

## Installing world bank data python package

In [1]: `pip install world_bank_data --upgrade`

```
Requirement already satisfied: world_bank_data in c:\users\ompra\anaconda3\lib\site-packages (0.1.3)
Requirement already satisfied: requests in c:\users\ompra\anaconda3\lib\site-packages (from world_bank_data) (2.26.0)
Requirement already satisfied: cachetools in c:\users\ompra\anaconda3\lib\site-packages (from world_bank_data) (5.0.0)
Requirement already satisfied: pandas in c:\users\ompra\anaconda3\lib\site-packages (from world_bank_data) (1.3.4)
Requirement already satisfied: pytz>=2017.3 in c:\users\ompra\anaconda3\lib\site-packages (from pandas->world_bank_data) (2021.3)
Requirement already satisfied: python-dateutil>=2.7.3 in c:\users\ompra\anaconda3\lib\site-packages (from pandas->world_bank_data) (2.8.2)
Requirement already satisfied: numpy>=1.17.3 in c:\users\ompra\anaconda3\lib\site-packages (from pandas->world_bank_data) (1.20.3)
Note: you may need to restart the kernel to use updated packages.
Requirement already satisfied: six>=1.5 in c:\users\ompra\anaconda3\lib\site-packages (from python-dateutil>=2.7.3->pandas->world_bank_data) (1.16.0)
Requirement already satisfied: charset-normalizer~=2.0.0 in c:\users\ompra\anaconda3\lib\site-packages (from requests->world_bank_data) (2.0.4)
Requirement already satisfied: idna<4,>=2.5 in c:\users\ompra\anaconda3\lib\site-packages (from requests->world_bank_data) (3.2)
Requirement already satisfied: urllib3<1.27,>=1.21.1 in c:\users\ompra\anaconda3\lib\site-packages (from requests->world_bank_data) (1.26.7)
Requirement already satisfied: certifi>=2017.4.17 in c:\users\ompra\anaconda3\lib\site-packages (from requests->world_bank_data) (2021.10.8)
```

In [2]: `pip install wbgapi`

```
Requirement already satisfied: wbgapi in c:\users\ompra\anaconda3\lib\site-packages (1.0.7)
Requirement already satisfied: requests in c:\users\ompra\anaconda3\lib\site-packages (from wbgapi) (2.26.0)
Requirement already satisfied: tabulate in c:\users\ompra\anaconda3\lib\site-packages (from wbgapi) (0.8.9)
Requirement already satisfied: PyYAML in c:\users\ompra\anaconda3\lib\site-packages (from wbgapi) (6.0)
Requirement already satisfied: urllib3<1.27,>=1.21.1 in c:\users\ompra\anaconda3\lib\site-packages (from requests->wbgapi) (1.26.7)
Requirement already satisfied: idna<4,>=2.5 in c:\users\ompra\anaconda3\lib\site-packages (from requests->wbgapi) (3.2)
Requirement already satisfied: charset-normalizer~=2.0.0 in c:\users\ompra\anaconda3\lib\site-packages (from requests->wbgapi) (2.0.4)
Requirement already satisfied: certifi>=2017.4.17 in c:\users\ompra\anaconda3\lib\site-packages (from requests->wbgapi) (2021.10.8)
Note: you may need to restart the kernel to use updated packages.
```

In [3]: `pip install wbdata`

```
Requirement already satisfied: wbdata in c:\users\ompra\anaconda3\lib\site-packages (0.3.0)
Requirement already satisfied: requests>=2.0 in c:\users\ompra\anaconda3\lib\site-packages (from wbdata) (2.26.0)
Requirement already satisfied: tabulate>=0.8.5 in c:\users\ompra\anaconda3\lib\site-packages (from wbdata) (0.8.9)
Requirement already satisfied: decorator>=4.0 in c:\users\ompra\anaconda3\lib\site-packages (from wbdata) (5.1.0)
Requirement already satisfied: appdirs<2.0,>=1.4 in c:\users\ompra\anaconda3\lib\site-packages (from wbdata) (1.4.4)
Requirement already satisfied: charset-normalizer~=2.0.0 in c:\users\ompra\anaconda3\lib\site-packages (from requests->wbdata) (2.0.4)
```

e-packages (from requests>=2.0->wbdata) (2.0.4)  
 Requirement already satisfied: idna<4,>=2.5 in c:\users\ompra\anaconda3\lib\site-packages (from requests>=2.0->wbdata) (3.2)  
 Requirement already satisfied: urllib3<1.27,>=1.21.1 in c:\users\ompra\anaconda3\lib\site-packages (from requests>=2.0->wbdata) (1.26.7)  
 Requirement already satisfied: certifi>=2017.4.17 in c:\users\ompra\anaconda3\lib\site-packages (from requests>=2.0->wbdata) (2021.10.8)  
 Note: you may need to restart the kernel to use updated packages.

## Importing the libraries

```
In [4]: import pandas as pd
import world_bank_data as wb
import wbapi as wb
import wbdata as wb
pd.set_option('display.max_rows', 6)
wb.get_topic()
```

```
Out[4]:
```

	id	value
	1	Agriculture & Rural Development
	2	Aid Effectiveness
	3	Economy & Growth
	4	Education
	5	Energy & Mining
	6	Environment
	7	Financial Sector
	8	Health
	9	Infrastructure
	10	Social Protection & Labor
	11	Poverty
	12	Private Sector
	13	Public Sector
	14	Science & Technology
	15	Social Development
	16	Urban Development
	17	Gender
	18	Millenium development goals
	19	Climate Change
	20	External Debt
	21	Trade

```
In [5]: wb.get_indicator(topic=5)
```

```
Out[5]:
```

id	name
EG.EGY.PRIM.PP.KD	Energy intensity level of primary energy (MJ/\$2011 PPP GDP)
EG.ELC.ACCS.RU.ZS	Access to electricity, rural (% of rural population)
EG.ELC.ACCS.UR.ZS	Access to electricity, urban (% of urban population)
EG.ELC.ACCS.ZS	Access to electricity (% of population)
EG.ELC.COAL.ZS	Electricity production from coal sources (% of total)
EG.ELC.FOSL.ZS	Electricity production from oil, gas and coal sources (% of total)
EG.ELC.HYRO.ZS	Electricity production from hydroelectric sources (% of total)
EG.ELC.LOSS.ZS	Electric power transmission and distribution losses (% of output)
EG.ELC.NGAS.ZS	Electricity production from natural gas sources (% of total)
EG.ELC.NUCL.ZS	Electricity production from nuclear sources (% of total)
EG.ELC.PETR.ZS	Electricity production from oil sources (% of total)
EG.ELC.RNEW.ZS	Renewable electricity output (% of total electricity output)
EG.ELC.RNWX.KH	Electricity production from renewable sources, excluding hydroelectric (kWh)
EG.ELC.RNWX.ZS	Electricity production from renewable sources, excluding hydroelectric (% of total)

EG.FEC.RNEW.ZS	Renewable energy consumption (% of total final energy consumption)
EG.GDP.PUSE.KO.PP	GDP per unit of energy use (PPP \$ per kg of oil equivalent)
EG.GDP.PUSE.KO.PP.KD	GDP per unit of energy use (constant 2017 PPP \$ per kg of oil equivalent)
EG.IMP.CONS.ZS	Energy imports, net (% of energy use)
EG.NSF.ACCS.RU.ZS	Access to non-solid fuel, rural (% of rural population)
EG.NSF.ACCS.UR.ZS	Access to non-solid fuel, urban (% of urban population)
EG.NSF.ACCS.ZS	Access to non-solid fuel (% of population)
EG.USE.COMM.CL.ZS	Alternative and nuclear energy (% of total energy use)
EG.USE.COMM.FO.ZS	Fossil fuel energy consumption (% of total)
EG.USE.COMM.GD.PP.KD	Energy use (kg of oil equivalent) per \$1,000 GDP (constant 2017 PPP)
EG.USE.CRNW.ZS	Combustible renewables and waste (% of total energy)
EG.USE.ELEC.KH.PC	Electric power consumption (kWh per capita)
EG.USE.PCAP.KG.OE	Energy use (kg of oil equivalent per capita)
EN.ATM.CO2E.GF.ZS	CO2 emissions from gaseous fuel consumption (% of total)
EN.ATM.CO2E.LF.KT	CO2 emissions from liquid fuel consumption (kt)
EN.ATM.METH.EG.KT.CE	Methane emissions in energy sector (thousand metric tons of CO2 equivalent)
EN.ATM.METH.EG.ZS	Energy related methane emissions (% of total)
EN.ATM.NOXE.EG.KT.CE	Nitrous oxide emissions in energy sector (thousand metric tons of CO2 equivalent)
EN.ATM.NOXE.EG.ZS	Nitrous oxide emissions in energy sector (% of total)
EP.PMP.DESL.CD	Pump price for diesel fuel (US\$ per liter)
EP.PMP.SGAS.CD	Pump price for gasoline (US\$ per liter)
IC.ELC.DURS	Time to obtain an electrical connection (days)
IC.ELC.TIME	Time required to get electricity (days)
IC.FRM.BNKS.ZS	Firms using banks to finance investment (% of firms)
IC.FRM.OUTG.ZS	Value lost due to electrical outages (% of sales for affected firms)
IE.PPI.ENGY.CD	Investment in energy with private participation (current US\$)
NY.ADJ.DMIN.CD	Adjusted savings: mineral depletion (current US\$)
NY.ADJ.DMIN.GN.ZS	Adjusted savings: mineral depletion (% of GNI)
NY.ADJ.DNGY.CD	Adjusted savings: energy depletion (current US\$)
NY.ADJ.DNGY.GN.ZS	Adjusted savings: energy depletion (% of GNI)
NY.ADJ.DRES.GN.ZS	Adjusted savings: natural resources depletion (% of GNI)
NY.GDP.MINR.RT.ZS	Mineral rents (% of GDP)
NY.GDP.NGAS.RT.ZS	Natural gas rents (% of GDP)
NY.GDP.PETR.RT.ZS	Oil rents (% of GDP)
NY.GDP.TOTL.RT.ZS	Total natural resources rents (% of GDP)
TM.VAL.FUEL.ZS.UN	Fuel imports (% of merchandise imports)
TM.VAL.MMTL.ZS.UN	Ores and metals imports (% of merchandise imports)
TX.VAL.FUEL.ZS.UN	Fuel exports (% of merchandise exports)
TX.VAL.MMTL.ZS.UN	Ores and metals exports (% of merchandise exports)

```
In [6]: indicators = {"EG.ELC.ACCS.ZS": "Access_to_electricity(% of population)", "EG.USE.ELEC.KH.PC":
```

```
In [7]: LAMcountries = ['ARG', 'BLZ', 'BOL', 'BRA', 'CHL', 'COL', 'CRI', 'CUB', 'SLV', 'GTM', 'GUY', 'HTI', 'HND',
                    'JAM', 'MEX', 'NIC', 'PAN', 'PRY', 'PER', 'PRI', 'SUR', 'URY', 'VEN']
NAMcountries=['USA', 'CAN']
#regions=['LCN', 'WLD', 'NAC'] #WORLD = WLD; LAT AM AND CARRIBEAN = LCN; NAC=North America
regions=['NAC', 'WLD', 'LCN']
allcountries=LAMcountries+NAMcountries
allregions=allcountries+regions
```

## Building the data frame

```
In [8]: df = wb.get_dataframe(indicators, country=allregions)
df
```

```
Out[8]:
```

		Access_to_electricity(% of population)	Electric power consumption (kWh per capita)	Energy use (kg of oil equivalent per capita)	CO2 emissions from gaseous fuel consumption (% of total)	CO2 emissions from liquid fuel consumption (kt)
country	date					
Argentina	2020	NaN	NaN	NaN	NaN	NaN
	2019	100.0	NaN	NaN	NaN	NaN
	2018	100.0	NaN	NaN	NaN	NaN
...	...	...	...	...	...	...
World	1962	NaN	NaN	NaN	11.413496	3.418103e+06
	1961	NaN	NaN	NaN	10.727979	3.185843e+06
	1960	NaN	NaN	NaN	10.110440	3.025727e+06

1708 rows × 5 columns

In [9]:

```
df.shape
df.columns.values
df.info()
df.describe()
df.head()
df.head(10)
df.tail()
df.tail(10)
```

```
<class 'wbdata.api.WBDataFrame'>
MultiIndex: 1708 entries, ('Argentina', '2020') to ('World', '1960')
Data columns (total 5 columns):
#   Column                                                                 Non-Null Count  Dtype
---  -
0   Access_to_electricity(% of population)                               767 non-null   float64
1   Electric power consumption (kWh per capita)                         1104 non-null  float64
2   Energy use (kg of oil equivalent per capita)                        1118 non-null  float64
3   CO2 emissions from gaseous fuel consumption (% of total)           1539 non-null  float64
4   CO2 emissions from liquid fuel consumption (kt)                     1539 non-null  float64
dtypes: float64(5)
memory usage: 138.5+ KB
```

Out[9]:

		Access_to_electricity(% of population)	Electric power consumption (kWh per capita)	Energy use (kg of oil equivalent per capita)	CO2 emissions from gaseous fuel consumption (% of total)	CO2 emissions from liquid fuel consumption (kt)
country	date					
World	1969	NaN	NaN	NaN	13.968217	5.673121e+06
	1968	NaN	NaN	NaN	13.642661	5.232388e+06
	1967	NaN	NaN	NaN	13.246163	4.808931e+06
	...	...	...	...	...	...
	1962	NaN	NaN	NaN	11.413496	3.418103e+06
	1961	NaN	NaN	NaN	10.727979	3.185843e+06
	1960	NaN	NaN	NaN	10.110440	3.025727e+06

10 rows × 5 columns

```
In [10]: data = df.copy()
data.reset_index(inplace=True)
```

```
In [11]: data['text']=data['country']+data['date'].map(str)
data
```

Out[11]:

	country	date	Access_to_electricity(% of population)	Electric power consumption (kWh per capita)	Energy use (kg of oil equivalent per capita)	CO2 emissions from gaseous fuel consumption (% of total)	CO2 emissions from liquid fuel consumption (kt)	
0	Argentina	2020	NaN	NaN	NaN	NaN	NaN	Argentina
1	Argentina	2019	100.0	NaN	NaN	NaN	NaN	Argentina
2	Argentina	2018	100.0	NaN	NaN	NaN	NaN	Argentina
...	...	...	...	...	...	...	...	
1705	World	1962	NaN	NaN	NaN	11.413496	3.418103e+06	World
1706	World	1961	NaN	NaN	NaN	10.727979	3.185843e+06	World
1707	World	1960	NaN	NaN	NaN	10.110440	3.025727e+06	World

1708 rows × 8 columns



```
In [12]: data1 = data.iloc[0:1647]
data1.tail(20)
```

Out[12]:

	country	date	Access_to_electricity(% of population)	Electric power consumption (kWh per capita)	Energy use (kg of oil equivalent per capita)	CO2 emissions from gaseous fuel consumption (% of total)	CO2 emissions from liquid fuel consumption (kt)	
1627	Venezuela, RB	1979	NaN	1835.725703	1935.398581	39.260572	38829.863	Venez RB
1628	Venezuela, RB	1978	NaN	1542.598167	1921.267277	37.817638	35826.590	Venez RB
1629	Venezuela, RB	1977	NaN	1400.420513	1830.998332	39.424626	30040.064	Venez RB
...	...	...	...	...	...	...	...	
1644	Venezuela, RB	1962	NaN	NaN	NaN	19.776347	8049.065	Venez RB
1645	Venezuela, RB	1961	NaN	NaN	NaN	19.490149	11551.050	Venez RB
1646	Venezuela, RB	1960	NaN	NaN	NaN	16.661312	14895.354	Venez RB

20 rows × 8 columns



```
In [13]: data1.head(10)
```

Out[13]:

	country	date	Access_to_electricity(% of population)	Electric power consumption (kWh per capita)	Energy use (kg of oil equivalent per capita)	CO2 emissions from gaseous fuel consumption (% of total)	CO2 emissions from liquid fuel consumption (kt)	
0	Argentina	2020	NaN	NaN	NaN	NaN	NaN	Argentina2
1	Argentina	2019	100.000000	NaN	NaN	NaN	NaN	Argentina2
2	Argentina	2018	100.000000	NaN	NaN	NaN	NaN	Argentina2
...	...	...	...	...	...	...	...	
7	Argentina	2013	99.356224	2967.376558	1967.021678	49.366988	90835.257	Argentina2
8	Argentina	2012	99.228859	3000.603523	1936.803540	53.469638	85657.453	Argentina2
9	Argentina	2011	99.080200	2929.075029	1952.051053	52.274299	86977.573	Argentina2

10 rows × 8 columns



```
In [14]: data2=data1.dropna()  
data2
```

Out[14]:

	country	date	Access_to_electricity(% of population)	Electric power consumption (kWh per capita)	Energy use (kg of oil equivalent per capita)	CO2 emissions from gaseous fuel consumption (% of total)	CO2 emissions from liquid fuel consumption (kt)	
6	Argentina	2014	100.000000	3074.702071	2029.922825	53.744348	87586.295	Argen
7	Argentina	2013	99.356224	2967.376558	1967.021678	49.366988	90835.257	Argen
8	Argentina	2012	99.228859	3000.603523	1936.803540	53.469638	85657.453	Argen
...	...	...	...	...	...	...	...	
1612	Venezuela, RB	1994	98.323334	2665.759795	2324.974122	45.391298	63494.105	Ve
1613	Venezuela, RB	1993	98.239532	2697.686148	2049.504018	36.197225	69302.633	Ve
1614	Venezuela, RB	1992	97.820203	2689.309414	2249.819897	18.476779	73831.378	Ve

537 rows × 8 columns



```
In [15]: data2.shape  
data2.describe()
```

Out[15]:

	Access_to_electricity(% of population)	Electric power consumption (kWh per capita)	Energy use (kg of oil equivalent per capita)	CO2 emissions from gaseous fuel consumption (% of total)	CO2 emissions from liquid fuel consumption (kt)
count	537.000000	537.000000	537.000000	537.000000	5.370000e+02
mean	88.330460	3201.388654	1943.707057	14.270363	3.015548e+05
std	14.929808	4566.648445	2385.349097	16.009996	6.743868e+05
...	...	...	...	...	...
50%	94.664978	1513.158055	1022.160344	7.076101	2.083589e+04
75%	98.820000	2602.250605	1599.516711	25.213470	2.057297e+05
max	100.000000	17264.736744	8455.547014	67.060262	2.699517e+06

8 rows × 5 columns

In [16]:

```
data3=data2.sort_values(by='date', ascending='True')
data3
```

Out[16]:

	country	date	Access_to_electricity(% of population)	Electric power consumption (kWh per capita)	Energy use (kg of oil equivalent per capita)	CO2 emissions from gaseous fuel consumption (% of total)	CO2 emissions from liquid fuel consumption (kt)	
396	Colombia	1990	89.900000	871.593984	731.743962	17.149192	31136.497	Colo
1006	North America	1990	100.000000	12158.089105	7667.611043	21.453940	2200111.992	Am
30	Argentina	1990	92.154800	1303.978019	1412.179959	43.317907	59236.718	Argei
...	...	...	...	...	...	...	...	
1470	Uruguay	2014	99.657085	3085.189883	1386.018099	1.410385	6233.900	Uru
555	Guatemala	2014	85.494371	601.189731	863.568935	0.000000	9739.552	Guate
6	Argentina	2014	100.000000	3074.702071	2029.922825	53.744348	87586.295	Argei

537 rows × 8 columns



In [17]:

```
data3.describe()
```

Out[17]:

	Access_to_electricity(% of population)	Electric power consumption (kWh per capita)	Energy use (kg of oil equivalent per capita)	CO2 emissions from gaseous fuel consumption (% of total)	CO2 emissions from liquid fuel consumption (kt)
count	537.000000	537.000000	537.000000	537.000000	5.370000e+02
mean	88.330460	3201.388654	1943.707057	14.270363	3.015548e+05
std	14.929808	4566.648445	2385.349097	16.009996	6.743868e+05

	Access_to_electricity(% of population)	Electric power consumption (kWh per capita)	Energy use (kg of oil equivalent per capita)	CO2 emissions from gaseous fuel consumption (% of total)	CO2 emissions from liquid fuel consumption (kt)
...	...	...	...	...	...
50%	94.664978	1513.158055	1022.160344	7.076101	2.083589e+04
75%	98.820000	2602.250605	1599.516711	25.213470	2.057297e+05
max	100.000000	17264.736744	8455.547014	67.060262	2.699517e+06

8 rows × 5 columns

In [18]:

```
dataw = data.iloc[1648:1708]
dataworld = dataw.dropna()
dataworld
```

Out[18]:

	country	date	Access_to_electricity(% of population)	Electric power consumption (kWh per capita)	Energy use (kg of oil equivalent per capita)	CO2 emissions from gaseous fuel consumption (% of total)	CO2 emissions from liquid fuel consumption (kt)	te
1653	World	2014	85.552847	3128.298712	1919.991765	20.336427	1.036714e+07	World20
1654	World	2013	85.031065	3106.981734	1894.112059	20.170828	1.020665e+07	World20
1655	World	2012	84.745819	3047.755531	1891.700426	20.216118	1.013381e+07	World20
...	...	...	...	...	...	...	...	...
1667	World	2000	78.736034	2386.825244	1637.205968	20.662066	9.274896e+06	World20
1668	World	1999	74.708477	2317.327280	1623.839599	20.716253	9.044144e+06	World19
1669	World	1998	73.406490	2284.670692	1611.335079	19.885250	8.898867e+06	World19

17 rows × 8 columns



In [19]:

```
dataworld.describe()
```

Out[19]:

	Access_to_electricity(% of population)	Electric power consumption (kWh per capita)	Energy use (kg of oil equivalent per capita)	CO2 emissions from gaseous fuel consumption (% of total)	CO2 emissions from liquid fuel consumption (kt)
count	17.000000	17.000000	17.000000	17.000000	1.700000e+01
mean	80.826483	2705.724699	1768.351915	20.263791	9.692576e+06
std	3.338761	288.516132	108.412889	0.286536	4.133745e+05
...	...	...	...	...	...
50%	81.251031	2732.640247	1796.215452	20.216118	9.779347e+06
75%	82.765005	2958.343199	1874.657688	20.438915	9.955128e+06
max	85.552847	3128.298712	1919.991765	20.716253	1.036714e+07



8 rows × 5 columns

In [20]:

```
data3.corr()
```

Out[20]:

	Access_to_electricity(% of population)	Electric power consumption (kWh per capita)	Energy use (kg of oil equivalent per capita)	CO2 emissions from gaseous fuel consumption (% of total)	CO2 emissions from liquid fuel consumption (kt)
Access_to_electricity(% of population)	1.000000	0.428142	0.426809	0.365172	0.307581
Electric power consumption (kWh per capita)	0.428142	1.000000	0.989543	0.366663	0.712610
Energy use (kg of oil equivalent per capita)	0.426809	0.989543	1.000000	0.397099	0.771272
CO2 emissions from gaseous fuel consumption (% of total)	0.365172	0.366663	0.397099	1.000000	0.247726
CO2 emissions from liquid fuel consumption (kt)	0.307581	0.712610	0.771272	0.247726	1.000000

In [21]:

```
data3['date']=pd.to_numeric(data3['date'])
```

In [22]:

```
print(data3['date'].corr(data3['Access_to_electricity(% of population)']))  
  
0.19089298637428545
```

In [23]:

```
dataworld.corr()
```

Out[23]:

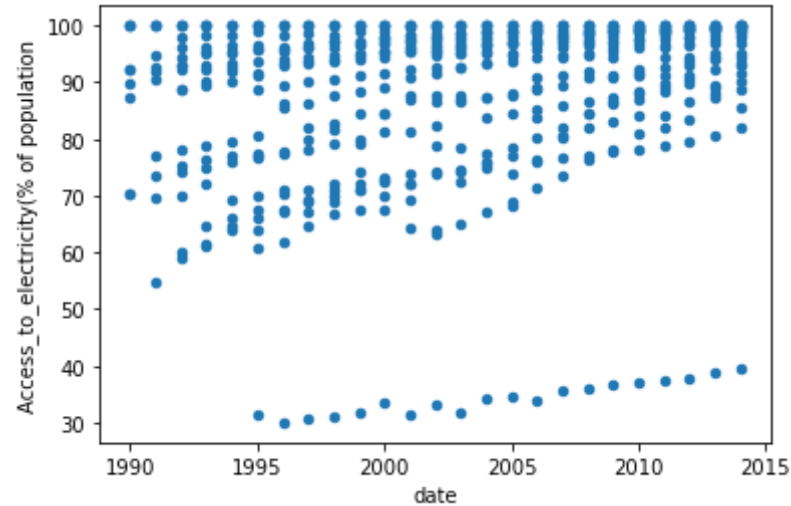
	Access_to_electricity(% of population)	Electric power consumption (kWh per capita)	Energy use (kg of oil equivalent per capita)	CO2 emissions from gaseous fuel consumption (% of total)	CO2 emissions from liquid fuel consumption (kt)
Access_to_electricity(% of population)	1.000000	0.927149	0.913741	-0.220728	0.954869
Electric power consumption (kWh per capita)	0.927149	1.000000	0.992909	-0.373846	0.956836
Energy use (kg of oil equivalent per capita)	0.913741	0.992909	1.000000	-0.427884	0.962349
CO2 emissions from gaseous fuel consumption (% of total)	-0.220728	-0.373846	-0.427884	1.000000	-0.336303

	Access_to_electricity(% of population)	Electric power consumption (kWh per capita)	Energy use (kg of oil equivalent per capita)	CO2 emissions from gaseous fuel consumption (% of total)	CO2 emissions from liquid fuel consumption (kt)
CO2 emissions from liquid fuel consumption (kt)	0.954869	0.956836	0.962349	-0.336303	1.000000

```
In [24]: import pandas as pd
import numpy as np
```

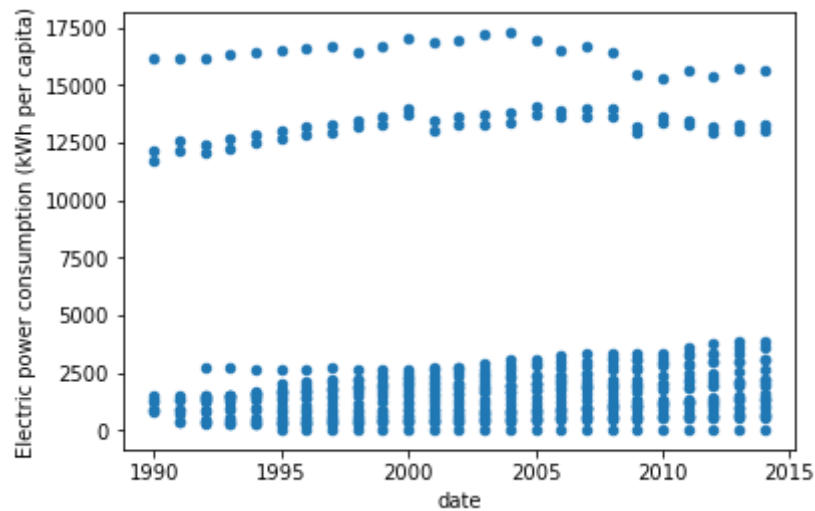
```
In [25]: data3.plot.scatter(x="date", y="Access_to_electricity(% of population)")
```

Out[25]: <AxesSubplot:xlabel='date', ylabel='Access\_to\_electricity(% of population)'



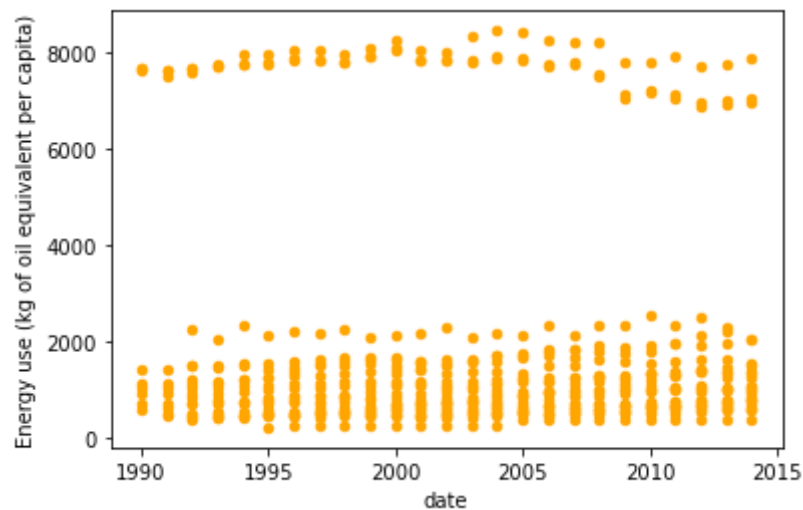
```
In [26]: data3.plot.scatter(x="date", y="Electric power consumption (kWh per capita)")
```

Out[26]: <AxesSubplot:xlabel='date', ylabel='Electric power consumption (kWh per capita)'



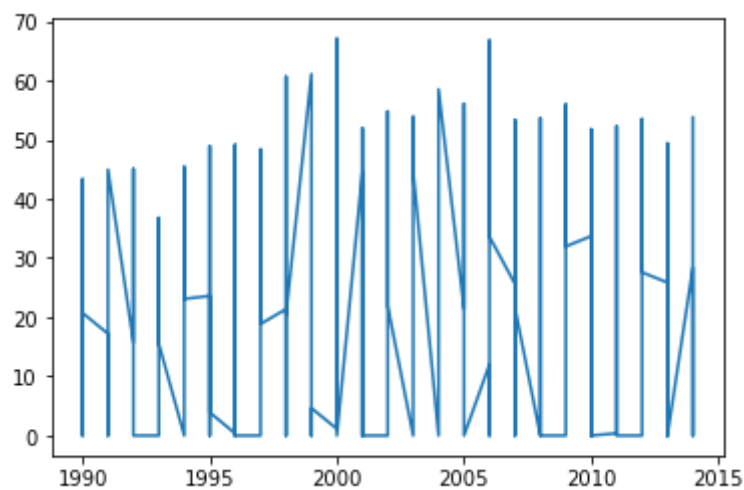
```
In [27]: data3.plot.scatter(x="date", y="Energy use (kg of oil equivalent per capita)", color="orange")
```

Out[27]: <AxesSubplot:xlabel='date', ylabel='Energy use (kg of oil equivalent per capita)'



In [28]: `import matplotlib.pyplot as plt  
plt.plot(data3['date'], data3['CO2 emissions from gaseous fuel consumption (% of total)'])`

Out[28]: [`matplotlib.lines.Line2D` at 0x101e7882b50>]



In [29]: `dataworld.plot.scatter(x="date", y="CO2 emissions from gaseous fuel consumption (% of total)",`

