

# Population Census In Ghana - Verification of Zipf's law

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## Data from World Population Ranking

```
In [5]: import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline

df = pd.read_csv('realTime.csv')
data = pd.DataFrame(df)
data.index = data['Country Name']
data.drop('Country Name', axis=1, inplace=True)
data.drop('Indicator Name', axis=1, inplace=True)
data.drop('Indicator Code', axis=1, inplace=True)
data.drop('Country Code', axis=1, inplace=True)
data
```

Out[5]:

	1960	1961	1962	1963	1964	1965	1966	196
Country Name								
Aruba	54211.0	55438.0	56225.0	56695.0	57032.0	57360.0	57715.0	58055.
Afghanistan	8996973.0	9169410.0	9351441.0	9543205.0	9744781.0	9956320.0	10174836.0	10399926.
Angola	5454933.0	5531472.0	5608539.0	5679458.0	5735044.0	5770570.0	5781214.0	5774243.
Albania	1608800.0	1659800.0	1711319.0	1762621.0	1814135.0	1864791.0	1914573.0	1965598.
Andorra	13411.0	14375.0	15370.0	16412.0	17469.0	18549.0	19647.0	20758.
...	...	...	...	...	...	...	...	.
Kosovo	947000.0	966000.0	994000.0	1022000.0	1050000.0	1078000.0	1106000.0	1135000.
Yemen, Rep.	5315355.0	5393036.0	5473671.0	5556766.0	5641597.0	5727751.0	5816247.0	5907874.
South Africa	17099840.0	17524533.0	17965725.0	18423161.0	18896307.0	19384841.0	19888250.0	20406864.
Zambia	3070776.0	3164329.0	3260650.0	3360104.0	3463213.0	3570464.0	3681955.0	3797873.
Zimbabwe	3776681.0	3905034.0	4039201.0	4178726.0	4322861.0	4471177.0	4623351.0	4779827.

264 rows x 60 columns

Imported dataFrame with slight edits for Original Documents

File will be uploaded to my github repository for further integrations

This is the population growth per annum in Ghana

In [6]:

```
data.loc['Ghana']
```

Out[6]:

1960	6635230.0
1961	6848295.0
1962	7071971.0
1963	7300116.0
1964	7524472.0
1965	7739473.0
1966	7941412.0
1967	8132803.0
1968	8321770.0
1969	8520015.0
1970	8735495.0
1971	8973244.0
1972	9229631.0
1973	9493556.0
1974	9749104.0
1975	9985946.0
1976	10199165.0
1977	10395452.0
1978	10590261.0
1979	10805314.0
1980	11056116.0
1981	11348289.0
1982	11676823.0
1983	12033575.0
1984	12405660.0
1985	12783613.0
1986	13164837.0
1987	13552021.0
1988	13947042.0
1989	14353410.0
1990	14773277.0
1991	15207367.0
1992	15653336.0
1993	16106765.0
1994	16561674.0
1995	17014057.0
1996	17462496.0
1997	17908985.0
1998	18357156.0
1999	18812359.0
2000	19278856.0
2001	19756928.0
2002	20246381.0
2003	20750299.0
2004	21272323.0
2005	21814642.0
2006	22379055.0
2007	22963946.0
2008	23563825.0
2009	24170940.0
2010	24779619.0
2011	25387710.0
2012	25996449.0
2013	26607642.0
2014	27224472.0
2015	27849205.0
2016	28481946.0
2017	29121471.0
2018	29767108.0

Unnamed: 63                      NaN  
Name: Ghana, dtype: float64

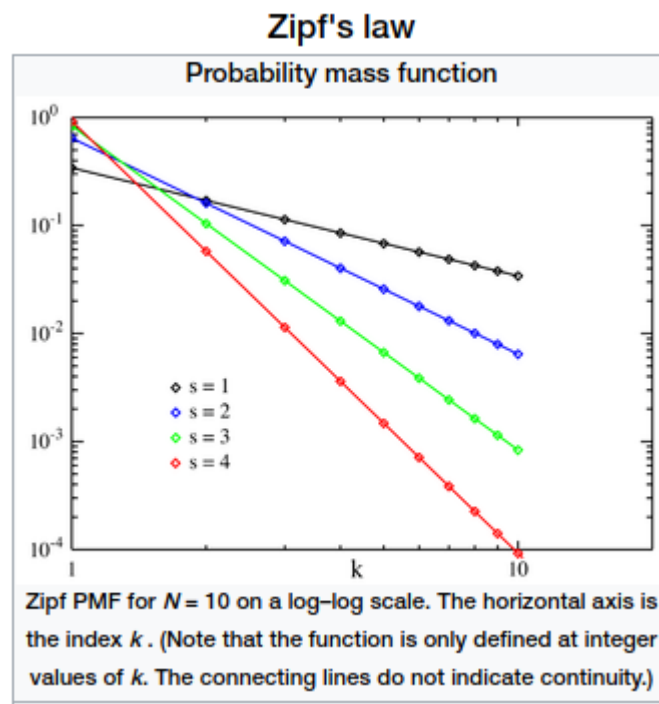
Zipf's law then predicts that out of a population of  $N$  elements, the normalized frequency of elements of rank  $k$ ,  $f(k; s, N)$ , is:

$$f(k; s, N) = \frac{1/k^s}{\sum_{n=1}^N (1/n^s)}$$

Zipf's law holds if the number of elements with a given frequency is a random variable with power law distribution  $p(f) = \alpha f^{-1-1/s}$ .<sup>[10]</sup>

# This can be used in population growth and probabilistic analysis

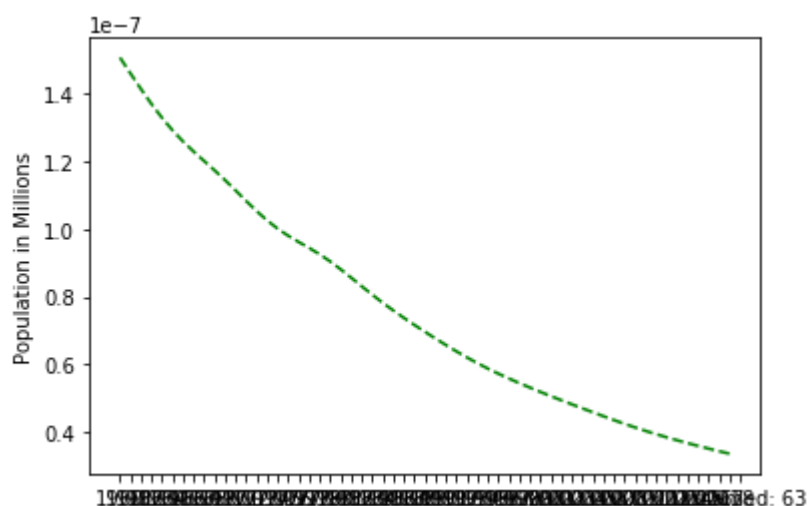
With an output graph similar to this below.



# GHANAIAN populis under the scope of the Zipf Law

```
In [7]: zipf = 1 / data.loc['Ghana']
plt.plot(zipf, 'g--')
plt.ylabel("Population in Millions")
# size is inversely proportional to its rank
```

```
Out[7]: Text(0, 0.5, 'Population in Millions')
```



# Original plot of Population against Years

```
In [11]: plt.plot(data.loc['Ghana'])  
plt.xlabel("Years")  
plt.ylabel("Population in Millions")
```

```
Out[11]: Text(0, 0.5, 'Population in Millions')
```

