

# Australia Population Overview Dataset

## Exploratory Data Analysis

In [1]:

```
# Importing Libraries
import numpy as np
import pandas as pd

import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
```

In [2]:

```
pd.set_option("display.max_rows", None, "display.max_columns", None, "display.width", No

# Importing dataset
data = pd.read_csv('1996-2016 Aus ERP_Original.csv')
data.head()
```

Out[2]:

	Sex	Age	Country.of.birth	Region	Time	Population
0	Males	0 - 4	Australia	New South Wales	1996	217870
1	Males	0 - 4	Australia	New South Wales	2001	216620
2	Males	0 - 4	Australia	New South Wales	2006	213520
3	Males	0 - 4	Australia	New South Wales	2011	233550
4	Males	0 - 4	Australia	New South Wales	2016	246710

In [3]:

```
data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 326656 entries, 0 to 326655
Data columns (total 6 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Sex                   326656 non-null object
1   Age                   326656 non-null object
2   Country.of.birth      326656 non-null object
3   Region                326656 non-null object
4   Time                  326656 non-null int64
5   Population             326656 non-null int64
dtypes: int64(2), object(4)
memory usage: 15.0+ MB
```

## Initial evaluation:

The dataset contains 6 features in total.

- 5 categorical variables: Time, Sex, Age, Country of Birth, Region
- 1 numerical variable: Population.

There is no null values found.

## Analysing numerical variable

In [4]:

```
data.Population.describe()
```

Out[4]:

```
count    326656.000000
mean      319.778881
std       5290.432716
min        0.000000
25%        0.000000
50%        0.000000
75%       10.000000
max      246710.000000
Name: Population, dtype: float64
```

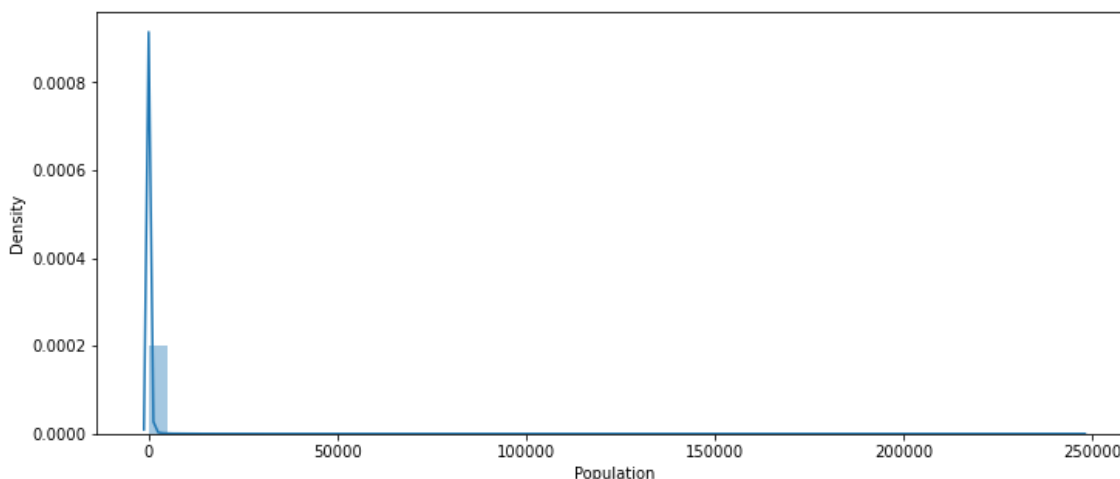
In [5]:

```
plt.figure(figsize=(12,5))
sns.distplot(data['Population'], hist=True, label = 'Population')
```

C:\Users\fresh\anaconda3\lib\site-packages\seaborn\distributions.py:2619:  
FutureWarning: `distplot` is a deprecated function and will be removed in  
a future version. Please adapt your code to use either `displot` (a figure  
-level function with similar flexibility) or `histplot` (an axes-level fun  
ction for histograms).  
warnings.warn(msg, FutureWarning)

Out[5]:

<AxesSubplot:xlabel='Population', ylabel='Density'>



The Population distribution is heavily right-skewed. Meaning that the huge percentage of population are somewhere between 0 and 7000, and a very small percentage of population is higher than that, causing the data to be skewed.

Looking at the statistical analysis, only 25% of the population entries is higher than 10.

## Analysing categorical variables

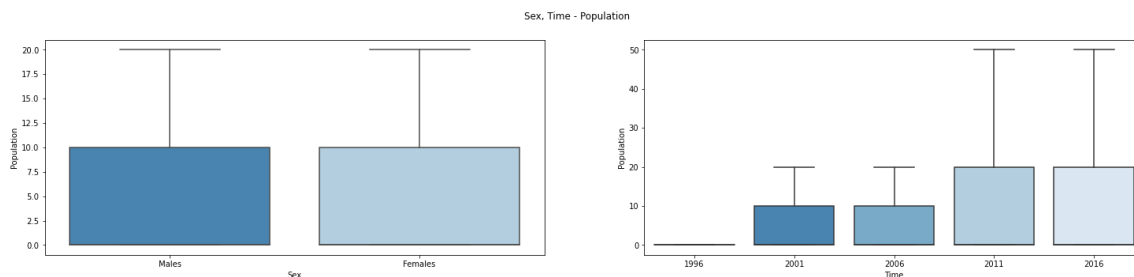
### Sex, Time

In [6]:

```
# Removing outliers
fig, axes = plt.subplots(1, 2, figsize=(25, 5))
fig.suptitle('Sex, Time - Population')
sns.boxplot(ax=axes[0], data=data, x='Sex', y='Population', palette = "Blues_r", showfliers=True)
sns.boxplot(ax=axes[1], data=data, x='Time', y='Population', palette = "Blues_r", showfliers=True)
```

Out[6]:

<AxesSubplot:xlabel='Time', ylabel='Population'>



In [7]:

```
data.Population[data.Time == 1996].describe()
```

Out[7]:

```
count      65536.000000
mean        278.089752
std         4845.388929
min          0.000000
25%          0.000000
50%          0.000000
75%          0.000000
max       217870.000000
Name: Population, dtype: float64
```

In [8]:

```
data.Population[data.Time == 1996].sum()
```

Out[8]:

18224890

Key findings

1. The population seems to distribute equally between 2 sexes
2. The more recent the data is, the larger the population range is.
3. The population by Time in 1996 is mostly 0 (at least 75%). However, the total population in 1996 is close to the national report (18,211,845), we can still consider trusting the population data in 1996 for the purpose of the report. Further research is suggested if possible.

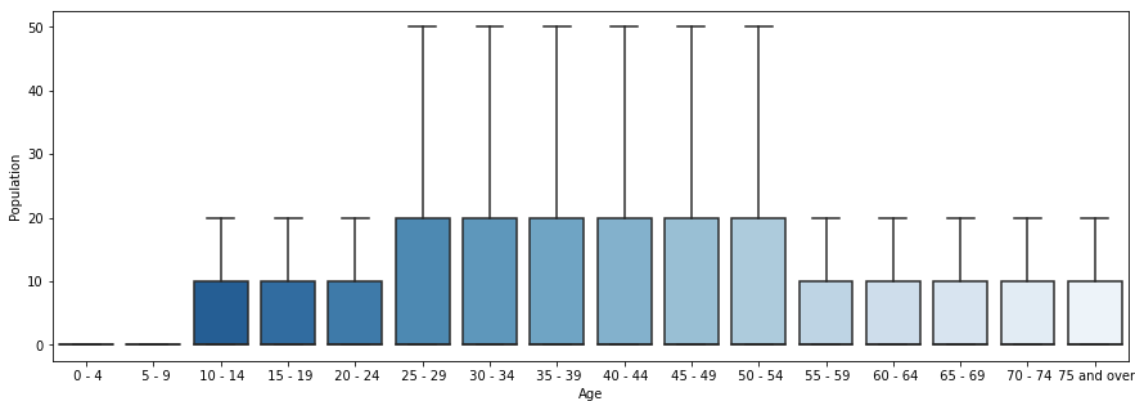
## Age

In [9]:

```
plt.figure(figsize=(15,5))
sns.boxplot(data=data, x='Age', y='Population', palette = "Blues_r", showfliers = False)
```

Out[9]:

<AxesSubplot:xlabel='Age', ylabel='Population'>



In [10]:

```
# Descriptive statistics for 0-4 age group
data.Population[data.Age == "0 - 4"].describe()
```

Out[10]:

```
count      20416.000000
mean         337.408405
std         6995.469695
min           0.000000
25%           0.000000
50%           0.000000
75%           0.000000
max        246710.000000
Name: Population, dtype: float64
```

Age feature contains no null values. After removing all outliers, we can see population come mostly from adults (25 to 54 years old). Some age groups that stay next to each other show the same range in population. Therefore, we might consider to group them into the same group. For example: 0-9 (Gen Alpha) or 10-24 (Gen Z)

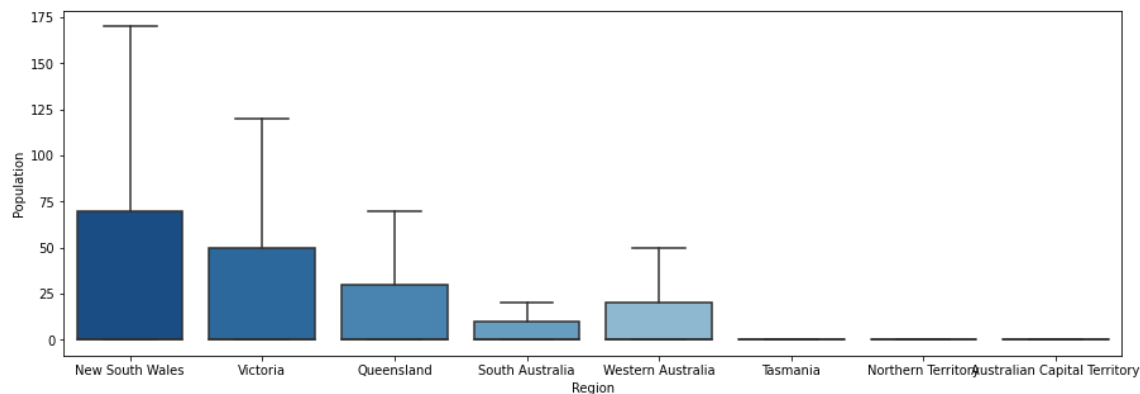
## Region

In [11]:

```
plt.figure(figsize=(15,5))
sns.boxplot(data=data, x='Region', y='Population', palette = "Blues_r", showfliers = False)
```

Out[11]:

<AxesSubplot:xlabel='Region', ylabel='Population'>



The region feature contains correctly 8 regions of Australia and no null values. Population varies greatly in different regions.

## Country of Birth

In [12]:

```
data1=data[['Country.of.birth', 'Population']]
```

In [13]:

```
data1.groupby('Country.of.birth').Population.sum()
```

Out[13]:

Country.of.birth	
Adelie Land (France)	0
Afghanistan	122510
Aland Islands	0
Albania	10330
Algeria	5340
Andorra	10
Angola	1700
Anguilla	10
Antigua and Barbuda	0
Argentina	67500
Argentinian Antarctic Territory	0
Armenia	4640
Aruba	30
Australia	77817590
Australian Antarctic Territory	0
Australian External Territories, nec	0
Austria	102390

In [14]:

```
data1['Country.of.birth'].nunique()
```

Out[14]:

256

Country of birth feature contains 256 unique values. Some countries do not have any people stay in Australia.