

# EDA ON REFUGE DATA BY CHUKWUEMEKA

June 30, 2021

## 1 GLOBAL FOOD PRICE PROJECT PROPOSAL

### 1.1 INTRODUCTION

Global Food Price is described as the average price of food commodities across countries, regions, and the global level. The level of food price usually depends on the food production process which includes food marketing and distribution. Fluctuations in the price of commodities can be multifactorial ranging from availability of natural resources for agriculture, market demand, energy cost, cost of production, exchange rate, government policy, and weather events amongst all. These factors have both positive and negative effects.

Starting from 2007-08 there was a surge in food prices, particularly in developing countries. This led to a global crisis causing political and economic instability as well as social unrest between poor and developed countries. The trend dropped and increased again in 2009 and 2010, reaching new heights in 2011 & 2012. Over the years prices dropped significantly reaching a lower point in March 2016 with a reduced Food and Agricultural Organization (FAO) food price index. The FAO Food Price Index (FFPI) is a measure of the monthly change in international prices of a basket of food commodities. In recent times, global food prices rose in March 2021, which marked the 10th consecutive monthly increase with products like vegetable oil and dairy products leading the rise.

It is needful to say that at this point, the impact of food prices not only provides an indicator of the balance of agricultural produce and market demands but also has an impact on the cost of living, food policies, and migration. While the producers benefit from the high food prices, consumers only benefit when the food prices are low. By implication, food prices now have an impact on food affordability, quality of a diet, undernourishment, and hunger.

In line with the United Nations Development Programme's (UNDP) sustainable development goal 2, i.e., zero hunger we will be taking a closer look at the trends in global food prices, possible causes and effects of increased global food prices and offer our solution.

### 1.2 PROBLEM STATEMENT

It can be observed from the previous discussion that global food price fluctuations can cause famine and large population shift. Hence, Identifying the drivers of global food prices and predicting future changes in global food prices, could help in understanding food prices and its causal effects.

### 1.3 AIM

Our research aims to understand and analyze fluctuations in global food prices and pair the outcome with currency fluctuations, weather patterns, and refugee movements. This will help us to build an end-to-end analysis and a food price prediction engine that will help the Government make better

decisions on food policy adjustments, International bodies with planning of food aid programmes, Individuals with planning and productivity in the advent of a potential food price crisis...

## **1.4 OBJECTIVES**

To achieve the above aim, we will: Analyze available datasets to observe and make inferences about changing food prices, fluctuations, and the trend they follow. Attempt to compare their correlation with factors such as currency fluctuation, weather patterns, and refugee movements. Investigate which food item controls the trends of the majority of the food markets. Use the best-performed model in predicting food prices and deploying it in a web application that can predict food prices.

## **1.5 REVIEW OF PAST LITERATURE**

Most of the recent research on Global Food Prices has centered around policy-making across nations and countries in addressing the issue. An article by ALNAP, it cited IFPRI/CGIAR, 2008 where it was stated that factors that have contributed to the global food price crisis are either cyclical, structural, or unique. Various World Organizations like WFP, UNOCHA/CERF, UNICEF, IMF, WORLD BANK, NEPAD, ADB, AU, WTO, etc have championed different policies towards mitigating the menace of the Global Food Prices crisis, especially through financial aids. Notable among them are FAO's Procurement and distribution of seeds, fertilizers, and other inputs which have been carried out in 54 countries under the Food and Agriculture Organisation (FAO) Initiative on Soaring Food Prices (ISFP). FAO is also urging governments and the International community to implement measures in support of poor countries hard hit by food price increases, specifically to provide small farmers with improved access to inputs like seeds and fertilizers to increase local crop production (RHVP/Wahenga brief, 2008).

From a micro perspective, Nigeria as a country has had several policies both in present and in the past regarding mitigation of the food prices crisis. Policies like Operation Feed the Nation, Green Revolution, and presently FADAMA programs. These policies and programs have contributed little or none to solving the challenge of the food price crisis.

With regards to predictive modeling technique, Artificial Neural Network(ANN) algorithm and Time Series Forecasting algorithms like ARIMA have been used recently by researchers in this Global Food prices crisis domain. A Machine Learning Approach to Forecasting Consumer Food Prices, J. Jay Harris(2017), applied ANN in modeling Global Food Prices, which was significantly insightful. In this project, we shall also be exploring predictive modeling techniques as well as time series models in forecasting food prices.

## **1.6 DATA COLLECTION**

Data related to global food prices will be collected from the Open source database compiled by the World Food Programme and distributed by the Humanitarian Data Exchange. Data on currency fluctuations will be gotten from the World Bank's open-source database on official exchange rates. Data on Refugee movements will be extracted from the Refugee statistics of the United Nations High Commissioner for Refugees. Data on Weather patterns will be excerpts from the World Meteorological Organization.

## 1.7 MACHINE LEARNING WORKFLOW

Data Volumes ↓ Data Ingestion ↓ Data Wrangling ↓ Data Cleaning ↓ Data preprocessing → Stationarity check → Time series modeling → forecasting ↓ Predictive modeling

## 1.8 WEB APPLICATION DEVELOPMENT FOR THE MODEL

The end product of this Global food prediction engine will be in the form of a web app that can be accessed from anywhere as long as there is an Internet connection, It will have a drop-down list to select the food categories, and a graph showing the trend of the price fluctuations over the years and the prediction over the next couple of months. The web app will be built using the streamlit service which makes deploying models quick and easy. The model which would have been worked on and perfected is saved as a pickle file and a python script is created for the usage of the model, then using streamlit, the interface stated above is created in python, then connected and deployed for use.

## 1.9 References:

“World Food Situation”. FAO. Archived from the original on 29 April 2011. Retrieved 24 April 2011. How do Food Prices Affect Producers and Consumers in Developing Countries?, ICTSD, Information Note Number 10, September 2009 UN Food and Agriculture Organization (2009). The State of Food Insecurity in the World 2009. Rome. Rahman, M. Mizanur (11 August 2011). “Food price inflation: Global and national problem”. The Daily Star. “FAO Food Price Index”. FAO. Retrieved 2 May 2017.

## 1.10 Group Trailblazers:

- Abiona Oluwafemi
- Roqeebat Olanrewaju
- Omeh Chukwuemeka
- Habeebullah Agbaje

## 1.11 Terms

### 1.11.1 Who is a refugee?

*Refugees are people who have fled war, violence, conflict or persecution and have crossed an international border to find safety in another country. They often have had to flee with little more than the clothes on their back, leaving behind homes, possessions, jobs and loved ones. Refugees are defined and protected in international law. The 1951 Refugee Convention is a key legal document and defines a refugee as: “someone who is unable or unwilling to return to their country of origin owing to a well-founded fear of being persecuted for reasons of race, religion, nationality, membership of a particular social group, or political opinion.”* [link](#)

A refugee is a person who has fled their own country because they are at risk of serious human rights violations and persecution there. The risks to their safety and life were so great that they felt they had no choice but to leave and seek safety outside their country because their own government cannot or will not protect them from those dangers. Refugees have a right to international protection.

### 1.11.2 Who is an asylum-seeker?

An asylum-seeker is a person who has left their country and is seeking protection from persecution and serious human rights violations in another country, but who hasn't yet been legally recognized as a refugee and is waiting to receive a decision on their asylum claim. Seeking asylum is a human right. This means everyone should be allowed to enter another country to seek asylum.

### 1.11.3 Who is a migrant?

There is no internationally accepted legal definition of a migrant. Like most agencies and organizations, we at Amnesty International understand migrants to be people staying outside their country of origin, who are not asylum-seekers or refugees.

Some migrants leave their country because they want to work, study or join family, for example. Others feel they must leave because of poverty, political unrest, gang violence, natural disasters or other serious circumstances that exist there.

## 2 Import Libraries

```
[1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import scipy
```

## 3 Setting display layout

```
[2]: pd.set_option("display.max_column", None)
pd.set_option("display.max_colwidth", None)
pd.set_option("display.max_row", None)
pd.set_option("display.float_format", lambda x: "%.2f" %x)
plt.style.use('ggplot')
plt.rcParams['font.size'] = 10
```

```
[3]: temp = pd.read_csv("https://raw.githubusercontent.com/Hab-eeb/Global_Food_Price/
↪main/Datasets/GlobalTemperatures.csv")
print(f"The number of rows is: {temp.shape[0]} and numbers of columns is: {temp.
↪shape[1]}")
```

The number of rows is: 3192 and numbers of columns is: 9

```
[4]: temp.head()
```

```
[4]:      dt  LandAverageTemperature  LandAverageTemperatureUncertainty  \
0  1750-01-01                3.03                      3.57
1  1750-02-01                3.08                      3.70
2  1750-03-01                5.63                      3.08
```

3	1750-04-01	8.49	2.45
4	1750-05-01	11.57	2.07

	LandMaxTemperature	LandMaxTemperatureUncertainty	LandMinTemperature	\
0	NaN	NaN	NaN	
1	NaN	NaN	NaN	
2	NaN	NaN	NaN	
3	NaN	NaN	NaN	
4	NaN	NaN	NaN	

	LandMinTemperatureUncertainty	LandAndOceanAverageTemperature	\
0	NaN	NaN	
1	NaN	NaN	
2	NaN	NaN	
3	NaN	NaN	
4	NaN	NaN	

	LandAndOceanAverageTemperatureUncertainty
0	NaN
1	NaN
2	NaN
3	NaN
4	NaN

```
[5]: ref = pd.read_csv("populations_countries.csv",delimiter = ',')
print(f"The number of rows is: {ref.shape[0]} and numbers of columns is: {ref.
↪shape[1]}")
```

The number of rows is: 90004 and numbers of columns is: 11

```
[6]: ref.head()
```

	Year	Country of origin	Country of origin (ISO)	\
0	2000	Afghanistan	AFG	
1	2000	Iraq	IRQ	
2	2000	Serbia and Kosovo: S/RES/1244 (1999)	SRB	
3	2000	Turkey	TUR	
4	2000	Chad	TCD	

	Country of asylum	Country of asylum (ISO)	Refugees under UNHCR's mandate	\
0	Afghanistan	AFG	0	
1	Albania	ALB	9	
2	Albania	ALB	507	
3	Albania	ALB	5	
4	Algeria	DZA	20	

	Asylum-seekers	IDPs of concern to UNHCR	Venezuelans displaced abroad	\
--	----------------	--------------------------	------------------------------	---

0	0	758625	NaN
1	0	0	NaN
2	5	0	NaN
3	0	0	NaN
4	19	0	NaN

	Stateless persons	Others of concern
0	0	0
1	0	0
2	0	0
3	0	0
4	0	0

```
[7]: """
=====No duplicate Entries=====
"""
dp = ref.duplicated()
print(f"The number of duplicated row and column are: {ref[dp].shape[0]} and
↳{ref[dp].shape[1]} respectively")
ref.duplicated().sum()
```

The number of duplicated row and column are: 0 and 11 respectively

```
[7]: 0
```

```
[8]: ref.columns
```

```
[8]: Index(['Year', 'Country of origin', 'Country of origin (ISO)',
        'Country of asylum', 'Country of asylum (ISO)',
        'Refugees under UNHCR's mandate', 'Asylum-seekers',
        'IDPs of concern to UNHCR', 'Venezuelans displaced abroad',
        'Stateless persons', 'Others of concern'],
        dtype='object')
```

```
[9]: ref.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 90004 entries, 0 to 90003
Data columns (total 11 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Year                                90004 non-null  int64
1   Country of origin                    90004 non-null  object
2   Country of origin (ISO)              88989 non-null  object
3   Country of asylum                    90004 non-null  object
4   Country of asylum (ISO)              90004 non-null  object
5   Refugees under UNHCR's mandate       90004 non-null  int64
6   Asylum-seekers                     90004 non-null  int64
```

```

7   IDPs of concern to UNHCR          90004 non-null  int64
8   Venezuelans displaced abroad      59 non-null    float64
9   Stateless persons                 90004 non-null  int64
10  Others of concern                 90004 non-null  int64
dtypes: float64(1), int64(6), object(4)
memory usage: 7.6+ MB

```

```
[10]: ref.dtypes
```

```

[10]: Year                                int64
      Country of origin                   object
      Country of origin (ISO)             object
      Country of asylum                   object
      Country of asylum (ISO)             object
      Refugees under UNHCR's mandate      int64
      Asylum-seekers                    int64
      IDPs of concern to UNHCR            int64
      Venezuelans displaced abroad        float64
      Stateless persons                    int64
      Others of concern                    int64
      dtype: object

```

```

[11]: ref_obj = ref.select_dtypes(include = 'object')
      ref_int = ref.select_dtypes(include = 'int')
      ref_float = ref.select_dtypes(include = 'float')
      ref_obj.shape, ref_int.shape, ref_float.shape

```

```
[11]: ((90004, 4), (90004, 6), (90004, 1))
```

```
[12]: ref.nunique().sort_values(ascending = False)
```

```

[12]: Refugees under UNHCR's mandate    6996
      Asylum-seekers                 3915
      Stateless persons                 675
      Others of concern                 640
      IDPs of concern to UNHCR         454
      Country of origin                 212
      Country of origin (ISO)           211
      Country of asylum                 189
      Country of asylum (ISO)           189
      Venezuelans displaced abroad       59
      Year                             21
      dtype: int64

```

```
[13]: ref.describe(include = 'all')
```

```

[13]:      Year Country of origin Country of origin (ISO) \
count 90004.00          90004          88989
unique      NaN          212          211
top        NaN      Somalia          SOM
freq       NaN          1990          1990
mean    2011.07          NaN          NaN
std        6.02          NaN          NaN
min     2000.00          NaN          NaN
25%     2006.00          NaN          NaN
50%     2012.00          NaN          NaN
75%     2016.00          NaN          NaN
max     2020.00          NaN          NaN

      Country of asylum Country of asylum (ISO) \
count          90004          90004
unique          189          189
top    United States of America          USA
freq          3572          3572
mean          NaN          NaN
std          NaN          NaN
min          NaN          NaN
25%          NaN          NaN
50%          NaN          NaN
75%          NaN          NaN
max          NaN          NaN

      Refugees under UNHCR's mandate Asylum-seekers \
count          90004.00          90004.00
unique          NaN          NaN
top            NaN          NaN
freq           NaN          NaN
mean          3077.74          393.77
std          46313.22          5586.88
min            0.00          0.00
25%            5.00          0.00
50%           14.00          6.00
75%           103.00          41.00
max          3641370.00          940668.00

      IDPs of concern to UNHCR Venezuelans displaced abroad \
count          90004.00          59.00
unique          NaN          NaN
top            NaN          NaN
freq           NaN          NaN
mean          4881.33          170025.02
std          124509.91          358009.88
min            0.00          11.00

```



25%	0.00	6568.00
50%	0.00	25686.00
75%	0.00	142565.50
max	8252788.00	1771237.00

	Stateless persons	Others of concern
count	90004.00	90004.00
unique	NaN	NaN
top	NaN	NaN
freq	NaN	NaN
mean	732.52	362.10
std	26468.49	16814.69
min	0.00	0.00
25%	0.00	0.00
50%	0.00	0.00
75%	0.00	0.00
max	3500000.00	2351313.00

```
[14]: plt.figure(figsize = (15,8))
ref_corr = ref.corr()
sns.heatmap(ref_corr, annot = True, square = True)
plt.show()
```



```
[15]: ref["Country of origin"].unique()
```

```
[15]: array(['Afghanistan', 'Iraq', 'Serbia and Kosovo: S/RES/1244 (1999)',
        'Turkey', 'Chad', 'Cameroon', 'Congo', 'Dem. Rep. of the Congo',
        'Palestinian', 'Guinea', 'Liberia', 'Libya', 'Mali', 'Morocco',
        'Nigeria', 'Rwanda', 'Sierra Leone', 'Somalia', 'Sudan',
        'Syrian Arab Rep.', 'Western Sahara', 'Unknown ', 'Angola',
        'Burundi', 'Comoros', 'Guinea-Bissau', 'United Rep. of Tanzania',
        'Zambia', 'Djibouti', 'Eritrea', 'Ethiopia', 'Russian Federation',
        'Yemen', 'Stateless', 'Albania', 'Algeria', 'Armenia', 'Benin',
        'Bangladesh', 'Bosnia and Herzegovina', 'Bulgaria', 'Chile',
        'Colombia', 'Cuba', 'Dominican Rep.', 'Ecuador', 'Estonia',
        'Georgia', 'Ghana', 'Haiti', 'India', 'Iran (Islamic Rep. of)',
        'Kazakhstan', 'Kyrgyzstan', 'Lao People's Dem. Rep.', 'Lebanon',
        'Sri Lanka', 'Nicaragua', 'Pakistan', 'Paraguay', 'Peru',
        'Romania', 'Senegal', 'Viet Nam', 'Tunisia', 'Ukraine',
```

```
'Azerbaijan', 'Egypt', 'Argentina', 'Austria', 'Bahrain',
'Belarus', 'Bolivia (Plurinational State of)', 'Brazil',
'Cambodia', 'China', 'Cyprus', 'Czechia', 'Fiji', 'France',
'United Kingdom of Great Britain and Northern Ireland', 'Germany',
'Guatemala', 'China, Hong Kong SAR', 'Croatia', 'Hungary',
'Indonesia', 'Israel', 'Italy', 'Jordan', 'Kenya', 'Rep. of Korea',
'Dem. People's Rep. of Korea', 'Kuwait', 'Lithuania', 'Latvia',
'North Macedonia', 'Rep. of Moldova', 'Mexico', 'Malaysia',
'Mongolia', 'Mauritius', 'Myanmar', 'Nepal', 'Niger',
'Philippines', 'Poland', 'Portugal', 'South Africa', 'El Salvador',
'Saudi Arabia', 'Singapore', 'Solomon Islands', 'Slovenia',
'Thailand', 'Timor-Leste', 'Tonga', 'United Arab Emirates',
'Uganda', 'Uruguay', 'Uzbekistan', 'Zimbabwe', 'Cote d'Ivoire',
'Tajikistan', 'Togo', 'Bhutan', 'Burkina Faso', 'Mauritania',
'Central African Rep.', 'Equatorial Guinea', 'Madagascar',
'Namibia', 'China, Macao SAR', 'Honduras', 'Antigua and Barbuda',
'Barbados', 'Belgium', 'Botswana', 'Costa Rica', 'Denmark',
'Dominica', 'Gabon', 'Gambia', 'Greece', 'Grenada', 'Guyana',
'Iceland', 'Ireland', 'Jamaica', 'Japan', 'Saint Lucia', 'Malawi',
'Mozambique', 'Malta', 'Netherlands', 'Oman', 'Panama', 'Qatar',
'Spain', 'Slovakia', 'Eswatini', 'Sweden', 'Switzerland',
'Trinidad and Tobago', 'United States of America',
'Saint Vincent and the Grenadines',
'Venezuela (Bolivarian Republic of)', 'Tibetan', 'Turkmenistan',
'Seychelles', 'Sao Tome and Principe', 'Papua New Guinea',
'Suriname', 'Tuvalu', 'Canada', 'Belize', 'Australia', 'Bahamas',
'Cabo Verde', 'Finland', 'Nauru', 'San Marino',
'Saint Kitts and Nevis', 'Samoa', 'Lesotho', 'Andorra',
'New Zealand', 'Norway', 'Micronesia (Federated States of)',
'Gibraltar', 'Turks and Caicos Islands', 'Kiribati', 'Maldives',
'Bermuda', 'Brunei Darussalam', 'New Caledonia', 'Monaco',
'Montenegro', 'Holy See', 'South Sudan', 'Niue', 'Palau',
'Cayman Islands', 'Marshall Islands', 'Curacao ', 'Guadeloupe',
'Vanuatu', 'French Guiana', 'Luxembourg', 'Liechtenstein',
'Anguilla', 'Martinique'], dtype=object)
```

## 4 Feature Engineering

- To create a continent and sub-region features of the country's of origin or country of asylumns

```
[16]: cont = pd.read_csv("countryContinent.csv")
      cont.shape
```

```
[16]: (249, 9)
```

```
[17]: cont.head()
```

```
[17]:      country code_2 code_3 country_code iso_3166_2 continent \
0    Afghanistan    AF    AFG           4 ISO 3166-2:AF    Asia
1    land Islands    AX    ALA        248 ISO 3166-2:AX    Europe
2      Albania     AL    ALB           8 ISO 3166-2:AL    Europe
3      Algeria     DZ    DZA          12 ISO 3166-2:DZ    Africa
4 American Samoa    AS    ASM          16 ISO 3166-2:AS  Oceania

      sub_region region_code sub_region_code
0    Southern Asia      142.00         34.00
1 Northern Europe      150.00        154.00
2 Southern Europe      150.00         39.00
3 Northern Africa         2.00         15.00
4      Polynesia         9.00         61.00
```

```
[18]: cont.nunique()
```

```
[18]: country          249
code_2              248
code_3              249
country_code        249
iso_3166_2          249
continent            5
sub_region           22
region_code          5
sub_region_code      22
dtype: int64
```

```
[19]: cont.continent.value_counts()
```

```
[19]: Africa          58
Americas           55
Asia               51
Europe             51
Oceania            25
Name: continent, dtype: int64
```

```
[20]: cont.sub_region.value_counts()
```

```
[20]: Caribbean          28
Eastern Africa         20
Western Asia           18
Western Africa         17
Northern Europe        16
Southern Europe        16
South America          14
South-Eastern Asia     11
Eastern Europe         10
```

Polynesia	10
Southern Asia	9
Western Europe	9
Middle Africa	9
Central America	8
Eastern Asia	8
Micronesia	7
Northern Africa	7
Central Asia	5
Southern Africa	5
Melanesia	5
Northern America	5
Australia and New Zealand	3

Name: sub\_region, dtype: int64

```
[21]: ref["Country of origin"].unique()
```

```
[21]: array(['Afghanistan', 'Iraq', 'Serbia and Kosovo: S/RES/1244 (1999)',
          'Turkey', 'Chad', 'Cameroon', 'Congo', 'Dem. Rep. of the Congo',
          'Palestinian', 'Guinea', 'Liberia', 'Libya', 'Mali', 'Morocco',
          'Nigeria', 'Rwanda', 'Sierra Leone', 'Somalia', 'Sudan',
          'Syrian Arab Rep.', 'Western Sahara', 'Unknown ', 'Angola',
          'Burundi', 'Comoros', 'Guinea-Bissau', 'United Rep. of Tanzania',
          'Zambia', 'Djibouti', 'Eritrea', 'Ethiopia', 'Russian Federation',
          'Yemen', 'Stateless', 'Albania', 'Algeria', 'Armenia', 'Benin',
          'Bangladesh', 'Bosnia and Herzegovina', 'Bulgaria', 'Chile',
          'Colombia', 'Cuba', 'Dominican Rep.', 'Ecuador', 'Estonia',
          'Georgia', 'Ghana', 'Haiti', 'India', 'Iran (Islamic Rep. of)',
          'Kazakhstan', 'Kyrgyzstan', 'Lao People's Dem. Rep.', 'Lebanon',
          'Sri Lanka', 'Nicaragua', 'Pakistan', 'Paraguay', 'Peru',
          'Romania', 'Senegal', 'Viet Nam', 'Tunisia', 'Ukraine',
          'Azerbaijan', 'Egypt', 'Argentina', 'Austria', 'Bahrain',
          'Belarus', 'Bolivia (Plurinational State of)', 'Brazil',
          'Cambodia', 'China', 'Cyprus', 'Czechia', 'Fiji', 'France',
          'United Kingdom of Great Britain and Northern Ireland', 'Germany',
          'Guatemala', 'China, Hong Kong SAR', 'Croatia', 'Hungary',
          'Indonesia', 'Israel', 'Italy', 'Jordan', 'Kenya', 'Rep. of Korea',
          'Dem. People's Rep. of Korea', 'Kuwait', 'Lithuania', 'Latvia',
          'North Macedonia', 'Rep. of Moldova', 'Mexico', 'Malaysia',
          'Mongolia', 'Mauritius', 'Myanmar', 'Nepal', 'Niger',
          'Philippines', 'Poland', 'Portugal', 'South Africa', 'El Salvador',
          'Saudi Arabia', 'Singapore', 'Solomon Islands', 'Slovenia',
          'Thailand', 'Timor-Leste', 'Tonga', 'United Arab Emirates',
          'Uganda', 'Uruguay', 'Uzbekistan', 'Zimbabwe', 'Cote d'Ivoire',
          'Tajikistan', 'Togo', 'Bhutan', 'Burkina Faso', 'Mauritania',
          'Central African Rep.', 'Equatorial Guinea', 'Madagascar',
          'Namibia', 'China, Macao SAR', 'Honduras', 'Antigua and Barbuda',
```

```
'Barbados', 'Belgium', 'Botswana', 'Costa Rica', 'Denmark',
'Dominica', 'Gabon', 'Gambia', 'Greece', 'Grenada', 'Guyana',
'Iceland', 'Ireland', 'Jamaica', 'Japan', 'Saint Lucia', 'Malawi',
'Mozambique', 'Malta', 'Netherlands', 'Oman', 'Panama', 'Qatar',
'Spain', 'Slovakia', 'Eswatini', 'Sweden', 'Switzerland',
'Trinidad and Tobago', 'United States of America',
'Saint Vincent and the Grenadines',
'Venezuela (Bolivarian Republic of)', 'Tibetan', 'Turkmenistan',
'Seychelles', 'Sao Tome and Principe', 'Papua New Guinea',
'Suriname', 'Tuvalu', 'Canada', 'Belize', 'Australia', 'Bahamas',
'Cabo Verde', 'Finland', 'Nauru', 'San Marino',
'Saint Kitts and Nevis', 'Samoa', 'Lesotho', 'Andorra',
'New Zealand', 'Norway', 'Micronesia (Federated States of)',
'Gibraltar', 'Turks and Caicos Islands', 'Kiribati', 'Maldives',
'Bermuda', 'Brunei Darussalam', 'New Caledonia', 'Monaco',
'Montenegro', 'Holy See', 'South Sudan', 'Niue', 'Palau',
'Cayman Islands', 'Marshall Islands', 'Curacao ', 'Guadeloupe',
'Vanuatu', 'French Guiana', 'Luxembourg', 'Liechtenstein',
'Anguilla', 'Martinique'], dtype=object)
```

```
[22]: ref["continent"] = ref["Country of origin"].map({'Afghanistan': 'Asia',
'Iraq': 'Asia',
'Serbia and Kosovo: S/RES/1244 (1999)': "Europe",
'Turkey': 'Asia',
'Chad': 'Africa',
'Cameroon': 'Africa',
'Congo': 'Africa',
'Dem. Rep. of the Congo': 'Africa',
'Palestinian': 'Asia',
'Guinea': 'Africa',
'Liberia': 'Africa',
'Libya': 'Africa',
'Mali': 'Africa',
'Morocco': 'Africa',
'Nigeria': 'Africa',
'Rwanda': 'Africa',
'Sierra Leone': 'Africa',
'Somalia': 'Africa',
'Sudan': 'Africa',
'Syrian Arab Rep.': "Asia",
'Western Sahara': 'Africa',
'Unknown ': "Unknown Continent",
'Angola': 'Africa',
'Burundi': 'Africa',
'Comoros': 'Africa',
'Guinea-Bissau': 'Africa',
'United Rep. of Tanzania': 'Africa',
```

'Zambia': 'Africa',  
'Djibouti': 'Africa',  
'Eritrea': 'Africa',  
'Ethiopia': 'Africa',  
'Russian Federation': "Europe",  
'Yemen': "Asia",  
'Stateless': "Asia",  
'Albania': "Europe",  
'Algeria': 'Africa',  
'Armenia': 'Asia',  
'Benin': 'Africa',  
'Bangladesh': 'Asia',  
'Bosnia and Herzegovina': "Europe",  
'Bulgaria': "Europe",  
'Chile': "Americas",  
'Colombia': "Americas",  
'Cuba': "Americas",  
'Dominican Rep.': "Americas",  
'Ecuador': "Americas",  
'Estonia': "Europe",  
'Georgia': 'Europe',  
'Ghana': 'Africa',  
'Haiti': 'Asia',  
'India': 'Asia',  
'Iran (Islamic Rep. of)': 'Asia',  
'Kazakhstan': "Asia",  
'Kyrgyzstan': "Asia",  
'Lao People's Dem. Rep.': "Asia",  
'Lebanon': "Asia",  
'Sri Lanka': 'Asia',  
'Nicaragua': "Americas",  
'Pakistan': 'Asia',  
'Paraguay': "Americas",  
'Peru': "Americas",  
'Romania': "Europe",  
'Senegal': 'Africa',  
'Viet Nam': "Asia",  
'Tunisia': 'Africa',  
'Ukraine': "Europe",  
'Azerbaijan': 'Asia',  
'Egypt': 'Africa',  
'Argentina': "Americas",  
'Austria': "Europe",  
'Bahrain': 'Asia',  
'Belarus': "Europe",  
'Bolivia (Plurinational State of)': "Americas",  
'Brazil': "Americas",

'Cambodia': 'Asia',  
'China': 'Asia',  
'Cyprus': 'Asia',  
'Czechia': "Europe",  
'Fiji': "Oceania",  
'France': "Europe",  
'United Kingdom of Great Britain and Northern Ireland': "Europe",  
'Germany': "Europe",  
'Guatemala': "Americas",  
'China': 'Asia',  
'Hong Kong SAR': 'Asia',  
'Croatia': "Europe",  
'Hungary': "Europe",  
'Indonesia': 'Asia',  
'Israel': 'Asia',  
'Italy': "Europe",  
'Jordan': 'Asia',  
'Kenya': 'Africa',  
'Rep. of Korea': 'Asia',  
'Dem. People's Rep. of Korea': 'Asia',  
'Kuwait': "Asia",  
'Lithuania': "Europe",  
'Latvia': "Europe",  
'North Macedonia': "Europe",  
'Rep. of Moldova': "Europe",  
'Mexico': "Americas",  
'Malaysia': 'Asia',  
'Mongolia': 'Asia',  
'Mauritius': "Africa",  
'Myanmar': 'Asia',  
'Nepal': 'Asia',  
'Niger': 'Africa',  
'Philippines': 'Asia',  
'Poland': "Europe",  
'Portugal': "Europe",  
'South Africa': 'Africa',  
'El Salvador': "Americas",  
'Saudi Arabia': 'Asia',  
'Singapore': 'Asia',  
'Solomon Islands': "Oceania",  
'Slovenia': "Europe",  
'Thailand': "Asia",  
'Timor-Leste': "Asia",  
'Tonga': "Oceania",  
'United Arab Emirates': 'Asia',  
'Uganda': 'Africa',  
'Uruguay': "Americas",



```
'Uzbekistan': "Asia",
'Zimbabwe': 'Africa',
"Cote d'Ivoire": 'Africa',
'Tajikistan': "Asia",
'Togo': 'Africa',
'Bhutan': "Asia",
'Burkina Faso': 'Africa',
'Mauritania': "Africa",
'Central African Rep.': 'Africa',
'Equatorial Guinea': 'Africa',
'Madagascar': 'Africa',
'Namibia': 'Africa',
'China, Macao SAR': "Asia",
'Honduras': "Americas",
'Antigua and Barbuda': "Americas",
'Barbados': "Americas",
'Belgium': "Europe",
'Botswana': 'Africa',
'Costa Rica': "Americas",
'Denmark': "Europe",
'Dominica': "Americas",
'Gabon': 'Africa',
'Gambia': 'Africa',
'Greece': "Europe",
'Grenada': "Americas",
'Guyana': "Americas",
'Iceland': "Europe",
'Ireland': "Europe",
'Jamaica': "Americas",
'Japan': 'Asia',
'Saint Lucia': "Americas",
'Malawi': "Africa",
'Mozambique': 'Africa',
'Malta': "Europe",
'Netherlands': "Europe",
'Oman': "Asia",
'Panama': "Americas",
'Qatar': 'Asia',
'Spain': "Europe",
'Slovakia': "Europe",
'Eswatini': "Africa",
'Sweden': "Europe",
'Switzerland': "Europe",
'Trinidad and Tobago': "Americas",
'United States of America': "Americas",
'Saint Vincent and the Grenadines': "Americas",
'Venezuela (Bolivarian Republic of)': "Americas",
```

```

'Tibetan': "Asia",
'Turkmenistan': "Asia",
'Seychelles': 'Africa',
'Sao Tome and Principe': 'Africa',
'Papua New Guinea': "Oceania",
'Suriname': "Americas",
'Tuvalu': "Oceania",
'Canada': "Americas",
'Belize': "Americas",
'Australia': "Oceania",
'Bahamas': "Americas",
'Cabo Verde': "Africa",
'Finland': "Europe",
'Nauru': "Oceania",
'San Marino': "Europe",
'Saint Kitts and Nevis': "Americas",
'Samoa': "Oceania",
'Lesotho': 'Africa',
'Andorra': "Europe",
'New Zealand': "Europe",
'Norway': "Europe",
'Micronesia (Federated States of)': "Oceania",
'Gibraltar': "Europe",
'Turks and Caicos Islands': "Americas",
'Kiribati': "Oceania",
'Maldives': "Asia",
'Bermuda': "Americas",
'Brunei Darussalam': "Asia",
'New Caledonia': "Oceania",
'Monaco': "Europe",
'Montenegro': "Europe",
'Holy See': "Europe",
'South Sudan': "Africa",
'Niue': "Oceania",
'Palau': "Oceania",
'Cayman Islands': "Americas",
'Marshall Islands': "Oceania",
'Curacao ': "Americas",
'Guadeloupe': "Americas",
'Vanuatu': "Oceania",
'French Guiana': "Americas",
'Luxembourg': "Europe",
'Liechtenstein': "Europe",
'Anguilla': "Americas",
'Martinique': "Americas"})
ref["continent"].head()

```

```
[22]: 0      Asia
      1      Asia
      2     Europe
      3      Asia
      4     Africa
      Name: continent, dtype: object
```

```
[23]: ref.head()
```

```
[23]:      Year      Country of origin Country of origin (ISO) \
0  2000      Afghanistan      AFG
1  2000      Iraq      IRQ
2  2000  Serbia and Kosovo: S/RES/1244 (1999)      SRB
3  2000      Turkey      TUR
4  2000      Chad      TCD

      Country of asylum Country of asylum (ISO)  Refugees under UNHCR's mandate \
0      Afghanistan      AFG      0
1      Albania      ALB      9
2      Albania      ALB      507
3      Albania      ALB      5
4      Algeria      DZA      20

      Asylum-seekers  IDPs of concern to UNHCR  Venezuelans displaced abroad \
0      0      758625      NaN
1      0      0      NaN
2      5      0      NaN
3      0      0      NaN
4      19      0      NaN

      Stateless persons  Others of concern  continent
0      0      0      Asia
1      0      0      Asia
2      0      0      Europe
3      0      0      Asia
4      0      0      Africa
```

## 5 Handling Missing Values

```
[24]: ref.isnull().sum()
```

```
[24]: Year      0
      Country of origin      0
      Country of origin (ISO)  1015
      Country of asylum      0
      Country of asylum (ISO)  0
```

Refugees under UNHCR's mandate	0
Asylum-seekers	0
IDPs of concern to UNHCR	0
Venezuelans displaced abroad	89945
Stateless persons	0
Others of concern	0
continent	61
dtype: int64	

```
[25]: ref.notnull().sum()
```

```
[25]: Year          90004
Country of origin  90004
Country of origin (ISO)  88989
Country of asylum  90004
Country of asylum (ISO)  90004
Refugees under UNHCR's mandate  90004
Asylum-seekers  90004
IDPs of concern to UNHCR  90004
Venezuelans displaced abroad    59
Stateless persons  90004
Others of concern  90004
continent          89943
dtype: int64
```

```
[26]: ref["Country of origin"].fillna(value = ref["Country of origin"].mode(),
    ↪inplace = True)
ref.isnull().sum()
```

```
[26]: Year          0
Country of origin  0
Country of origin (ISO)  1015
Country of asylum  0
Country of asylum (ISO)  0
Refugees under UNHCR's mandate  0
Asylum-seekers  0
IDPs of concern to UNHCR  0
Venezuelans displaced abroad  89945
Stateless persons  0
Others of concern  0
continent          61
dtype: int64
```

```
[27]: ref["continent"].value_counts()
```

```
[27]: Africa          38918
Asia              30089
```

Europe	11467
Americas	8063
Unknown Continent	1015
Oceania	391

Name: continent, dtype: int64

```
[28]: ref["continent"].value_counts(normalize = True)*100
```

```
[28]: Africa          43.27
Asia             33.45
Europe          12.75
Americas         8.96
Unknown Continent 1.13
Oceania          0.43
Name: continent, dtype: float64
```

## 5.1 Creating Continent and sub-region

```
[29]: # continent1 = (cont.continent == "Africa")
# Africa_Country = cont.country[continent1]
# for countries1 in Africa_Country:
#     if countries1 in [cout for cout in ref["Country of origin"]]:
#         ref["continent"] = "Africa"
#     else:
#         pass

# continent2 = (cont.continent == "Asia")
# Asia_country = cont.country[continent2]
# for countries2 in Asia_country:
#     if countries2 in [cout for cout in ref["Country of origin"]]:
#         ref["continent"].append(continent2 = "Asia")
#     else:
#         pass

# continent3 = (cont.continent == "Europe")
# Europe_country = cont.country[continent3]
# for countries3 in Europe_country:
#     if countries3 in [cout for cout in ref["Country of origin"]]:
#         ref["continent"].append(continent3 = "Europe")
#     else:
#         pass

# continent4 = (cont.continent == "Americas")
# American_country = cont.country[continent4]
# for countries4 in American_country:
#     if countries4 in [cout for cout in ref["Country of origin"]]:
#         ref["continent"].append(continent4 = "Americas")
```

```
#     else:
#         pass
# continent5 = (cont.continent == "Oceania")
# Oceania_country = cont.country[continent5]
# for countries5 in Oceania_country:
#     if countries5 in [cout for cout in ref["Country of origin"]]:
#         ref["continent"].append(continent5 = "Oceania")
#     else:
#         pass
# ref.head()
```

```
[30]: ref.groupby("continent")["Country of origin"].value_counts().head()
```

```
[30]: continent  Country of origin
Africa        Somalia                1990
              Dem. Rep. of the Congo  1892
              Sudan                  1813
              Ethiopia                1587
              Nigeria                1455
Name: Country of origin, dtype: int64
```

```
[31]: ref.continent.unique()
```

```
[31]: array(['Asia', 'Europe', 'Africa', 'Unknown Continent', 'Americas',
            'Oceania', nan], dtype=object)
```

## 6 Exploring column by column

### 6.1 Year

- The years range from 2000 - 2020
- The year 2020 has highest frequency
- The year 2000 has the least frequency

```
[32]: ref.Year.describe(include = 'all')
```

```
[32]: count    90004.00
mean       2011.07
std         6.02
min        2000.00
25%        2006.00
50%        2012.00
75%        2016.00
max        2020.00
Name: Year, dtype: float64
```

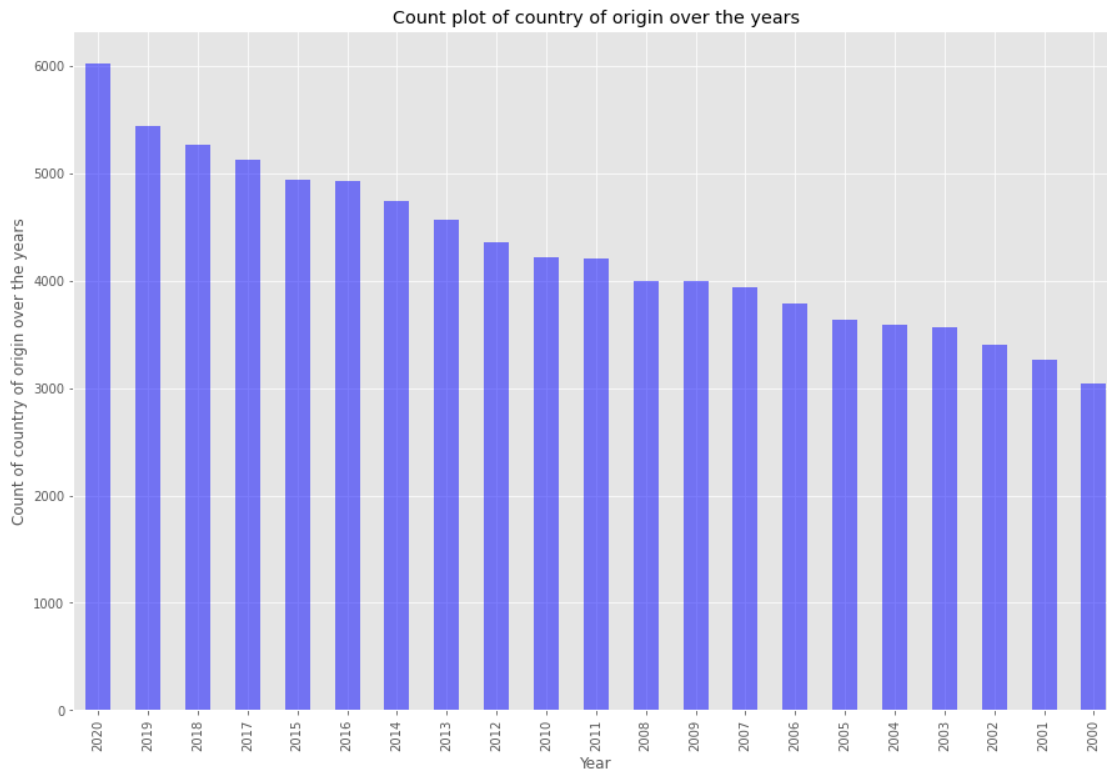
```
[33]: ref.Year.unique()
```

```
[33]: array([2000, 2001, 2002, 2003, 2004, 2005, 2006, 2007, 2008, 2009, 2010,
          2011, 2012, 2013, 2014, 2015, 2016, 2017, 2018, 2019, 2020])
```

```
[34]: ref.Year.value_counts()
```

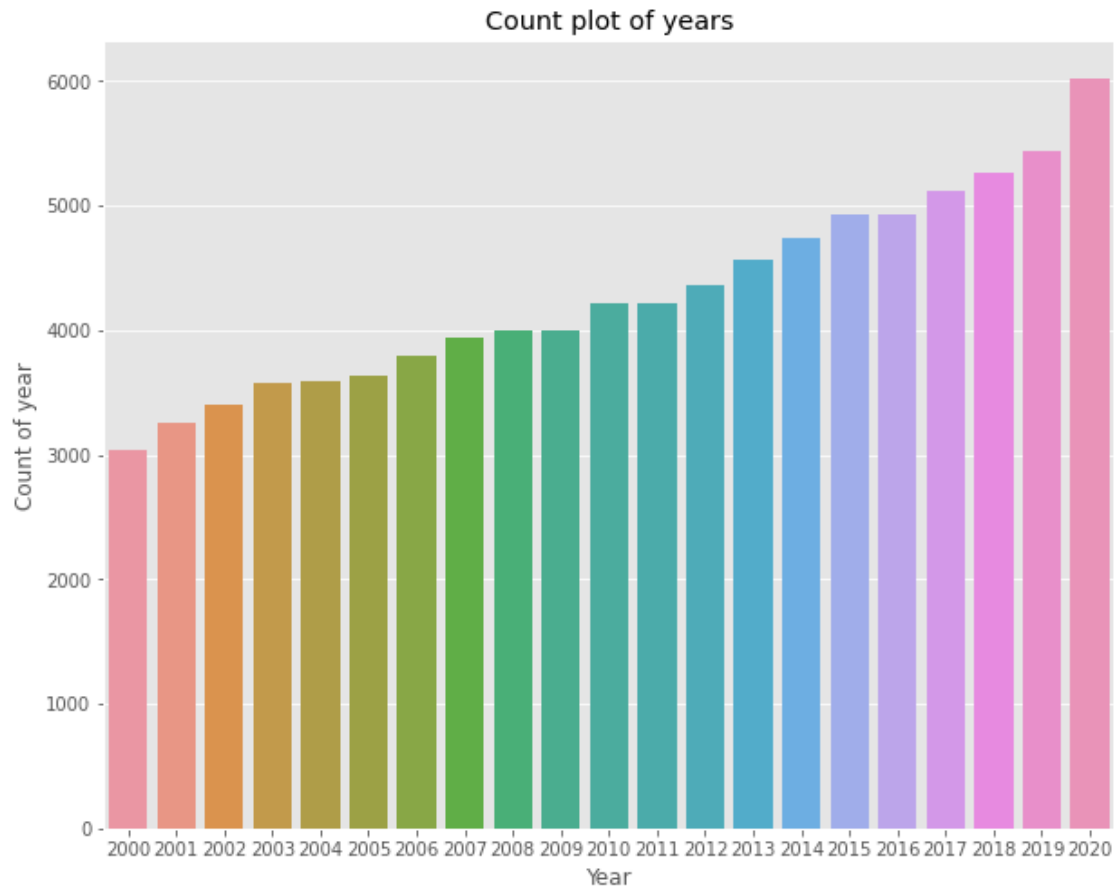
```
[34]: 2020    6024
      2019    5438
      2018    5263
      2017    5120
      2015    4933
      2016    4923
      2014    4744
      2013    4561
      2012    4356
      2010    4212
      2011    4208
      2008    4001
      2009    4000
      2007    3937
      2006    3789
      2005    3640
      2004    3586
      2003    3570
      2002    3403
      2001    3259
      2000    3037
      Name: Year, dtype: int64
```

```
[35]: ref.groupby("Year")["Country of origin"].count().sort_values(ascending=False).
      ↪plot.bar(figsize = (15,10), alpha = 0.5,
      color = 'blue', ylabel = "Count of_
      ↪country of origin over the years",
      title = "Count plot of country of_
      ↪origin over the years");
```



```
[36]: plt.figure(figsize = (10,8))
sns.countplot(x = "Year", data = ref)
plt.ylabel("Count of year")
plt.title("Count plot of years")
plt.show()
```





## 6.2 Country of origin

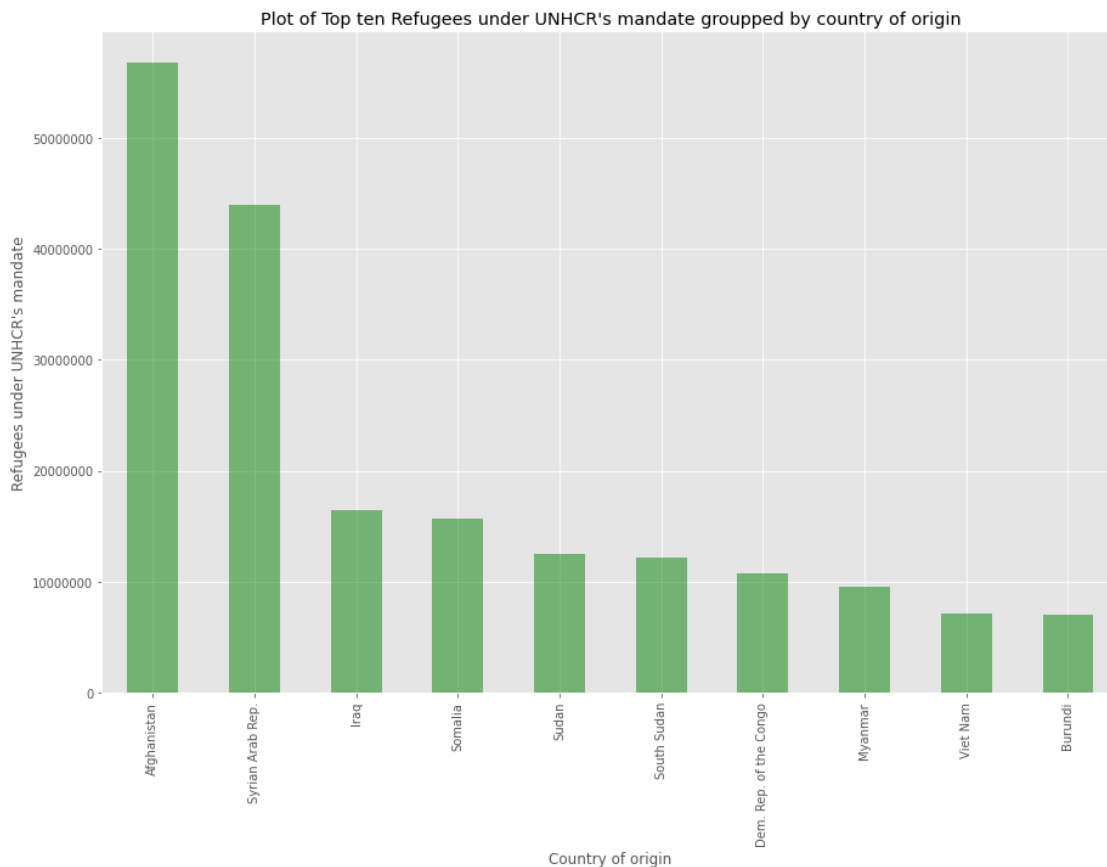
\*The country of origin feature represents the country where those seeking refugee are coming from.

- Somalia is country with the highest number in terms of refugees country's of origin.
- There are 212 countries where refugees originated.
- The data didn't give any information why people from these countries are seeking refugee and asylum in other countries, but we know that most of these countries are in war. Countries like Somalia, Afghanistan, Syria etc.
- Refugees from Afghanistan as country of origin has the highest number of Refugees under UNHCR's mandate

```
[37]: ref["Country of origin"].describe(include = 'all')
```

```
[37]: count      90004
      unique       212
      top        Somalia
      freq        1990
      Name: Country of origin, dtype: object
```

```
[38]: """
Afghanistan has the number of Refugees under UNHCR's mandate
"""
plt.ticklabel_format(style='plain',useOffset=False)
ref.groupby("Country of origin")["Refugees under UNHCR's mandate"].sum().
↳sort_values(ascending=False).head(10).plot.bar(figsize = (15,10), alpha = 0.
↳5,
                                color = 'green', ylabel = "Refugees under UNHCR's",
↳mandate",
                                title="Plot of Top ten
↳Refugees under UNHCR's mandate grouped by country of origin");
```

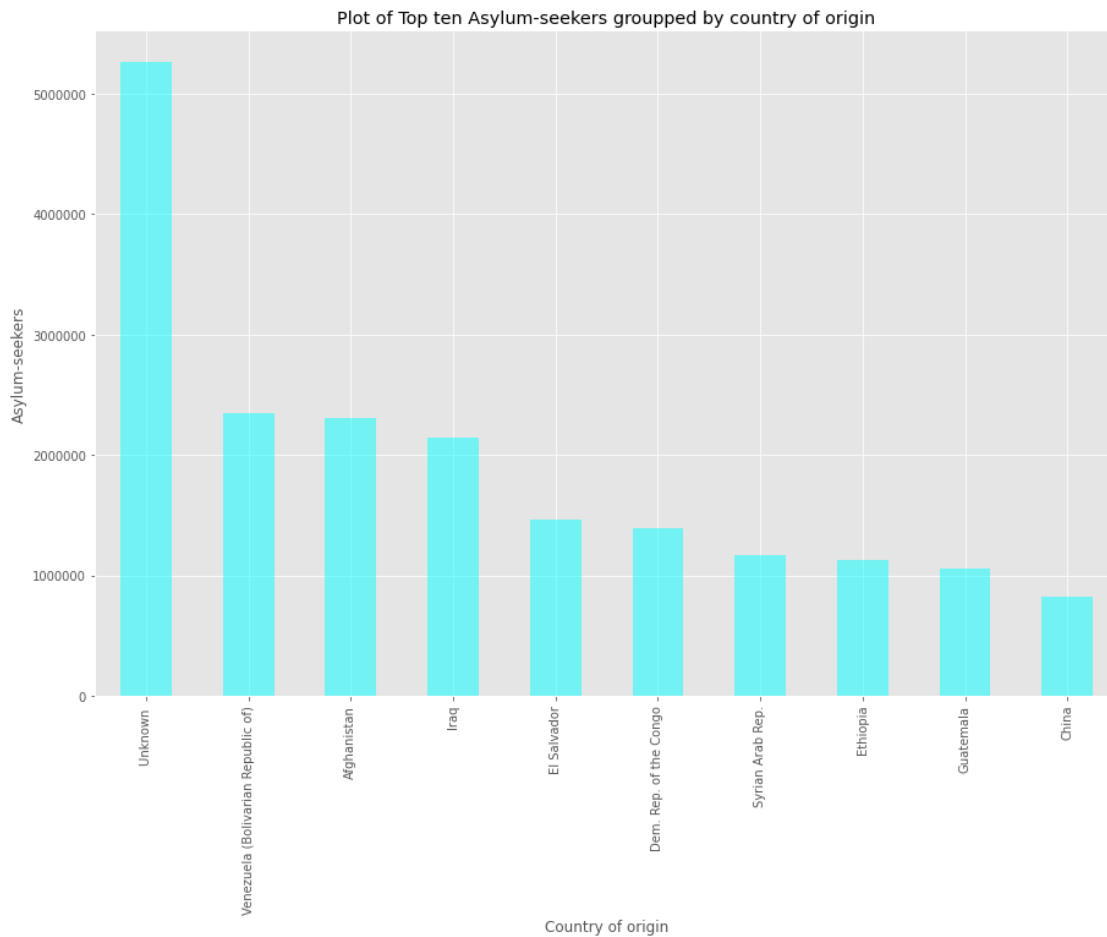


```
[39]: """
Asylum seekers from unknown countries are highest followed by Venezuela
"""
plt.ticklabel_format(style='plain',useOffset=False)
ref.groupby("Country of origin")["Asylum-seekers"].sum().sort_values(ascending
↳False).head(10).plot.bar(figsize = (15,10), alpha = 0.5,
                                color = 'cyan', ylabel = "Asylum-seekers",
```

```

title="Plot of Top ten Asylum-seekers_
↳grouped by country of origin");

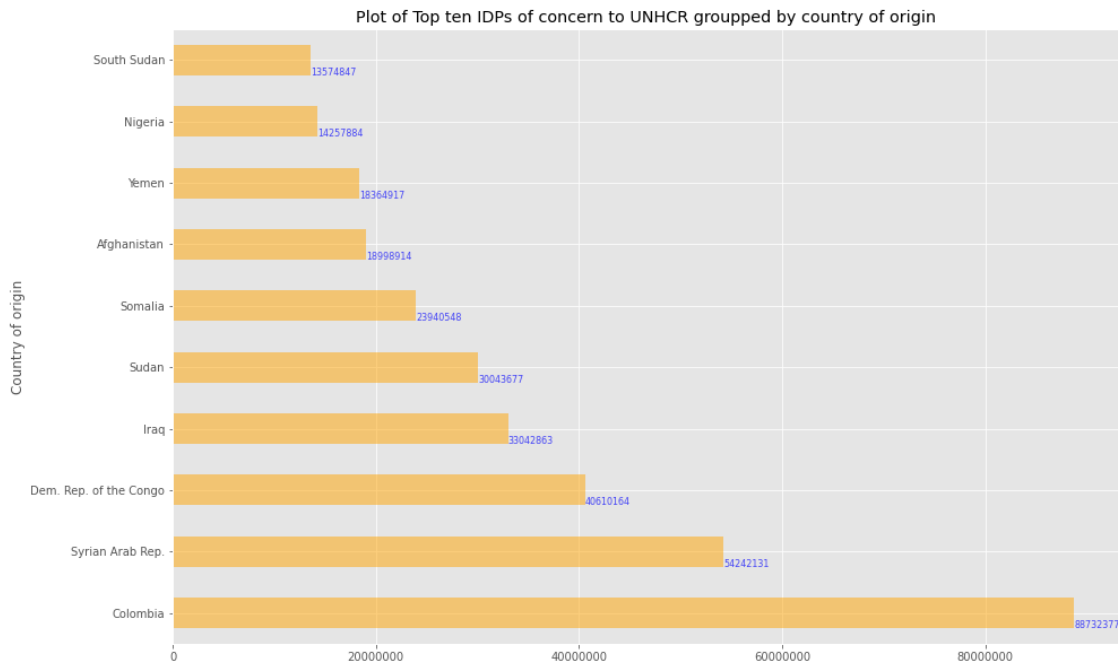
```



```

[41]: """
Colombia has the highest number of IDPs of concern to UNHCR based on country of
↳origin
"""
plt.ticklabel_format(style='plain',useOffset=False)
ax2 = ref.groupby("Country of origin")["IDPs of concern to UNHCR"].sum().
↳sort_values(ascending = False).head(10).plot.barh(figsize = (15,10), alpha =
↳0.5,
color = 'orange', ylabel = "IDPs of concern to
↳UNHCR",
title="Plot of Top ten IDPs of
↳concern to UNHCR grouped by country of origin");
for i in ax2.patches:
ax2.text(i.get_width()+0.005, i.get_y(), str(int(round(i.get_width(),2))),
↳fontsize=8, color='b', alpha=0.7);

```



```
[42]: ref.columns
```

```
[42]: Index(['Year', 'Country of origin', 'Country of origin (ISO)',
        'Country of asylum', 'Country of asylum (ISO)',
        'Refugees under UNHCR's mandate', 'Asylum-seekers',
        'IDPs of concern to UNHCR', 'Venezuelans displaced abroad',
        'Stateless persons', 'Others of concern', 'continent'],
        dtype='object')
```

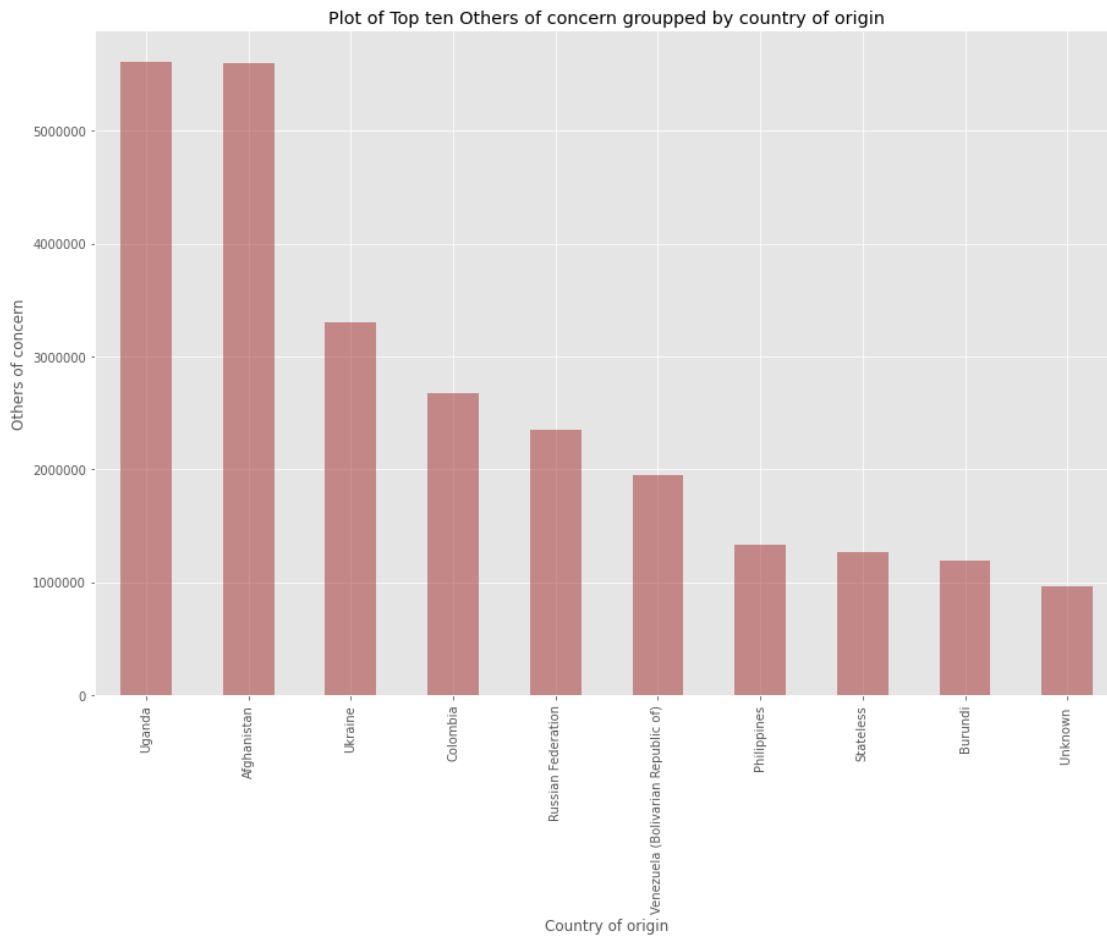
```
[43]: ref.groupby("Country of origin")["Venezuelans displaced abroad"].sum().
        ↪sort_values(ascending = False).head(10)
```

```
[43]: Country of origin
Venezuela (Bolivarian Republic of)    10031476.00
Afghanistan                           0.00
Panama                                0.00
New Zealand                           0.00
Nicaragua                             0.00
Niger                                 0.00
Nigeria                              0.00
Niue                                  0.00
North Macedonia                       0.00
Norway                                0.00
Name: Venezuelans displaced abroad, dtype: float64
```

```
[44]: ref.groupby("Country of origin")["Stateless persons"].sum().
      ↪sort_values(ascending = False).head(10)
```

```
[44]: Country of origin
Stateless          65929313
Afghanistan         0
New Caledonia       0
Nicaragua           0
Niger               0
Nigeria            0
Niue                0
North Macedonia    0
Norway              0
Oman                0
Name: Stateless persons, dtype: int64
```

```
[45]: plt.ticklabel_format(style='plain',useOffset=False)
ref.groupby("Country of origin")["Others of concern"].sum().
      ↪sort_values(ascending = False).head(10).plot(figsize = (15,10), alpha = 0.5,
      color = 'brown', ylabel = "Others of concern",
      title="Plot of Top ten Others of concern grouped by country of origin");
```



### 6.3 Country of asylum & Asylum-seekers

The Country of asylum represent the country where people from various countries of origin are seeking for asylum outside their home country. There are 189 country of asylum seekers \* United States of America has the highest number of asylum seekers with 3572 frequency.

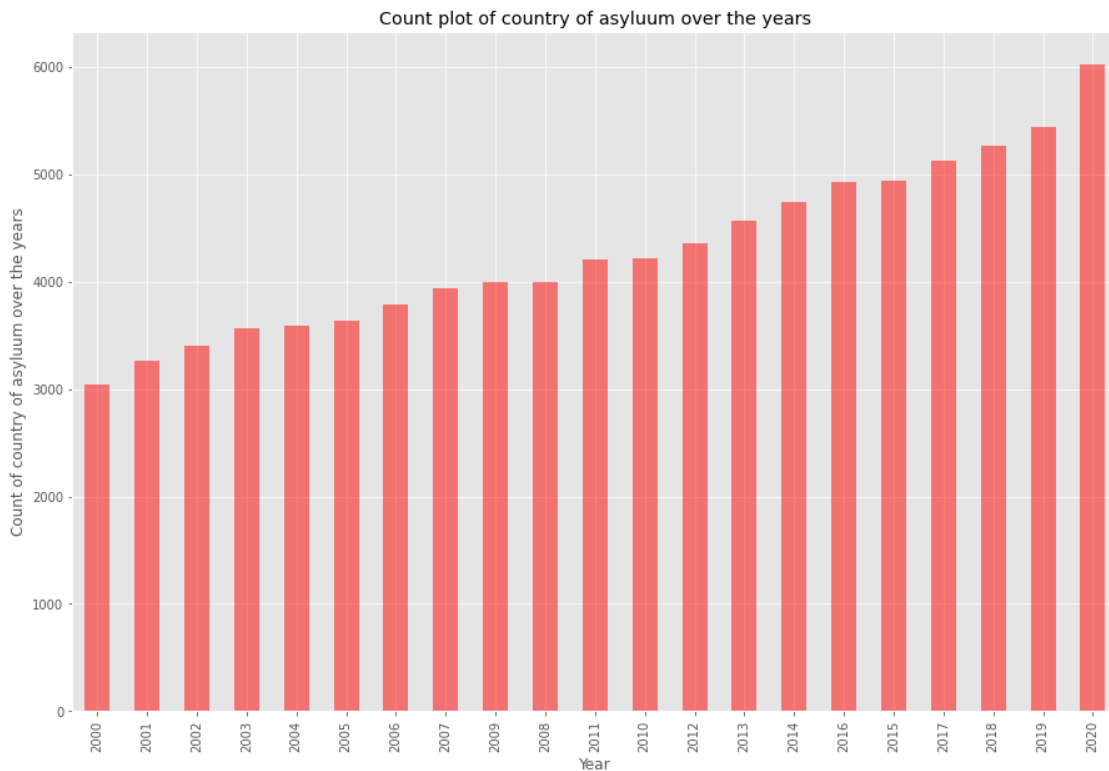
```
[46]: ref[["Country of asylum", "Asylum-seekers"]].describe(include = 'all')
```

```
[46]:
```

	Country of asylum	Asylum-seekers
count	90004	90004.00
unique	189	NaN
top	United States of America	NaN
freq	3572	NaN
mean	NaN	393.77
std	NaN	5586.88
min	NaN	0.00
25%	NaN	0.00
50%	NaN	6.00

75%	NaN	41.00
max	NaN	940668.00

```
[47]: """
Year 2020 has the highest number of people seeking asylum in various countries_
↳ represented
"""
ref.groupby("Year")["Country of asylum"].count().sort_values(ascending =True).
↳plot.bar(figsize = (15,10), alpha = 0.5,
color = 'red', ylabel = "Count of_
↳country of asylum over the years",
title="Count plot of country of_
↳asylum over the years");
```



```
[48]: ref.groupby("Country of asylum")["Asylum-seekers"].sum().sort_values(ascending_
↳= False).head(10)
```

```
[48]: Country of asylum
United States of America    6545156
South Africa                4631417
Germany                    3686650
Turkey                     1978137
```

```

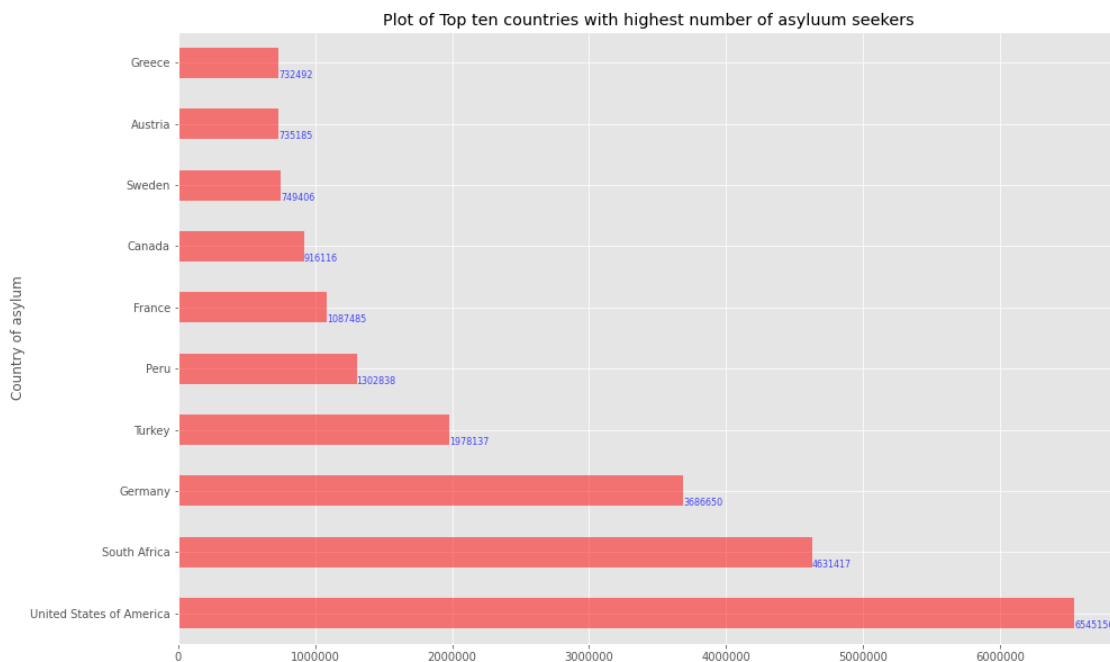
Peru                1302838
France              1087485
Canada              916116
Sweden              749406
Austria             735185
Greece              732492
Name: Asylum-seekers, dtype: int64

```

```

[49]: plt.ticklabel_format(style='plain',useOffset=False)
ax1 = ref.groupby("Country of asylum")["Asylum-seekers"].sum().
    ↪sort_values(ascending = False).head(10).plot.barh(figsize = (15,10), alpha = 0.5,
    ↪0.5,
        color = 'red', ylabel = "Asylum seekers",
        title="Plot of Top ten countries with_
    ↪highest number of asylum seekers");
for i in ax1.patches:
    ax1.text(i.get_width()+0.005, i.get_y(), str(int(round(i.get_width(),2))),
    ↪fontsize=8, color='b', alpha=0.7);

```



## 6.4 Refugees under UNHCR's mandate

```

[50]: ref["Refugees under UNHCR's mandate"].describe(include = 'all')

```

```

[50]: count    90004.00
      mean      3077.74

```



```

std      46313.22
min      0.00
25%      5.00
50%     14.00
75%    103.00
max    3641370.00
Name: Refugees under UNHCR's mandate, dtype: float64

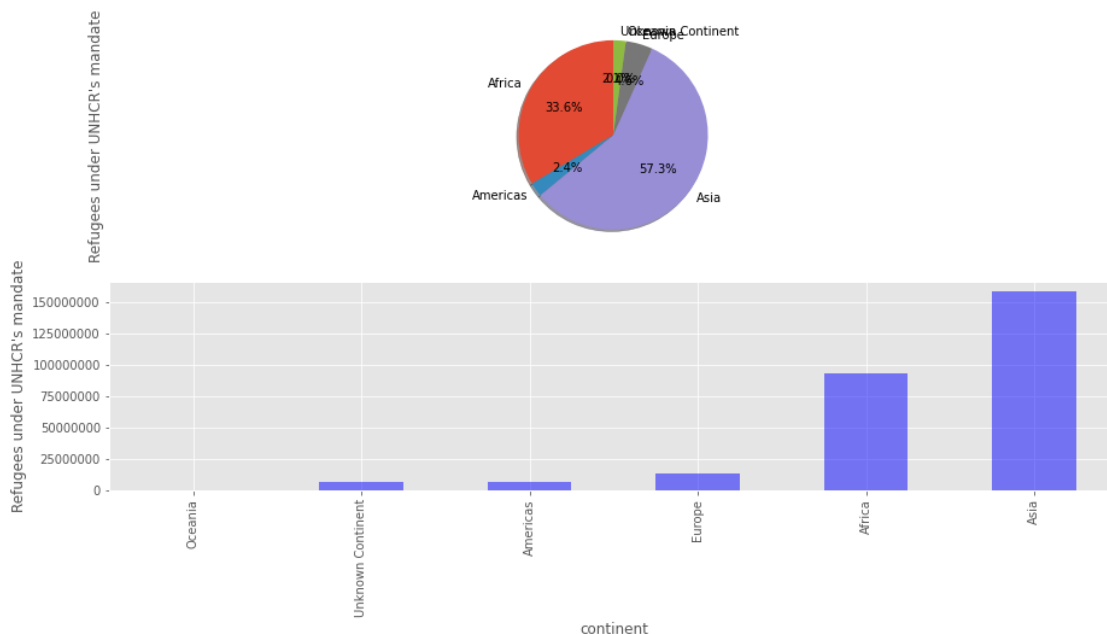
```

```

[51]: plt.suptitle("Pie plot and Barplot of Refugees under UNHCR's mandate grouppe_d_
      ↪by Continents")
plt.subplot(2,1,1)
ax = ref.groupby('continent')['Refugees under UNHCR's mandate'].sum().plot.
      ↪pie(autopct = "%1.1f%%",
          shadow = True, startangle = 90, figsize = (15,7))
ax.axis("equal")
plt.subplot(2,1,2)
plt.ticklabel_format(style = 'plain', useOffset=False)
ax1 = ref.groupby('continent')['Refugees under UNHCR's mandate'].sum().
      ↪sort_values(ascending = True).plot.bar(figsize = (15,7),
          alpha = 0.5,color = 'blue', ylabel = "Refugees_
      ↪under UNHCR's mandate");
plt.show()

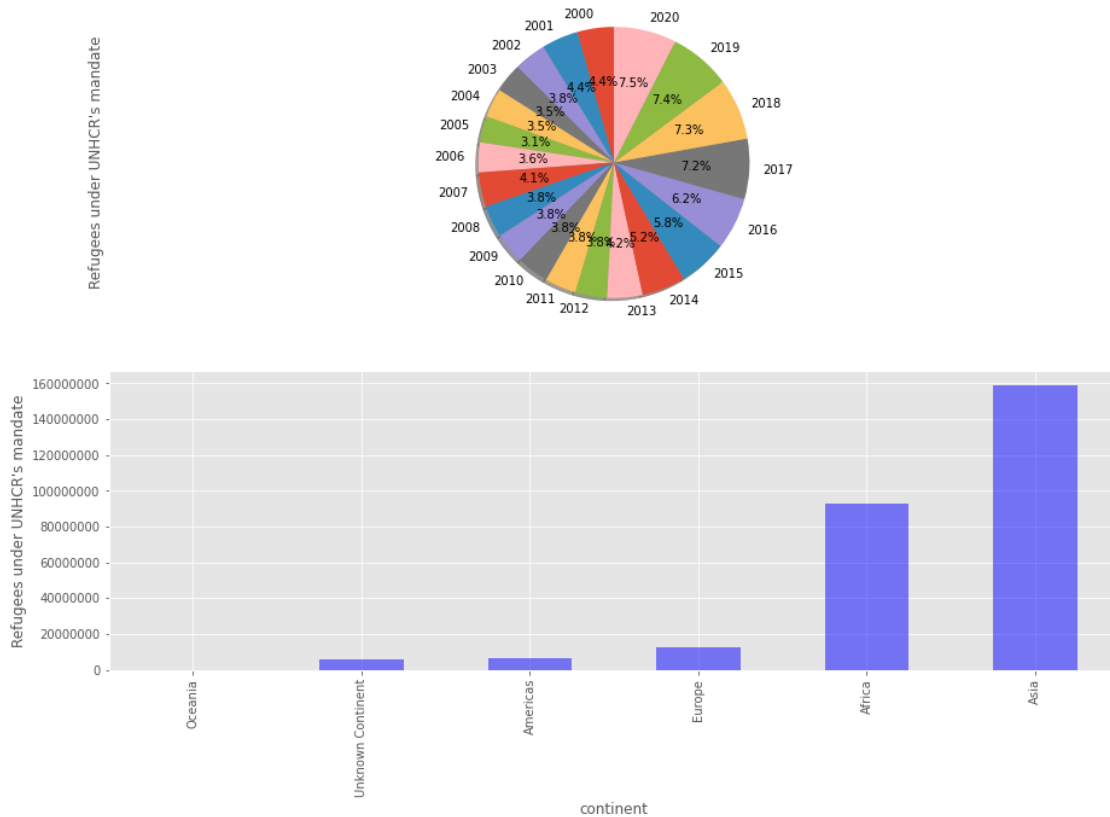
```

Pie plot and Barplot of Refugees under UNHCR's mandate grouped by Continents



```
[52]: plt.suptitle("Pie plot and Barplot of Refugees under UNHCR's mandate grouppe_d_
↳by Year and continent")
plt.subplot(2,1,1)
ax = ref.groupby('Year')['Refugees under UNHCR's mandate'].sum().plot.
↳pie(autopct = "%1.1f%%", shadow = True, startangle = 90, figsize = (15,10))
ax.axis("equal")
plt.subplot(2,1,2)
plt.ticklabel_format(style = 'plain', useOffset=False)
ax1 = ref.groupby('continent')['Refugees under UNHCR's mandate'].sum().
↳sort_values(ascending = True).plot.bar(figsize = (15,10), alpha = 0.5,
color = 'blue', ylabel = "Refugees_
↳under UNHCR's mandate");
plt.show()
```

Pie plot and Barplot of Refugees under UNHCR's mandate grouped by Year and continent



```
[53]: ref.columns
```

```
[53]: Index(['Year', 'Country of origin', 'Country of origin (ISO)',
'Country of asylum', 'Country of asylum (ISO)',
```

```

'Refugees under UNHCR's mandate', 'Asylum-seekers',
'IDPs of concern to UNHCR', 'Venezuelans displaced abroad',
'Stateless persons', 'Others of concern', 'continent'],
dtype='object')

```

## 6.5 IDPs of concern to UNHCR

```
[54]: ref["IDPs of concern to UNHCR"].describe(include = 'all')
```

```

[54]: count      90004.00
      mean       4881.33
      std      124509.91
      min         0.00
      25%         0.00
      50%         0.00
      75%         0.00
      max      8252788.00
      Name: IDPs of concern to UNHCR, dtype: float64

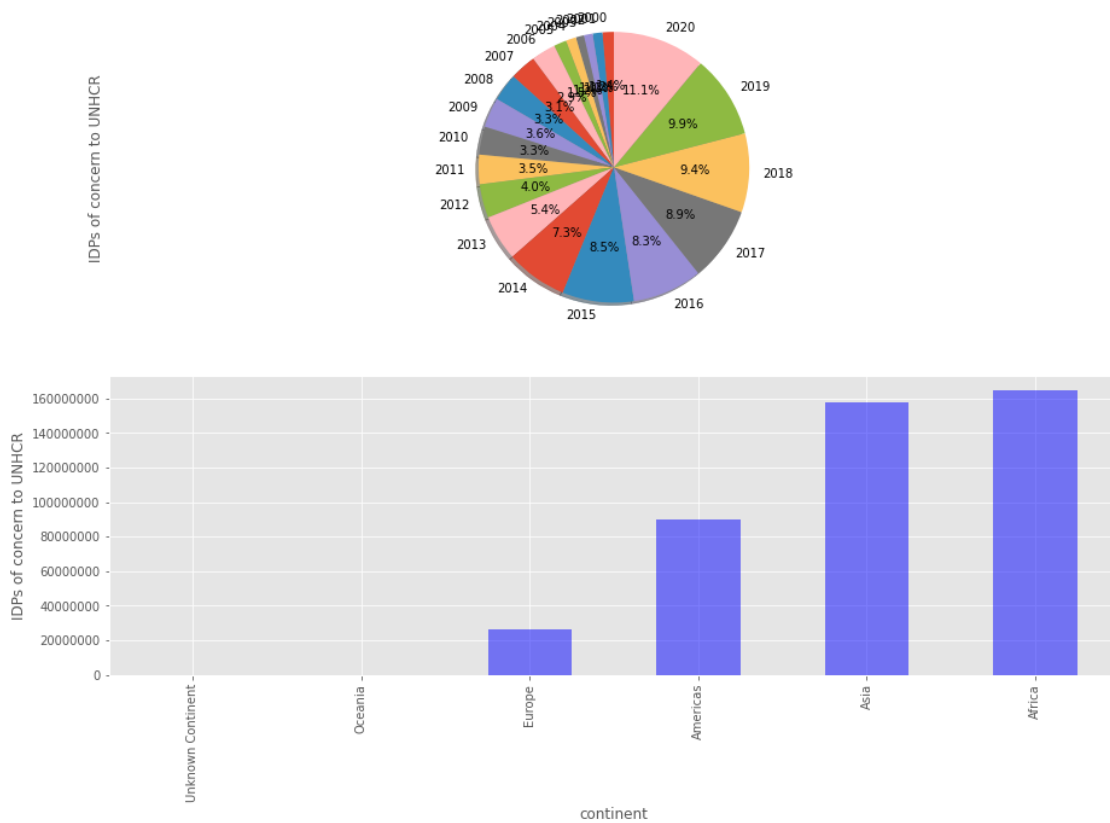
```

```

[55]: plt.suptitle("Pie plot and Barplot of IDPs of concern to UNHCR groupped by Year
      ↪and Continents")
      plt.subplot(2,1,1)
      ax = ref.groupby('Year')['IDPs of concern to UNHCR'].sum().plot.pie(autopct =
      ↪"%1.1f%%",
                                     shadow = True, startangle =
      ↪90, figsize = (15,10))
      ax.axis("equal")
      plt.subplot(2,1,2)
      plt.ticklabel_format(style = 'plain', useOffset=False)
      ax1 = ref.groupby('continent')['IDPs of concern to UNHCR'].sum().
      ↪sort_values(ascending = True).plot.bar(figsize = (15,10),
      alpha = 0.5,color = 'blue', ylabel = "IDPs of
      ↪concern to UNHCR");
      plt.show()

```

Pie plot and Barplot of IDPs of concern to UNHCR grouped by Year and Continents



## 6.6 Others of concern

```
[56]: ref["Others of concern"].describe(include = 'all')
```

```
[56]: count      90004.00
      mean        362.10
      std      16814.69
      min         0.00
      25%         0.00
      50%         0.00
      75%         0.00
      max     2351313.00
      Name: Others of concern, dtype: float64
```

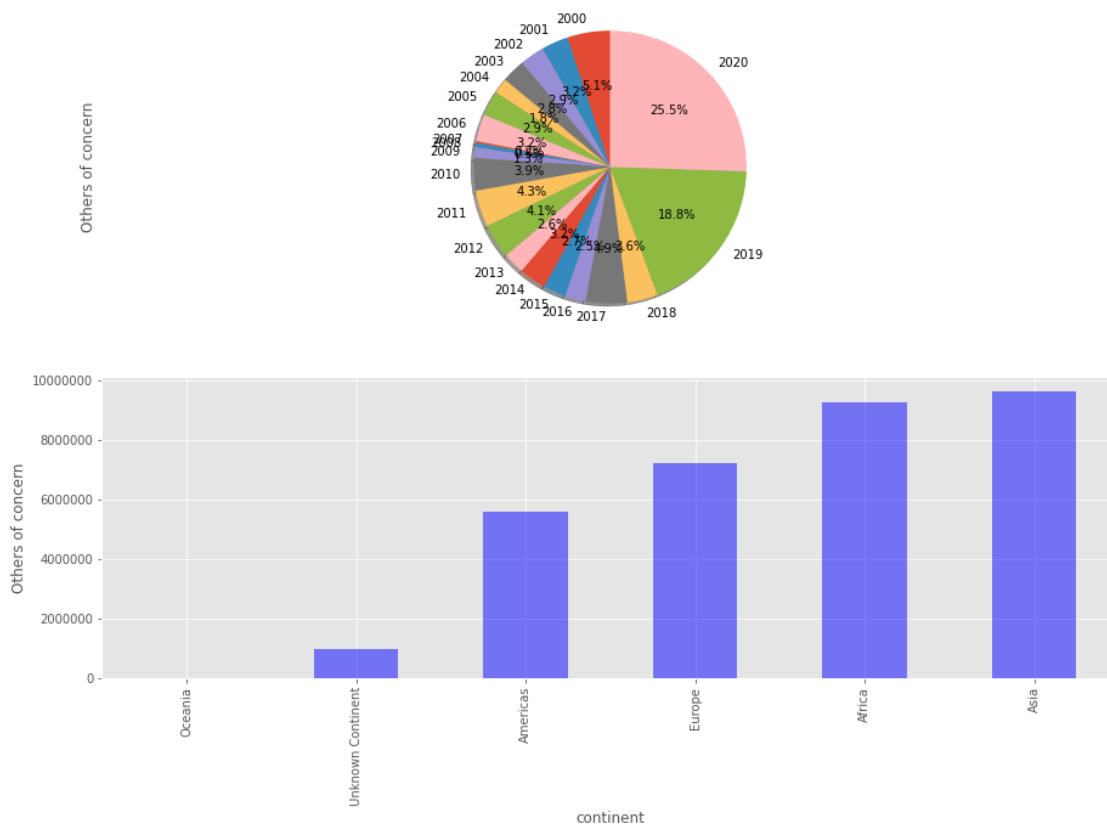
```
[57]: plt.suptitle("Pie plot and Barplot of Others of concern grouped by Year and_
      ↪Continents")
      plt.subplot(2,1,1)
      ax = ref.groupby('Year')['Others of concern'].sum().plot.pie(autopct = "%1.
      ↪1f%%",
```

```

shadow = True, startangle = 90, figsize = (15,10))
ax.axis("equal")
plt.subplot(2,1,2)
plt.ticklabel_format(style = 'plain', useOffset=False)
ax1 = ref.groupby('continent')['Others of concern'].sum().sort_values(ascending=True)
ax1.plot.bar(figsize = (15,10),
             alpha = 0.5,color = 'blue', ylabel = "Others of concern");
plt.show()

```

Pie plot and Barplot of Others of concern grouped by Year and Continents



## 7 Conclusion

The EDA shows a true picture of what we see in our world. The Asia world battling a lot of wars comes out top in term of refugees and asylum seekers. Though, there was no exact data information on relation of global food prices to refugee movement, but war, civil unrest comes with hunger. This makes people leave their country of origin to another country as a refugee or asylum seeker. Therefore, I can categorically say there is a positive correlation with refugee movement to

food prices. If there conflicts or war, there will be lack of food and even if there is food, it will scarce and costly. This will encourage migration of people to avoid dying of starvation.