

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
df=pd.read_csv("train.csv")
```

```
df
```

	ID	age	job	marital	education	default	balance
housing \							
0	26110	56	admin.	married	unknown	no	1933
no							
1	40576	31	unknown	married	secondary	no	3
no							
2	15320	27	services	married	secondary	no	891
yes							
3	43962	57	management	divorced	tertiary	no	3287
no							
4	29842	31	technician	married	secondary	no	119
yes							
...	...	...	...	...	...	...	...
...							
31642	36483	29	management	single	tertiary	no	0
yes							
31643	40178	53	management	divorced	tertiary	no	380
no							
31644	19710	32	management	single	tertiary	no	312
no							
31645	38556	57	technician	married	secondary	no	225
yes							
31646	14156	55	management	divorced	secondary	no	204
yes							

	loan	contact	day	month	duration	campaign	pdays	previous
\								
0	no	telephone	19	nov	44	2	-1	0
1	no	cellular	20	jul	91	2	-1	0
2	no	cellular	18	jul	240	1	-1	0
3	no	cellular	22	jun	867	1	84	3
4	no	cellular	4	feb	380	1	-1	0
...	...	...	...	...	...	...	...	...
31642	no	cellular	12	may	116	2	-1	0

31643	yes	cellular	5	jun	438	2	-1	0
31644	no	cellular	7	aug	37	3	-1	0
31645	no	telephone	15	may	22	7	337	12
31646	no	cellular	11	jul	1973	2	-1	0

	poutcome	subscribed
0	unknown	no
1	unknown	no
2	unknown	no
3	success	yes
4	unknown	no
...	...	...
31642	unknown	no
31643	unknown	yes
31644	unknown	no
31645	failure	no
31646	unknown	yes

[31647 rows x 18 columns]

df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 31647 entries, 0 to 31646
Data columns (total 18 columns):
#   Column          Non-Null Count  Dtype
---  -
0   ID               31647 non-null  int64
1   age              31647 non-null  int64
2   job              31647 non-null  object
3   marital          31647 non-null  object
4   education        31647 non-null  object
5   default          31647 non-null  object
6   balance          31647 non-null  int64
7   housing          31647 non-null  object
8   loan             31647 non-null  object
9   contact          31647 non-null  object
10  day              31647 non-null  int64
11  month            31647 non-null  object
12  duration         31647 non-null  int64
13  campaign         31647 non-null  int64
14  pdays            31647 non-null  int64
15  previous         31647 non-null  int64
16  poutcome         31647 non-null  object
17  subscribed       31647 non-null  object
```

```
dtypes: int64(8), object(10)
memory usage: 4.3+ MB
```

```
df.head()
```

	ID	age	job	marital	education	default	balance
housing	loan	\					
0	26110	56	admin.	married	unknown	no	1933
no	no						
1	40576	31	unknown	married	secondary	no	3
no	no						
2	15320	27	services	married	secondary	no	891
yes	no						
3	43962	57	management	divorced	tertiary	no	3287
no	no						
4	29842	31	technician	married	secondary	no	119
yes	no						

	contact	day	month	duration	campaign	pdays	previous	poutcome
\								
0	telephone	19	nov	44	2	-1	0	unknown
1	cellular	20	jul	91	2	-1	0	unknown
2	cellular	18	jul	240	1	-1	0	unknown
3	cellular	22	jun	867	1	84	3	success
4	cellular	4	feb	380	1	-1	0	unknown

	subscribed
0	no
1	no
2	no
3	yes
4	no

```
df.columns
```

```
Index(['ID', 'age', 'job', 'marital', 'education', 'default',
      'balance',
      'housing', 'loan', 'contact', 'day', 'month', 'duration',
      'campaign',
      'pdays', 'previous', 'poutcome', 'subscribed'],
      dtype='object')
```

```
df.drop_duplicates()
```

	ID	age	job	marital	education	default	balance
housing	\						

0 no	26110	56	admin.	married	unknown	no	1933
1 no	40576	31	unknown	married	secondary	no	3
2 yes	15320	27	services	married	secondary	no	891
3 no	43962	57	management	divorced	tertiary	no	3287
4 yes	29842	31	technician	married	secondary	no	119
...	...	...	...	...	...	...	...
...							
31642 yes	36483	29	management	single	tertiary	no	0
31643 no	40178	53	management	divorced	tertiary	no	380
31644 no	19710	32	management	single	tertiary	no	312
31645 yes	38556	57	technician	married	secondary	no	225
31646 yes	14156	55	management	divorced	secondary	no	204

\	loan	contact	day	month	duration	campaign	pdays	previous
0	no	telephone	19	nov	44	2	-1	0
1	no	cellular	20	jul	91	2	-1	0
2	no	cellular	18	jul	240	1	-1	0
3	no	cellular	22	jun	867	1	84	3
4	no	cellular	4	feb	380	1	-1	0
...	...	...	...	...	...	...	...	...
31642	no	cellular	12	may	116	2	-1	0
31643	yes	cellular	5	jun	438	2	-1	0
31644	no	cellular	7	aug	37	3	-1	0
31645	no	telephone	15	may	22	7	337	12
31646	no	cellular	11	jul	1973	2	-1	0

poutcome subscribed

```

0      unknown      no
1      unknown      no
2      unknown      no
3      success      yes
4      unknown      no
...      ...      ...
31642  unknown      no
31643  unknown      yes
31644  unknown      no
31645  failure      no
31646  unknown      yes

```

[31647 rows x 18 columns]

```
df.describe()
```

	ID	age	balance	day
duration \				
count	31647.000000	31647.000000	31647.000000	31647.000000
mean	22563.972162	40.957247	1363.890258	15.835466
std	258.113534	10.625134	3028.304293	8.337097
min	0.000000	18.000000	-8019.000000	1.000000
25%	11218.000000	33.000000	73.000000	8.000000
50%	22519.000000	39.000000	450.000000	16.000000
75%	33879.500000	48.000000	1431.000000	21.000000
max	45211.000000	95.000000	102127.000000	31.000000

	campaign	pdays	previous
count	31647.000000	31647.000000	31647.000000
mean	2.765697	39.576042	0.574272
std	3.113830	99.317592	2.422529
min	1.000000	-1.000000	0.000000
25%	1.000000	-1.000000	0.000000
50%	2.000000	-1.000000	0.000000
75%	3.000000	-1.000000	0.000000
max	63.000000	871.000000	275.000000

```
df['subscribed'].unique()
```

```
array(['no', 'yes'], dtype=object)
```

```
df['balance'].unique()
```

```
array([1933,    3, 891, ..., 2787, 8741, 2968])
df['campaign'].unique()
array([ 2,  1,  3,  4,  7,  5, 33, 12,  8,  9,  6, 24, 17, 11, 20, 25,
        19,
        29, 21, 10, 27, 38, 16, 18, 14, 30, 13, 15, 63, 23, 31, 43, 35,
        22,
        34, 28, 26, 41, 37, 50, 55, 32, 44, 36, 39])
```

```
df['day'].unique()
array([19, 20, 18, 22,  4,  2,  3,  8, 15,  5, 28,  6, 14,  7, 24, 13,
        9,
        11, 21, 12, 30, 27, 17, 16, 25, 10,  1, 29, 26, 31, 23])
```

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

```
31647
```

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

```
6326
```

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

```
45
```

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

```
31
```

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

```
1454
```

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

```
509
```

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

```
2
```

```
df
```

	ID	age	job	marital	education	default	balance
housing \							
0	26110	56	admin.	married	unknown	no	1933
no							
1	40576	31	unknown	married	secondary	no	3

no							
2	15320	27	services	married	secondary	no	891
yes							
3	43962	57	management	divorced	tertiary	no	3287
no							
4	29842	31	technician	married	secondary	no	119
yes							
...	...	...	...	...	...	...	...
...							
31642	36483	29	management	single	tertiary	no	0
yes							
31643	40178	53	management	divorced	tertiary	no	380
no							
31644	19710	32	management	single	tertiary	no	312
no							
31645	38556	57	technician	married	secondary	no	225
yes							
31646	14156	55	management	divorced	secondary	no	204
yes							

	loan	contact	day	month	duration	campaign	pdays	previous
\								
0	no	telephone	19	nov	44	2	-1	0
1	no	cellular	20	jul	91	2	-1	0
2	no	cellular	18	jul	240	1	-1	0
3	no	cellular	22	jun	867	1	84	3
4	no	cellular	4	feb	380	1	-1	0
...	...	...	...	...	...	...	...	...
31642	no	cellular	12	may	116	2	-1	0
31643	yes	cellular	5	jun	438	2	-1	0
31644	no	cellular	7	aug	37	3	-1	0
31645	no	telephone	15	may	22	7	337	12
31646	no	cellular	11	jul	1973	2	-1	0

	poutcome	subscribed
0	unknown	no
1	unknown	no
2	unknown	no

```

3      success      yes
4      unknown      no
...      ...      ...
31642  unknown      no
31643  unknown      yes
31644  unknown      no
31645  failure      no
31646  unknown      yes

```

```
[31647 rows x 18 columns]
```

```
df['job'].unique()
```

```

array(['admin.', 'unknown', 'services', 'management', 'technician',
      'retired', 'blue-collar', 'housemaid', 'self-employed',
      'student',
      'entrepreneur', 'unemployed'], dtype=object)

```

```
df['marital'].unique()
```

```
array(['married', 'divorced', 'single'], dtype=object)
```

```
df['education'].unique()
```

```
array(['unknown', 'secondary', 'tertiary', 'primary'], dtype=object)
```

```
df['contact'].unique()
```

```
array(['telephone', 'cellular', 'unknown'], dtype=object)
```

```
df.dtypes
```

```

ID          int64
age         int64
job         object
marital     object
education   object
default     object
balance     int64
housing     object
loan        object
contact     object
day         int64
month       object
duration    int64
campaign    int64
pdays      int64
previous    int64
poutcome    object
subscribed  object
dtype: object

```



```
table = pd.crosstab(df['job'], df['subscribed'])
print(table)
```

subscribed	no	yes
job		
admin.	3179	452
blue-collar	6353	489
entrepreneur	923	85
housemaid	795	79
management	5716	923
retired	1212	362
self-employed	983	140
services	2649	254
student	453	182
technician	4713	594
unemployed	776	129
unknown	180	26

```
table = pd.crosstab(df['marital'], df['subscribed'])
print(table)
```

subscribed	no	yes
marital		
divorced	3185	445
married	17176	1919
single	7571	1351

```
table = pd.crosstab(df['education'], df['subscribed'])
print(table)
```

subscribed	no	yes
education		
primary	4381	427
secondary	14527	1697
tertiary	7886	1415
unknown	1138	176

```
table = pd.crosstab(df['default'], df['subscribed'])
print(table)
```

subscribed	no	yes
default		
no	27388	3674
yes	544	41

```
table = pd.crosstab(df['housing'], df['subscribed'])
print(table)
```

subscribed	no	yes
housing		
no	11698	2365
yes	16234	1350

```
table = pd.crosstab(df['loan'], df['subscribed'])
print(table)
```

subscribed	no	yes
loan		
no	23132	3384
yes	4800	331

```
table = pd.crosstab(df['contact'], df['subscribed'])
print(table)
```

subscribed	no	yes
contact		
cellular	17352	3071
telephone	1779	268
unknown	8801	376

```
table = pd.crosstab(df['month'], df['subscribed'])
print(table)
```

subscribed	no	yes
month		
apr	1671	384
aug	3813	520
dec	85	72
feb	1522	305
jan	880	97
jul	4403	441
jun	3355	383
mar	168	174
may	9020	649
nov	2508	275
oct	288	224
sep	219	191

```
table = pd.crosstab(df['poutcome'], df['subscribed'])
print(table)
```

subscribed	no	yes
poutcome		
failure	2931	431
other	1071	217
success	374	694
unknown	23556	2373

```
df.groupby('balance').mean().head(10)
```

```
<ipython-input-41-e5dda727ac69>:1: FutureWarning: The default value of
numeric_only in DataFrameGroupBy.mean is deprecated. In a future
version, numeric_only will default to False. Either specify
numeric_only or select only columns which should be valid for the
function.
```

```
df.groupby('balance').mean().head(10)
```

	ID	age	day	duration	campaign	pdays	previous
balance							
-8019	12910.0	26.0	7.0	299.0	3.0	-1.0	0.0
-6847	15683.0	49.0	21.0	206.0	1.0	-1.0	0.0
-4057	38737.0	60.0	18.0	242.0	6.0	-1.0	0.0
-3372	7414.0	43.0	29.0	386.0	2.0	-1.0	0.0
-3058	32714.0	39.0	17.0	882.0	3.0	-1.0	0.0
-2712	31510.0	52.0	2.0	253.0	1.0	-1.0	0.0
-2604	25120.0	49.0	18.0	142.0	1.0	-1.0	0.0
-2282	14435.0	51.0	14.0	301.0	6.0	-1.0	0.0
-2122	25241.0	43.0	18.0	141.0	3.0	-1.0	0.0
-2082	17160.0	51.0	28.0	123.0	6.0	-1.0	0.0

```
grouped = df.groupby('marital')
```

```
grouped
```

```
<pandas.core.groupby.generic.DataFrameGroupBy object at
0x7f9939eala80>
```

```
grouped_mean = grouped_marital.mean()
```

```
<ipython-input-48-9652e5ae1168>:1: FutureWarning: The default value of
numeric_only in DataFrameGroupBy.mean is deprecated. In a future
version, numeric_only will default to False. Either specify
numeric_only or select only columns which should be valid for the
function.
```

```
grouped_mean = grouped_marital.mean()
```

```
grouped_sum = grouped['balance'].sum()
```

```
grouped_count = grouped['poutcome'].count()
```

```
print("Mean balance:")
```

```
print(grouped_mean)
```

```
print("\nTotal balance:")
```

```
print(grouped_sum)
```

```
print("\n Poutcome:")
```

```
print(grouped_count)
```

```
Mean balance:
```

	ID	age	balance	day	duration
\					
marital					

divorced	21759.828375	45.945455	1212.203030	15.899449	263.644628
married	21732.145954	43.401885	1410.935847	15.875622	253.023828
single	24671.432751	33.695696	1324.917956	15.723492	266.756221

	campaign	pdays	previous
marital			
divorced	2.578788	40.983471	0.562259
married	2.866195	37.067976	0.547840
single	2.626653	44.371217	0.635732

Total balance:

marital	
divorced	4400297
married	26941820
single	11820918

Name: balance, dtype: int64

Poutcome:

marital	
divorced	3630
married	19095
single	8922

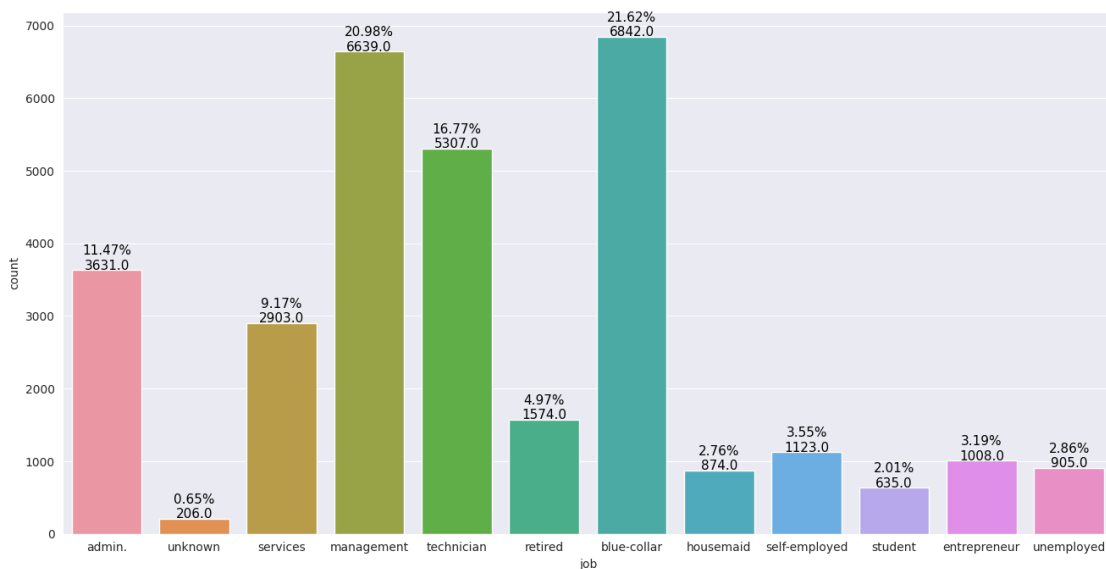
Name: poutcome, dtype: int64

df.dtypes

ID	int64
age	int64
job	object
marital	object
education	object
default	object
balance	int64
housing	object
loan	object
contact	object
day	int64
month	object
duration	int64
campaign	int64
pdays	int64
previous	int64
poutcome	object
subscribed	object
dtype:	object

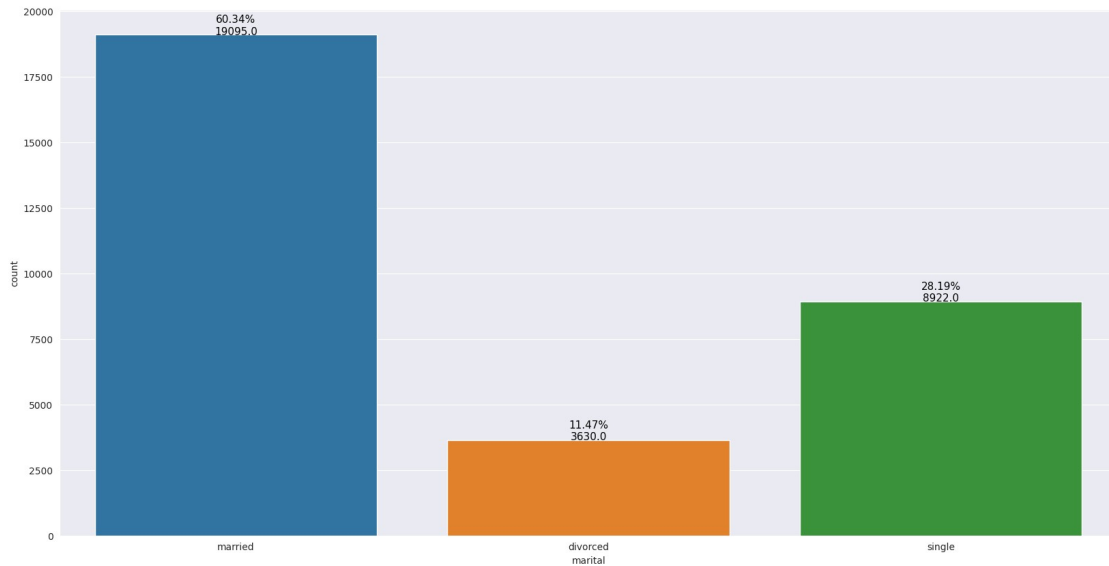
```
plt.figure(figsize=(16, 8))
ax=sns.countplot(df , x="job")
for p in ax.patches:
    height = p.get_height()
    ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
                xy=(p.get_x() + p.get_width() / 2., height),
                ha='center', va='center', fontsize=11, color='black',
xytext=(0, 10),
                textcoords='offset points')
```

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



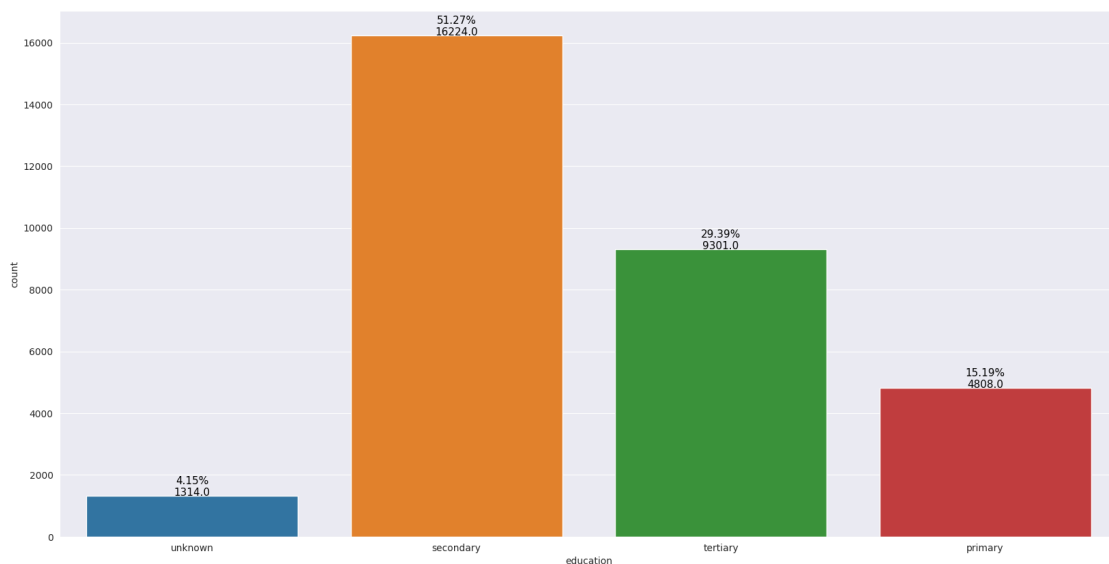
```
plt.figure(figsize=(20, 10))
ax=sns.countplot(df , x="marital")
for p in ax.patches:
    height = p.get_height()
    ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
                xy=(p.get_x() + p.get_width() / 2., height),
                ha='center', va='center', fontsize=11, color='black',
xytext=(0, 10),
                textcoords='offset points')
```

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



```
plt.figure(figsize=(20, 10))
ax=sns.countplot(df , x="education")
for p in ax.patches:
    height = p.get_height()
    ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
                xy=(p.get_x() + p.get_width() / 2., height),
                ha='center', va='center', fontsize=11, color='black',
xytext=(0, 10),
                textcoords='offset points')
```

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



```
plt.figure(figsize=(20, 10))
ax=sns.countplot(df , x="default")
for p in ax.patches:
```

```

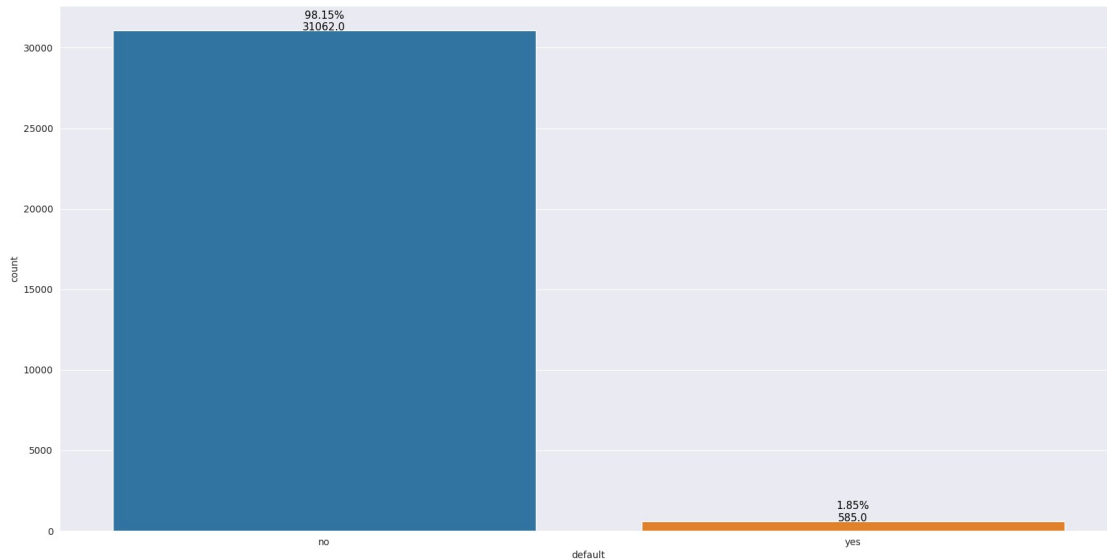
height = p.get_height()
ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
            xy=(p.get_x() + p.get_width() / 2., height),
            ha='center', va='center', fontsize=11, color='black',
xytext=(0, 10),
            textcoords='offset points')

```

```

# Display the plot
plt.show()

```



```

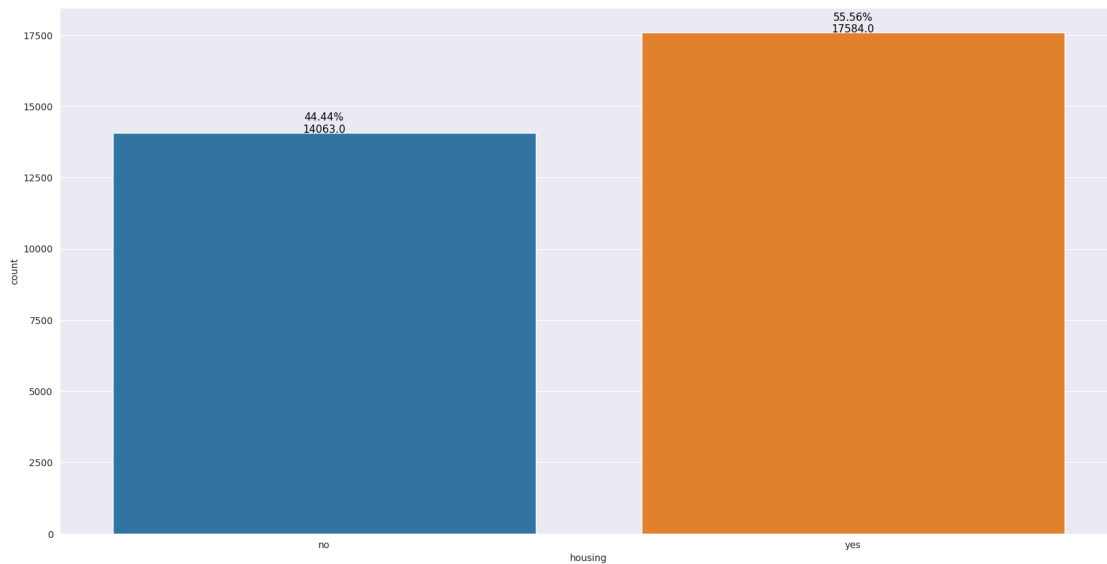
plt.figure(figsize=(20, 10))
ax=sns.countplot(df , x="housing")
for p in ax.patches:
    height = p.get_height()
    ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
                xy=(p.get_x() + p.get_width() / 2., height),
                ha='center', va='center', fontsize=11, color='black',
xytext=(0, 10),
                textcoords='offset points')

```

```

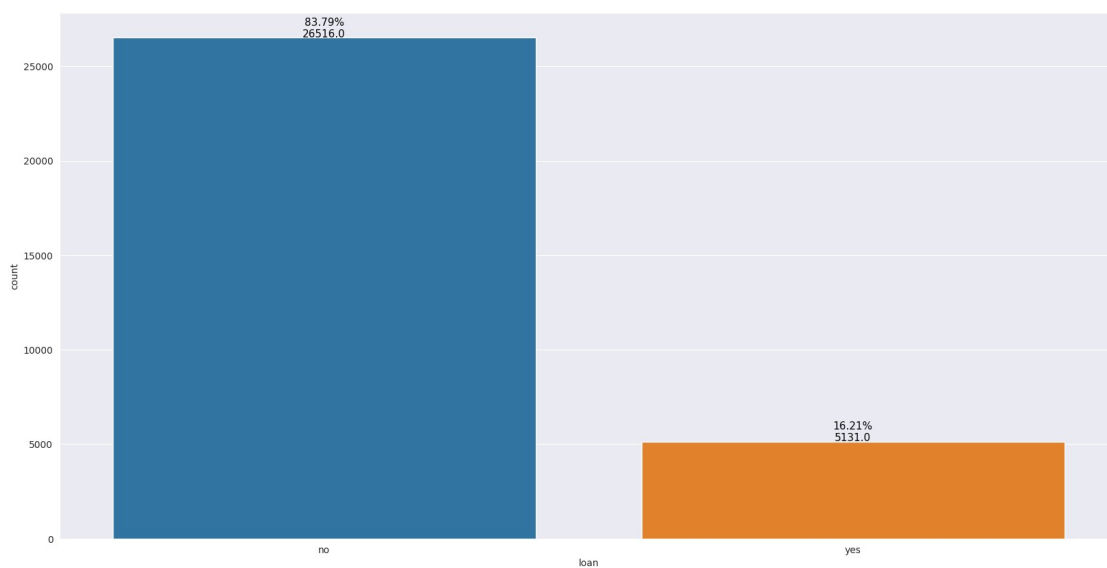
# Display the plot
plt.show()

```



```
plt.figure(figsize=(20, 10))
ax=sns.countplot(df , x="loan")
for p in ax.patches:
    height = p.get_height()
    ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
                xy=(p.get_x() + p.get_width() / 2., height),
                ha='center', va='center', fontsize=11, color='black',
xytext=(0, 10),
textcoords='offset points')
```

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



```
plt.figure(figsize=(20, 10))
ax=sns.countplot(df , x="contact")
for p in ax.patches:
```



```

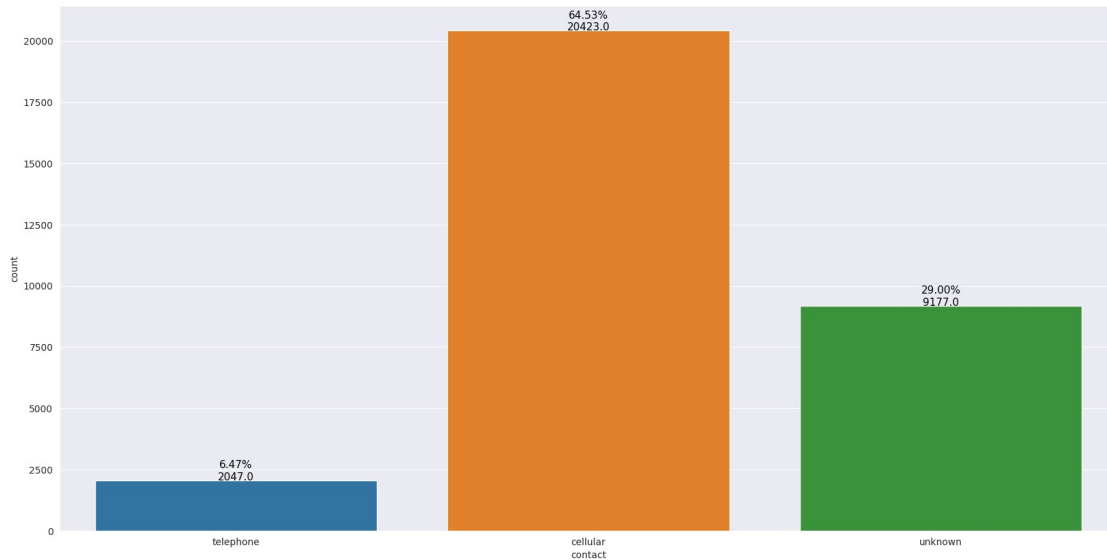
height = p.get_height()
ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
            xy=(p.get_x() + p.get_width() / 2., height),
            ha='center', va='center', fontsize=11, color='black',
xytext=(0, 10),
            textcoords='offset points')

```

```

# Display the plot
plt.show()

```



```

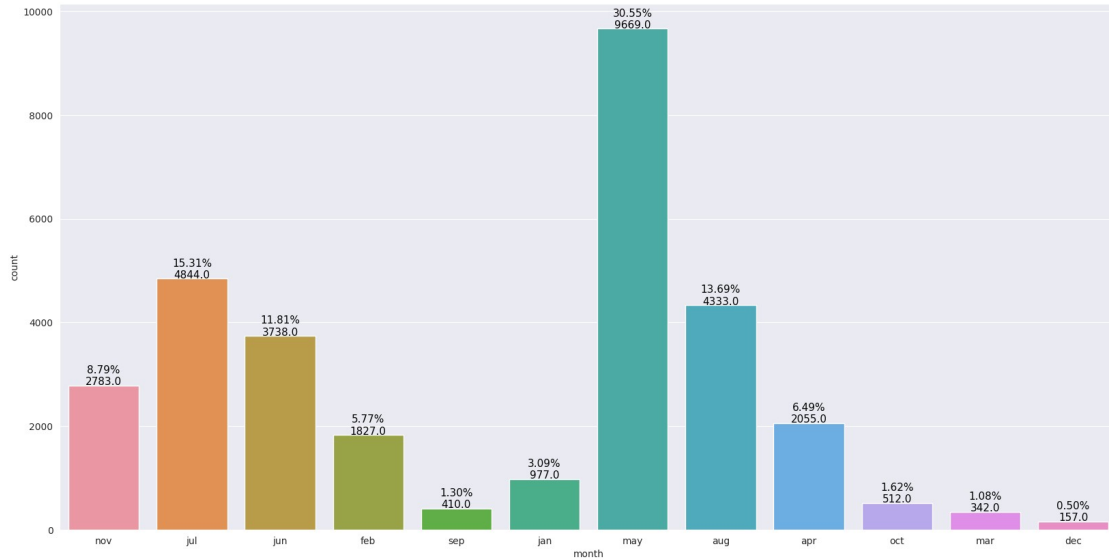
plt.figure(figsize=(20, 10))
ax=sns.countplot(df , x="month")
for p in ax.patches:
    height = p.get_height()
    ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
                xy=(p.get_x() + p.get_width() / 2., height),
                ha='center', va='center', fontsize=11, color='black',
xytext=(0, 10),
                textcoords='offset points')

```

```

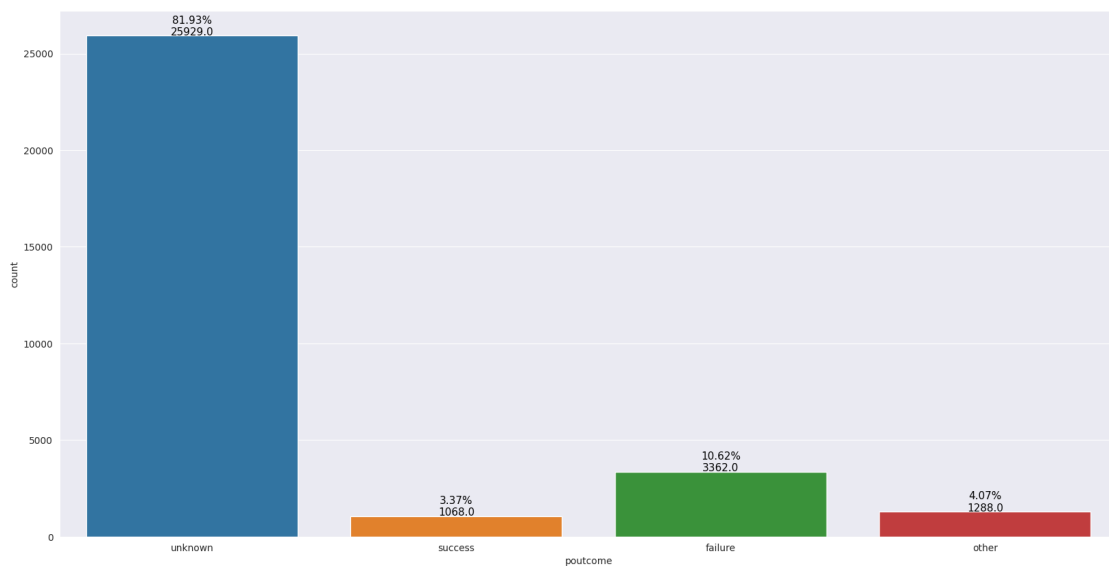
# Display the plot
plt.show()

```



```
plt.figure(figsize=(20, 10))
ax=sns.countplot(df , x="poutcome")
for p in ax.patches:
    height = p.get_height()
    ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
                xy=(p.get_x() + p.get_width() / 2., height),
                ha='center', va='center', fontsize=11, color='black',
xytext=(0, 10),
textcoords='offset points')
```

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



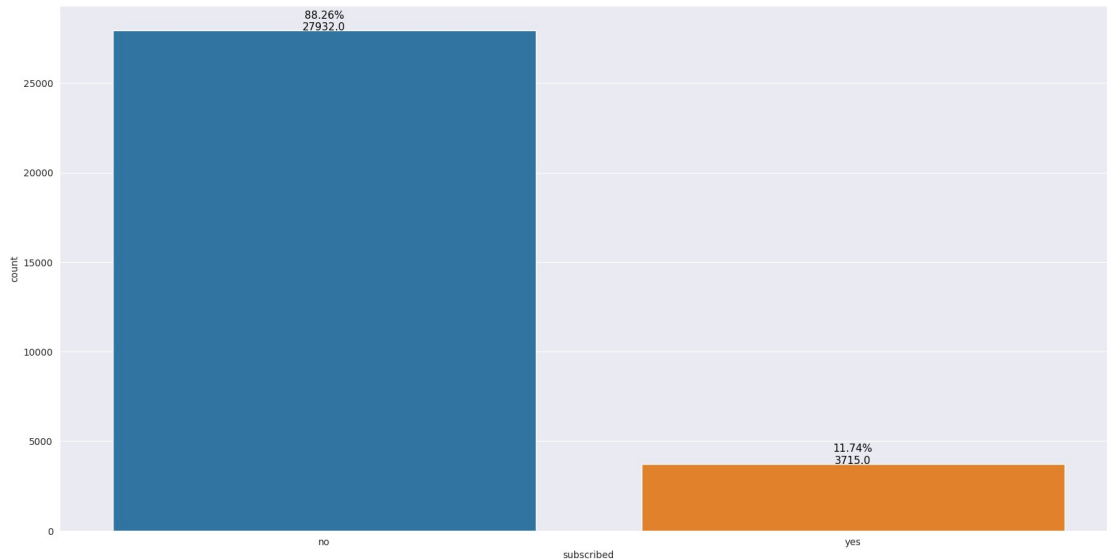
```
plt.figure(figsize=(20, 10))
ax=sns.countplot(df , x="subscribed")
for p in ax.patches:
```

```

height = p.get_height()
ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
            xy=(p.get_x() + p.get_width() / 2., height),
            ha='center', va='center', fontsize=11, color='black',
xytext=(0, 10),
            textcoords='offset points')

# Display the plot
plt.show()

```



df.dtypes

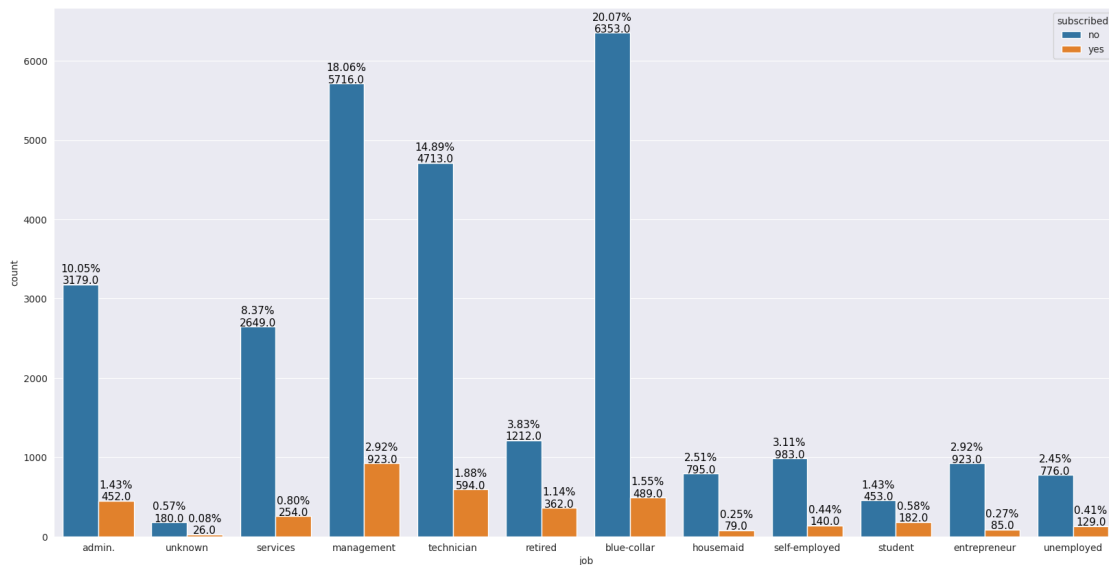
```

ID                int64
age              int64
job              object
marital          object
education        object
default          object
balance          int64
housing          object
loan             object
contact          object
day              int64
month            object
duration         int64
campaign         int64
pdays          int64
previous         int64
poutcome        object
subscribed       object
dtype: object

```

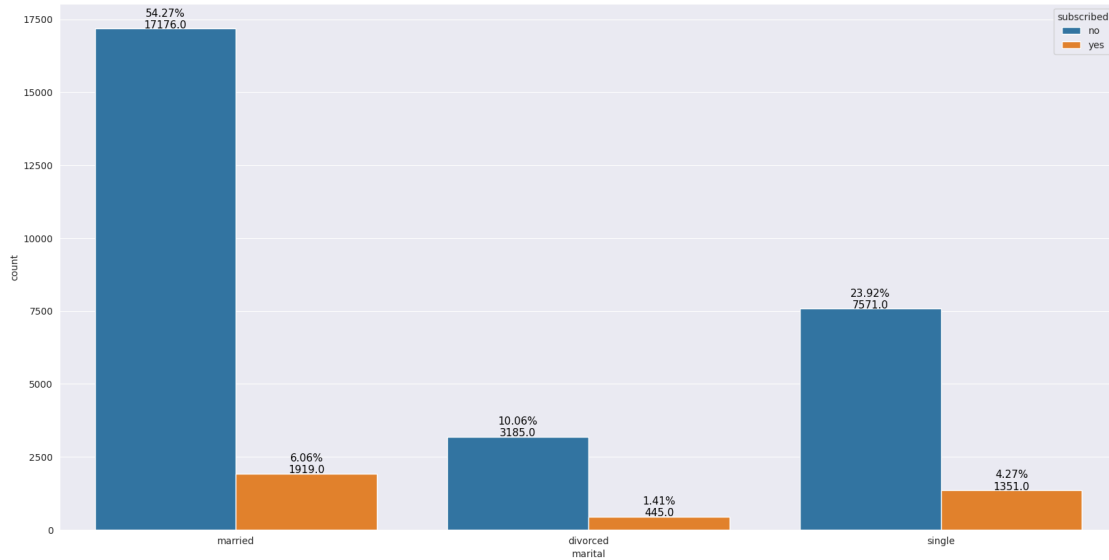
```
plt.figure(figsize=(20, 10))
ax=sns.countplot(df , x="job" , hue="subscribed")
for p in ax.patches:
    height = p.get_height()
    ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
                xy=(p.get_x() + p.get_width() / 2., height),
                ha='center', va='center', fontsize=11, color='black',
xytext=(0, 10),
                textcoords='offset points')

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



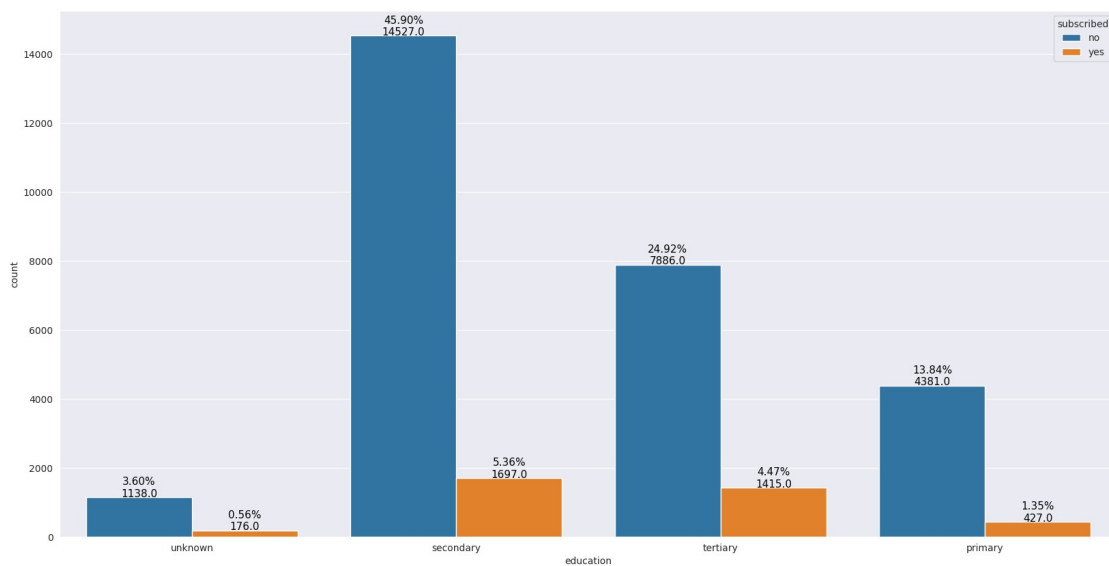
```
plt.figure(figsize=(20, 10))
ax=sns.countplot(df , x="marital" , hue="subscribed")
for p in ax.patches:
    height = p.get_height()
    ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
                xy=(p.get_x() + p.get_width() / 2., height),
                ha='center', va='center', fontsize=11, color='black',
xytext=(0, 10),
                textcoords='offset points')

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



```
plt.figure(figsize=(20, 10))
ax=sns.countplot(df , x="education" , hue="subscribed")
for p in ax.patches:
    height = p.get_height()
    ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
                xy=(p.get_x() + p.get_width() / 2., height),
                ha='center', va='center', fontsize=11, color='black',
                xytext=(0, 10),
                textcoords='offset points')
```

*# Display the plot*  
plt.show()



```
plt.figure(figsize=(20, 10))
ax=sns.countplot(df , x="default" , hue="subscribed")
for p in ax.patches:
```

```

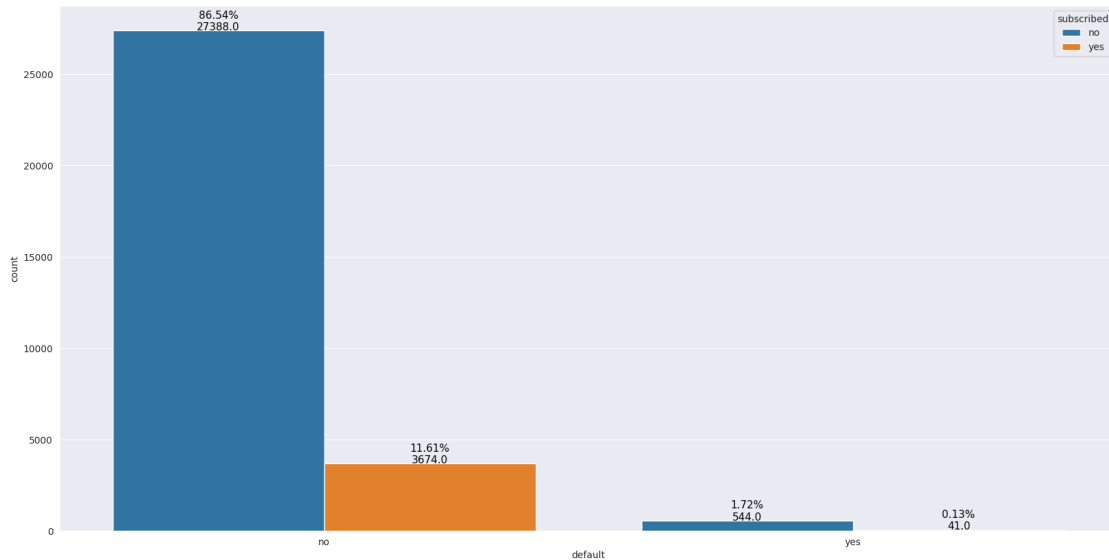
height = p.get_height()
ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
            xy=(p.get_x() + p.get_width() / 2., height),
            ha='center', va='center', fontsize=11, color='black',
xytext=(0, 10),
            textcoords='offset points')

```

```

# Display the plot
plt.show()

```



```

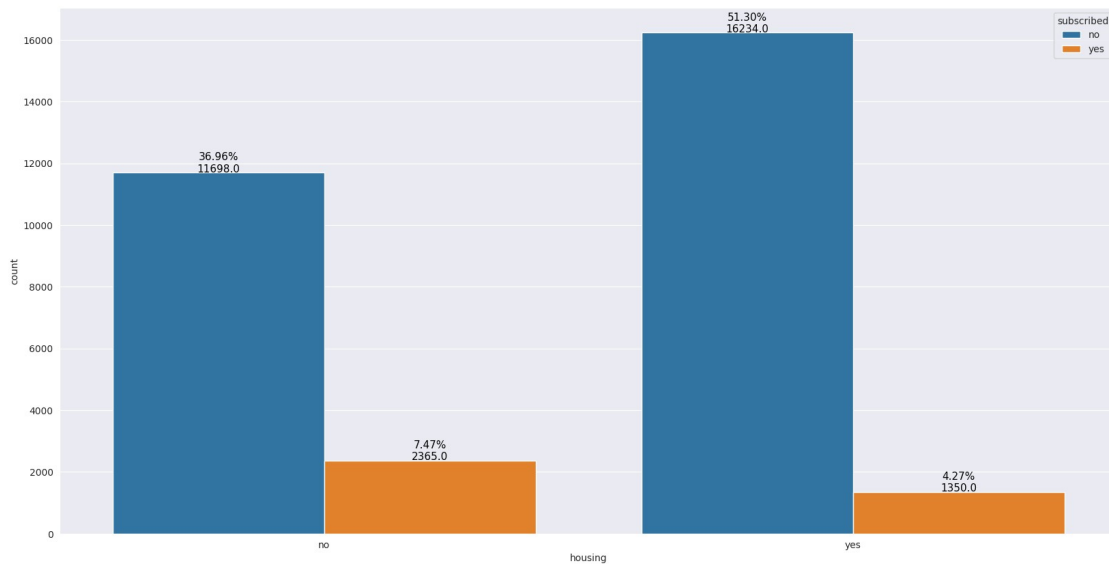
plt.figure(figsize=(20, 10))
ax=sns.countplot(df , x="housing" , hue="subscribed")
for p in ax.patches:
    height = p.get_height()
    ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
                xy=(p.get_x() + p.get_width() / 2., height),
                ha='center', va='center', fontsize=11, color='black',
xytext=(0, 10),
                textcoords='offset points')

```

```

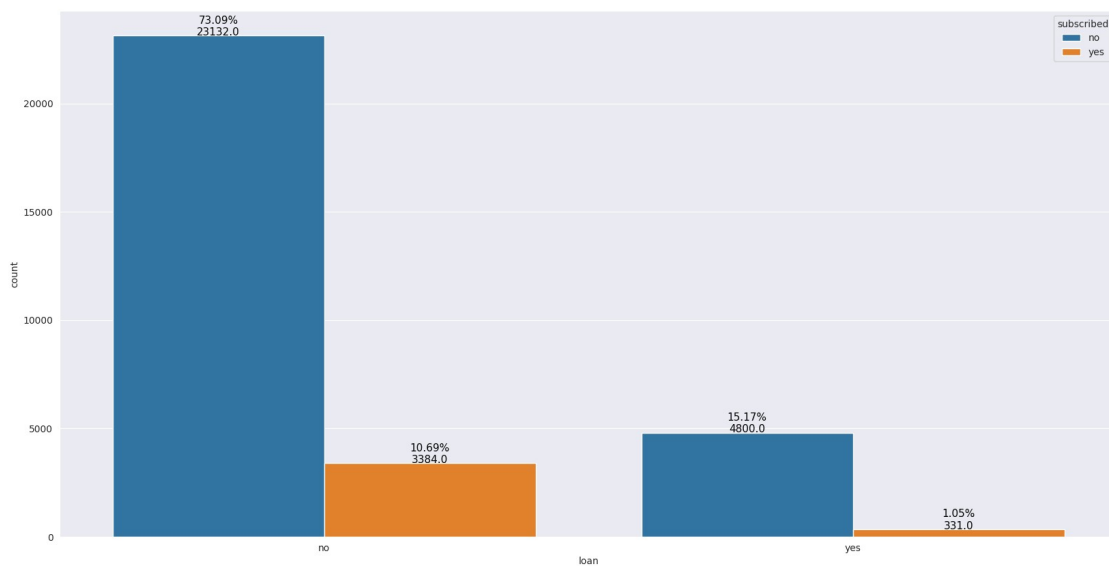
# Display the plot
plt.show()

```



```
plt.figure(figsize=(20, 10))
ax=sns.countplot(df , x="loan" , hue="subscribed")
for p in ax.patches:
    height = p.get_height()
    ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
                xy=(p.get_x() + p.get_width() / 2., height),
                ha='center', va='center', fontsize=11, color='black',
xytext=(0, 10),
textcoords='offset points')
```

*# Display the plot*  
plt.show()



```
plt.figure(figsize=(20, 10))
ax=sns.countplot(df , x="contact" , hue="subscribed")
for p in ax.patches:
```

```

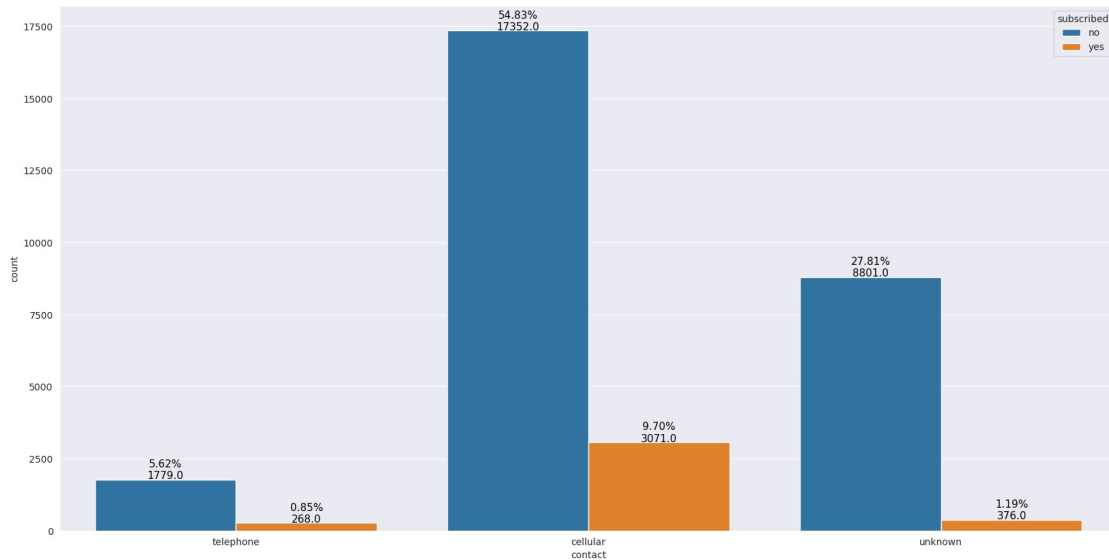
height = p.get_height()
ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
            xy=(p.get_x() + p.get_width() / 2., height),
            ha='center', va='center', fontsize=11, color='black',
xytext=(0, 10),
            textcoords='offset points')

```

```

# Display the plot
plt.show()

```



```

plt.figure(figsize=(20, 10))
ax=sns.countplot(df , x="month" , hue="subscribed")
for p in ax.patches:
    height = p.get_height()
    ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
                xy=(p.get_x() + p.get_width() / 2., height),
                ha='center', va='center', fontsize=11, color='black',
xytext=(0, 10),
                textcoords='offset points')

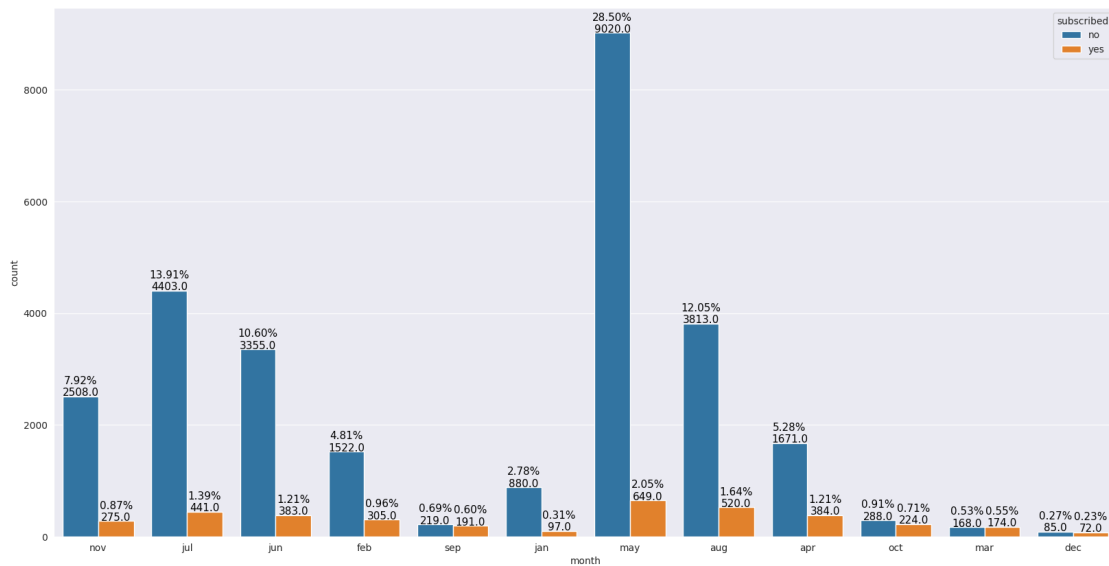
```

```

# Display the plot
plt.show()

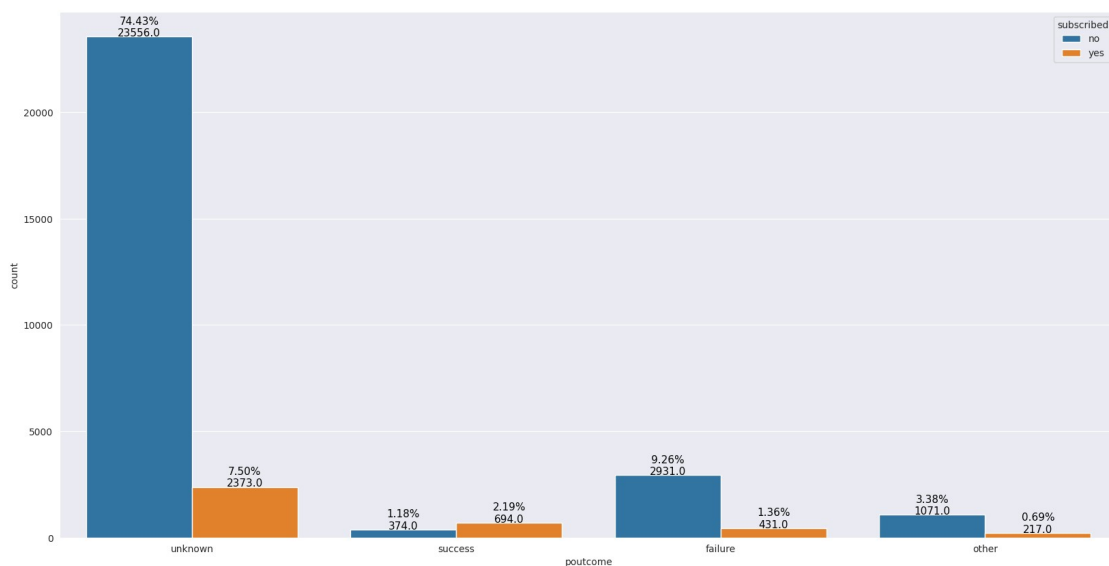
```





```
plt.figure(figsize=(20, 10))
ax=sns.countplot(df , x="outcome" , hue="subscribed")
for p in ax.patches:
    height = p.get_height()
    ax.annotate(f'{height / df.shape[0]:.2%}\n{height}',
                xy=(p.get_x() + p.get_width() / 2., height),
                ha='center', va='center', fontsize=11, color='black',
xytext=(0, 10),
                textcoords='offset points')

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



```
corr_matrix=df.corr()
```

```
<ipython-input-86-a4bfebfd3231>:1: FutureWarning: The default value of
numeric_only in DataFrame.corr is deprecated. In a future version, it
will default to False. Select only valid columns or specify the value
of numeric_only to silence this warning.
```

```
corr_matrix=df.corr()
```

```
plt.figure(figsize=(16, 8))
```

```
sns.heatmap(corr_matrix, cmap='coolwarm', annot=True, linewidths=0.5)
```

```
# set plot title and axis labels
```

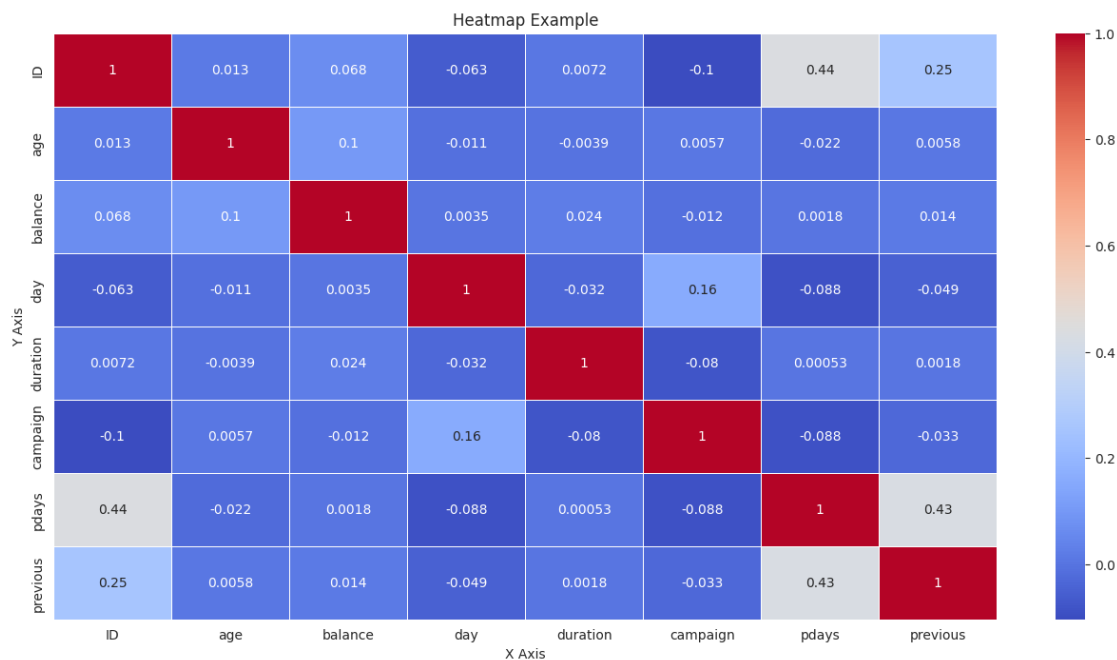
```
plt.title('Heatmap Example')
```

```
plt.xlabel('X Axis')
```

```
plt.ylabel('Y Axis')
```

```
# show the plot
```

```
plt.show()
```



```
df.dtypes
```

```
ID                int64
age               int64
job              object
marital          object
education        object
default          object
balance          int64
housing          object
loan             object
contact          object
day              int64
```

```
month          object
duration       int64
campaign       int64
pdays         int64
previous       int64
poutcome       object
subscribed     object
dtype: object
```

```
sns.distplot(df["age"])
```

```
<ipython-input-92-eef84e7ff8f0>:1: UserWarning:
```

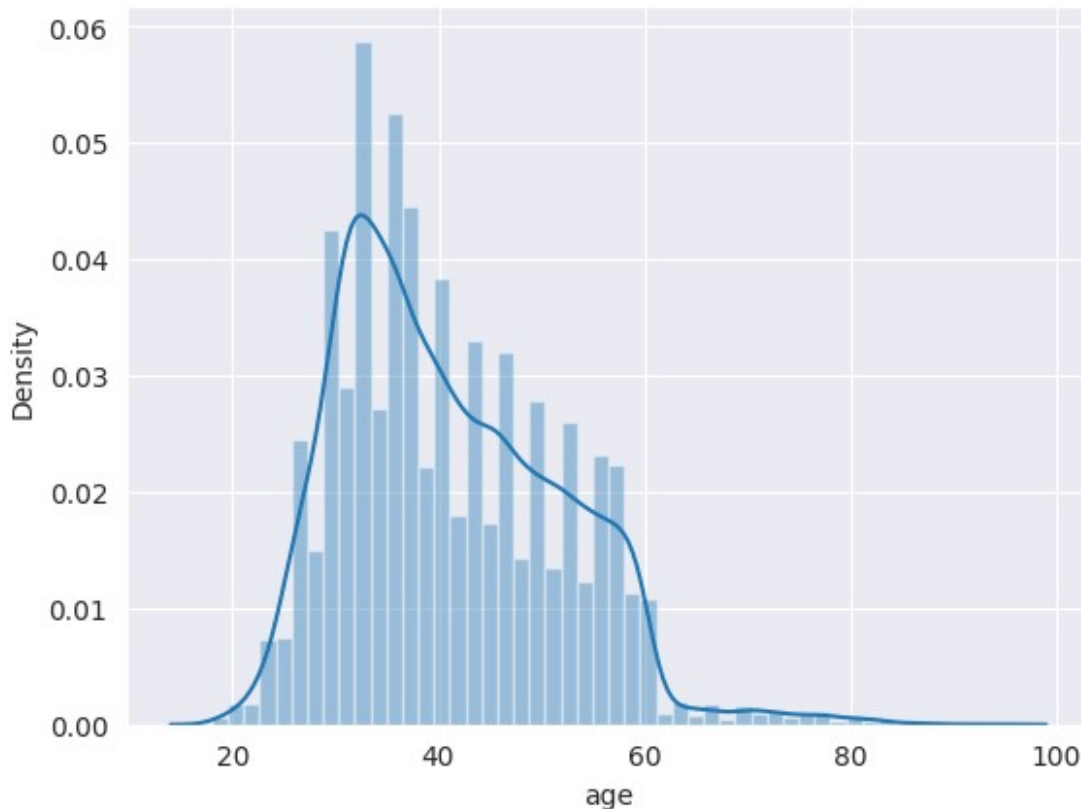
```
`distplot` is a deprecated function and will be removed in seaborn  
v0.14.0.
```

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(df["age"])
```

```
<Axes: xlabel='age', ylabel='Density'>
```



```
sns.distplot(df["balance"])
```

<ipython-input-94-adebfed082c1>:1: UserWarning:

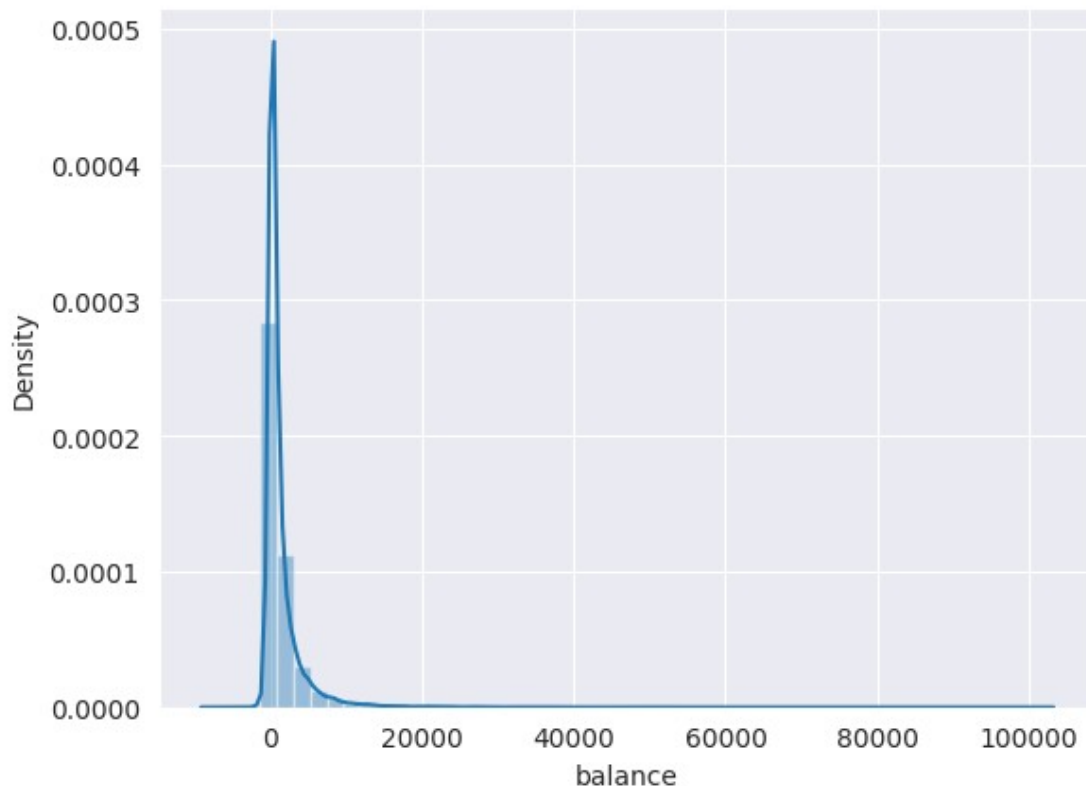
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(df["balance"])
```

<Axes: xlabel='balance', ylabel='Density'>



```
sns.distplot(df["day"])
```

```
<ipython-input-95-d8169acdb438>:1: UserWarning:
```

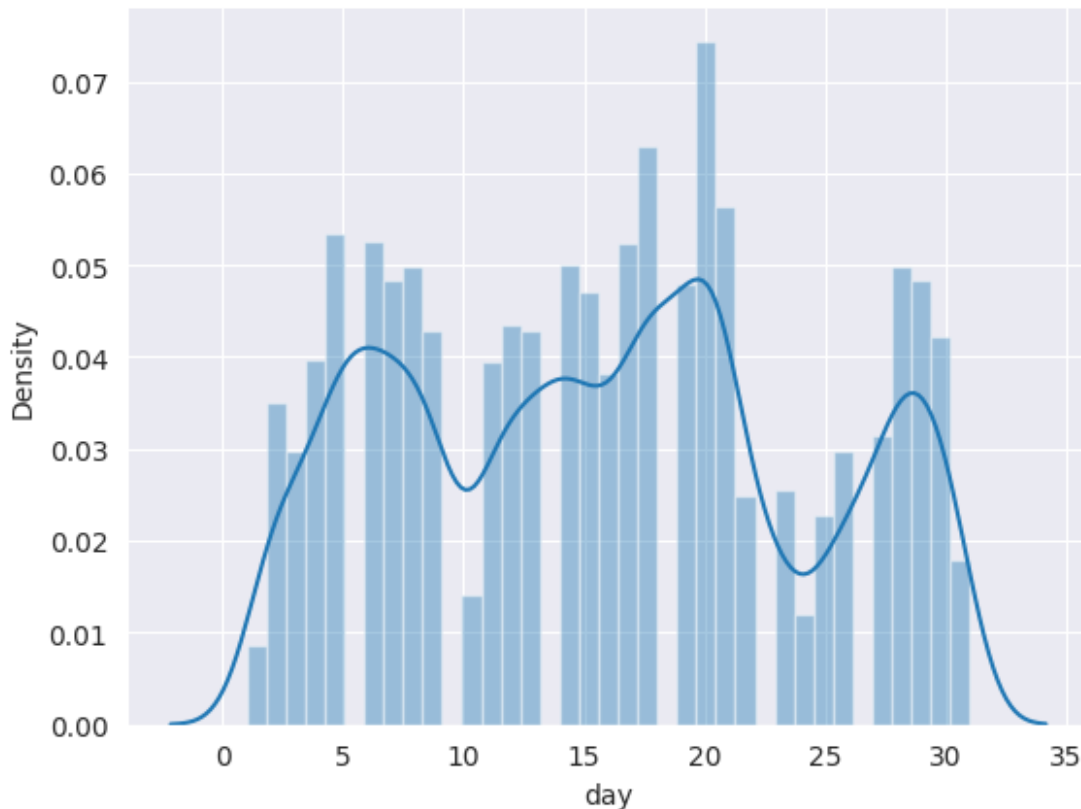
```
`distplot` is a deprecated function and will be removed in seaborn  
v0.14.0.
```

Please adapt your code to use either ``displot`` (a figure-level function with similar flexibility) or ``histplot`` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(df["day"])
```

```
<Axes: xlabel='day', ylabel='Density'>
```



```
sns.distplot(df["duration"])
```

<ipython-input-96-07d4faf592e6>:1: UserWarning:

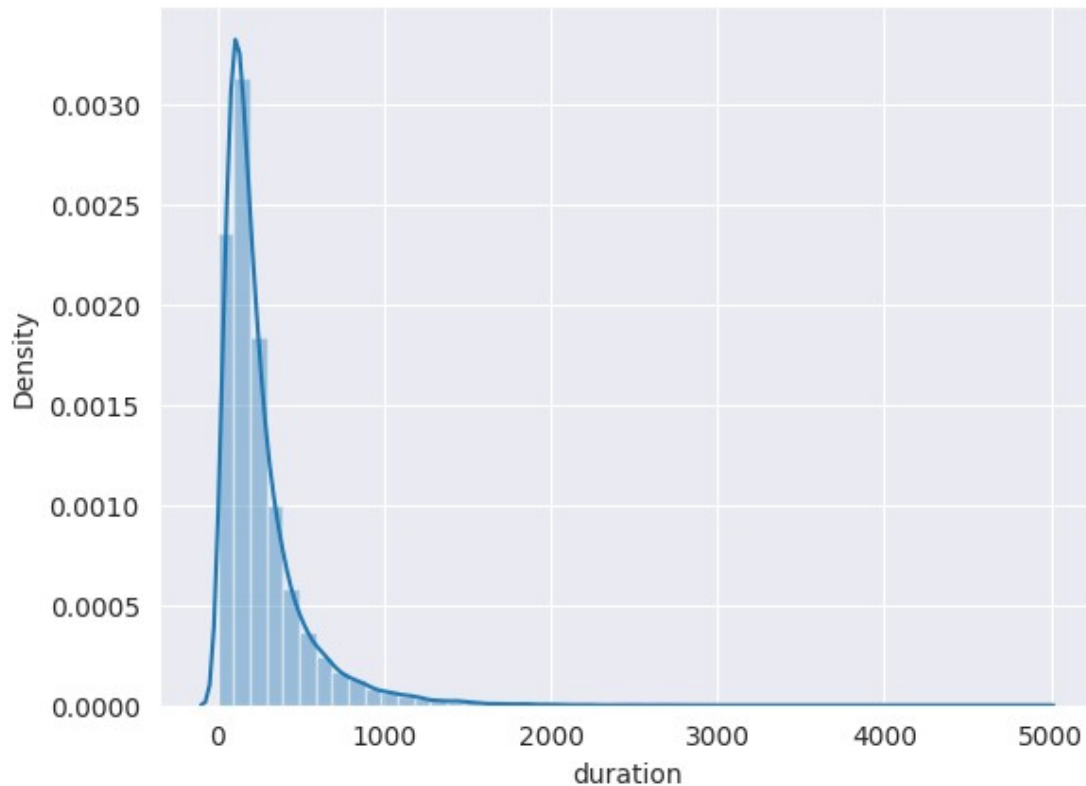
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(df["duration"])
```

<Axes: xlabel='duration', ylabel='Density'>



```
sns.distplot(df["campaign"])
```

```
<ipython-input-97-1570079ca3f4>:1: UserWarning:
```

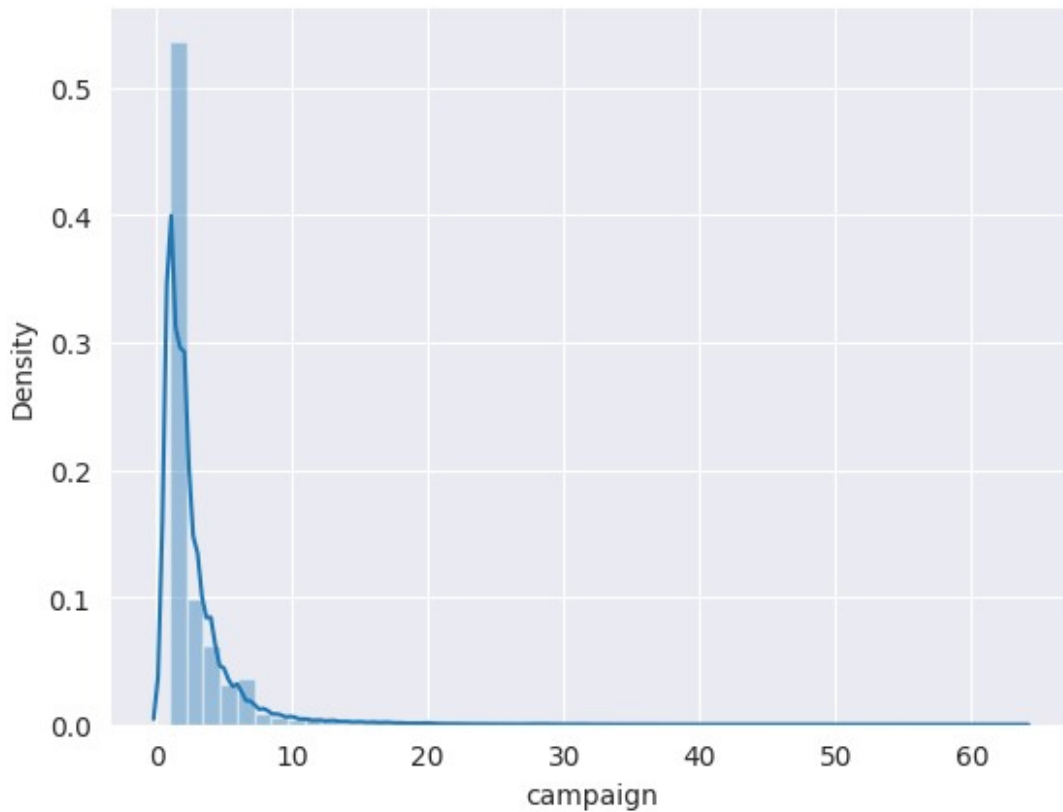
```
`distplot` is a deprecated function and will be removed in seaborn  
v0.14.0.
```

Please adapt your code to use either ``displot`` (a figure-level function with similar flexibility) or ``histplot`` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(df["campaign"])
```

```
<Axes: xlabel='campaign', ylabel='Density'>
```



```
sns.distplot(df["pdays"])
```

```
<ipython-input-98-973f1250fe67>:1: UserWarning:
```

```
`distplot` is a deprecated function and will be removed in seaborn  
v0.14.0.
```

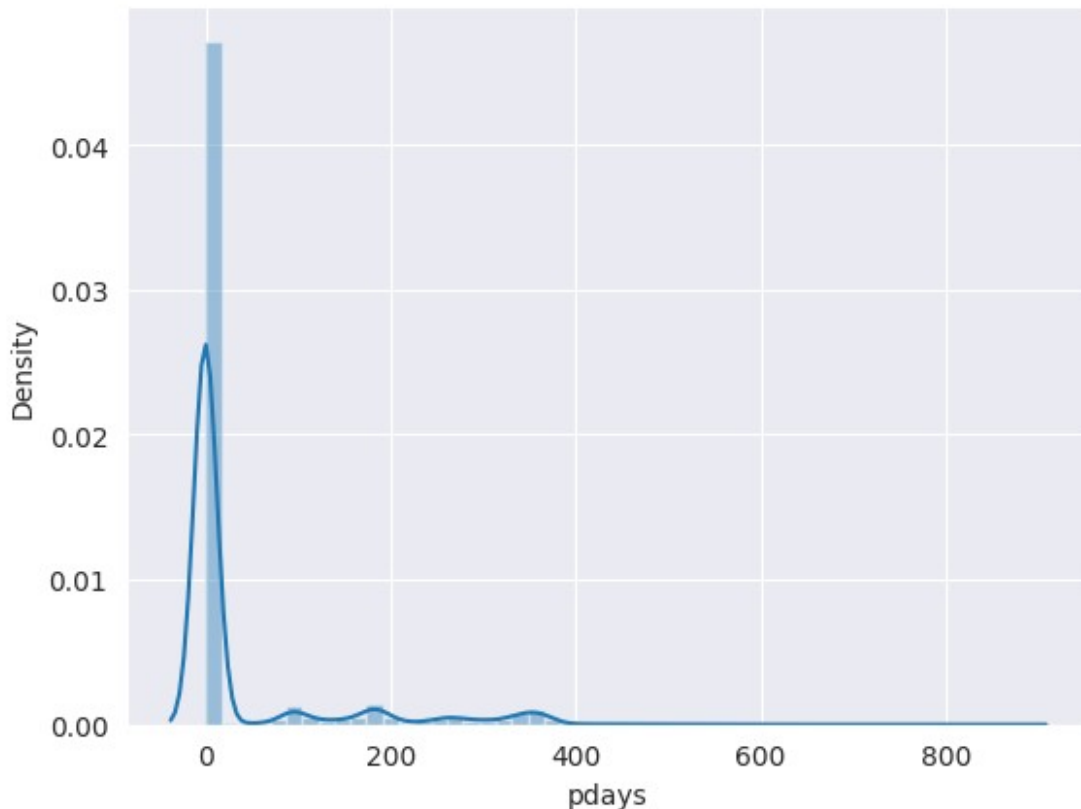
Please adapt your code to use either ``displot`` (a figure-level function with similar flexibility) or ``histplot`` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(df["pdays"])
```

```
<Axes: xlabel='pdays', ylabel='Density'>
```





```
sns.distplot(df["previous"])
```

```
<ipython-input-99-615a65952c4c>:1: UserWarning:
```

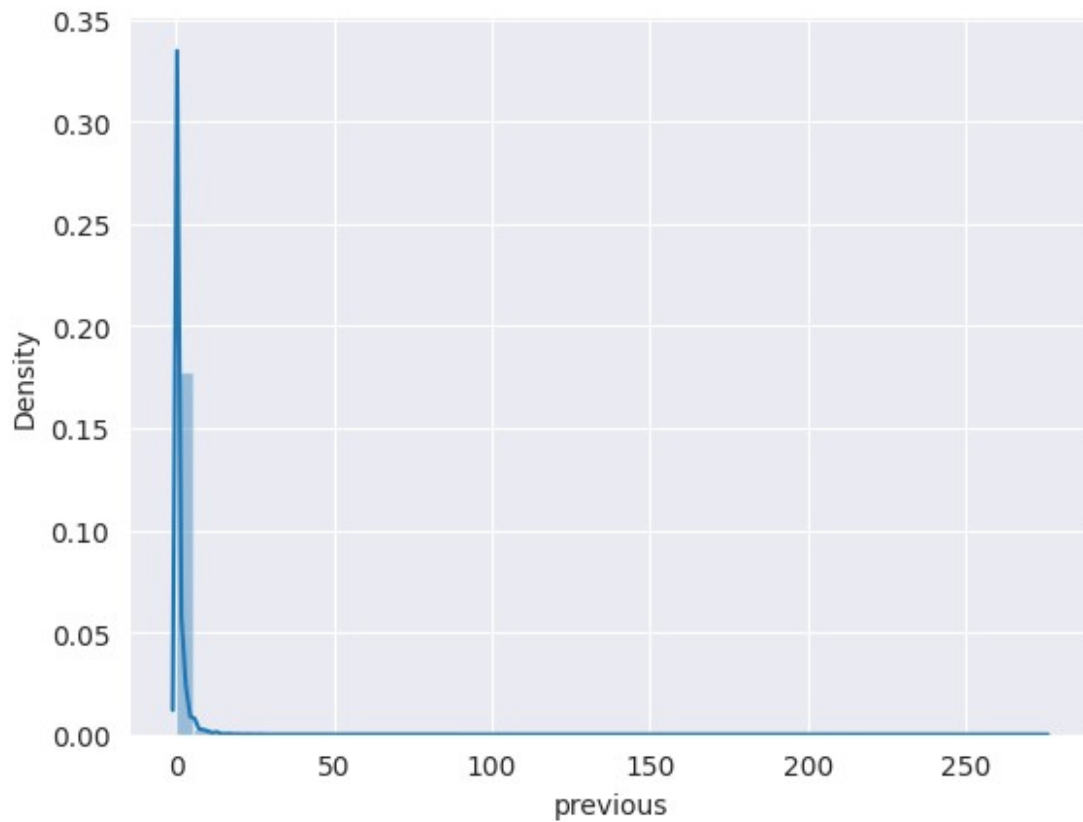
```
`distplot` is a deprecated function and will be removed in seaborn  
v0.14.0.
```

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(df["previous"])
```

```
<Axes: xlabel='previous', ylabel='Density'>
```



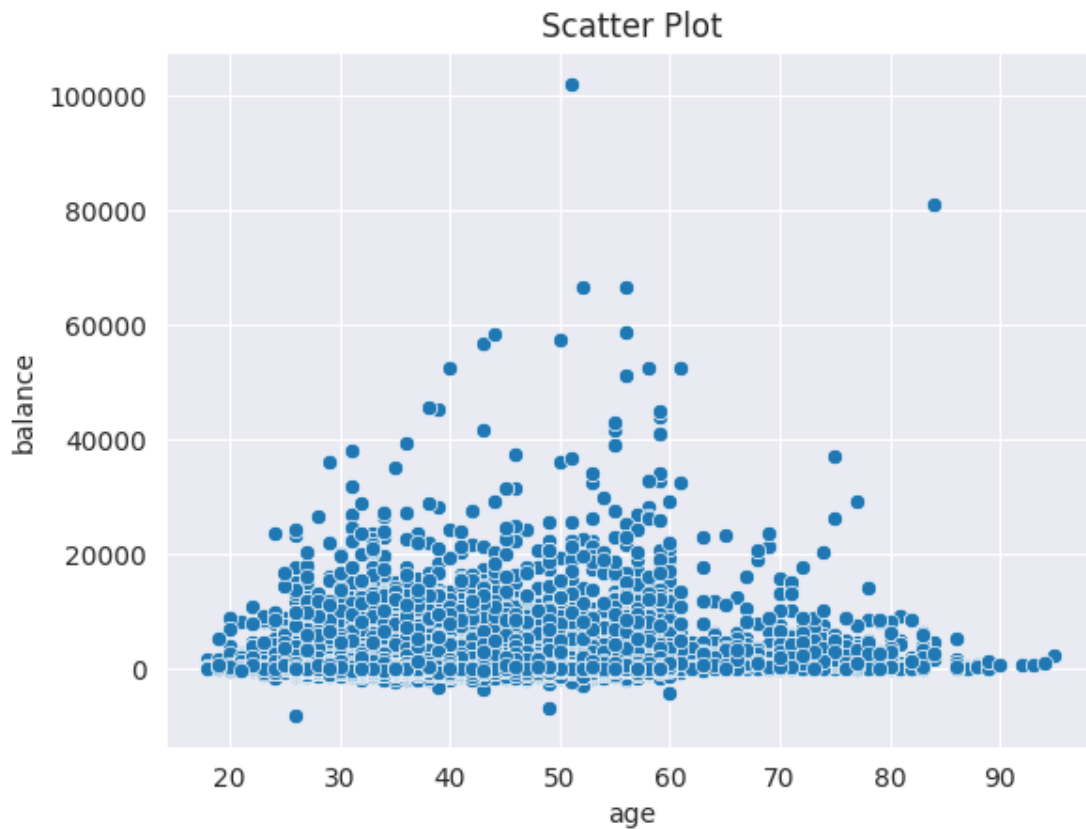
```
df.dtypes
```

```
ID          int64
age          int64
job          object
marital      object
education    object
default      object
balance      int64
housing      object
loan         object
contact      object
day          int64
month        object
duration     int64
campaign     int64
pdays       int64
previous     int64
poutcome     object
subscribed   object
dtype: object
```

```
sns.scatterplot(data=df, x='age', y='balance')
```

```
# Add labels and title
plt.xlabel('age')
plt.ylabel('balance')
plt.title('Scatter Plot')
```

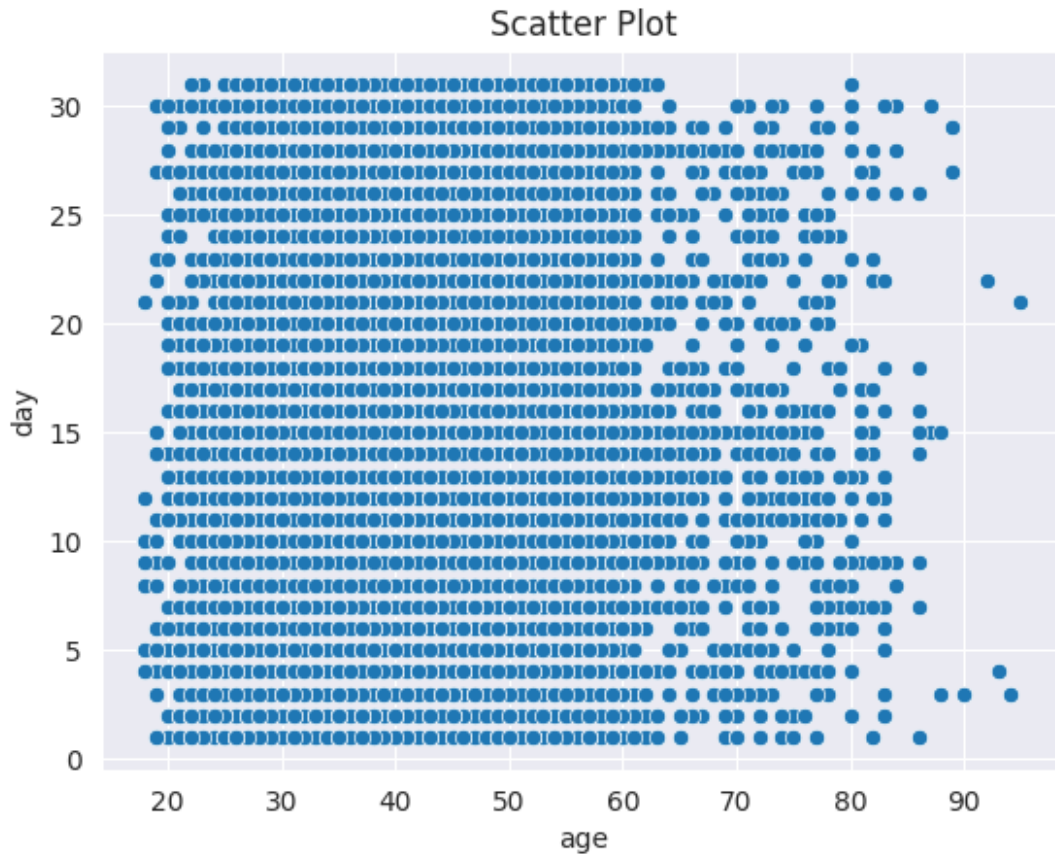
```
# Show the plot
plt.show()
```



```
sns.scatterplot(data=df, x='age', y='day')
```

```
# Add labels and title
plt.xlabel('age')
plt.ylabel('day')
plt.title('Scatter Plot')
```

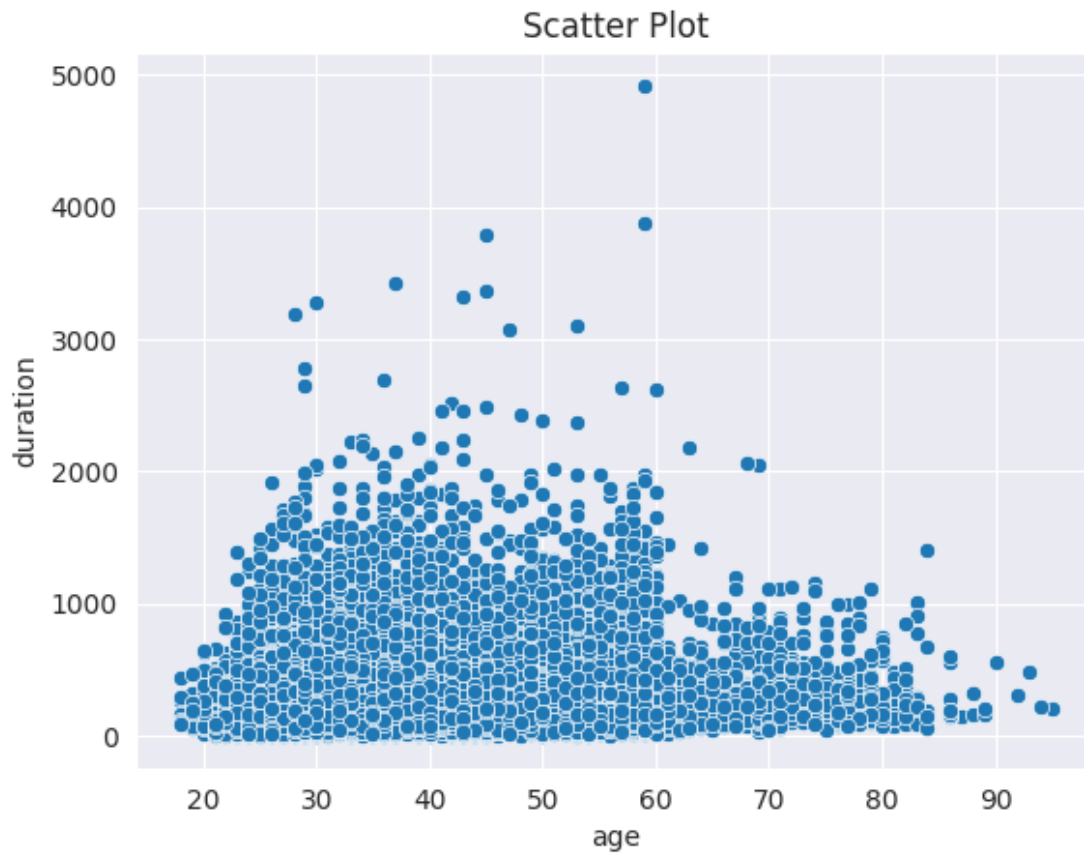
```
# Show the plot
plt.show()
```



```
sns.scatterplot(data=df, x='age', y='duration')
```

```
# Add labels and title
plt.xlabel('age')
plt.ylabel('duration')
plt.title('Scatter Plot')
```

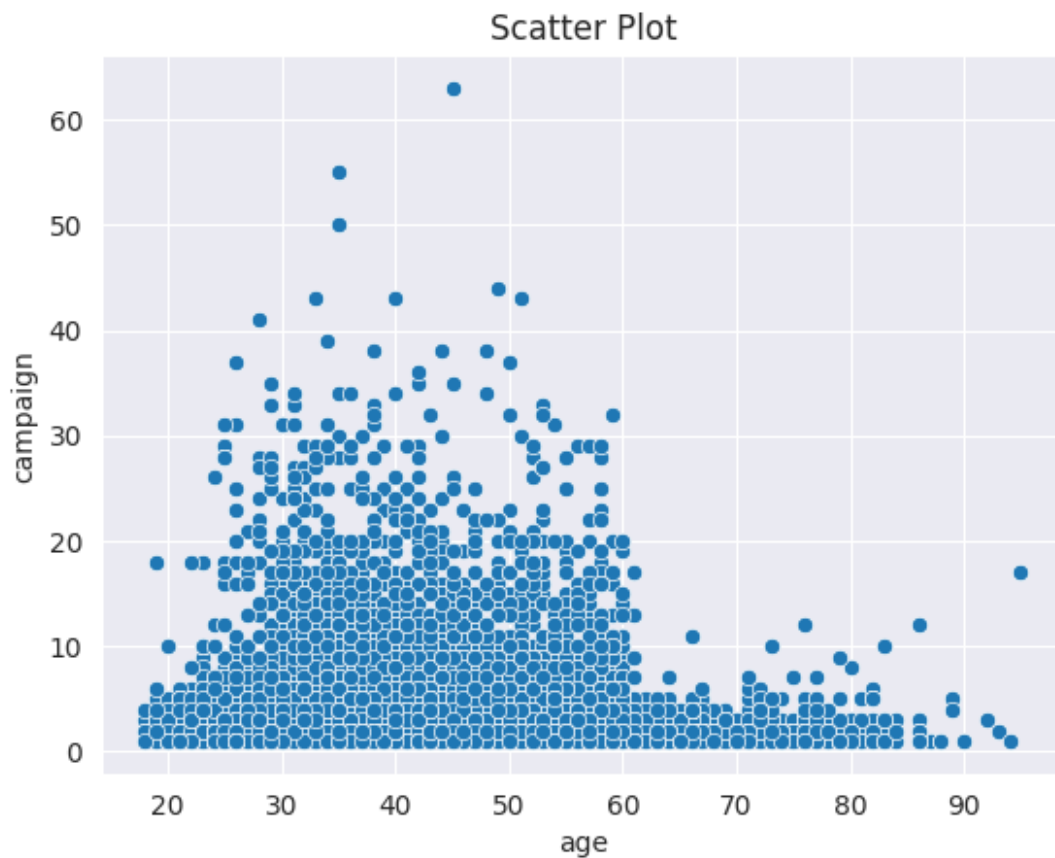
```
# Show the plot
plt.show()
```



```
sns.scatterplot(data=df, x='age', y='campaign')
```

```
# Add labels and title  
plt.xlabel('age')  
plt.ylabel('campaign')  
plt.title('Scatter Plot')
```

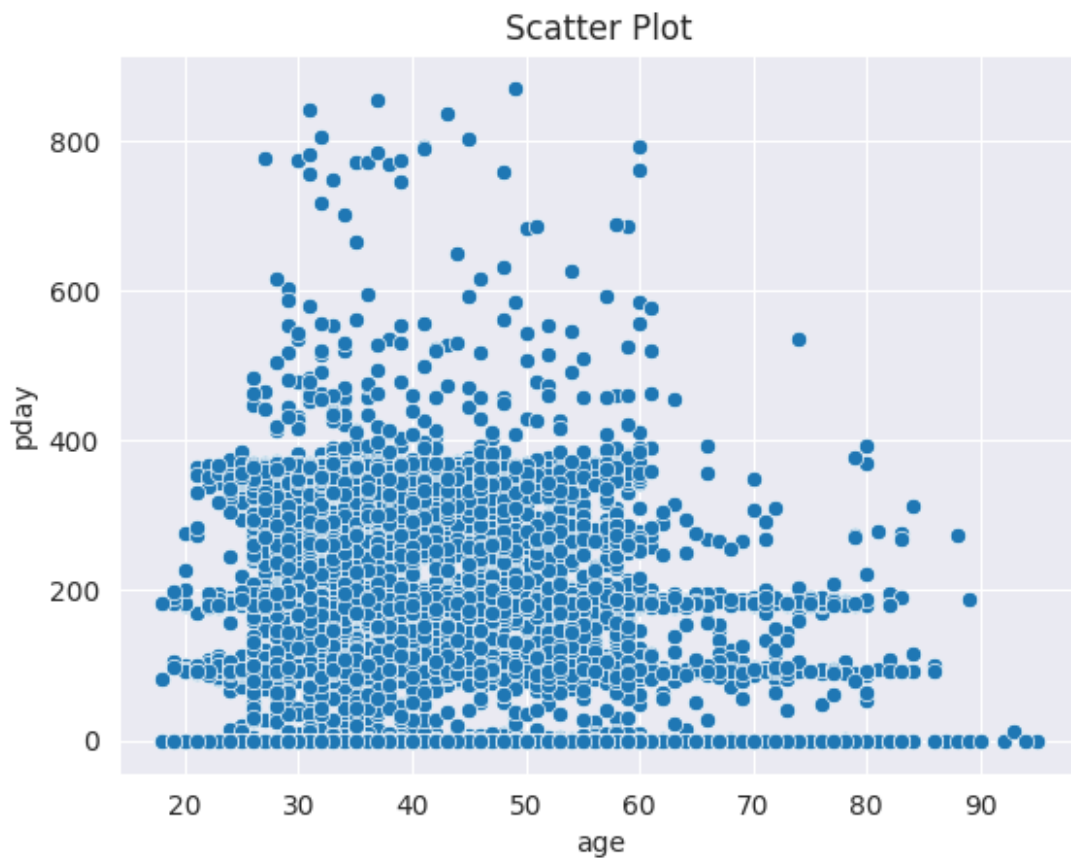
```
# Show the plot  
plt.show()
```



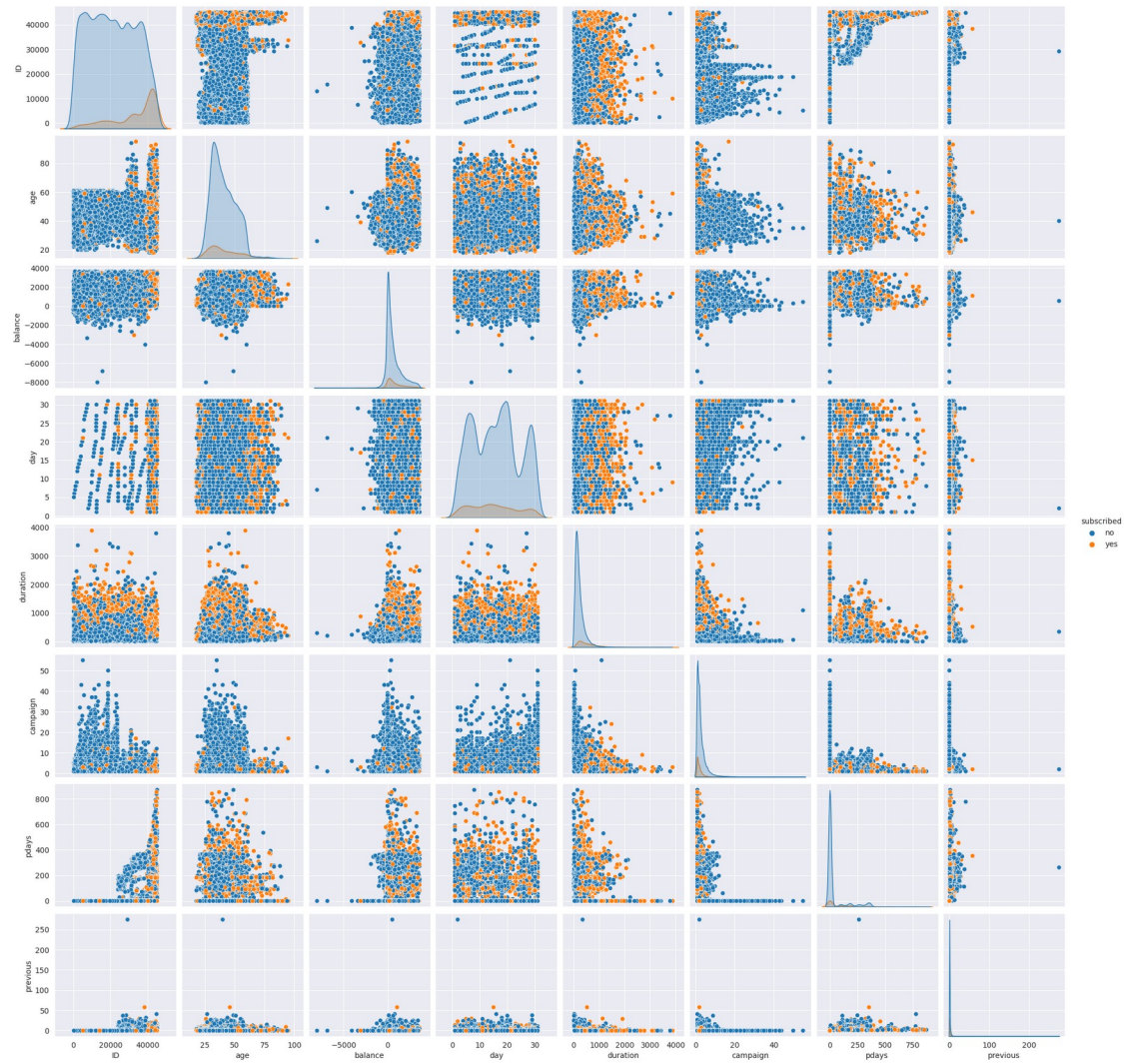
```
sns.scatterplot(data=df, x='age', y='pdays')
```

```
# Add labels and title  
plt.xlabel('age')  
plt.ylabel('pday')  
plt.title('Scatter Plot')
```

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



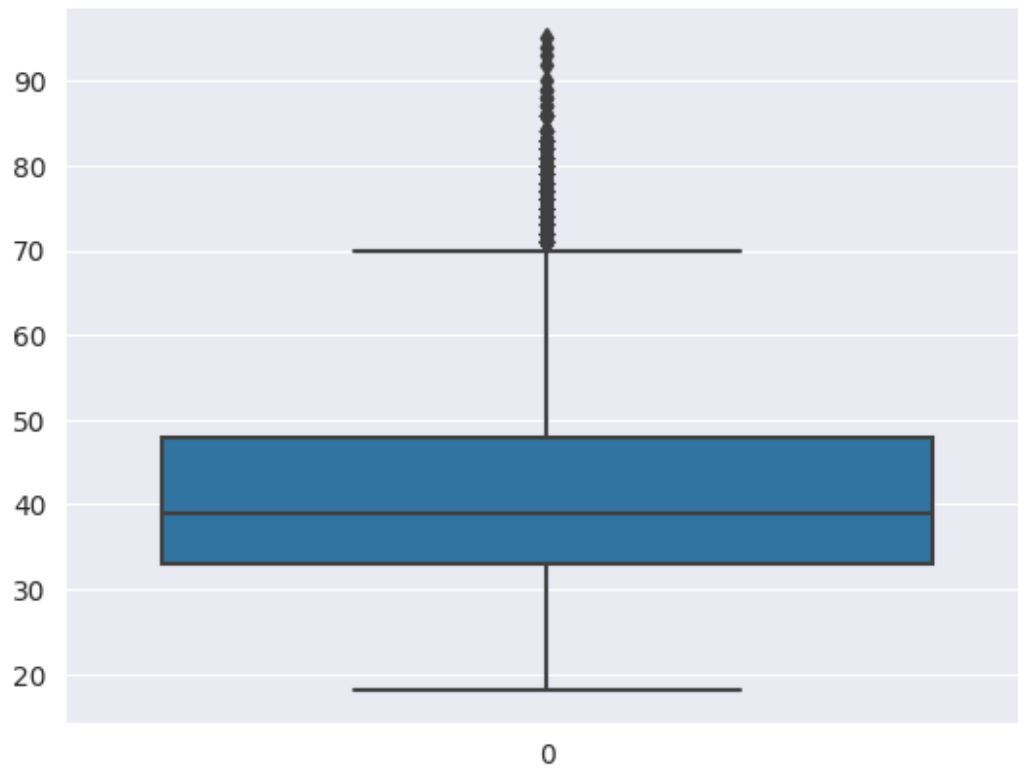
```
plt.figure(figsize=(10,6))  
sns.pairplot(df , hue="subscribed")  
  
<seaborn.axisgrid.PairGrid at 0x7f9927aafdf0>  
  
<Figure size 1000x600 with 0 Axes>
```



```
sns.boxplot(df['age'])
```

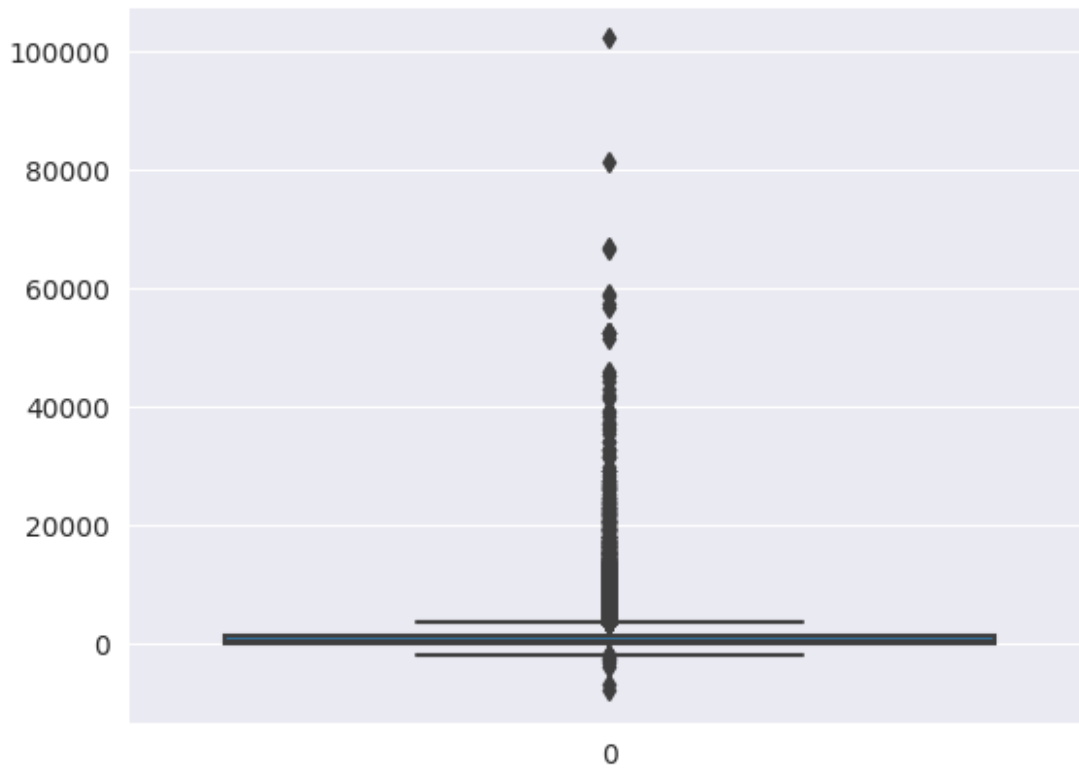
```
<Axes: >
```





```
sns.boxplot(df['balance'])
```

<Axes: >

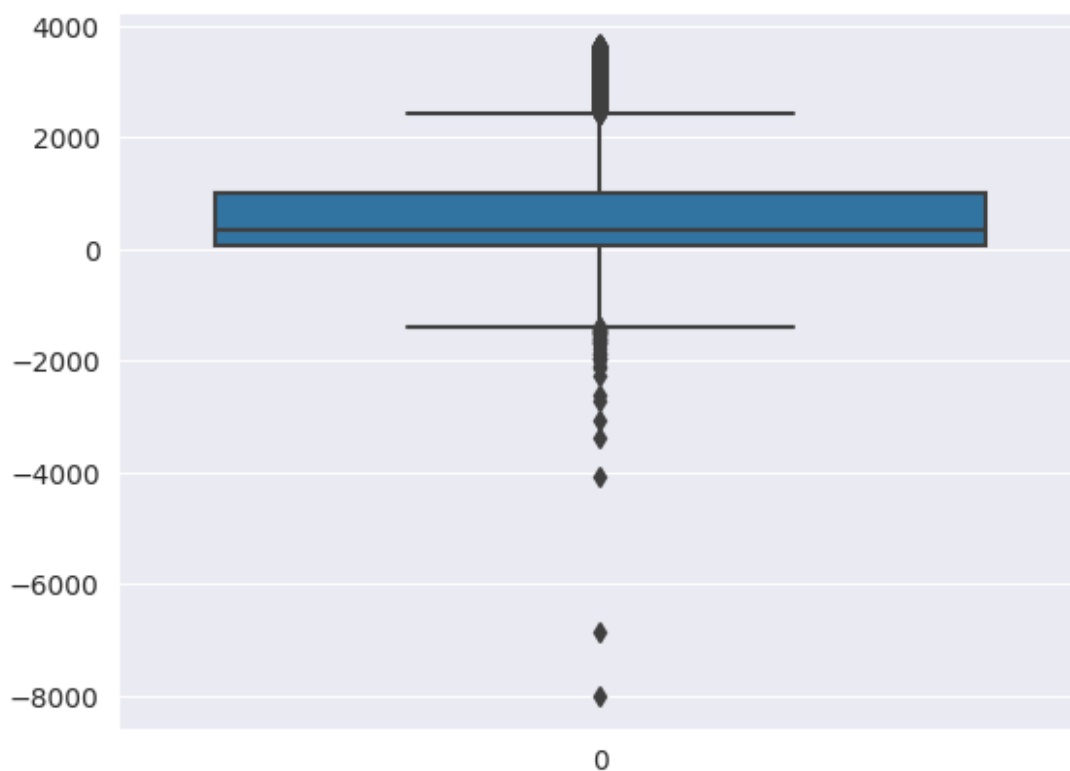


```
q = df['balance'].quantile(0.95)
```

```
# filter the DataFrame to exclude values above the 95th percentile  
df = df[df['balance'] < q]
```

```
sns.boxplot(df['balance'])
```

```
<Axes: >
```

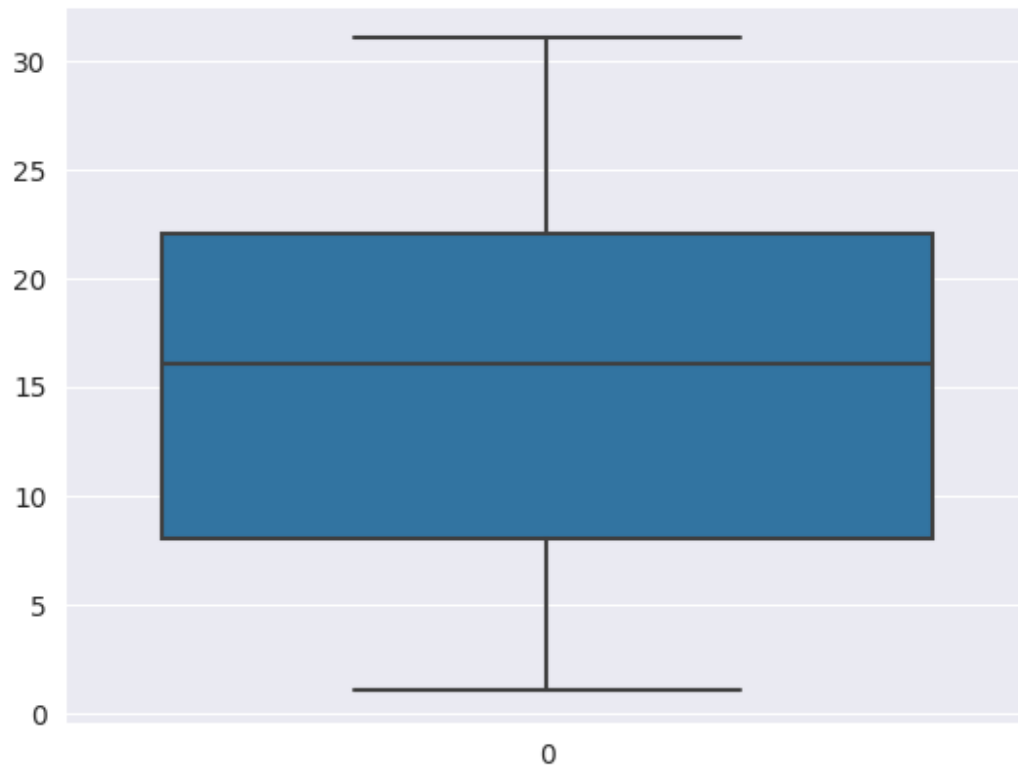


```
df.dtypes
```

```
ID          int64
age          int64
job          object
marital      object
education    object
default      object
balance      int64
housing      object
loan         object
contact      object
day          int64
month        object
duration     int64
campaign     int64
pdays       int64
previous     int64
poutcome    object
subscribed   object
dtype: object
```

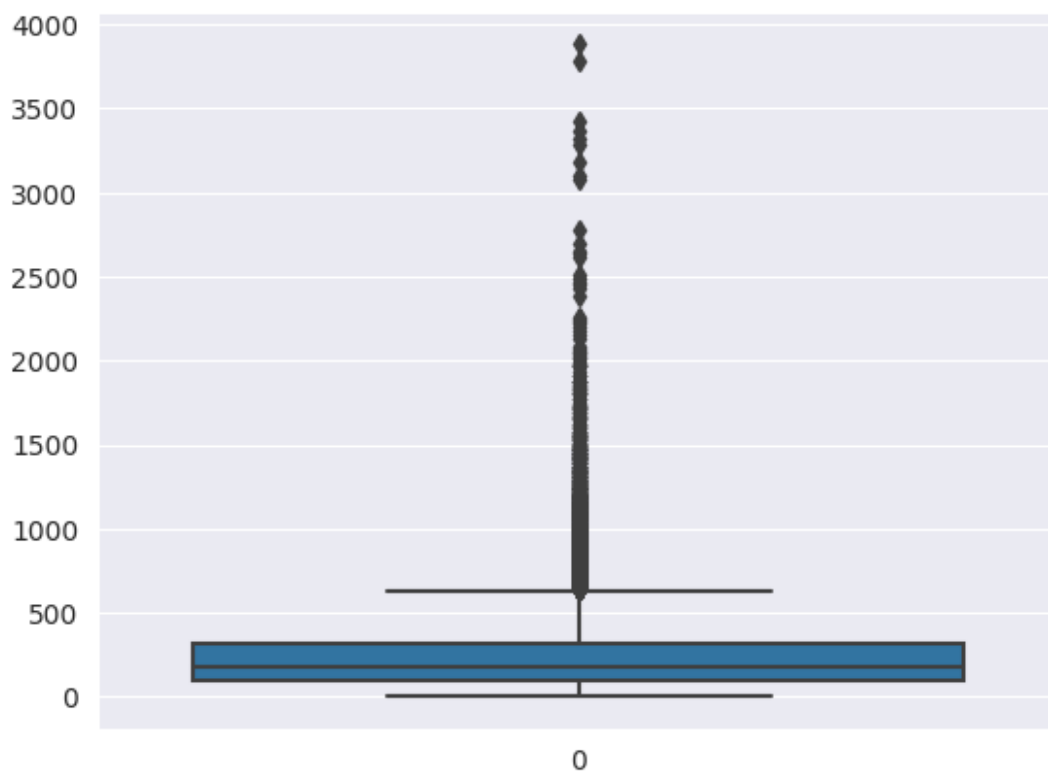
```
sns.boxplot(df['day'])
```

```
<Axes: >
```



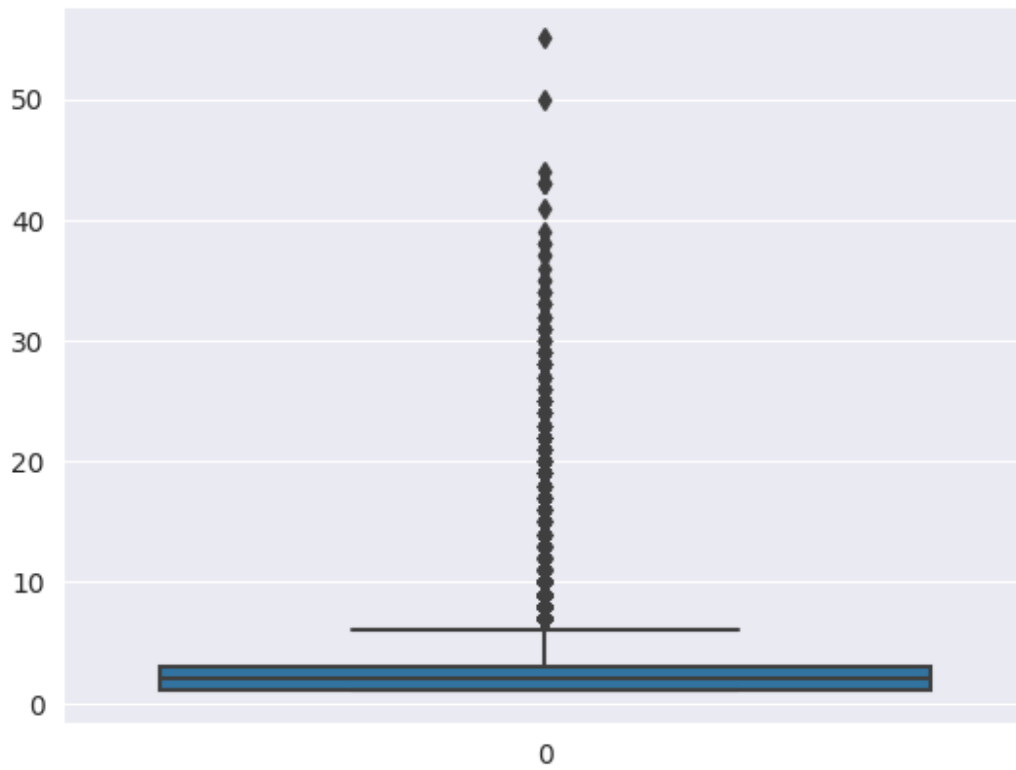
```
sns.boxplot(df['duration'])
```

<Axes: >



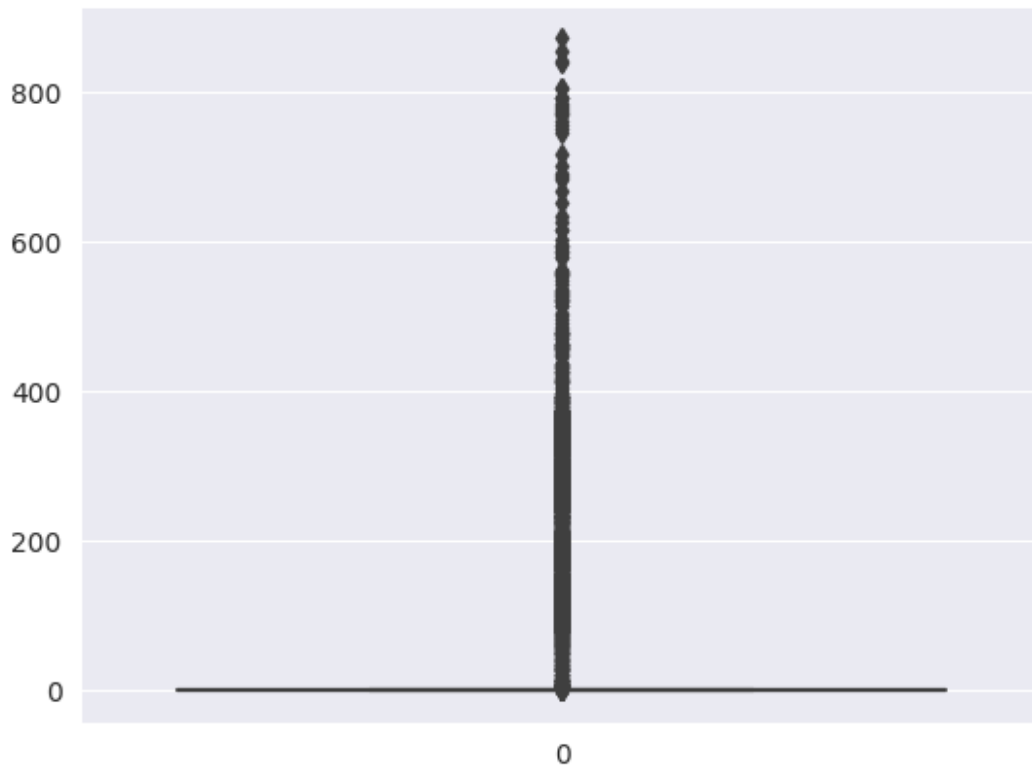
```
sns.boxplot(df['campaign'])
```

<Axes: >



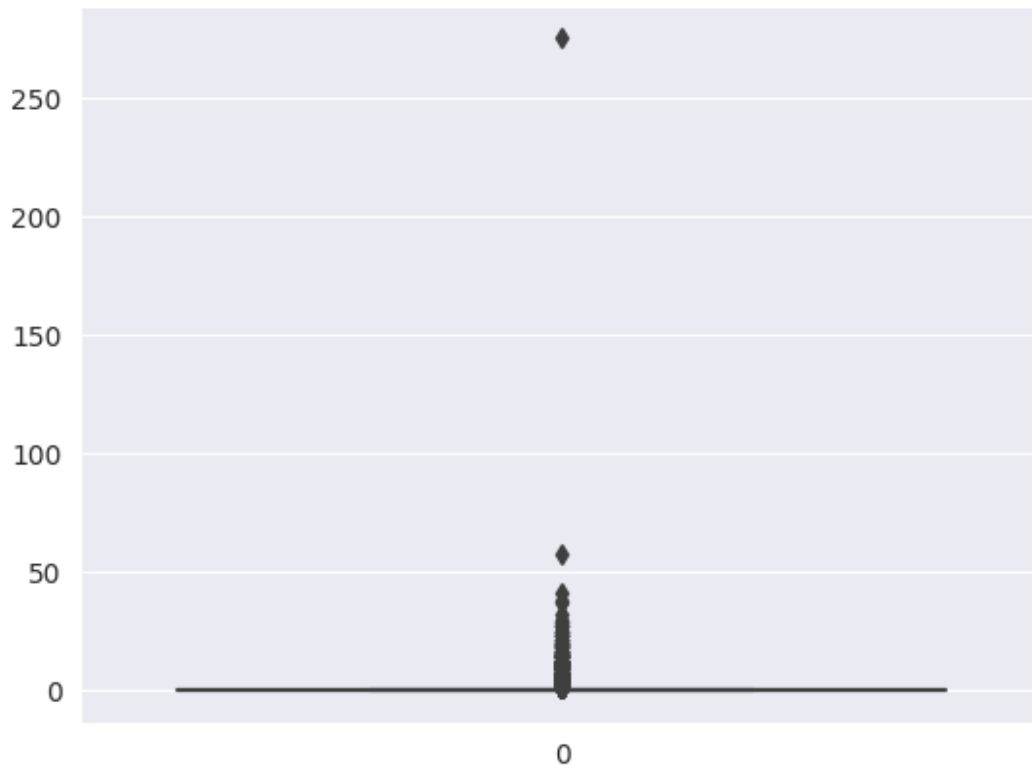
```
sns.boxplot(df['pdays'])
```

<Axes: >



```
sns.boxplot(df['previous'])
```

```
<Axes: >
```



```
df.dtypes
```

```
ID          int64
age          int64
job          object
marital      object
education    object
default      object
balance      int64
housing      object
loan         object
contact      object
day          int64
month        object
duration     int64
campaign     int64
pdays       int64
previous     int64
poutcome     object
subscribed   object
dtype: object
```

```
skewness=df[['ID','age','balance','day','duration','campaign','pdays',
'previous']]
```

```
print(skewness.skew())
```



```
ID          0.022241
age         0.678898
balance     1.249584
day         0.101965
duration    3.145838
campaign    4.765892
pdays      2.644174
previous    52.063050
dtype: float64
```

```
kurtosis=df[['ID','age','balance','day','duration','campaign','pdays',
'previous']]
print(kurtosis.kurt())
```

```
ID          -1.209371
age          0.301654
balance      1.942254
day         -1.087342
duration     17.626100
campaign     34.985834
pdays       7.087155
previous    5496.152091
dtype: float64
```

```
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
```

```
df_encoded = df.apply(le.fit_transform)
```

```
# print the encoded DataFrame
print(df_encoded)
```

	ID	age	job	marital	education	default	balance	housing
loan \								
0	16728	38	0	1	3	0	2719	0
0								
1	25764	13	11	1	1	0	822	0
0								
2	9923	9	7	1	1	0	1709	1
0								
3	27785	39	4	0	2	0	3815	0
0								
4	18961	13	9	1	1	0	938	1
0								
...	...	...	...	...	...	...	...	...
...								
31642	23164	11	4	2	2	0	819	1
0								
31643	25536	35	4	0	2	0	1199	0
1								
31644	12812	14	4	2	2	0	1131	0

```

0
31645 24507 39 9 1 1 0 1044 1
0
31646 9148 37 4 0 1 0 1023 1
0

```

```

      contact  day  month  duration  campaign  pdays  previous
poutcome \
0          1  18     9        43         1      0         0
3
1          0  19     5        90         1      0         0
3
2          0  17     5       239         0      0         0
3
3          0  21     6       860         0     77         3
2
4          0   3     3       379         0      0         0
3
...      ...  ...   ...      ...      ...      ...      ...
...
31642      0  11     8       115         1      0         0
3
31643      0   4     6       437         1      0         0
3
31644      0   6     1        36         2      0         0
3
31645      1  14     8        21         6    330        12
0
31646      0  10     5      1371         1      0         0
3

```

```

      subscribed
0          0
1          0
2          0
3          1
4          0
...      ...
31642      0
31643      1
31644      0
31645      0
31646      1

```

[28556 rows x 18 columns]

df.dtypes

```

ID          int64
age         int64

```

job	object
marital	object
education	object
default	object
balance	int64
housing	object
loan	object
contact	object
day	int64
month	object
duration	int64
campaign	int64
pdays	int64
previous	int64
poutcome	object
subscribed	object
dtype:	object

```
cat_columns =
['job', 'marital', 'education', 'default', 'housing', 'loan', 'contact', 'mon
th', 'poutcome', 'subscribed']
```

```
# Apply label encoding to categorical columns
```

```
for col in cat_columns:
    le = LabelEncoder()
    df[col] = le.fit_transform(df[col])
```

```
# Print the encoded dataframe
```

```
print(df)
```

loan	ID	age	job	marital	education	default	balance	housing
000203040	26110	56	0	1	3	0	1933	0
00010	40576	31	11	1	1	0	3	0
00020	15320	27	7	1	1	0	891	1
00030	43962	57	4	0	2	0	3287	0
00040	29842	31	9	1	1	0	119	1
...	...	...	...	...	...	...	...	...
00010	36483	29	4	2	2	0	0	1
00011	40178	53	4	0	2	0	380	0
00010	19710	32	4	2	2	0	312	0

```

31645  38556   57   9       1       1       0       225       1
0
31646  14156   55   4       0       1       0       204       1
0

```

```

      contact  day  month  duration  campaign  pdays  previous
poutcome \
0          1   19     9        44         2    -1         0
3
1          0   20     5        91         2    -1         0
3
2          0   18     5       240         1    -1         0
3
3          0   22     6       867         1    84         3
2
4          0    4     3       380         1    -1         0
3
...          ...   ...   ...       ...       ...    ...       ...
...
31642          0   12     8       116         2    -1         0
3
31643          0    5     6       438         2    -1         0
3
31644          0    7     1        37         3    -1         0
3
31645          1   15     8        22         7   337        12
0
31646          0   11     5      1973         2    -1         0
3

```

```

      subscribed
0          0
1          0
2          0
3          1
4          0
...
31642          0
31643          1
31644          0
31645          0
31646          1

```

```
[28556 rows x 18 columns]
```

```

# Import libraries
from sklearn.model_selection import train_test_split

```

```

X = df.drop('subscribed', axis=1)
y = df['subscribed']

```

```

# Split data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.2, random_state=42)

# Print the shape of the training and testing sets
print("Training set shape:", X_train.shape, y_train.shape)
print("Testing set shape:", X_test.shape, y_test.shape)

Training set shape: (22844, 17) (22844,)
Testing set shape: (5712, 17) (5712,)

from sklearn.ensemble import RandomForestClassifier
import matplotlib.pyplot as plt

rf_model = RandomForestClassifier(n_estimators=100, random_state=42)
rf_model.fit(X, y)

importances = rf_model.feature_importances_

sorted_indices = importances.argsort()[::-1]
sorted_importances = importances[sorted_indices]

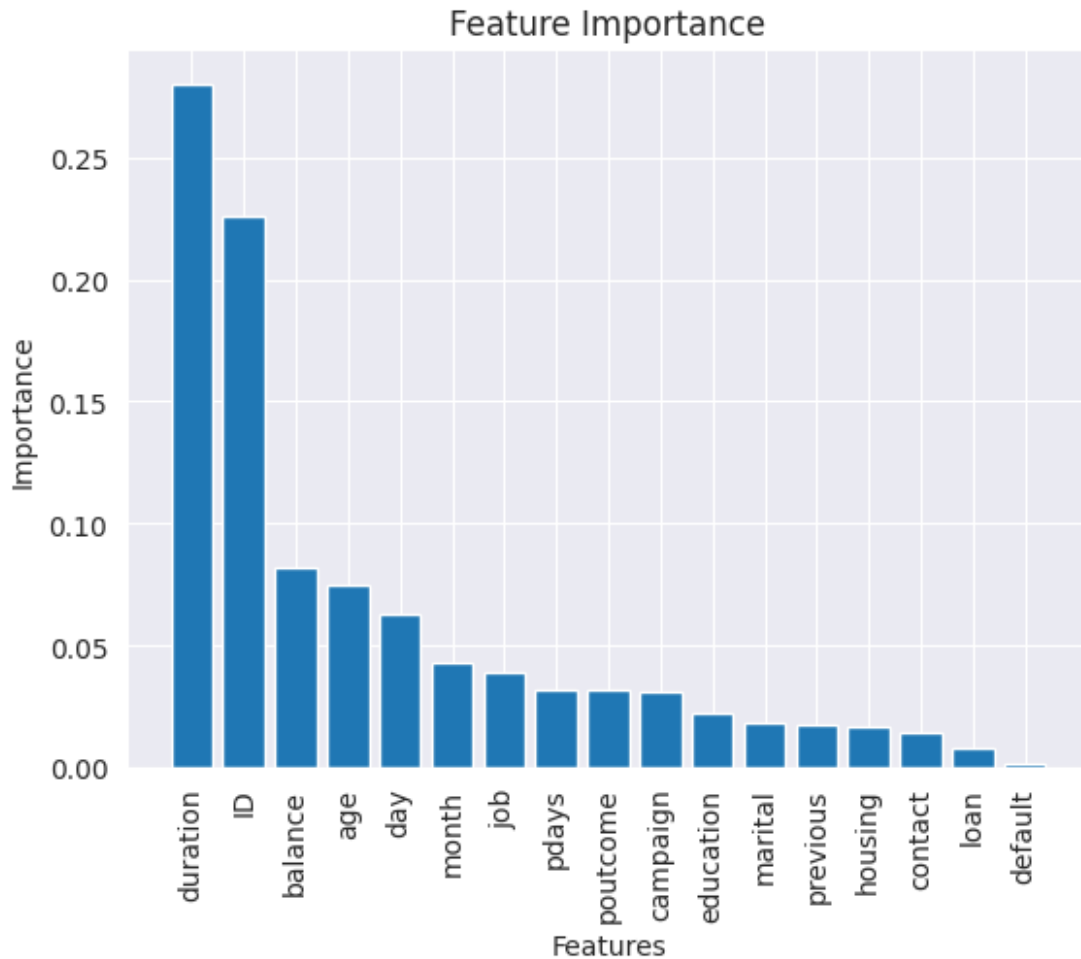
print("Feature ranking:")
for i in range(X.shape[1]):
    print("%d. %s (%f)" % (i + 1, X.columns[sorted_indices[i]],
sorted_importances[i]))

plt.bar(range(X.shape[1]), sorted_importances)
plt.xticks(range(X.shape[1]), X.columns[sorted_indices], rotation=90)
plt.xlabel('Features')
plt.ylabel('Importance')
plt.title('Feature Importance')
plt.show()

Feature ranking:
1. duration (0.279997)
2. ID (0.225587)
3. balance (0.082023)
4. age (0.074897)
5. day (0.062483)
6. month (0.042858)
7. job (0.038827)
8. pdays (0.031894)

```

```
9. poutcome (0.031562)
10. campaign (0.031325)
11. education (0.022226)
12. marital (0.018302)
13. previous (0.017872)
14. housing (0.016435)
15. contact (0.014297)
16. loan (0.007569)
17. default (0.001847)
```



```
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score
```

```
X=df.drop(['default','loan','contact','subscribed'] , axis=1)
y=df["subscribed"]
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.2, random_state=42)
```

```
clf = RandomForestClassifier(n_estimators=100, random_state=42)
clf.fit(X_train, y_train)
```

```
y_pred = clf.predict(X_test)
```

```
# Compute the accuracy score of the classifier
```

```
accuracy = accuracy_score(y_test, y_pred)
```

```
print("Accuracy: {:.2f}%".format(accuracy*100))
```

```
Accuracy: 90.91%
```

```
from sklearn.model_selection import train_test_split
```

```
from sklearn.naive_bayes import GaussianNB
```

```
from sklearn.metrics import accuracy_score
```

```
X=df.drop(['default', 'loan', 'contact', 'subscribed'] , axis=1)
```

```
y=df["subscribed"]
```

```
# Split the data into training and test sets
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.2, random_state=42)
```

```
# Train a Gaussian Naive Bayes classifier
```

```
gnb = GaussianNB()
```

```
gnb.fit(X_train, y_train)
```

```
# Make predictions on the test set
```

```
y_pred = gnb.predict(X_test)
```

```
# Calculate the accuracy score
```

```
acc = accuracy_score(y_test, y_pred)
```

```
print("Accuracy score:", acc)
```

```
Accuracy score: 0.8419117647058824
```