

Intitulé du projet : Analyse des données de santé « Cas du Paludisme en Afrique »

Contenu du Dataset « Africa Malaria » :

- L'ensemble de données "Africa Malaria" comprend des données sur les pays africains de 2007 à 2017, avec les caractéristiques suivantes :
- Code de pays ISO-3 unique : Chaque pays est identifié par un code de pays ISO-3, qui est un code standardisé utilisé pour représenter les pays dans les données internationales.
 - Latitude et longitude : Pour chaque pays, l'ensemble de données fournit également les coordonnées de latitude et de longitude, qui donnent la position géographique approximative du pays.
 - Cas de paludisme signalés : L'ensemble de données comprend des informations sur les cas de paludisme signalés dans chaque pays et chaque année. Ces données peuvent inclure le nombre total de cas, le nombre de cas selon le sexe, l'âge ou d'autres caractéristiques démographiques, ainsi que la gravité des cas signalés.
 - Mesures préventives : L'ensemble de données fournit également des données sur les mesures préventives prises pour lutter contre le paludisme dans chaque pays. Cela peut inclure des informations sur les campagnes de sensibilisation, les programmes de distribution de moustiquaires, les traitements médicaux administrés, etc.

1. Importation des bibliothèques pandas, numpy et matplotlib, pour manipuler les données et effectuer des analyses statistiques

```
In [101]:
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

2. Chargement du fichier du Dataset dans un dataframe pandas

Lien de téléchargement du Dataset : <https://www.kaggle.com/datasets/lydia70/malaria-in-africa> (<https://www.kaggle.com/datasets/lydia70/malaria-in-africa>)

```
In [102]:
df = pd.read_csv("DatasetAfricaMalaria.csv")
```

3. Nettoyage et structuration des données pour l'analyse

3.1. Copie du Dataset

```
In [103]:
dfc = df.copy()
```

3.2. Affichage des 5 premières lignes du Dataset

```
In [104]:
dfc.head()
```

Out[104]:

	Country Name	Year	Country Code	Incidence of malaria (per 1,000 population at risk)	Malaria cases reported	Use of insecticide-treated bed nets (% of under-5 population)	Children with fever receiving antimalarial drugs (% of children under age 5 with fever)	Intermittent preventive treatment (IPT) of malaria in pregnancy (% of pregnant women)	People using safely managed drinking water services (% of population)	People using safely managed drinking water services, rural (% of rural population)	Urban population growth (annual %)	People using at least basic drinking water services (% of population)	People using at least basic drinking water services, rural (% of rural population)
0	Algeria	2007	DZA	0.01	26.0	NaN	NaN	NaN	NaN	NaN	2.71	91.68	85.83
1	Angola	2007	AGO	286.72	1533485.0	18.0	29.8	1.5	NaN	NaN	5.01	47.96	23.77
2	Benin	2007	BEN	480.24	0.0	NaN	NaN	NaN	NaN	NaN	4.09	63.78	54.92
3	Botswana	2007	BWA	1.03	390.0	NaN	NaN	NaN	NaN	NaN	4.80	78.89	57.60
4	Burkina Faso	2007	BFA	503.80	44246.0	NaN	NaN	NaN	NaN	NaN	5.91	52.27	45.13

5 rows × 27 columns

3.3. Affichage des 5 dernières lignes du Dataset

In [105]:

dfc.tail()

Out[105]:

	Country Name	Year	Country Code	Incidence of malaria (per 1,000 population at risk)	Malaria cases reported	Use of insecticide-treated bed nets (% of under-5 population)	Children with fever receiving antimalarial drugs (% of children under age 5 with fever)	Intermittent preventive treatment (IPT) of malaria in pregnancy (% of pregnant women)	People using safely managed drinking water services (% of population)	People using safely managed drinking water services, rural (% of rural population)	...	Urban population growth (annual %)	People using at least basic drinking water services (% of population)	People using at least basic sanitation services, rural (% of rural population)
589	Togo	2017	TGO	278.20	1755577.0	69.7	31.1	41.7	NaN	NaN	...	3.79	65.13	41
590	Tunisia	2017	TUN	NaN	NaN	NaN	NaN	NaN	92.66	NaN	...	1.57	96.25	81
591	Uganda	2017	UGA	336.76	11667831.0	NaN	NaN	NaN	7.07	4.46	...	6.25	49.10	41
592	Zambia	2017	ZMB	160.05	5505639.0	NaN	NaN	NaN	NaN	NaN	...	4.21	59.96	41
593	Zimbabwe	2017	ZWE	108.55	467508.0	NaN	NaN	NaN	NaN	NaN	...	1.28	64.05	41

5 rows × 27 columns

3.4. Examen des différentes colonnes et types de données dans le Dataset

In [106]:

dfc.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 594 entries, 0 to 593
Data columns (total 27 columns):
Column Non-Null Count Dtype
--- -
0 Country Name 594 non-null object
1 Year 594 non-null int64
2 Country Code 594 non-null object
3 Incidence of malaria (per 1,000 population at risk) 550 non-null float64
4 Malaria cases reported 550 non-null float64
5 Use of insecticide-treated bed nets (% of under-5 population) 132 non-null float64
6 Children with fever receiving antimalarial drugs (% of children under age 5 with fever) 122 non-null float64
7 Intermittent preventive treatment (IPT) of malaria in pregnancy (% of pregnant women) 106 non-null float64
8 People using safely managed drinking water services (% of population) 99 non-null float64
9 People using safely managed drinking water services, rural (% of rural population) 88 non-null float64
10 People using safely managed drinking water services, urban (% of urban population) 176 non-null float64
11 People using safely managed sanitation services (% of population) 132 non-null float64
12 People using safely managed sanitation services, rural (% of rural population) 110 non-null float64
13 People using safely managed sanitation services, urban (% of urban population) 132 non-null float64
14 Rural population (% of total population) 588 non-null float64
15 Rural population growth (annual %) 588 non-null float64
16 Urban population (% of total population) 588 non-null float64
17 Urban population growth (annual %) 588 non-null float64
18 People using at least basic drinking water services (% of population) 588 non-null float64
19 People using at least basic drinking water services, rural (% of rural population) 566 non-null float64
20 People using at least basic drinking water services, urban (% of urban population) 566 non-null float64
21 People using at least basic sanitation services (% of population) 588 non-null float64
22 People using at least basic sanitation services, rural (% of rural population) 566 non-null float64
23 People using at least basic sanitation services, urban (% of urban population) 566 non-null float64
24 latitude 594 non-null float64
25 longitude 594 non-null float64
26 geometry 594 non-null object
dtypes: float64(23), int64(1), object(3)
memory usage: 125.4+ KB

Nombre d'index dans notre Dataset

In [107]:

dfc.index

Out[107]:

RangeIndex(start=0, stop=594, step=1)

Conversion du type de données de la colonne "Malaria cases reported" en nombre entier

In [108]:

dfc['Malaria cases reported']

Out[108]:

```

0          26.0
1    1533485.0
2           0.0
3         390.0
4    44246.0
...
589    1755577.0
590          NaN
591    11667831.0
592    5505639.0
593    467508.0
Name: Malaria cases reported, Length: 594, dtype: float64

```

In [109]:

dfc['Malaria cases reported'] = dfc['Malaria cases reported'].astype('Int64')

In [110]:

dfc['Malaria cases reported']

Out[110]:

```

0          26
1    1533485
2           0
3         390
4    44246
...
589    1755577
590         <NA>
591    11667831
592    5505639
593    467508
Name: Malaria cases reported, Length: 594, dtype: Int64

```

Renommeons certaines colonnes de notre Dataset

In [111]:

```

### Listons Les noms de colonnes du Dataset avant renommage
list(dfc)

```

Out[111]:

```

['Country Name',
 'Year',
 'Country Code',
 'Incidence of malaria (per 1,000 population at risk)',
 'Malaria cases reported',
 'Use of insecticide-treated bed nets (% of under-5 population)',
 'Children with fever receiving antimalarial drugs (% of children under age 5 with fever)',
 'Intermittent preventive treatment (IPT) of malaria in pregnancy (% of pregnant women)',
 'People using safely managed drinking water services (% of population)',
 'People using safely managed drinking water services, rural (% of rural population)',
 'People using safely managed drinking water services, urban (% of urban population)',
 'People using safely managed sanitation services (% of population)',
 'People using safely managed sanitation services, rural (% of rural population)',
 'People using safely managed sanitation services, urban (% of urban population)',
 'Rural population (% of total population)',
 'Rural population growth (annual %)',
 'Urban population (% of total population)',
 'Urban population growth (annual %)',
 'People using at least basic drinking water services (% of population)',
 'People using at least basic drinking water services, rural (% of rural population)',
 'People using at least basic drinking water services, urban (% of urban population)',
 'People using at least basic sanitation services (% of population)',
 'People using at least basic sanitation services, rural (% of rural population)',
 'People using at least basic sanitation services, urban (% of urban population)',
 'latitude',
 'longitude',
 'geometry']

```

Renommage des colonnes du Dataset

```
In [112]:
n (% of urban population)':'% of urban population using at least basic sanitation services', 'geometry':'Localisation'}, inplace = True)
```

```
In [113]:
list(dfc)
```

Out[113]:

```
['Country Name',
'Year',
'Country Code',
'Incidence of malaria (per 1,000 population at risk)',
'Malaria cases reported',
'Use of insecticide-treated bed nets (% of under-5 population)',
'% of children under age 5 with fever receiving antimalarial drugs',
'% of pregnant women using Intermittent preventive treatment (IPT) of malaria in pregnancy',
'% of population using safely managed drinking water services',
'% of rural population using safely managed drinking water services',
'% of urban population using safely managed drinking water services',
'% of population using safely managed sanitation services',
'% of rural population using safely managed sanitation services',
'% of urban population using safely managed sanitation services',
'Rural population (% of total population)',
'Rural population growth (annual %)',
'Urban population (% of total population)',
'Urban population growth (annual %)',
'% of population using at least basic drinking water services',
'% of rural population using at least basic drinking water services',
'% of urban population using at least basic drinking water services',
' % of population using at least basic sanitation services',
'% of rural population using at least basic sanitation services',
'% of urban population using at least basic sanitation services',
'latitude',
'longitude',
'Localisation']
```

4. Analyse exploratoire des données

4.1. Statistiques descriptives du Dataset

```
In [114]:
df.describe()
```

Out[114]:

	Year	Incidence of malaria (per 1,000 population at risk)	Malaria cases reported	Use of insecticide-treated bed nets (% of under-5 population)	Children with fever receiving antimalarial drugs (% of children under age 5 with fever)	Intermittent preventive treatment (IPT) of malaria in pregnancy (% of pregnant women)	People using safely managed drinking water services (% of population)	People using safely managed drinking water services, rural (% of rural population)	People using safely managed drinking water services, urban (% of urban population)	People using safely managed sanitation services (% of population)	...	Urban population (% of total population)	po
count	594.000000	550.000000	5.500000e+02	132.000000	122.000000	106.000000	99.000000	88.000000	176.000000	132.000000	...	588.000000	588.000000
mean	2012.000000	190.087491	1.068330e+06	42.530303	30.201639	15.013958	33.478990	12.470568	51.549545	28.768939	...	43.164116	3
std	3.164943	163.054527	2.192802e+06	20.157059	18.903198	12.389166	26.678321	10.078371	24.157416	18.631510	...	18.086118	3
min	2007.000000	0.000000	0.000000e+00	1.000000	0.500000	0.000000	5.770000	0.930000	11.200000	6.370000	...	9.860000	-0
25%	2009.000000	30.857500	2.211750e+03	26.675000	17.275000	5.763285	8.975000	4.185000	34.125000	16.532500	...	28.795000	2
50%	2012.000000	174.775000	1.130260e+05	42.900000	29.300000	11.500000	28.390000	10.675000	51.365000	25.410000	...	41.560000	3
75%	2015.000000	347.637500	1.154808e+06	56.325000	42.625000	21.850000	43.890000	16.887500	70.747500	35.725000	...	56.945000	4
max	2017.000000	585.540000	1.682113e+07	95.500000	76.900000	59.600000	92.660000	39.930000	89.540000	78.120000	...	88.980000	7

8 rows × 24 columns

4.2. Analyses statistiques pour identifier les tendances des maladies, les facteurs de risque

* Tendances annuelles des cas de paludisme, classées par année du nombre de cas le plus élevé de paludisme au nombre de cas le plus faible

```
In [115]:
som_tendances_annuelles = dfc.groupby("Year")["Malaria cases reported"].sum().sort_values(ascending=False)
```

In [117]:

```
som_tendances_annuelles
```

Out[117]:

```
Year
2017    128146255
2016    123330818
2015     92968526
2014     77028814
2013     47174959
2012     39719221
2011     23002865
2010     22184760
2009     14414616
2007     10102341
2008      9508374
Name: Malaria cases reported, dtype: Int64
```

In [118]:

```
moy_tendances_annuelles = dfc.groupby("Year")["Malaria cases reported"].mean().sort_values(ascending=False)
```

In [119]:

```
moy_tendances_annuelles
```

Out[119]:

```
Year
2017    2562925.1
2016    2466616.36
2015    1859370.52
2014    1540576.28
2013     943499.18
2012     794384.42
2011     460057.3
2010     443695.2
2009     288292.32
2007     202046.82
2008     190167.48
Name: Malaria cases reported, dtype: Float64
```

*** Tendances par pays des cas de paludisme de 2007-2017, classé par pays ayant le nombre de cas de paludisme le plus élevé au nombre de cas le plus faible**

In [120]:

```
som_tendances_pays = dfc.groupby("Country Name")["Malaria cases reported"].sum().sort_values(ascending=False)
```

In [121]:

```
som_tendances_pays
```

Out[121]:

Country Name	
Congo, Dem. Rep.	77555524
Mozambique	43984158
Burkina Faso	41601963
Uganda	41126230
Burundi	40249714
Nigeria	40066570
Ghana	28008309
Tanzania	24283088
Angola	24034182
Malawi	19445640
Niger	18866709
Zambia	18619166
Cote d'Ivoire	17978339
Rwanda	17498217
Kenya	17353851
Ethiopia	16266669
Liberia	12237026
Mali	12106257
Sierra Leone	11919725
Togo	11162390
Benin	8579720
Sudan	7265570
Chad	6198112
Madagascar	4616832
Guinea	4580444
Cameroon	4373628
Zimbabwe	3344410
Senegal	3283980
South Sudan	3143581
Central African Republic	2561261
Gambia, The	1675997
Congo, Rep.	932292
Guinea-Bissau	749386
Equatorial Guinea	336825
Eritrea	306876
Comoros	289016
Gabon	264144
Somalia	254342
Namibia	120908
Mauritania	117335
South Africa	93089
Djibouti	61718
Sao Tome and Principe	54276
Botswana	8756
Eswatini	4531
Cabo Verde	699
Algeria	94
Seychelles	0
Libya	0
Egypt, Arab Rep.	0
Tunisia	0
Morocco	0
Mauritius	0
Lesotho	0

Name: Malaria cases reported, dtype: Int64

In [122]:

```
moy_tendances_pays = dfc.groupby("Country Name")["Malaria cases reported"].mean().sort_values(ascending=False)
```

In [123]:

```
moy_tendances_pays
```

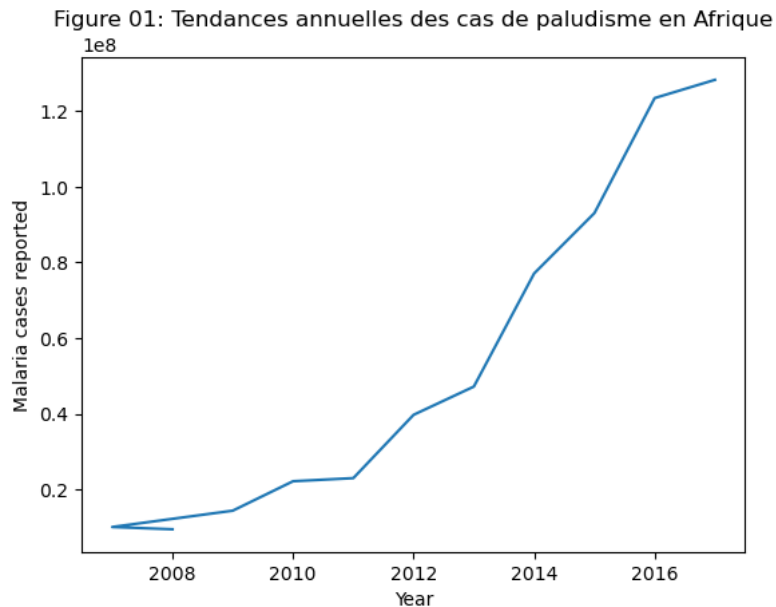
Out[123]:

Country Name	
Congo, Dem. Rep.	7050502.181818
Mozambique	3998559.818182
Burkina Faso	3781996.636364
Uganda	3738748.181818
Burundi	3659064.909091
Nigeria	3642415.454545
Ghana	2546209.909091
Tanzania	2207553.454545
Angola	2184925.636364
Malawi	1767785.454545
Niger	1715155.363636
Zambia	1692651.454545
Cote d'Ivoire	1634394.454545
Rwanda	1590747.0
Kenya	1577622.818182
Ethiopia	1478788.090909
Liberia	1112456.909091
Mali	1100568.818182
Sierra Leone	1083611.363636
Togo	1014762.727273
Benin	779974.545455
Sudan	660506.363636
Chad	563464.727273
Madagascar	419712.0
Guinea	416404.0
Cameroon	397602.545455
Zimbabwe	304037.272727
Senegal	298543.636364
South Sudan	285780.090909
Central African Republic	232841.909091
Gambia, The	152363.363636
Congo, Rep.	84753.818182
Guinea-Bissau	68126.0
Equatorial Guinea	30620.454545
Eritrea	27897.818182
Comoros	26274.181818
Gabon	24013.090909
Somalia	23122.0
Namibia	10991.636364
Mauritania	10666.818182
South Africa	8462.636364
Djibouti	5610.727273
Sao Tome and Principe	4934.181818
Botswana	796.0
Eswatini	411.909091
Cabo Verde	63.545455
Algeria	8.545455
Egypt, Arab Rep.	0.0
Morocco	0.0
Libya	0.0
Lesotho	<NA>
Mauritius	<NA>
Seychelles	<NA>
Tunisia	<NA>
Name: Malaria cases reported, dtype: Float64	

*** Visualisation des résultats des tendances annuelles**

In [135]:

```
som_tendances_annuelles.plot(kind="line")
plt.xlabel("Year")
plt.ylabel("Malaria cases reported")
plt.title("Figure 01: Tendances annuelles des cas de paludisme en Afrique")
plt.show()
```



Lecture de la Figure 01 : la tendance est que le nombre de cas de paludisme en Afrique n'a fait qu'augmenter d'année en année entre 2007 et 2017

*** Tendances des cas de paludisme en fonction de l'utilisation de moustiquaires imprégnées d'insecticide (% de la population de moins de 5 ans)**

In [131]:

```
tendances_treated_bed_nets = dfc.groupby("Use of insecticide-treated bed nets (% of under-5 population)")["Malaria cases reported"].sum()
```

In [132]:

```
som_tendances_treated_bed_nets
```

Out[132]:

```
Use of insecticide-treated bed nets (% of under-5 population)
```

```
1.0      84
1.5     268
4.5    33405
5.0    44518
5.6     4911
...
```

```
75.3   5428655
```

```
76.5   224498
```

```
79.3   2454508
```

```
80.6    93431
```

```
95.5   2392108
```

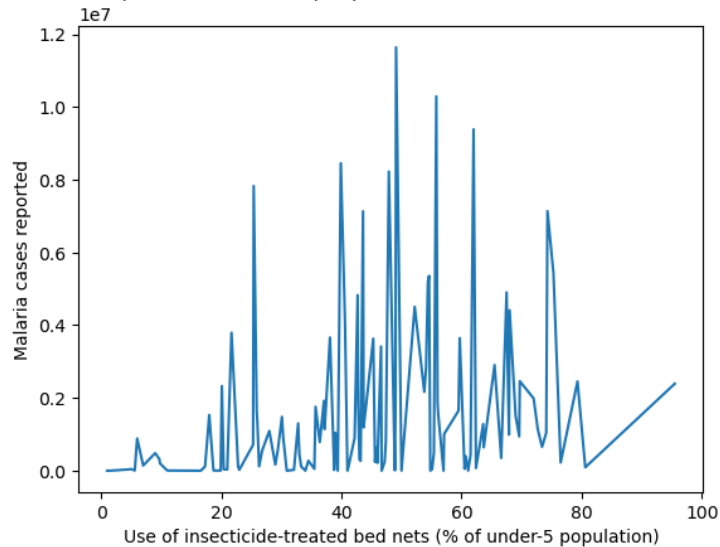
```
Name: Malaria cases reported, Length: 117, dtype: Int64
```

*** Visualisation de la tendance des cas de paludisme en Afrique pour enfants de -5 ans dormant sous moustiquaires imprégnés**

In [136]:

```
som_tendances_treated_bed_nets.plot(kind="line")
plt.xlabel("Use of insecticide-treated bed nets (% of under-5 population)")
plt.ylabel("Malaria cases reported")
plt.title("Figure 02 : Tendances des cas de paludisme en Afrique pour enfants de -5 ans dormant sous moustiquaires imprégnés")
plt.show()
```

Figure 02 : Tendances des cas de paludisme en Afrique pour enfants de -5 ans dormant sous moustiquaires imprégnés



Lecture de la Figure 02 : la tendance est que plus les enfants de moins de 5 ans dorment sous moustiquaires imprégnés, moins ils ont le paludisme.

4.3. Corrélations entre les variables pour analyser les relations entre les facteurs de risque et les cas de paludisme

Definition : La relation statistique entre deux variables est appelée leur corrélation.

In [138]:

```
correlation_matrix = dfc.corr()
```

C:\Users\HP_ZBOOK\AppData\Local\Temp\ipykernel_28904\2508308341.py: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.

```
correlation_matrix = dfc.corr()
```

In [139]:

```
correlation_matrix
```

Out[139]:

	Year	Incidence of malaria (per 1,000 population at risk)	Malaria cases reported	Use of insecticide-treated bed nets (% of under-5 population)	% of children under age 5 with fever receiving antimalarial drugs	% of pregnant women using Intermittent preventive treatment (IPT) of malaria in pregnancy	% of population using safely managed drinking water services	% of rural population using safely managed drinking water services	% of urban population using safely managed drinking water services	% of population using safely managed sanitation services	...	Urban population (% of total population)	po
Year	1.000000e+00	-0.070605	0.371363	0.531014	-0.043820	0.454461	0.089462	0.171102	0.050082	0.081409	...	0.084292	-C
Incidence of malaria (per 1,000 population at risk)	-7.060502e-02	1.000000	0.288509	0.009608	0.532327	0.154911	-0.501461	-0.459303	-0.678977	-0.613592	...	-0.253803	C
Malaria cases reported	3.713634e-01	0.288509	1.000000	0.271116	0.301719	0.238335	-0.278893	-0.197603	-0.408585	-0.172106	...	-0.219683	C
Use of insecticide-treated bed nets (% of under-5 population)	5.310142e-01	0.009608	0.271116	1.000000	0.079162	0.307991	0.051592	-0.099520	-0.121495	-0.293166	...	-0.136613	C
% of children under age 5 with fever receiving antimalarial drugs	-4.381980e-02	0.532327	0.301719	0.079162	1.000000	0.314459	-0.208050	-0.156504	-0.616924	-0.153175	...	-0.090201	C
% of pregnant women using Intermittent preventive treatment (IPT) of malaria in pregnancy	4.544615e-01	0.154911	0.238335	0.307991	0.314459	1.000000	0.184018	-0.272432	0.123014	-0.201184	...	0.178811	-C
% of population using safely managed drinking water services	8.946214e-02	-0.501461	-0.278893	0.051592	-0.208050	0.184018	1.000000	0.921974	0.946457	0.927503	...	0.822392	-C
% of rural population using safely managed drinking water services	1.711018e-01	-0.459303	-0.197603	-0.099520	-0.156504	-0.272432	0.921974	1.000000	0.814306	0.951713	...	0.695415	-C
% of urban population using safely managed drinking water services	5.008197e-02	-0.678977	-0.408585	-0.121495	-0.616924	0.123014	0.946457	0.814306	1.000000	0.904425	...	0.507907	-C
% of population using safely managed sanitation services	8.140907e-02	-0.613592	-0.172106	-0.293166	-0.153175	-0.201184	0.927503	0.951713	0.904425	1.000000	...	0.427114	-C
% of rural population using safely managed sanitation services	2.196390e-01	-0.205144	0.103046	0.038775	0.042544	-0.231655	0.950821	0.978063	-0.046646	0.650613	...	0.588600	-C
% of urban population using safely managed sanitation services	5.671254e-02	-0.499475	-0.106211	-0.198089	0.218336	-0.112146	0.898606	0.977609	0.947106	0.971281	...	0.406288	-C
Rural population (% of total population)	-8.429034e-02	0.253799	0.219674	0.136593	0.090178	-0.178830	-0.822383	-0.695418	-0.507889	-0.427094	...	-1.000000	C

	Year	Incidence of malaria (per 1,000 population at risk)	Malaria cases reported	Use of insecticide-treated bed nets (% of under-5 population)	% of children under age 5 with fever receiving antimalarial drugs	% of pregnant women using Intermittent preventive treatment (IPT) of malaria in pregnancy	% of population using safely managed drinking water services	% of rural population using safely managed drinking water services	% of urban population using safely managed drinking water services	% of population using safely managed sanitation services	...	Urban population (% of total population)	po (ar
Rural population growth (annual %)	-4.988517e-02	0.411329	0.254232	0.198001	0.289761	-0.043661	-0.826812	-0.826590	-0.729787	-0.368766	...	-0.651653	C
Urban population (% of total population)	8.429226e-02	-0.253803	-0.219683	-0.136613	-0.090201	0.178811	0.822392	0.695415	0.507907	0.427114	...	1.000000	-C
Urban population growth (annual %)	-5.670150e-02	0.350415	0.284905	0.214265	0.363973	-0.012669	-0.856482	-0.723214	-0.654459	-0.618049	...	-0.228287	1
% of population using at least basic drinking water services	1.194164e-01	-0.410028	-0.257250	0.034067	-0.207770	0.241572	0.870408	0.754185	0.735460	0.666937	...	0.631472	-C
% of rural population using at least basic drinking water services	1.138409e-01	-0.352018	-0.199931	0.146298	-0.154126	0.186519	0.774201	0.592671	0.661666	0.649492	...	0.290278	-C
% of urban population using at least basic drinking water services	1.034695e-01	-0.430043	-0.177815	0.004874	-0.394741	0.140941	0.936218	0.821300	0.942218	0.656757	...	0.280251	-C
% of population using at least basic sanitation services	6.887078e-02	-0.512391	-0.202425	-0.042730	-0.461137	-0.190158	0.886794	0.932769	0.742667	0.677943	...	0.518670	-C
% of rural population using at least basic sanitation services	7.023027e-02	-0.431779	-0.128802	0.057677	-0.396981	-0.224373	0.837097	0.867757	0.656613	0.703158	...	0.281079	-C
% of urban population using at least basic sanitation services	6.026910e-02	-0.479482	-0.231405	-0.150451	-0.463726	-0.172149	0.873025	0.925027	0.749673	0.736721	...	0.465425	-C

Après les corrélations et identification des facteurs de risque potentiels en examinant les valeurs de corrélation

*** Corrélation entre les cas de paludisme et les variables démographiques**

lat	1.512567e-15	0.000000	0.000000	0.000000	0.000000	-0.098717	0.809231	0.761962	-0.309760	0.555364	...	0.227344	C
longitude	-1.679748e-15	-0.287883	0.072659	-0.011117	-0.110809	-0.203676	-0.321583	-0.520107	-0.050205	0.317706	...	-0.305732	-C

- Avec incidence du paludisme (pour 1 000 habitants à risque)

24 rows x 24 columns

In [140]:

```
corr_cases_population_1 = correlation_matrix["Malaria cases reported"]["Incidence of malaria (per 1,000 population at risk)"]
```

In [141]:

```
corr_cases_population_1
```

Out[141]:

0.28850886831479716

- Avec % d'enfants de moins de 5 ans ayant de la fièvre recevant des médicaments antipaludiques

In [142]:

```
rr_cases_population_2 = correlation_matrix["Malaria cases reported"]["% of children under age 5 with fever receiving antimalarial drugs"]
```

In [143]:

```
corr_cases_population_2
```

Out[143]:

0.30171947845151337

- Avec l'utilisation de moustiquaires imprégnées d'insecticide (% de la population de moins de 5 ans)

In [144]:

```
corr_cases_population_3 = correlation_matrix["Malaria cases reported"]["Use of insecticide-treated bed nets (% of under-5 population)"]
```

In [145]:

```
corr_cases_population_3
```

Out[145]:

```
0.2711160562803437
```

- Avec la Population urbaine (% de la population totale)

In [146]:

```
corr_cases_population_4 = correlation_matrix["Malaria cases reported"]["Urban population growth (annual %)"]
```

In [147]:

```
corr_cases_population_4
```

Out[147]:

```
0.28490501349417036
```

Forte Corrélation entre "% of rural population using safely managed sanitation services" et "% of rural population using safely managed drinking water services"

In [148]:

```
x["% of rural population using safely managed sanitation services"]["% of rural population using safely managed drinking water services"]
```

In [149]:

```
corr_cases_population_5
```

Out[149]:

```
0.9780628308867829
```

Machine Learning : Modélisation des données pour identifier des modèles et des relations importantes dans les données

In [184]:

```
## Utilisons La bibliothèque scikit-Learn pour effectuer une régression linéaire et importons Le modèle de régression linéaire
from sklearn.linear_model import LinearRegression
## Le module Impute de Sklearn (scikit-Learn) permet de nettoyer notre dataset des valeurs manquantes qui le composent.
## SimpleImputer remplace toute valeur manquante par une statistique ou une constante donnée
from sklearn.impute import SimpleImputer
```

In [201]:

```
## X représente des variables indépendantes : dans notre cas "% d'enfants de moins de 5 ans dormant sous moustiquaires imprégnées" et "% d'enfants de moins de 5 ans utilisant des services d'assainissement gérés en toute sécurité"
X = dfc[["Use of insecticide-treated bed nets (% of under-5 population)", "% of pregnant women using Intermittent preventive treatment (IPT) for malaria prevention"]]
## y représente la variable cible : dans notre cas "Year (Année)"
y = dfc["Year"]
```

In [203]:

```
## Utilisons SimpleImputer pour remplacer toute valeur manquante par une statistique ou une constante donnée. Dans notre cas, la moyenne de la variable cible
imputer = SimpleImputer(strategy='mean', missing_values=np.nan)
## Transformons dans les variables indépendantes contenues dans X, les valeurs manquantes par np.nan
X_imputed = imputer.fit_transform(X)
```

In [206]:

```
## Utilisons SimpleImputer pour remplacer toute valeur manquante par une statistique ou une constante donnée. Dans notre cas, la moyenne de la variable cible
imputer_y = SimpleImputer(strategy='mean')
## Appelons la méthode `reshape` sur `y.values`, pour remodeler le tableau en un tableau 2D avec une seule colonne,
## ce qui est le format d'entrée attendu pour la méthode `fit_transform` de la classe `SimpleImputer`.
y_imputed = imputer_y.fit_transform(y.values.reshape(-1, 1))
```

In [207]:

```
## Instancions ensuite un objet modèle de régression linéaire en utilisant la classe LinearRegression().
model = LinearRegression()
## À l'aide de la méthode fit(), entraînons le modèle en utilisant les données X et y.
model.fit(X_imputed, y_imputed)
## Utilisons la méthode predict() pour effectuer des prédictions sur les données d'entraînement X et stockez les prédictions
## résultantes dans la variable predictions.
predictions = model.predict(X_imputed)
```

Machine Learning : Evaluation du modèle de Regression Linéaire

In [210]:

```
## Importation de la fonction "mean_squared_error" du module "sklearn.metrics". Cela permettra de calculer l'erreur
## quadratique moyenne ou Mean Squared Error (MSE).
from sklearn.metrics import mean_squared_error

## Calcul de l'erreur quadratique moyenne (MSE) entre les valeurs réelles 'y' et les valeurs prédites 'predictions'
## en utilisant la fonction 'mean_squared_error'.
mse = mean_squared_error(y, predictions)

## Calcul de la racine carrée de l'erreur quadratique moyenne (MSE) en utilisant la fonction 'np.sqrt' de numpy.
rmse = np.sqrt(mse)

## Affichage de la valeur du RMSE en utilisant la fonction 'print'
## Habituellement, un score RMSE inférieur à 180 est considéré comme un bon score pour un algorithme qui fonctionne modérément
## ou bien. Dans le cas où la valeur RMSE dépasse 180, nous devons effectuer une sélection de caractéristiques et un réglage
## des paramètres hyper sur les paramètres du modèle.
print("RMSE:", rmse)
```

RMSE: 3.0520908649792458

Conclusion RMSE : Le résultat du RMSE est inférieur à 180, donc notre algorithme de Machine Learning fonctionne bien

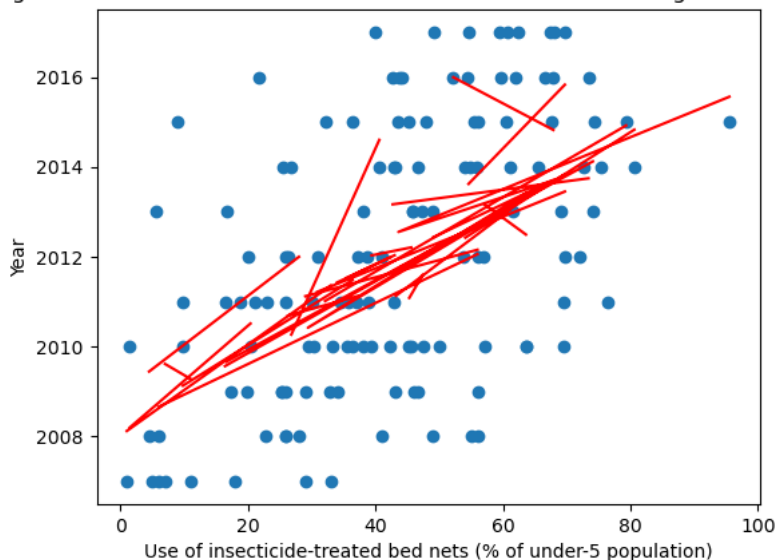
Machine Learning : Communication des résultats

Visualisation des résultats

In [214]:

```
plt.scatter(X['Use of insecticide-treated bed nets (% of under-5 population)'], y)
plt.plot(X['Use of insecticide-treated bed nets (% of under-5 population)'], predictions, color='red')
plt.xlabel('Use of insecticide-treated bed nets (% of under-5 population)')
plt.ylabel('Year')
plt.title("Figure 03 : Visualisation des résultats de notre modèle de Regression Linéaire")
plt.show()
```

Figure 03 : Visualisation des résultats de notre modèle de Regression Linéaire



Lecture de la Figure 03 : le taux d'enfants de moins de 5 ans dormant sous moustiquaires imprégnés, continuera à augmenter d'année en année afin de réduire le nombre de cas de paludisme.

In []: