basic.4y	no	no	no	telephone
1	57	services	married	high.school	unknown	no	no	telephone
2	37	services	married	high.school	no	yes	no	telephone
3	40	admin.	married	basic.6y	no	no	no	telephone
4	56	services	married	high.school	no	no	yes	telephone

	month	day_of_week	...	campaign	pdays	previous	poutcome	emp.var.rate
0	may	mon	...	1	999	0	nonexistent	1.1
1	may	mon	...	1	999	0	nonexistent	1.1
2	may	mon	...	1	999	0	nonexistent	1.1
3	may	mon	...	1	999	0	nonexistent	1.1
4	may	mon	...	1	999	0	nonexistent	1.1

	cons.price.idx	cons.conf.idx	euribor3m	nr.employed	y
0	93.994	-36.4	4.857	5191.0	no
1	93.994	-36.4	4.857	5191.0	no
2	93.994	-36.4	4.857	5191.0	no
3	93.994	-36.4	4.857	5191.0	no
4	93.994	-36.4	4.857	5191.0	no

[5 rows x 21 columns]

--- DataFrame df1 Info ---

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 41188 entries, 0 to 41187

Data columns (total 21 columns):

#	Column	Non-Null Count	Dtype
0	age	41188 non-null	int64
1	job	41188 non-null	object
2	marital	41188 non-null	object
3	education	41188 non-null	object
4	default	41188 non-null	object
5	housing	41188 non-null	object
6	loan	41188 non-null	object
7	contact	41188 non-null	object
8	month	41188 non-null	object
9	day_of_week	41188 non-null	object
10	duration	41188 non-null	int64
11	campaign	41188 non-null	int64
12	pdays	41188 non-null	int64
13	previous	41188 non-null	int64
14	poutcome	41188 non-null	object
15	emp.var.rate	41188 non-null	float64

```

16  cons.price.idx  41188 non-null  float64
17  cons.conf.idx  41188 non-null  float64
18  euribor3m      41188 non-null  float64
19  nr.employed    41188 non-null  float64
20  y              41188 non-null  object
dtypes: float64(5), int64(5), object(11)
memory usage: 6.6+ MB

```

Congratulations!

You have completed this lab, and you can now end the lab by following the lab guide instructions.

In [24]: `df1.shape`

Out[24]: (41188, 21)

In [25]: `df.info()`

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 310 entries, 0 to 309
Data columns (total 7 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   pelvic_incidence                      310 non-null    float64
1   pelvic_tilt                           310 non-null    float64
2   lumbar_lordosis_angle                 310 non-null    float64
3   sacral_slope                         310 non-null    float64
4   pelvic_radius                        310 non-null    float64
5   degree_spondylolisthesis             310 non-null    float64
6   class                                310 non-null    int64
dtypes: float64(6), int64(1)
memory usage: 17.1 KB

```

In [26]: `df.describe()`

Out[26]:

	pelvic_incidence	pelvic_tilt	lumbar_lordosis_angle	sacral_slope	pelvic_radius	degree_s
count	310.000000	310.000000	310.000000	310.000000	310.000000	
mean	60.496653	17.542822	51.930930	42.953831	117.920655	
std	17.236520	10.008330	18.554064	13.423102	13.317377	
min	26.147921	-6.554948	14.000000	13.366931	70.082575	
25%	46.430294	10.667069	37.000000	33.347122	110.709196	
50%	58.691038	16.357689	49.562398	42.404912	118.268178	
75%	72.877696	22.120395	63.000000	52.695888	125.467674	
max	129.834041	49.431864	125.742385	121.429566	163.071041	

In [29]: `df1.columns`

```
Out[29]: Index(['age', 'job', 'marital', 'education', 'default', 'housing', 'loan',
               'contact', 'month', 'day_of_week', 'duration', 'campaign', 'pdays',
               'previous', 'poutcome', 'emp.var.rate', 'cons.price.idx',
               'cons.conf.idx', 'euribor3m', 'nr.employed', 'y'],
              dtype='object')
```

```
In [30]: df1.dtypes
```

```
Out[30]: age                int64
job                object
marital            object
education          object
default            object
housing            object
loan               object
contact            object
month              object
day_of_week        object
duration           int64
campaign           int64
pdays             int64
previous           int64
poutcome           object
emp.var.rate       float64
cons.price.idx     float64
cons.conf.idx      float64
euribor3m          float64
nr.employed        float64
y                  object
dtype: object
```

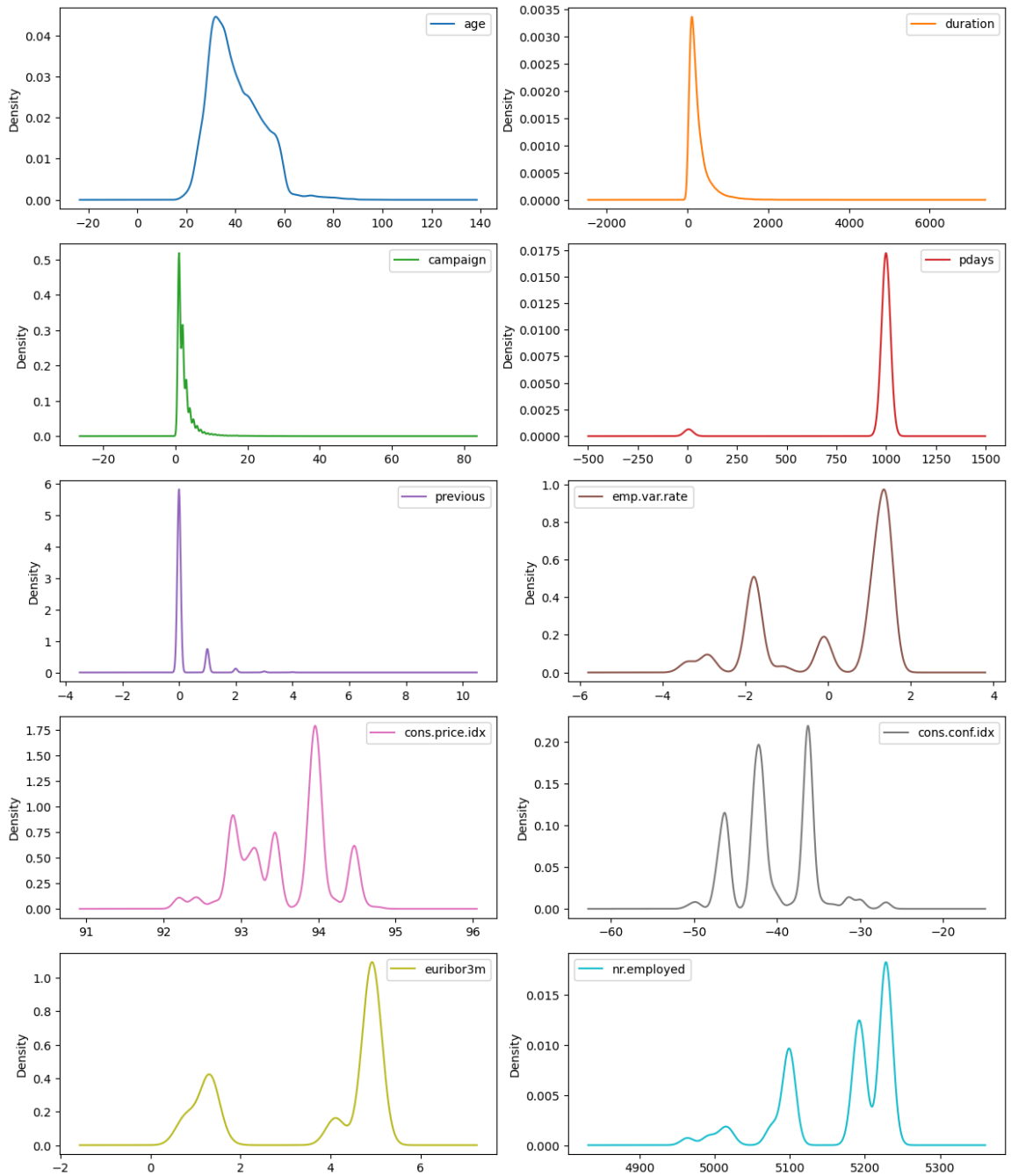
```
In [33]: import matplotlib.pyplot as plt

# Overwrite df1 with a version containing only its numeric columns
df1 = df1.select_dtypes(include='number')

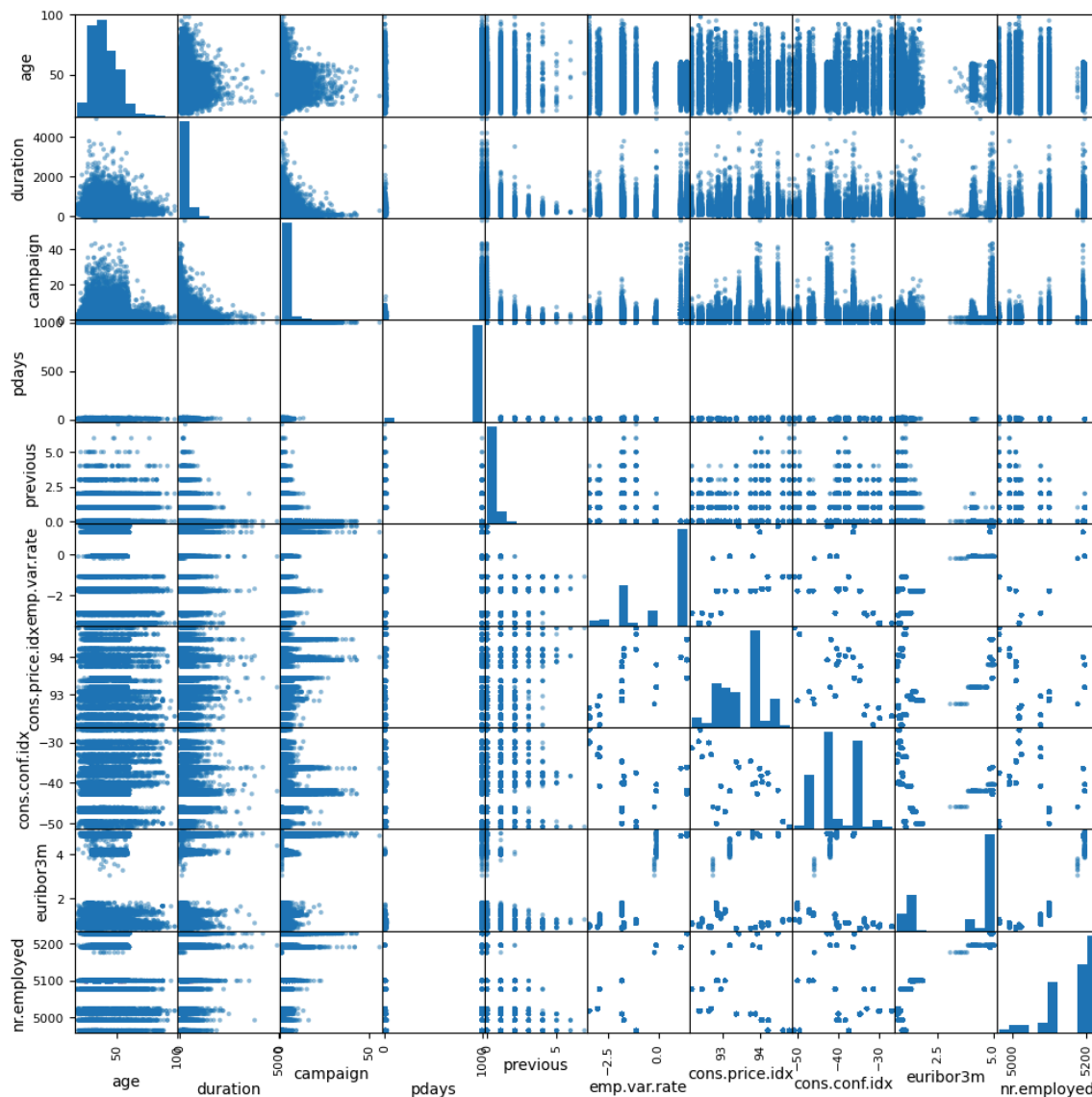
# Plot the new df1, adjusting layout to fit all columns (5 rows, 2 columns)
df1.plot(kind='density', subplots=True, layout=(5, 2), figsize=(12, 14), sharex=

# Adjust layout to prevent labels from overlapping
plt.tight_layout()

# Show the plot
plt.show()
```



```
In [36]: pd.plotting.scatter_matrix(df1,figsize=(12,12))
plt.show()
```



```
In [38]: import seaborn as sns
import matplotlib.pyplot as plt

# Ensure df1 contains only numeric data (from the previous step)
df1 = df1.select_dtypes(include='number')

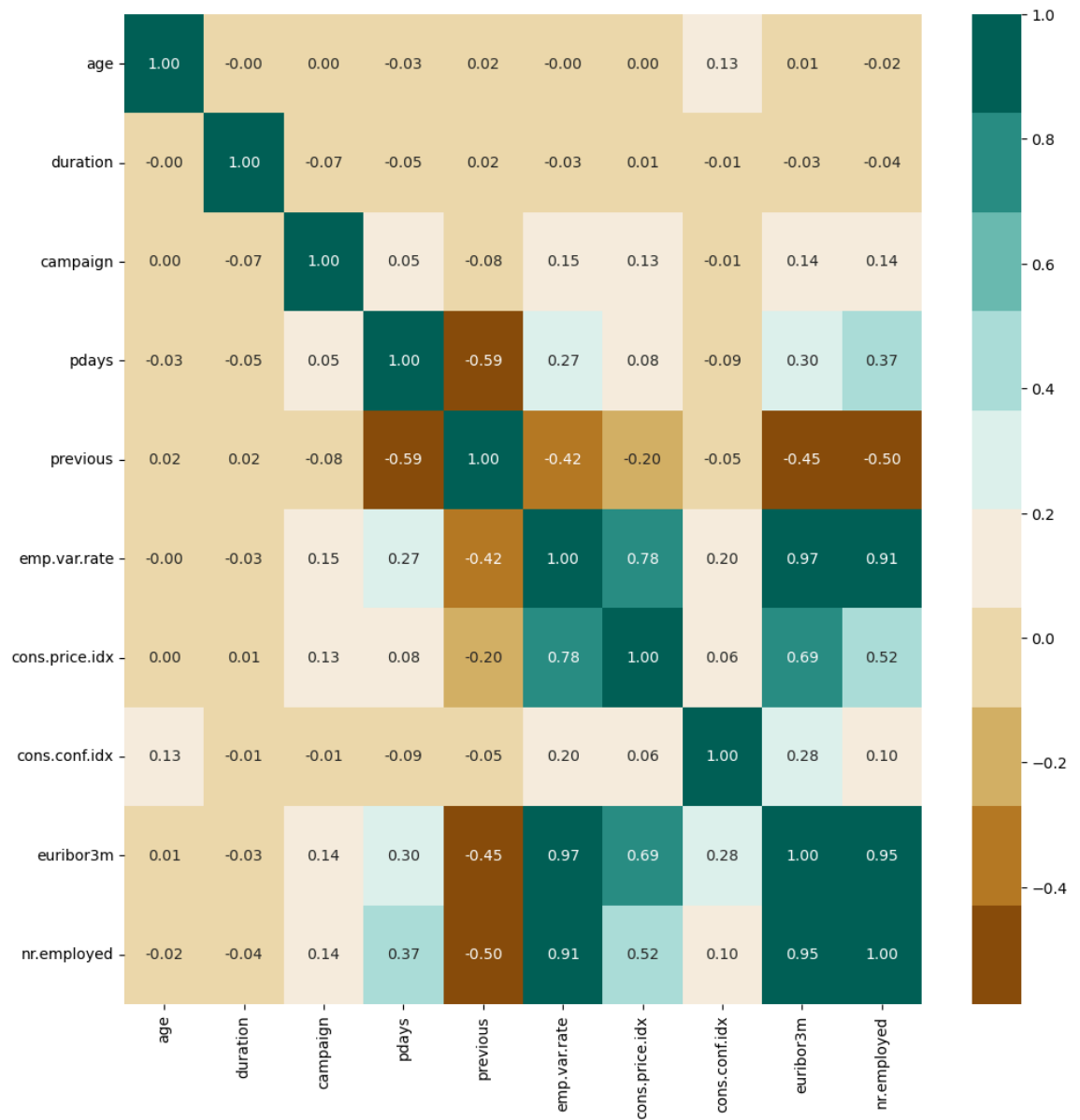
# 1. Calculate the correlation matrix for df1
corr_matrix = df1.corr()

# 2. Set up the plot
fig, ax = plt.subplots(figsize=(12, 12)) # Increased size for better readability

# 3. Generate the colormap
colormap = sns.color_palette("BrBG", 10)

# 4. Generate the Heatmap
# - Pass the calculated corr_matrix
# - Use the specified colormap
# - Enable annotations (annot=True) and format them to two decimal places (fmt=".2f")
sns.heatmap(corr_matrix, cmap=colormap, annot=True, fmt=".2f")

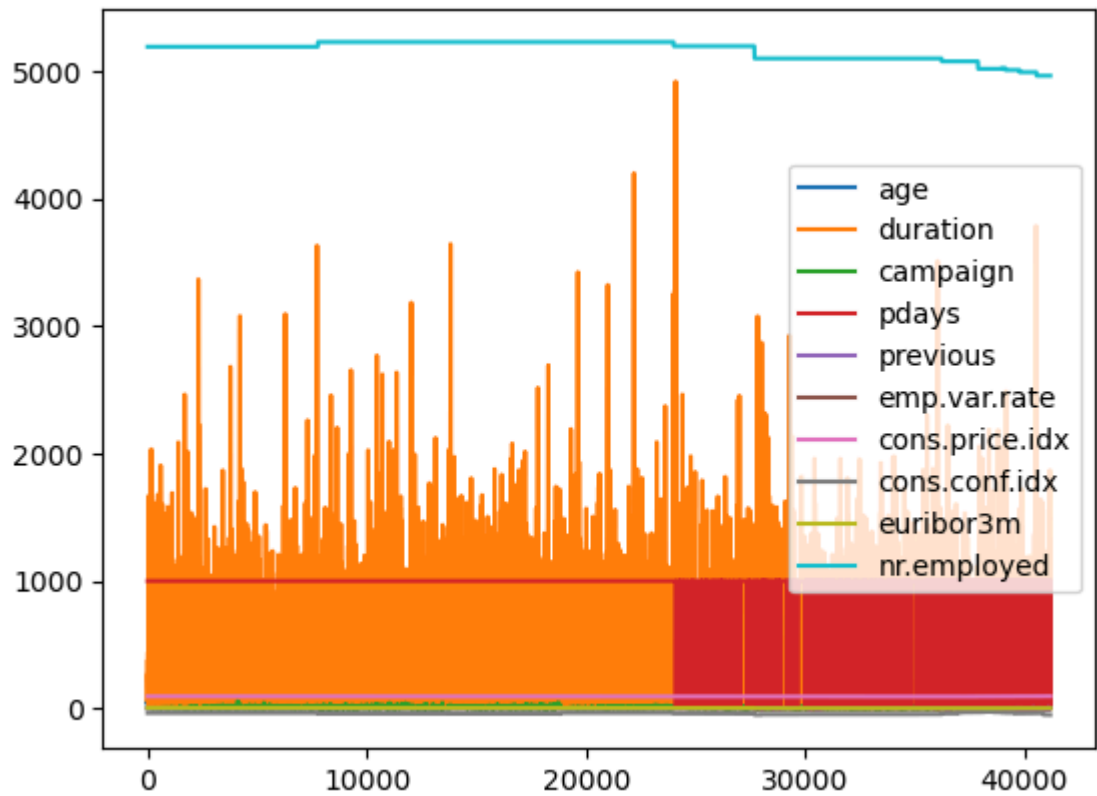
# 5. Display the plot
plt.show()
```



In []:

```
In [22]: import matplotlib.pyplot as plt
%matplotlib inline
df1.plot()
```

Out[22]: <Axes: >



In []: