

```
import numpy as np
import pandas as pd
```

```
iris = 'https://gist.githubusercontent.com/curran/a08a1080b88344b0c8a7/raw/639388c2cbc2120a14dcf466e85730eb8be498bb/iris.csv'
df_iris = pd.read_csv(iris,sep=',')
print(type(df_iris))
```

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

```
spotify = '/content/spotify_top_songs_audio_features.csv'
df_spotify = pd.read_csv(spotify,sep=',')
print(type(df_spotify))
```

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

```
s=pd.Series([1,3,5,6,8])
print(type(s))
s
```

```
<class 'pandas.core.series.Series'>
```

```
0    1
1    3
2    5
3    6
4    8
dtype: int64
```

```
d=pd.DataFrame({'col':[1,2,3,4,5,6], 'col2':[1,2,3,4,5,6], 'col3':['1','2','3','4','5',None]})
print(d)
```

	col	col2	col3
0	1	1	1
1	2	2	2
2	3	3	3
3	4	4	4
4	5	5	5
5	6	6	None

```
d.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6 entries, 0 to 5
Data columns (total 3 columns):
#   Column  Non-Null Count  Dtype
---  ---
0    col      6 non-null       int64
1   col2      6 non-null       int64
2   col3      5 non-null       object
dtypes: int64(2), object(1)
memory usage: 272.0+ bytes
```

```
df_iris.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 5 columns):
#   Column          Non-Null Count  Dtype
---  ---
0   sepal_length    150 non-null    float64
1   sepal_width     150 non-null    float64
2   petal_length    150 non-null    float64
3   petal_width     150 non-null    float64
4   species         150 non-null    object
dtypes: float64(4), object(1)
memory usage: 6.0+ KB
```

```
df_iris.head(10)
```

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa
5	5.4	3.9	1.7	0.4	setosa
6	4.6	3.4	1.4	0.3	setosa
7	5.0	3.4	1.5	0.2	setosa
8	4.4	2.9	1.4	0.2	setosa
9	4.9	3.1	1.5	0.1	setosa

```
df_spotify.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6513 entries, 0 to 6512
Data columns (total 19 columns):
#   Column                Non-Null Count  Dtype
---  -
0   id                    6513 non-null  object
1   artist_names          6513 non-null  object
2   track_name            6513 non-null  object
3   source                6513 non-null  object
4   key                   6513 non-null  object
5   mode                  6513 non-null  object
6   time_signature        6513 non-null  object
7   danceability           6513 non-null  float64
8   energy                6513 non-null  float64
9   speechiness           6513 non-null  float64
10  acousticness           6513 non-null  float64
11  instrumentalness       6513 non-null  float64
12  liveness               6513 non-null  float64
13  valence                6513 non-null  float64
14  loudness               6513 non-null  float64
15  tempo                  6513 non-null  float64
16  duration_ms            6513 non-null  int64
17  weeks_on_chart         6513 non-null  int64
18  streams                6513 non-null  int64
dtypes: float64(9), int64(3), object(7)
memory usage: 966.9+ KB
```

```
df_spotify.head(10)
```

	id	artist_names	track_name	source	key	mode	time_signature	danceability	energy	sp
0	000xQL6tZNLJzIrtlgxqSl	ZAYN, PARTYNEXTDOOR	Still Got Time (feat. PARTYNEXTDOOR)	RCA Records Label	G	Major	4 beats	0.748	0.627	
1	003eolwxETJujVWmNFMoZy	Alessia Cara	Growing Pains	Def Jam Recordings	C#/Db	Minor	4 beats	0.353	0.755	
2	003vvx7Niy0yvvhvHt4a68B	The Killers	Mr. Brightside	Island Records	C#/Db	Major	4 beats	0.352	0.911	
3	00B7TZ0Xawar6NZ00JFomN	Cardi B, Chance the Rapper	Best Life (feat. Chance The Rapper)	Atlantic/KSR	A	Major	4 beats	0.620	0.625	
4	00Blm7zeNqgYLPtW6zg8cj	Post Malone, The Weeknd	One Right Now (with The Weeknd)	Republic Records	C#/Db	Major	4 beats	0.687	0.781	
5	00EPIEnX1JFjff8sC6bccd	Thalia, NATTI NATASHA	No Me Acuerdo	Sony Music Latin	G	Minor	4 beats	0.836	0.799	
6	00ETaeHUQ6lops3oWU1Wrt	Kygo, Donna Summer	Hot Stuff	RCA Records Label	F	Major	4 beats	0.681	0.773	
7	00ZKeP47bZtswtANKvxz2j	Tropa do Bruxo, DJ Ws da Igrejinha, SMU, Triz,...	Baile do Bruxo	Tropa Do Bruxo	G	Minor	5 beats	0.734	0.228	
8	00gpGR84M27moP7AFuqHlx	YBN Nahmir	Bounce Out With That	2018	G#/Ab	Major	4 beats	0.857	0.560	
9	00imgaPIYRrMGn9o83hfmk	Brent Faiyaz	LOOSE CHANGE	Lost Kids LLC., Marketed by Venice / Stem	C#/Db	Minor	4 beats	0.574	0.369	

```
df_iris.tail()
```

	sepal_length	sepal_width	petal_length	petal_width	species
145	6.7	3.0	5.2	2.3	virginica
146	6.3	2.5	5.0	1.9	virginica
147	6.5	3.0	5.2	2.0	virginica
148	6.2	3.4	5.4	2.3	virginica
149	5.9	3.0	5.1	1.8	virginica

```
df_spotify.tail()
```

	id	artist_names	track_name	source	key	mode	time_signature	danceability	
6508	7zgqtptZvhf8GEmdsM2vp2	Taylor Swift	...Ready For It?	Big Machine Records, LLC	D	Major	4 beats	0.615	
6509	7zjEyeBsaw9gV0jofJLfOM	Young Thug, A\$AP Rocky, Post Malone	Livin It Up (with Post Malone & A\$AP Rocky)	300 Entertainment/Atl	G	Major	4 beats	0.767	
6510	7zl7kehxesNEo2pYkKXTSe	Eminem, Jack Harlow, Cordae	Killer (feat. Jack Harlow & Cordae) – Remix	Shady/Aftermath/Interscope Records	B	Minor	4 beats	0.924	
6511	7zvfDihYiJ8RQ1nRcpKBF5	Kendrick Lamar, Tanna Leone	Mr. Morale	pgLang/Top Dawg Entertainment/Aftermath/Inters...	A	Major	3 beats	0.727	
6512	7zxRMhXxJMQCeDDg0rKAVo	NAV, The Weeknd	Some Way	XO Records	C	Major	4 beats	0.744	

```
df_iris.describe()
```

	sepal_length	sepal_width	petal_length	petal_width
count	150.000000	150.000000	150.000000	150.000000
mean	5.843333	3.054000	3.758667	1.198667
std	0.828066	0.433594	1.764420	0.763161
min	4.300000	2.000000	1.000000	0.100000
25%	5.100000	2.800000	1.600000	0.300000
50%	5.800000	3.000000	4.350000	1.300000
75%	6.400000	3.300000	5.100000	1.800000
max	7.900000	4.400000	6.900000	2.500000

```
df_spotify.describe()
```

	danceability	energy	speechiness	acousticness	instrumentalness	liveness	valence	loudness	tempo
count	6513.000000	6513.000000	6513.000000	6513.000000	6513.000000	6513.000000	6513.000000	6513.000000	6513.000000
mean	0.681731	0.636522	0.121933	0.236761	0.012469	0.180168	0.492412	-6.350667	122.117244
std	0.141787	0.164813	0.113441	0.244784	0.075151	0.138054	0.227001	2.536114	29.416097
min	0.150000	0.021800	0.023200	0.000008	0.000000	0.019700	0.032000	-34.475000	46.718000
25%	0.591000	0.534000	0.044000	0.044400	0.000000	0.097400	0.316000	-7.564000	98.007000
50%	0.698000	0.651000	0.072200	0.145000	0.000000	0.124000	0.489000	-5.983000	120.034000
75%	0.785000	0.759000	0.163000	0.356000	0.000041	0.219000	0.669000	-4.673000	142.025000
max	0.985000	0.989000	0.966000	0.994000	0.953000	0.977000	0.982000	1.509000	212.117000

```
df_iris.dtypes
```

```
sepal_length    float64
sepal_width     float64
petal_length    float64
petal_width     float64
species         object
dtype: object
```

```
df_spotify.dtypes
```

```
id              object
artist_names    object
track_name      object
source          object
key             object
mode            object
time_signature  object
danceability    float64
energy          float64
speechiness     float64
acousticness    float64
instrumentalness float64
liveness        float64
valence         float64
loudness        float64
tempo           float64
duration_ms     int64
weeks_on_chart  int64
streams         int64
dtype: object
```

```
print(df_iris.index)
```

```
RangeIndex(start=0, stop=150, step=1)
```

```
print(df_spotify.index)
```

```
RangeIndex(start=0, stop=6513, step=1)
```

```
df_iris.columns
```

```
Index(['sepal_length', 'sepal_width', 'petal_length', 'petal_width',
       'species'],
      dtype='object')
```

```
df_spotify.columns
```

```
Index(['id', 'artist_names', 'track_name', 'source', 'key', 'mode',
       'time_signature', 'danceability', 'energy', 'speechiness',
       'acousticness', 'instrumentalness', 'liveness', 'valence', 'loudness',
       'tempo', 'duration_ms', 'weeks_on_chart', 'streams'],
      dtype='object')
```

```
attributes = ['sepal_length','sepal_width','petal_length','petal_width','additional']
df_iris.columns = attributes
df_iris.head(1)#重新命名
```

```
sepal_length sepal_width petal_length petal_width additional
0           5.1         3.5         1.4         0.2       setosa
```

```
concat#用于左右合并
duplicated(keep=False去重)
```

```
import seaborn as sns #处理outlier的时候需要使用seaborn函数
dt_outlier = np.concatenate([np.random.randn(1000),np.random.normal(7,1,10)])
sns.set_style('whitegrid')
sns.distplot(dt_outlier)
```

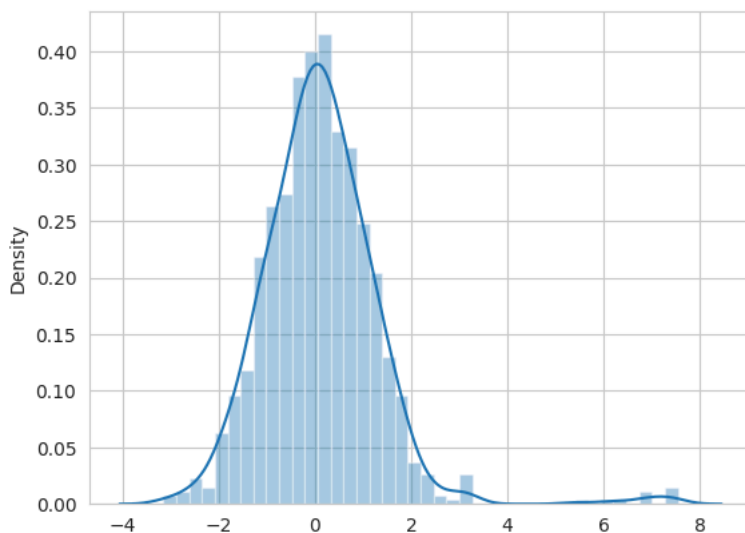
```
<ipython-input-35-aabf72eeb118>:4: UserWarning:
```

```
`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).
```

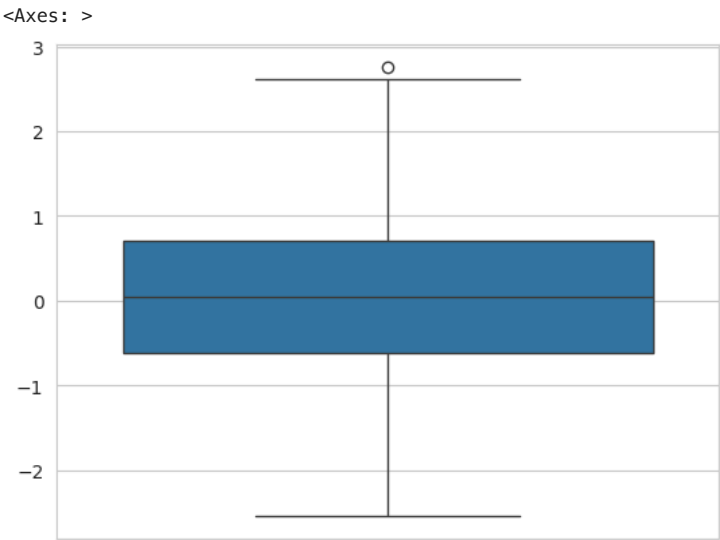
```
For a guide to updating your code to use the new functions, please see
https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751
```

```
sns.distplot(dt_outlier)
<Axes: ylabel='Density'>
```

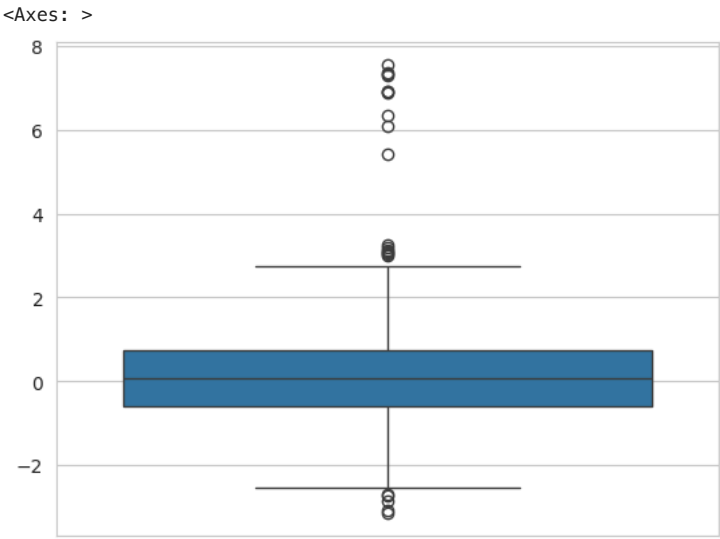


```
def iqr_outlier_rm(dt_input):
    lq,uq = np.percentile(dt_input,[25,75])
    lower_l = lq - 1.5*(uq-lq)
    upper_l = uq + 1.5*(uq-lq)
    return dt_input[(dt_input >= lower_l)&(dt_input <= upper_l)]
```

```
dt_outlier_ws = iqr_outlier_rm(dt_outlier)
sns.boxplot(dt_outlier_ws,orient='v')
```



```
sns.boxplot(dt_outlier,orient = 'v')
```



```
raw_data = {'name': ['Jason', np.nan, 'Mike', 'Rayman', 'Alex', 'Meimei'],
            'age': [36, np.nan, 36, 18, 36, 16],
            'gender': ['m', np.nan, 'm', np.nan, 'f', 'f'],
            'preMLScore': [1, np.nan, np.nan, 2, 3, 90],
            'postMLScore': [65, np.nan, np.nan, 62, 70, 100]}

# create a dataframe by passing a dictionary
df = pd.DataFrame(raw_data, columns = ['name', 'age', 'gender', 'preMLScore', 'postMLScore'])
```

df

	name	age	gender	preMLScore	postMLScore
0	Jason	36.0	m	1.0	65.0
1	NaN	NaN	NaN	NaN	NaN
2	Mike	36.0	m	NaN	NaN
3	Rayman	18.0	NaN	2.0	62.0
4	Alex	36.0	f	3.0	70.0
5	Meimei	16.0	f	90.0	100.0

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6 entries, 0 to 5
Data columns (total 5 columns):
#   Column      Non-Null Count  Dtype
---  ---
0   name        5 non-null     object
1   age         5 non-null     float64
2   gender      4 non-null     object
3   preMLScore  4 non-null     float64
4   postMLScore 4 non-null     float64
dtypes: float64(3), object(2)
memory usage: 368.0+ bytes
```

```
df.isnull()
```

	name	age	gender	preMLScore	postMLScore
0	False	False	False	False	False
1	True	True	True	True	True
2	False	False	False	True	True
3	False	False	True	False	False
4	False	False	False	False	False
5	False	False	False	False	False

```
df.isnull().sum()
```

```
name      1
age       1
gender    2
preMLScore 2
postMLScore 2
dtype: int64
```

```
df.isnull().any(axis = 0)
```

```
name      True
age       True
gender    True
preMLScore True
postMLScore True
dtype: bool
```

```
df
```

	name	age	gender	preMLScore	postMLScore
0	Jason	36.0	m	1.0	65.0
1	NaN	NaN	NaN	NaN	NaN
2	Mike	36.0	m	NaN	NaN
3	Rayman	18.0	NaN	2.0	62.0
4	Alex	36.0	f	3.0	70.0
5	Meimei	16.0	f	90.0	100.0

```
df.dropna(axis = 0,how = 'any')#去除NA缺失值
```

	name	age	gender	preMLScore	postMLScore
0	Jason	36.0	m	1.0	65.0
4	Alex	36.0	f	3.0	70.0
5	Meimei	16.0	f	90.0	100.0

```
df.dropna(axis =1,how = 'any')
```

```
0
1
2
3
4
5

df=df.dropna(how = 'all',inplace = False)
df

name age gender preMLScore postMLScore
```