

```
# 1. Ładowanie biblioteki Pandas
```

```
import pandas as pd
```

```
# 2. Tworzenie ramki danych ze słownika
```

```
dane_sownik = {  
    "Kraj": ["Polska", "Niemcy", "Francja"],  
    "Ludność (mln)": [38, 83, 67],  
    "PKB (mld)": [600, 4000, 3000]  
}  
ramka_danych_sownik = pd.DataFrame(dane_sownik)
```

```
print("Ramka danych utworzona ze słownika:")
```

```
print(ramka_danych_sownik)
```

Ramka danych utworzona ze słownika:

	Kraj	Ludność (mln)	PKB (mld)
0	Polska	38	600
1	Niemcy	83	4000
2	Francja	67	3000

```
# 3. Zachowanie ramki danych pobranych z pliku w formacie .csv
```

```
plik_csv = "IHME_GDP_1960_2050_Y2021M09D22.CSV"
```

```
ramka_danych_csv = pd.read_csv(plik_csv)
```

```
print("\nRamka danych wczytana z pliku CSV:")
```

```
print(ramka_danych_csv.head())
```

Ramka danych wczytana z pliku CSV:

	location_id	location_name	iso3	level	year	gdp_ppp_mean
gdp_ppp_lower \						
0	1	Global	G	Global	1960	1.748345e+13
1.601915e+13						
1	1	Global	G	Global	1961	1.813537e+13
1.659537e+13						
2	1	Global	G	Global	1962	1.895328e+13
1.739039e+13						
3	1	Global	G	Global	1963	1.965662e+13
1.811706e+13						
4	1	Global	G	Global	1964	2.100575e+13
1.935664e+13						

	gdp_ppp_upper	gdp_usd_mean	gdp_usd_lower	gdp_usd_upper
0	1.911586e+13	1.296863e+13	1.266890e+13	1.334177e+13
1	1.982493e+13	1.346097e+13	1.314767e+13	1.383021e+13
2	2.061477e+13	1.406576e+13	1.376060e+13	1.443746e+13
3	2.134993e+13	1.461831e+13	1.432132e+13	1.497693e+13
4	2.276791e+13	1.552986e+13	1.523498e+13	1.587998e+13

```
# 4. Tworzenie ramki danych z listy list
```

```
dane_lista = [  
    ["Polska", 38, 600],
```

```

    ["Niemcy", 83, 4000],
    ["Francja", 67, 3000]
]
ramka_danych_lista = pd.DataFrame(dane_lista, columns=["Kraj",
"Ludność (mln)", "PKB (mld)"])

print("\nRamka danych utworzona z listy list:")
print(ramka_danych_lista)

```

Ramka danych utworzona z listy list:

	Kraj	Ludność (mln)	PKB (mld)
0	Polska	38	600
1	Niemcy	83	4000
2	Francja	67	3000

5. Transponowanie (wymiana kolumny a wierszy)

```

ramka_danych_sownik_transponowany = ramka_danych_sownik.T
print(ramka_danych_sownik_transponowany)

```

	0	1	2
Kraj	Polska	Niemcy	Francja
Ludność (mln)	38	83	67
PKB (mld)	600	4000	3000

6. Wyświetlić pierwsze 10 wierszy ramki danych

```

print(ramka_danych_csv.head(10))

```

	location_id	location_name	iso3	level	year	gdp_ppp_mean
gdp_ppp_lower \						
0	1	Global	G	Global	1960	1.748345e+13
1.601915e+13						
1	1	Global	G	Global	1961	1.813537e+13
1.659537e+13						
2	1	Global	G	Global	1962	1.895328e+13
1.739039e+13						
3	1	Global	G	Global	1963	1.965662e+13
1.811706e+13						
4	1	Global	G	Global	1964	2.100575e+13
1.935664e+13						
5	1	Global	G	Global	1965	2.202459e+13
2.034585e+13						
6	1	Global	G	Global	1966	2.306193e+13
2.136085e+13						
7	1	Global	G	Global	1967	2.391268e+13
2.217842e+13						
8	1	Global	G	Global	1968	2.516723e+13
2.340479e+13						
9	1	Global	G	Global	1969	2.642403e+13
2.464521e+13						

	gdp_ppp_upper	gdp_usd_mean	gdp_usd_lower	gdp_usd_upper
0	1.911586e+13	1.296863e+13	1.266890e+13	1.334177e+13
1	1.982493e+13	1.346097e+13	1.314767e+13	1.383021e+13
2	2.061477e+13	1.406576e+13	1.376060e+13	1.443746e+13
3	2.134993e+13	1.461831e+13	1.432132e+13	1.497693e+13
4	2.276791e+13	1.552986e+13	1.523498e+13	1.587998e+13
5	2.382275e+13	1.628972e+13	1.598727e+13	1.663310e+13
6	2.489782e+13	1.708885e+13	1.678223e+13	1.742396e+13
7	2.577837e+13	1.770884e+13	1.740660e+13	1.804193e+13
8	2.698215e+13	1.865379e+13	1.833216e+13	1.898399e+13
9	2.831984e+13	1.955395e+13	1.921164e+13	1.987990e+13

```
# 7. Wyświetlić ostatnie 10 wierszy ramki danych
print(ramka_danych_csv.tail(10))
```

year \	location_id	location_name	iso3	level
19828	44578	Low income	NaN	World Bank Income Group 2041
19829	44578	Low income	NaN	World Bank Income Group 2042
19830	44578	Low income	NaN	World Bank Income Group 2043
19831	44578	Low income	NaN	World Bank Income Group 2044
19832	44578	Low income	NaN	World Bank Income Group 2045
19833	44578	Low income	NaN	World Bank Income Group 2046
19834	44578	Low income	NaN	World Bank Income Group 2047
19835	44578	Low income	NaN	World Bank Income Group 2048
19836	44578	Low income	NaN	World Bank Income Group 2049
19837	44578	Low income	NaN	World Bank Income Group 2050

	gdp_ppp_mean	gdp_ppp_lower	gdp_ppp_upper	gdp_usd_mean \
19828	3.120963e+12	2.724077e+12	3.582807e+12	9.752426e+11
19829	3.216988e+12	2.801335e+12	3.686394e+12	1.008813e+12
19830	3.314031e+12	2.886768e+12	3.815672e+12	1.042881e+12
19831	3.413020e+12	2.968361e+12	3.933135e+12	1.077714e+12
19832	3.514244e+12	3.055623e+12	4.049325e+12	1.113207e+12
19833	3.617310e+12	3.140835e+12	4.166469e+12	1.149318e+12
19834	3.724063e+12	3.225849e+12	4.292403e+12	1.186597e+12
19835	3.831942e+12	3.307609e+12	4.424674e+12	1.224062e+12
19836	3.941856e+12	3.398884e+12	4.560961e+12	1.262129e+12
19837	4.053883e+12	3.482933e+12	4.713596e+12	1.300764e+12

gdp_usd_lower	gdp_usd_upper
---------------	---------------

19828	8.875033e+11	1.068693e+12
19829	9.169149e+11	1.107239e+12
19830	9.461940e+11	1.147550e+12
19831	9.735487e+11	1.188093e+12
19832	1.003241e+12	1.228145e+12
19833	1.031500e+12	1.271992e+12
19834	1.061313e+12	1.318836e+12
19835	1.092874e+12	1.365610e+12
19836	1.122895e+12	1.413991e+12
19837	1.151548e+12	1.457362e+12

8. Wyświetlić informacje o ramce danych

```
print(ramka_danych_csv.info())
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 19838 entries, 0 to 19837
Data columns (total 11 columns):
#   Column                Non-Null Count  Dtype
---  -
0   location_id            19838 non-null  int64
1   location_name          19838 non-null  object
2   iso3                   18655 non-null  object
3   level                  19838 non-null  object
4   year                   19838 non-null  int64
5   gdp_ppp_mean           19838 non-null  float64
6   gdp_ppp_lower          19838 non-null  float64
7   gdp_ppp_upper          19838 non-null  float64
8   gdp_usd_mean           19838 non-null  float64
9   gdp_usd_lower          19838 non-null  float64
10  gdp_usd_upper          19838 non-null  float64
dtypes: float64(6), int64(2), object(3)
memory usage: 1.7+ MB
None
```

9. Wyświetlić ile wierszy i kolumn znajduje się w ramce danych

```
print(ramka_danych_csv.shape)
```

```
(19838, 11)
```

10. Wyświetlić informacje statystyczną o kolumnach liczbowych (wartości niepowtarzalne, średnia, odchylenie standardowe, minimum, kwartyle, maksimum)

```
print(ramka_danych_csv.describe())
```

	location_id	year	gdp_ppp_mean	gdp_ppp_lower
count	19838.000000	19838.000000	1.983800e+04	1.983800e+04
mean	949.871560	2005.000000	1.334543e+12	1.235788e+12
std	5965.433243	26.268513	9.148287e+12	8.610030e+12

9.789327e+12				
min	1.000000	1960.000000	1.448063e+02	6.299026e+01
2.621094e+02				
25%	63.000000	1982.000000	3.678736e+03	2.639116e+03
4.829886e+03				
50%	125.500000	2005.000000	1.103640e+04	8.105541e+03
1.346178e+04				
75%	183.000000	2028.000000	2.949281e+04	2.308992e+04
3.562660e+04				
max	44578.000000	2050.000000	1.827414e+14	1.667007e+14
2.025062e+14				

	gdp_usd_mean	gdp_usd_lower	gdp_usd_upper
count	1.983800e+04	1.983800e+04	1.983800e+04
mean	8.554096e+11	8.197528e+11	8.967612e+11
std	6.286364e+12	6.041288e+12	6.585419e+12
min	1.174979e+02	8.318772e+01	1.270468e+02
25%	1.624411e+03	1.395430e+03	1.828575e+03
50%	4.863298e+03	4.279291e+03	5.465731e+03
75%	1.997525e+04	1.795003e+04	2.223434e+04
max	1.119468e+14	1.017185e+14	1.239708e+14

11. Wyświetlić informacje statyczną o kolumnach kategoryzowanych (ile unikalnych wartości, top - jaka jest najpopularniejsza wartość, freq - jak często najpopularniejsza)

```
print(ramka_danych_csv.describe(include=['object']))
```

	location_name	iso3	level
count	19838	18655	19838
unique	216	205	4
top	South Asia	G	Country
freq	182	91	18564

12. Usunąć brakujące wartości w ramce danych

```
ramka_danych_csv = ramka_danych_csv.dropna()
```

13. Przedstawić wybór wierszy i kolumny używając nazw oraz indeksów na różne sposoby

```
print(ramka_danych_csv.loc[0:5, ['location_name', 'gdp_ppp_mean']]) # Nazwy kolumn
print(ramka_danych_csv.iloc[0:5, 1:3]) # Indeksy kolumn
```

	location_name	gdp_ppp_mean
0	Global	1.748345e+13
1	Global	1.813537e+13
2	Global	1.895328e+13
3	Global	1.965662e+13
4	Global	2.100575e+13
5	Global	2.202459e+13
	location_name	iso3

```
0      Global      G
1      Global      G
2      Global      G
3      Global      G
4      Global      G
```

14. Przedstawić wybór wierszy z ramki danych pod warunkiem określonej wartości kolumny

```
print(ramka_danych_csv[ramka_danych_csv['location_name'] == 'Poland'])
```

	location_id	location_name	iso3	level	year	gdp_ppp_mean	\
3913	51	Poland	POL	Country	1960	6477.852541	
3914	51	Poland	POL	Country	1961	6854.160388	
3915	51	Poland	POL	Country	1962	6696.268564	
3916	51	Poland	POL	Country	1963	6981.908383	
3917	51	Poland	POL	Country	1964	7192.051449	
...	
3999	51	Poland	POL	Country	2046	38341.602497	
4000	51	Poland	POL	Country	2047	38361.930630	
4001	51	Poland	POL	Country	2048	38342.127372	
4002	51	Poland	POL	Country	2049	38291.223003	
4003	51	Poland	POL	Country	2050	38239.561147	

	gdp_ppp_lower	gdp_ppp_upper	gdp_usd_mean	gdp_usd_lower	gdp_usd_upper
3913	3974.353115	8867.341510	3340.866044	3013.073267	3767.183362
3914	4209.334194	9266.617081	3528.521854	3181.314168	3955.470420
3915	4137.676353	9056.936779	3447.774153	3104.590497	3867.083569
3916	4381.353870	9390.144726	3589.554120	3242.049335	4016.402456
3917	4537.064677	9677.839801	3693.222182	3344.989514	4126.100385
...
...
3999	28096.663520	51449.445517	17392.703365	12862.490537	23265.109294
4000	27796.090370	52186.079877	17401.902620	12665.251631	23581.749549
4001	27466.932992	52654.694631	17393.027956	12498.169215	23783.922827
4002	27105.252343	53056.023696	17369.859621	12305.183049	24051.240960
4003	26728.543779	53156.930801	17346.349035	12139.994778	24206.609198

```
[91 rows x 11 columns]
```

```
# 15. Przedstawić wybór wierszy z ramki danych pod warunkiem
spełnienia kilku warunków jednocześnie
print(ramka_danych_csv[(ramka_danych_csv['location_name'] == 'Poland')
& (ramka_danych_csv['year'] > 2024)])
```

	location_id	location_name	iso3	level	year	gdp_ppp_mean	\
3978	51	Poland	POL	Country	2025	32746.999437	
3979	51	Poland	POL	Country	2026	32914.035363	
3980	51	Poland	POL	Country	2027	33150.861860	
3981	51	Poland	POL	Country	2028	33466.534617	
3982	51	Poland	POL	Country	2029	33845.943995	
3983	51	Poland	POL	Country	2030	34262.156546	
3984	51	Poland	POL	Country	2031	34684.507200	
3985	51	Poland	POL	Country	2032	35094.834819	
3986	51	Poland	POL	Country	2033	35489.219953	
3987	51	Poland	POL	Country	2034	35879.588365	
3988	51	Poland	POL	Country	2035	36263.317214	
3989	51	Poland	POL	Country	2036	36623.408529	
3990	51	Poland	POL	Country	2037	36958.569309	
3991	51	Poland	POL	Country	2038	37248.918811	
3992	51	Poland	POL	Country	2039	37497.939670	
3993	51	Poland	POL	Country	2040	37703.604057	
3994	51	Poland	POL	Country	2041	37879.536633	
3995	51	Poland	POL	Country	2042	38038.938282	
3996	51	Poland	POL	Country	2043	38167.733724	
3997	51	Poland	POL	Country	2044	38255.709502	
3998	51	Poland	POL	Country	2045	38310.991457	
3999	51	Poland	POL	Country	2046	38341.602497	
4000	51	Poland	POL	Country	2047	38361.930630	
4001	51	Poland	POL	Country	2048	38342.127372	
4002	51	Poland	POL	Country	2049	38291.223003	
4003	51	Poland	POL	Country	2050	38239.561147	

	gdp_ppp_lower	gdp_ppp_upper	gdp_usd_mean	gdp_usd_lower
gdp_usd_upper				
3978	30092.195731	35503.313225	14854.849644	14056.398773
	15663.528879			
3979	30046.680048	35995.808629	14930.623781	13960.885206
	15925.104161			
3980	30041.903293	36610.693734	15038.027521	13912.362504
	16230.319850			
3981	30109.057428	37248.220741	15181.254112	13888.157933
	16582.456688			
3982	30090.325852	38041.964433	15353.331066	13889.111499
	16968.669700			
3983	30186.653655	38853.442510	15542.161674	13915.431093
	17385.081872			
3984	30209.883184	39742.317576	15733.699224	13858.453725
	17760.333322			
3985	30242.053023	40568.105501	15919.873491	13830.697742

18222.190232				
3986	30241.783438	41443.235244	16098.718258	13797.457684
18617.895687				
3987	30266.595933	42462.257986	16275.785255	13745.584028
19099.749770				
3988	30205.053229	43340.288613	16449.871159	13742.795346
19506.562473				
3989	30174.544960	44185.732988	16613.238344	13656.622696
19947.251805				
3990	30087.477029	45077.191565	16765.299346	13687.881303
20306.946127				
3991	30023.264809	45975.391029	16897.055787	13648.635010
20730.760082				
3992	29889.878587	46915.521299	17010.031047	13616.621123
21116.041100				
3993	29784.460066	47441.689278	17103.321801	13514.969487
21467.346446				
3994	29534.567189	48331.668321	17183.191423	13383.267219
21827.744208				
3995	29205.795211	49230.732911	17255.501259	13298.015373
22242.787940				
3996	28876.433219	49729.333605	17313.833385	13152.190918
22513.534843				
3997	28612.932167	50457.779813	17353.713535	13095.302179
22773.395829				
3998	28210.802867	50856.938961	17378.830577	12966.356828
22982.886727				
3999	28096.663520	51449.445517	17392.703365	12862.490537
23265.109294				
4000	27796.090370	52186.079877	17401.902620	12665.251631
23581.749549				
4001	27466.932992	52654.694631	17393.027956	12498.169215
23783.922827				
4002	27105.252343	53056.023696	17369.859621	12305.183049
24051.240960				
4003	26728.543779	53156.930801	17346.349035	12139.994778
24206.609198				

16. Wybrać wiersze które zawierają w kolumnie skategoryzowanej określone słowo

```
print(ramka_danych_csv[ramka_danych_csv['location_name'].str.contains('United', na=False)])
```

	location_id	location_name	iso3	level	
year \					
7553	95	United Kingdom	GBR	Country	1960
7554	95	United Kingdom	GBR	Country	1961
7555	95	United Kingdom	GBR	Country	1962

7556	95	United Kingdom	GBR	Country	1963
7557	95	United Kingdom	GBR	Country	1964
...
19287	422	United States Virgin Islands	VIR	Country	2046
19288	422	United States Virgin Islands	VIR	Country	2047
19289	422	United States Virgin Islands	VIR	Country	2048
19290	422	United States Virgin Islands	VIR	Country	2049
19291	422	United States Virgin Islands	VIR	Country	2050

	gdp_ppp_mean	gdp_ppp_lower	gdp_ppp_upper	gdp_usd_mean	\
7553	16310.982717	13739.359462	18211.080013	14468.928557	
7554	16637.378306	14021.255691	18552.357712	14738.960806	
7555	16699.652950	14116.793663	18618.780049	14794.176887	
7556	17204.352407	14568.220451	19153.680713	15283.800698	
7557	17964.368546	15412.086655	19860.371971	15932.807887	
...	
19287	41287.177057	30556.895863	55066.904811	41234.500216	
19288	41596.380683	30505.069557	56047.487106	41543.041145	
19289	41923.883728	30508.427873	56639.931939	41869.856709	
19290	42269.179899	30601.194839	57629.728086	42214.585846	
19291	42600.038587	30535.154812	58469.665785	42545.067618	

	gdp_usd_lower	gdp_usd_upper
7553	13698.701744	15172.657255
7554	13978.985853	15434.739721
7555	14036.102767	15483.073848
7556	14647.369479	15921.312162
7557	15383.515617	16539.728393
...
19287	31638.150835	53139.473363
19288	31569.471257	53287.294060
19289	31488.955147	54291.408201
19290	31839.581250	54988.746887
19291	31823.556963	55743.351459

[455 rows x 11 columns]

```
# 17. Wybrać wiersze które nie zawierają w kolumnie skategoryzowanej
określonego słowa
print(ramka_danych_csv[~ramka_danych_csv['location_name'].str.contains
('United', na=False)])
```

	location_id	location_name	iso3	level	year	gdp_ppp_mean	\
0	1	Global	G	Global	1960	1.748345e+13	
1	1	Global	G	Global	1961	1.813537e+13	
2	1	Global	G	Global	1962	1.895328e+13	
3	1	Global	G	Global	1963	1.965662e+13	
4	1	Global	G	Global	1964	2.100575e+13	
...	
19469	522	Sudan	SDN	Country	2046	6.656899e+03	
19470	522	Sudan	SDN	Country	2047	6.729027e+03	
19471	522	Sudan	SDN	Country	2048	6.796123e+03	
19472	522	Sudan	SDN	Country	2049	6.866343e+03	
19473	522	Sudan	SDN	Country	2050	6.935555e+03	

	gdp_ppp_lower	gdp_ppp_upper	gdp_usd_mean	gdp_usd_lower	\
0	1.601915e+13	1.911586e+13	1.296863e+13	1.266890e+13	
1	1.659537e+13	1.982493e+13	1.346097e+13	1.314767e+13	
2	1.739039e+13	2.061477e+13	1.406576e+13	1.376060e+13	
3	1.811706e+13	2.134993e+13	1.461831e+13	1.432132e+13	
4	1.935664e+13	2.276791e+13	1.552986e+13	1.523498e+13	
...	
19469	3.356042e+03	1.155051e+04	1.459547e+03	9.801683e+02	
19470	3.374504e+03	1.171206e+04	1.475378e+03	9.886902e+02	
19471	3.398699e+03	1.184386e+04	1.490021e+03	9.935248e+02	
19472	3.417444e+03	1.196204e+04	1.505368e+03	1.002889e+03	
19473	3.429198e+03	1.208179e+04	1.520564e+03	1.002953e+03	

	gdp_usd_upper
0	1.334177e+13
1	1.383021e+13
2	1.443746e+13
3	1.497693e+13
4	1.587998e+13
...	...
19469	2.269566e+03
19470	2.286933e+03
19471	2.322390e+03
19472	2.362591e+03
19473	2.408108e+03

[18200 rows x 11 columns]

```
# 18. Utwórz kolumnę na podstawie istniejącej
ramka_danych_csv['gdp_ppp_diff'] = ramka_danych_csv['gdp_ppp_upper'] -
ramka_danych_csv['gdp_ppp_lower'] # Utworzenie kolumny z różnicą
między gdp_ppp_upper i gdp_ppp_lower
print(ramka_danych_csv['gdp_ppp_diff'].head(10))
```

0	3.096716e+12
1	3.229556e+12
2	3.224381e+12

```

3      3.232877e+12
4      3.411270e+12
5      3.476908e+12
6      3.536968e+12
7      3.599954e+12
8      3.577357e+12
9      3.674624e+12
Name: gdp_ppp_diff, dtype: float64

# 19. Usuń kolumnę
# ramka_danych_csv = ramka_danych_csv.drop(columns=['gdp_ppp_diff']) #
Nie usuwamy kolumny, bo operujemy na niej niżej!

# 20. Zmień nazwę kolumny
ramka_danych_csv = ramka_danych_csv.rename(columns={'gdp_ppp_diff':
'gdp_ppp_difference'})

# 21. Zachowaj ramkę danych jako plik csv na komputerze
ramka_danych_csv.to_csv('updated_data.csv', index=False)

# 22. Wyświetlić liczbę wierszy
print(len(ramka_danych_csv))

18655

# 23. Wyświetlić wartości unikatowe w kolumnie
print(ramka_danych_csv['location_name'].unique())

['Global' 'China' "Democratic People's Republic of Korea"
'Taiwan (Province of China)' 'Cambodia' 'Indonesia'
'Lao People's Democratic Republic' 'Malaysia' 'Maldives' 'Myanmar'
'Philippines' 'Sri Lanka' 'Thailand' 'Timor-Leste' 'Viet Nam' 'Fiji'
'Kiribati' 'Marshall Islands' 'Micronesia (Federated States of)'
'Papua New Guinea' 'Samoa' 'Solomon Islands' 'Tonga' 'Vanuatu'
'Armenia'
'Azerbaijan' 'Georgia' 'Kazakhstan' 'Kyrgyzstan' 'Mongolia'
'Tajikistan'
'Turkmenistan' 'Uzbekistan' 'Albania' 'Bosnia and Herzegovina'
'Bulgaria'
'Croatia' 'Czechia' 'Hungary' 'North Macedonia' 'Montenegro' 'Poland'
'Romania' 'Serbia' 'Slovakia' 'Slovenia' 'Belarus' 'Estonia' 'Latvia'
'Lithuania' 'Republic of Moldova' 'Russian Federation' 'Ukraine'
'Brunei Darussalam' 'Japan' 'Republic of Korea' 'Singapore'
'Australia'
'New Zealand' 'Andorra' 'Austria' 'Belgium' 'Cyprus' 'Denmark'
'Finland'
'France' 'Germany' 'Greece' 'Iceland' 'Ireland' 'Israel' 'Italy'
'Luxembourg' 'Malta' 'Netherlands' 'Norway' 'Portugal' 'Spain'
'Sweden'
'Switzerland' 'United Kingdom' 'Argentina' 'Chile' 'Uruguay' 'Canada'
'United States of America' 'Antigua and Barbuda' 'Bahamas' 'Barbados'

```

```
'Belize' 'Cuba' 'Dominica' 'Dominican Republic' 'Grenada' 'Guyana'
'Haiti' 'Jamaica' 'Saint Lucia' 'Saint Vincent and the Grenadines'
'Suriname' 'Trinidad and Tobago' 'Bolivia (Plurinational State of)'
'Ecuador' 'Peru' 'Colombia' 'Costa Rica' 'El Salvador' 'Guatemala'
'Honduras' 'Mexico' 'Nicaragua' 'Panama'
'Venezuela (Bolivarian Republic of)' 'Brazil' 'Paraguay' 'Algeria'
'Bahrain' 'Egypt' 'Iran (Islamic Republic of)' 'Iraq' 'Jordan'
'Kuwait'
'Lebanon' 'Libya' 'Morocco' 'Palestine' 'Oman' 'Qatar' 'Saudi Arabia'
'Syrian Arab Republic' 'Tunisia' 'Turkey' 'United Arab Emirates'
'Yemen'
'Afghanistan' 'Bangladesh' 'Bhutan' 'India' 'Nepal' 'Pakistan'
'Angola'
'Central African Republic' 'Congo' 'Democratic Republic of the Congo'
'Equatorial Guinea' 'Gabon' 'Burundi' 'Comoros' 'Djibouti' 'Eritrea'
'Ethiopia' 'Kenya' 'Madagascar' 'Malawi' 'Mauritius' 'Mozambique'
'Rwanda' 'Seychelles' 'Somalia' 'United Republic of Tanzania'
'Uganda'
'Zambia' 'Botswana' 'Lesotho' 'Namibia' 'South Africa' 'Eswatini'
'Zimbabwe' 'Benin' 'Burkina Faso' 'Cameroon' 'Cabo Verde' 'Chad'
'Côte d'Ivoire' 'Gambia' 'Ghana' 'Guinea' 'Guinea-Bissau' 'Liberia'
'Mali' 'Mauritania' 'Niger' 'Nigeria' 'Sao Tome and Principe'
'Senegal'
'Sierra Leone' 'Togo' 'American Samoa' 'Bermuda' 'Cook Islands'
'Greenland' 'Guam' 'Monaco' 'Nauru' 'Niue' 'Northern Mariana Islands'
'Palau' 'Puerto Rico' 'Saint Kitts and Nevis' 'San Marino' 'Tokelau'
'Tuvalu' 'United States Virgin Islands' 'South Sudan' 'Sudan']
```

24. Wyświetlić liczby rekordów odpowiadających do wartości

```
print(ramka_danych_csv['location_name'].value_counts())
```

```
location_name
Global                91
China                 91
Democratic People's Republic of Korea  91
Taiwan (Province of China)  91
Cambodia              91
..
Tokelau               91
Tuvalu                91
United States Virgin Islands  91
South Sudan           91
Sudan                 91
Name: count, Length: 205, dtype: int64
```

25. Sortowanie wierszy ramki danych według wartości określonej kolumny (malejąco, rosnąco)

```
ramka_danych_csv = ramka_danych_csv.sort_values(by='location_name',
ascending=False)
```

```
# 26. Wyświetlić wierszy dla 10 największych (najmniejszych) wartości
określonej kolumny
print(ramka_danych_csv.nlargest(10, 'gdp_ppp_difference'))
print(ramka_danych_csv.nsmallest(10, 'gdp_ppp_difference'))
```

	location_id	location_name	iso3	level	year	gdp_ppp_mean
gdp_ppp_lower \						
90	1	Global	G	Global	2050	1.827414e+14
1.667007e+14						
89	1	Global	G	Global	2049	1.811701e+14
1.657675e+14						
88	1	Global	G	Global	2048	1.795422e+14
1.647031e+14						
87	1	Global	G	Global	2047	1.778053e+14
1.635681e+14						
86	1	Global	G	Global	2046	1.759560e+14
1.622744e+14						
85	1	Global	G	Global	2045	1.740498e+14
1.608327e+14						
84	1	Global	G	Global	2044	1.720934e+14
1.594056e+14						
83	1	Global	G	Global	2043	1.701152e+14
1.579438e+14						
82	1	Global	G	Global	2042	1.681175e+14
1.566207e+14						
81	1	Global	G	Global	2041	1.661209e+14
1.552230e+14						

	gdp_ppp_upper	gdp_usd_mean	gdp_usd_lower	gdp_usd_upper \
90	2.025062e+14	1.119468e+14	1.017185e+14	1.239708e+14
89	2.003282e+14	1.110748e+14	1.012670e+14	1.226294e+14
88	1.978349e+14	1.101656e+14	1.008704e+14	1.212579e+14
87	1.952850e+14	1.091923e+14	1.003097e+14	1.197614e+14
86	1.928964e+14	1.081513e+14	9.968511e+13	1.180625e+14
85	1.903320e+14	1.070717e+14	9.903290e+13	1.164315e+14
84	1.874514e+14	1.059643e+14	9.831993e+13	1.147651e+14
83	1.847172e+14	1.048522e+14	9.771637e+13	1.129875e+14
82	1.817886e+14	1.037319e+14	9.697921e+13	1.115238e+14
81	1.792966e+14	1.026157e+14	9.630379e+13	1.098151e+14

	gdp_ppp_difference
90	3.580549e+13
89	3.456061e+13
88	3.313177e+13
87	3.171689e+13
86	3.062192e+13
85	2.949930e+13
84	2.804583e+13
83	2.677341e+13
82	2.516788e+13

```

81          2.407368e+13
location_id location_name iso3 level year gdp_ppp_mean \
17791      218          Togo TGO Country 2006 1256.106656
15701      194        Lesotho LS0 Country 2009 2386.813789
15891      196  South Africa ZAF Country 2017 13415.109285
17790      218          Togo TGO Country 2005 1251.135277
17794      218          Togo TGO Country 2009 1269.161805
15066      185          Rwanda RWA Country 2011 1626.420947
14974      184    Mozambique MOZ Country 2010 1060.720788
15067      185          Rwanda RWA Country 2012 1717.968641
17795      218          Togo TGO Country 2010 1308.887516
17798      218          Togo TGO Country 2013 1441.917427

```

```

gdp_ppp_lower gdp_ppp_upper gdp_usd_mean gdp_usd_lower \
17791 1239.451885 1272.811018 525.427266 524.712203
15701 2364.955982 2408.208729 762.378717 729.814377
15891 13392.134704 13438.793149 5331.268573 5327.630613
17790 1228.017491 1278.123473 525.655949 524.957683
17794 1235.694959 1303.017996 525.753280 525.026947
15066 1592.229066 1663.254347 562.630729 558.618670
14974 1029.574562 1100.995605 382.037004 370.488013
15067 1678.204824 1753.122750 595.431231 591.130483
17795 1276.145386 1351.422620 542.979145 542.113686
17798 1412.922361 1488.496965 634.980682 594.900379

```

```

gdp_usd_upper gdp_ppp_difference
17791 525.969435 33.359132
15701 789.724274 43.252747
15891 5334.127410 46.658445
17790 526.226948 50.105982
17794 526.332654 67.323037
15066 568.725657 71.025280
14974 390.499028 71.421043
15067 600.995223 74.917925
17795 543.708334 75.277235
17798 706.144537 75.574603

```

```

# 27. Wyświetlić wierszy dla 10 największych wartości określonej
kolumny pod warunkiem określonych wartości innej kolumny
filtered_data = ramka_danych_csv[ramka_danych_csv['location_name'] ==
'Global']
print(filtered_data.nlargest(10, 'gdp_ppp_mean'))

```

```

location_id location_name iso3 level year gdp_ppp_mean
gdp_ppp_lower \
90      1      Global      G Global 2050 1.827414e+14
1.667007e+14
89      1      Global      G Global 2049 1.811701e+14
1.657675e+14
88      1      Global      G Global 2048 1.795422e+14

```

1.647031e+14						
87	1	Global	G	Global	2047	1.778053e+14
1.635681e+14						
86	1	Global	G	Global	2046	1.759560e+14
1.622744e+14						
85	1	Global	G	Global	2045	1.740498e+14
1.608327e+14						
84	1	Global	G	Global	2044	1.720934e+14
1.594056e+14						
83	1	Global	G	Global	2043	1.701152e+14
1.579438e+14						
82	1	Global	G	Global	2042	1.681175e+14
1.566207e+14						
81	1	Global	G	Global	2041	1.661209e+14
1.552230e+14						

	gdp_ppp_upper	gdp_usd_mean	gdp_usd_lower	gdp_usd_upper	\
90	2.025062e+14	1.119468e+14	1.017185e+14	1.239708e+14	
89	2.003282e+14	1.110748e+14	1.012670e+14	1.226294e+14	
88	1.978349e+14	1.101656e+14	1.008704e+14	1.212579e+14	
87	1.952850e+14	1.091923e+14	1.003097e+14	1.197614e+14	
86	1.928964e+14	1.081513e+14	9.968511e+13	1.180625e+14	
85	1.903320e+14	1.070717e+14	9.903290e+13	1.164315e+14	
84	1.874514e+14	1.059643e+14	9.831993e+13	1.147651e+14	
83	1.847172e+14	1.048522e+14	9.771637e+13	1.129875e+14	
82	1.817886e+14	1.037319e+14	9.697921e+13	1.115238e+14	
81	1.792966e+14	1.026157e+14	9.630379e+13	1.098151e+14	

	gdp_ppp_difference
90	3.580549e+13
89	3.456061e+13
88	3.313177e+13
87	3.171689e+13
86	3.062192e+13
85	2.949930e+13
84	2.804583e+13
83	2.677341e+13
82	2.516788e+13
81	2.407368e+13

```
# 28. Grupowanie wierszy według wartości kolumny skategoryzowanej,
potem uśrednienie wartości wszystkich kolumn w grupie - MultiIndex
grouped_mean = ramka_danych_csv.groupby(['location_name',
'year']).mean(numeric_only=True)
print(grouped_mean)
```

	location_id	gdp_ppp_mean	gdp_ppp_lower
gdp_ppp_upper	\		
location_name	year		

Afghanistan	1960	160.0	2221.335586	1353.292858
3082.415995	1961	160.0	2192.653614	1336.349002
3012.241102	1962	160.0	2178.869688	1327.326777
2963.810432	1963	160.0	2169.572283	1322.003248
2940.315793	1964	160.0	2155.861769	1324.331042
2932.818437
...
Zimbabwe	2046	198.0	3086.223490	1856.652268
4531.850929	2047	198.0	3116.294363	1866.152043
4625.577546	2048	198.0	3145.941464	1857.642354
4693.442459	2049	198.0	3175.312716	1861.027391
4766.207444	2050	198.0	3204.717710	1852.077740
4849.920213				
		gdp_usd_mean	gdp_usd_lower	gdp_usd_upper \
location_name	year			
Afghanistan	1960	690.992776	516.513170	964.233914
	1961	682.493782	507.833855	958.141591
	1962	678.584622	510.383482	950.032579
	1963	676.055466	509.234425	943.496288
	1964	672.157496	506.548073	932.652043
...
Zimbabwe	2046	1283.050542	990.146349	1666.231011
	2047	1295.530479	992.038533	1691.789642
	2048	1307.813498	982.320381	1719.039709
	2049	1319.994426	986.284349	1749.976131
	2050	1332.173722	981.546944	1767.718479
		gdp_ppp_difference		
location_name	year			
Afghanistan	1960	1729.123137		
	1961	1675.892100		
	1962	1636.483654		
	1963	1618.312546		
	1964	1608.487395		
...		
Zimbabwe	2046	2675.198661		
	2047	2759.425503		
	2048	2835.800105		
	2049	2905.180053		

2050 2997.842472

[18655 rows x 8 columns]

29. Grupowanie wierszy według wartości kolumny skategoryzowanej, potem uśrednienie wartości dla pewnych kolumn, liczba wartości i mediana dla pozostałych kolumn w grupach

```
grouped_stats = ramka_danych_csv.groupby('location_name').agg({  
    'gdp_ppp_mean': ['mean', 'count'],  
    'gdp_usd_mean': 'median',  
    'year': 'max'  
})
```

```
print(grouped_stats)
```

	gdp_ppp_mean		gdp_usd_mean
year			
	mean	count	median
max			
location_name			
Afghanistan	1941.160286	91	515.274036
2050			
Albania	9092.515182	91	3098.516205
2050			
Algeria	8820.271149	91	3163.885729
2050			
American Samoa	15340.365197	91	13620.772462
2050			
Andorra	25139.562251	91	38178.372791
2050			
...
...			
Venezuela (Bolivarian Republic of)	10594.142490	91	5823.785745
2050			
Viet Nam	5737.873614	91	1437.919432
2050			
Yemen	2637.237249	91	828.806903
2050			
Zambia	3107.029470	91	1078.009951
2050			
Zimbabwe	2925.918096	91	1069.856772
2050			

[205 rows x 4 columns]

30. Wyświetlić nazwy kolumn indeksu złożonego

```
print(grouped_mean.index.names)
```

['location_name', 'year']

31. Posortować kolumnę indeksu złożonego

```
sorted_index = grouped_mean.sort_index()
```

```
print(sorted_index)
```

		location_id	gdp_ppp_mean	gdp_ppp_lower
gdp_ppp_upper \	location_name year			
Afghanistan	1960	160.0	2221.335586	1353.292858
3082.415995	1961	160.0	2192.653614	1336.349002
3012.241102	1962	160.0	2178.869688	1327.326777
2963.810432	1963	160.0	2169.572283	1322.003248
2940.315793	1964	160.0	2155.861769	1324.331042
2932.818437				

...	
...				
Zimbabwe	2046	198.0	3086.223490	1856.652268
4531.850929	2047	198.0	3116.294363	1866.152043
4625.577546	2048	198.0	3145.941464	1857.642354
4693.442459	2049	198.0	3175.312716	1861.027391
4766.207444	2050	198.0	3204.717710	1852.077740
4849.920213				

		gdp_usd_mean	gdp_usd_lower	gdp_usd_upper \
location_name	year			
Afghanistan	1960	690.992776	516.513170	964.233914
	1961	682.493782	507.833855	958.141591
	1962	678.584622	510.383482	950.032579
	1963	676.055466	509.234425	943.496288
	1964	672.157496	506.548073	932.652043
...	
Zimbabwe	2046	1283.050542	990.146349	1666.231011
	2047	1295.530479	992.038533	1691.789642
	2048	1307.813498	982.320381	1719.039709
	2049	1319.994426	986.284349	1749.976131
	2050	1332.173722	981.546944	1767.718479

		gdp_ppp_difference
location_name	year	
Afghanistan	1960	1729.123137
	1961	1675.892100
	1962	1636.483654

	1963	1618.312546
	1964	1608.487395
...		...
Zimbabwe	2046	2675.198661
	2047	2759.425503
	2048	2835.800105
	2049	2905.180053
	2050	2997.842472

[18655 rows x 8 columns]

32. Stworzyć tabelę przestawną (pivot table) na podstawie ramki danych

```

pivot_table = pd.pivot_table(ramka_danych_csv, values='gdp_ppp_mean',
index=['location_name'], columns=['year'], aggfunc='mean')
print(pivot_table)

```

year	1960	1961
1962 \		
location_name		
Afghanistan	2221.335586	2192.653614
2178.869688		
Albania	3158.241995	3179.711325
3272.046583		
Algeria	5006.438360	4453.400566
3658.175321		
American Samoa	22109.859273	21946.482938
21884.955783		
Andorra	15636.337352	15796.419665
16031.522138		
...
...		
Venezuela (Bolivarian Republic of)	10432.796342	10430.234336
10804.655654		
Viet Nam	1301.479002	1319.501206
1427.159417		
Yemen	1207.599807	1213.637071
1220.327712		
Zambia	2685.864457	2613.932835
2518.024702		
Zimbabwe	2408.747887	2460.565228
2405.779753		

year	1963	1964
1965 \		
location_name		
Afghanistan	2169.572283	2155.861769
2151.064936		

Albania	3370.096538	3471.383451	
3583.125647			
Algeria	4362.291504	4467.247662	
4635.281973			
American Samoa	21746.150959	22039.103093	
21967.292708			
Andorra	16212.717854	16721.236248	
16962.500876			
...	
...			
Venezuela (Bolivarian Republic of)	11116.532476	11810.328016	
11935.494850			
Viet Nam	1421.953513	1440.696855	
1413.406360			
Yemen	1230.605680	1236.624591	
1245.455329			
Zambia	2545.033066	2794.094677	
3042.546710			
Zimbabwe	2377.944795	2361.629168	
2451.921918			
year	1966	1967	
1968 \			
location_name			
Afghanistan	2121.159148	2125.293873	
2141.203832			
Albania	3701.606998	3826.761339	
3947.787571			
Algeria	4234.622602	4531.823753	
4803.430268			
American Samoa	21941.986732	21778.478625	
21809.608764			
Andorra	17243.441075	17419.093114	
17753.186591			
...	
...			
Venezuela (Bolivarian Republic of)	11691.544995	11714.768598	
12148.279419			
Viet Nam	1386.091408	1196.087362	
1147.824135			
Yemen	1257.818357	1271.555576	
1282.432072			
Zambia	3136.130606	3299.617291	
3305.972107			
Zimbabwe	2371.081395	2426.346281	
2334.537805			
year	1969	...	2041 \
location_name		...	

Afghanistan	2120.287941	...	2342.098650
Albania	4065.934215	...	14819.211029
Algeria	5051.997456	...	11030.573193
American Samoa	21868.789321	...	11893.422465
Andorra	18116.808178	...	29227.359814
...
Venezuela (Bolivarian Republic of)	12262.570715	...	5936.593472
Viet Nam	1207.247597	...	11978.148704
Yemen	1296.296950	...	2687.929912
Zambia	3412.255898	...	4528.583497
Zimbabwe	2607.252194	...	2920.094735

year	2042	2043
------	------	------

2044 \

location_name		
---------------	--	--

Afghanistan	2373.513925	2408.692690
2447.457933		
Albania	15023.009454	15229.411365
15463.602305		
Algeria	11103.336795	11165.459200
11210.295359		
American Samoa	11930.178447	11982.538279
12045.045051		
Andorra	28842.858958	28498.492075
28158.292062		

...
...		

Venezuela (Bolivarian Republic of)	5910.458362	5877.138076
5836.249945		
Viet Nam	12143.050046	12305.079272
12460.181246		
Yemen	2721.105116	2752.205471
2782.981501		
Zambia	4596.447849	4663.435135
4728.373411		
Zimbabwe	2953.357891	2987.035398
3019.468770		

year	2045	2046
------	------	------

2047 \

location_name		
---------------	--	--

Afghanistan	2484.684464	2526.771339
2568.918404		
Albania	15675.489103	15891.474827
16093.835400		
Algeria	11241.399380	11262.030649
11271.942839		
American Samoa	12116.603188	12194.337343

```

12275.051475
Andorra                27813.855276  27554.062720
27352.447739
...
...
Venezuela (Bolivarian Republic of)  5796.897698  5752.656452
5711.373515
Viet Nam                12612.891371  12763.816982
12904.398221
Yemen                  2813.127885  2842.694538
2871.818130
Zambia                 4794.522235  4860.353288
4926.102992
Zimbabwe              3052.770960  3086.223490
3116.294363

year                2048                2049
2050
location_name

Afghanistan           2612.639879  2655.415849
2702.860491
Albania              16272.721532  16445.334011
16597.323356
Algeria             11275.241774  11262.595293
11244.318633
American Samoa       12365.951829  12458.781807
12552.708157
Andorra              27197.498599  27089.146046
27037.722215
...
...
Venezuela (Bolivarian Republic of)  5673.687788  5629.163661
5586.185557
Viet Nam            13041.740748  13175.413001
13307.742671
Yemen              2900.489905  2928.794443
2955.558976
Zambia            4989.723480  5050.614585
5113.775149
Zimbabwe          3145.941464  3175.312716
3204.717710

```

```
[205 rows x 91 columns]
```

```

# 33. Wyświetlić indeksy i kolumny tabeli przestawnej
print(pivot_table.index)
print(pivot_table.columns)

```

```

Index(['Afghanistan', 'Albania', 'Algeria', 'American Samoa',
      'Andorra',
      'Angola', 'Antigua and Barbuda', 'Argentina', 'Armenia',
      'Australia',
      ...,
      'United States Virgin Islands', 'United States of America',
      'Uruguay',
      'Uzbekistan', 'Vanuatu', 'Venezuela (Bolivarian Republic of)',
      'Viet Nam', 'Yemen', 'Zambia', 'Zimbabwe'],
      dtype='object', name='location_name', length=205)
Index([1960, 1961, 1962, 1963, 1964, 1965, 1966, 1967, 1968, 1969,
      1970, 1971,
      1972, 1973, 1974, 1975, 1976, 1977, 1978, 1979, 1980, 1981,
      1982, 1983,
      1984, 1985, 1986, 1987, 1988, 1989, 1990, 1991, 1992, 1993,
      1994, 1995,
      1996, 1997, 1998, 1999, 2000, 2001, 2002, 2003, 2004, 2005,
      2006, 2007,
      2008, 2009, 2010, 2011, 2012, 2013, 2014, 2015, 2016, 2017,
      2018, 2019,
      2020, 2021, 2022, 2023, 2024, 2025, 2026, 2027, 2028, 2029,
      2030, 2031,
      2032, 2033, 2034, 2035, 2036, 2037, 2038, 2039, 2040, 2041,
      2042, 2043,
      2044, 2045, 2046, 2047, 2048, 2049, 2050],
      dtype='int64', name='year')

```

34. Utwórz indeks złożony tabeli przestawnej i wyświetl go

```

pivot_table = pivot_table.reset_index()
pivot_table.set_index(['location_name'] +
list(pivot_table.columns[1:]), inplace=True)
print(pivot_table.index)

```

```

MultiIndex([(
      'Afghanistan',
      2221.33558642915, ...),
      (
      'Albania',
      3158.24199462837, ...),
      (
      'Algeria',
      5006.43835997237, ...),
      (
      'American Samoa',
      22109.859272757, ...),
      (
      'Andorra',
      15636.3373517101, ...),
      (
      'Angola',
      4424.22337093641, ...),
      (
      'Antigua and Barbuda',
      3939.96456079785, ...),
      (
      'Argentina',
      9947.85853860087, ...),
      (
      'Armenia',

```

```

4717.49893052611, ...),
    ('Australia',
17521.8499818484, ...),
    ...
    ('United States Virgin Islands',
11402.6973448973, ...),
    ('United States of America',
21498.7249723008, ...),
    ('Uruguay',
8032.29318246791, ...),
    ('Uzbekistan',
2971.75653904325, ...),
    ('Vanuatu',
1761.93293871854, ...),
    ('Venezuela (Bolivarian Republic of)',
10432.796342326, ...),
    ('Viet Nam',
1301.47900194441, ...),
    ('Yemen',
1207.59980696746, ...),
    ('Zambia',
2685.86445730912, ...),
    ('Zimbabwe',
2408.74788699075, ...)],
    names=['location_name', 1960, 1961, 1962, 1963, 1964, 1965,
1966, 1967, 1968, 1969, 1970, 1971, 1972, 1973, 1974, 1975, 1976,
1977, 1978, 1979, 1980, 1981, 1982, 1983, 1984, 1985, 1986, 1987,
1988, 1989, 1990, 1991, 1992, 1993, 1994, 1995, 1996, 1997, 1998,
1999, 2000, 2001, 2002, 2003, 2004, 2005, 2006, 2007, 2008, 2009,
2010, 2011, 2012, 2013, 2014, 2015, 2016, 2017, 2018, 2019, 2020,
2021, 2022, 2023, 2024, 2025, 2026, 2027, 2028, 2029, 2030, 2031,
2032, 2033, 2034, 2035, 2036, 2037, 2038, 2039, 2040, 2041, 2042,
2043, 2044, 2045, 2046, 2047, 2048, 2049, 2050], length=205)

```

35. Zaimportuj moduł pyplot z biblioteki matplotlib i 36. Wskazać, że wykresy należy rysować bezpośrednio w zeszycie, a nie w osobnej zakładce

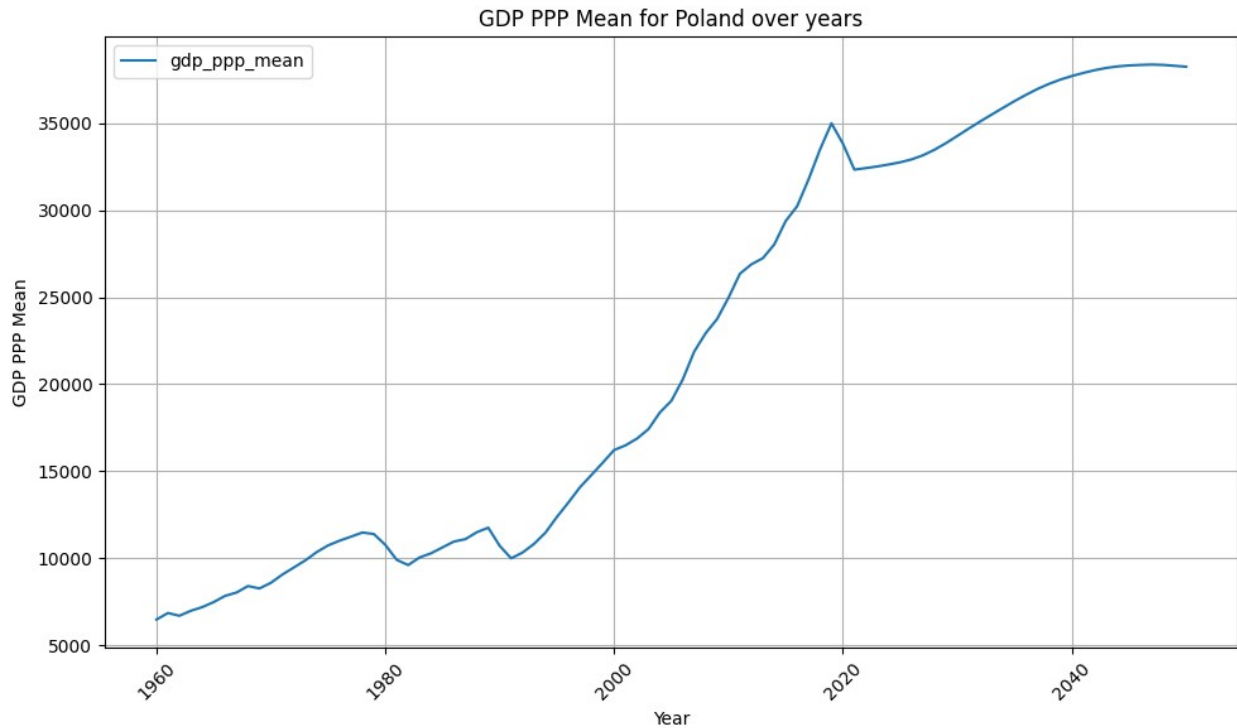
```

import matplotlib.pyplot as plt
%matplotlib inline
# 37. Wyświetlić wykres na podstawie tabeli przestawnej
pivot_table_reset = pivot_table.reset_index()
pivot_table_melted = pd.melt(pivot_table_reset,
id_vars=["location_name"], var_name="year", value_name="gdp_ppp_mean")
# Meltujemy (przetapiamy) dane, aby 'year' stał się kolumną,
location_name = 'Poland'
data = pivot_table_melted[pivot_table_melted['location_name'] ==
location_name]
data.plot(x='year', y='gdp_ppp_mean', kind='line', title=f'GDP PPP
Mean for {location_name} over years', figsize=(10,6))
plt.xlabel('Year')

```



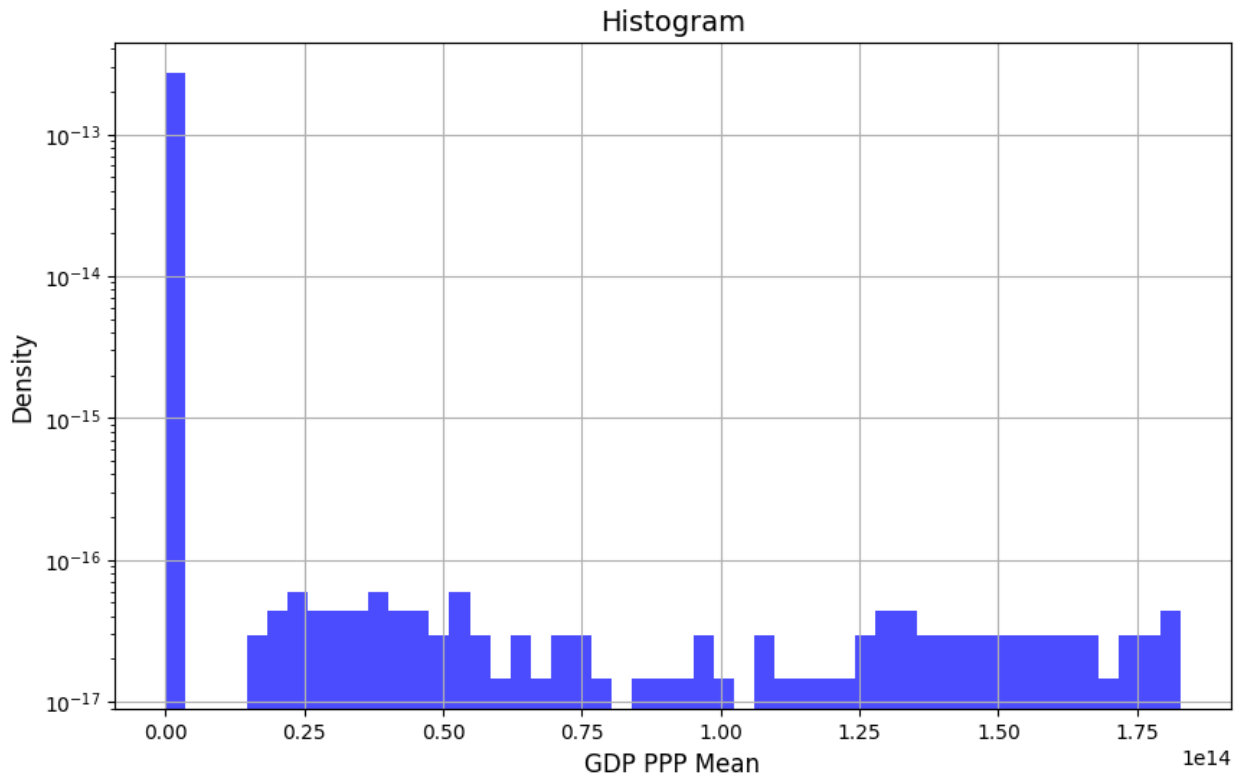
```
plt.ylabel('GDP PPP Mean')
plt.grid(True)
plt.xticks(rotation=45) # Rotacja etykiet osi X dla lepszej
czytelności
plt.tight_layout()
plt.show()
```



```
# 38. Narysować histogram na podstawie wartości kolumny
# Usunięcie wierszy z brakującymi danymi
ramka_danych_csv = ramka_danych_csv.dropna(subset=['gdp_ppp_mean'])

# Narysowanie histogramu z mniejszą liczbą przedziałów i zastosowanie
skali logarytmicznej
plt.figure(figsize=(10,6))
ramka_danych_csv['gdp_ppp_mean'].hist(bins=50, density=True,
alpha=0.7, color='blue')

plt.title('Histogram', fontsize=14)
plt.xlabel('GDP PPP Mean', fontsize=12)
plt.ylabel('Density', fontsize=12)
plt.yscale('log')
plt.show()
```



39. Przedstawić sposoby łączenia ramek danych za pomocą metod merge i concat

Metoda merge:

```
df1 = pd.DataFrame({'key': ['A', 'B', 'C'], 'value1': [1, 2, 3]})
```

```
df2 = pd.DataFrame({'key': ['A', 'B', 'D'], 'value2': [4, 5, 6]})
```

```
merged_df = pd.merge(df1, df2, on='key', how='inner')
```

```
print(merged_df)
```

Metoda konkatencji:

```
df3 = pd.DataFrame({'key': ['A', 'B'], 'value1': [7, 8]})
```

```
df4 = pd.DataFrame({'key': ['C', 'D'], 'value1': [9, 10]})
```

```
concatenated_df = pd.concat([df3, df4], ignore_index=True)
```

```
print(concatenated_df)
```

	key	value1	value2
0	A	1	4
1	B	2	5

	key	value1
0	A	7
1	B	8
2	C	9
3	D	10

40. Pokazać dodawanie nowych kolumn za pomocą operacji matematycznych

```
ramka_danych_csv['gdp_growth'] = ramka_danych_csv['gdp_ppp_mean'] /  
ramka_danych_csv['gdp_ppp_mean'].shift(1) - 1
```

```
print(ramka_danych_csv[['location_name', 'year',
'gdp_growth']].head())
```

	location_name	year	gdp_growth
16017	Zimbabwe	1961	NaN
16084	Zimbabwe	2028	0.033804
16082	Zimbabwe	2026	-0.019321
16081	Zimbabwe	2025	-0.008344
16080	Zimbabwe	2024	-0.007727

41. Przedstawić na przykładzie dodawanie nowych kolumn z pomocą funkcji lambda

```
ramka_danych_csv['gdp_per_capita'] = ramka_danych_csv.apply(lambda
row: row['gdp_ppp_mean'] / row['gdp_usd_mean'], axis=1)
print(ramka_danych_csv[['location_name', 'year',
'gdp_per_capita']].head())
```

	location_name	year	gdp_per_capita
16017	Zimbabwe	1961	3.002179
16084	Zimbabwe	2028	2.404670
16082	Zimbabwe	2026	2.404573
16081	Zimbabwe	2025	2.404522
16080	Zimbabwe	2024	2.404459

42. Przedstawić możliwości pracy z dużymi plikami przy użyciu argumentu chunksize

```
chunksize = 10000
for chunk in pd.read_csv('IHME_GDP_1960_2050_Y2021M09D22.CSV',
chunksize=chunksize):
    print(chunk.head(4))
```

	location_id	location_name	iso3	level	year	gdp_ppp_mean
gdp_ppp_lower \						
0	1	Global	G	Global	1960	1.748345e+13
1.601915e+13						
1	1	Global	G	Global	1961	1.813537e+13
1.659537e+13						
2	1	Global	G	Global	1962	1.895328e+13
1.739039e+13						
3	1	Global	G	Global	1963	1.965662e+13
1.811706e+13						

	gdp_ppp_upper	gdp_usd_mean	gdp_usd_lower	gdp_usd_upper
0	1.911586e+13	1.296863e+13	1.266890e+13	1.334177e+13
1	1.982493e+13	1.346097e+13	1.314767e+13	1.383021e+13
2	2.061477e+13	1.406576e+13	1.376060e+13	1.443746e+13
3	2.134993e+13	1.461831e+13	1.432132e+13	1.497693e+13

	location_id	location_name	iso3	level	year	gdp_ppp_mean
10000	126	Costa Rica	CRI	Country	2041	21222.968139
10001	126	Costa Rica	CRI	Country	2042	21384.771030

10002	126	Costa Rica	CRI	Country	2043	21531.353078
10003	126	Costa Rica	CRI	Country	2044	21654.766957

	gdp_ppp_lower	gdp_ppp_upper	gdp_usd_mean	gdp_usd_lower	\
10000	16351.045982	26688.622084	12731.983537	10072.290068	
10001	16373.548021	27237.275872	12828.873351	10014.947690	
10002	16295.954034	27650.146027	12916.696284	9925.971971	
10003	16183.900733	28059.557778	12990.534044	9872.441363	

	gdp_usd_upper
10000	15839.573177
10001	16094.157651
10002	16384.105693
10003	16584.745555