

1. Business Understanding

1.1 Overview

Immunization is an essential, cost-effective strategy to reduce childhood morbidity and mortality which saves an estimated 2 to 3 million lives each year. In their 2018 Strategy for Immunization and PHC System Strengthening, Nigeria committed to investing in strengthening immunization service delivery, expanding cold chain capacity, improving data quality, introducing new vaccines and addressing significant risk of vaccine-preventable diseases including measles, yellow fever and meningitis over the Gavi transition period 2018 – 2028. In 2017, routine immunization was declared a Public Health Concern, which led to operationalizing the National Emergency Routine Immunization Coordination Centres (NERICC) in low performing states and LGAs.

1.2 Objective

The objective of this data science project is to analyze and derive insights from the vaccination rates of children aged 12-23 months against preventable childhood diseases in Nigeria.

2. Data Understanding

This data was collected from the National Nutrition and Health Survey conducted in 37 domains, 36 states and Federal Capital Territory (FCT) between February 19 and June 2, 2018. The data will be extracted from the [opendataAfrica website](https://africaopendata.org/dataset/vaccination_coverage_2018/resource/e2bcc139-4c7a-4096-945b-912a0f8b5385) (https://africaopendata.org/dataset/vaccination_coverage_2018/resource/e2bcc139-4c7a-4096-945b-912a0f8b5385), an API is provided to extract the data for an analysis. The data includes information about the states in Nigeria, the types of vaccine administered and the total percentage of children immunized.

This phase is broken down into four task that include;

- Collection of the Initial Data
- Data Description
- Data Cleaning
- Exploratory Data Analysis

2.1 Data Overview

2.1.1 Loading the dataset

```
In [28]: #import the necessary libraries
## Libraries for loading the dataset
import requests
import json
## For data analysis and preparation
import pandas as pd
import numpy as np
## For data visualisation
# For data visualization
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
```

```
In [2]: # Define the URL to access the data
url = 'https://openafrica.net/api/3/action/datastore_search?resource_id=e2b'

# Send a GET request to the URL and retrieve the response
response = requests.get(url)

# Extract the JSON data from the response
data = response.json()

# Extract the records from the JSON data
records = data['result']['records']

# Create a DataFrame from the records
df = pd.DataFrame(records)
print(df.set_index('_id'))
```

	state	vaccine	total
_id			
1	Abia	Any	95.5
2	Abia	Measles	86.4
3	Abia	Penta 1	95.5
4	Abia	Penta 2	90.9
5	Abia	Penta 3	86.4
..
96	Kano	Any	63.3
97	Kano	Measles	49.0
98	Kano	Penta 1	55.1
99	Kano	Penta 2	47.6
100	Kano	Penta 3	36.1

[100 rows x 3 columns]

In [3]: `## Checking the information about the dataframe`
`df.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100 entries, 0 to 99
Data columns (total 4 columns):
#   Column      Non-Null Count  Dtype
---  ---
0   _id          100 non-null    int64
1   state        100 non-null    object
2   vaccine      100 non-null    object
3   total        100 non-null    float64
dtypes: float64(1), int64(1), object(2)
memory usage: 3.2+ KB
```

From checking the summary information about the dataset, the dataframe has 4 columns and 100 rows.

In [4]: `#Checking the names of the columns and their types in the dataframe`
`df.columns`
`df.dtypes`

```
Out[4]: _id          int64
state        object
vaccine      object
total        float64
dtype: object
```

From the above information, we can see that we have two categorical; 'state' and 'vaccine' and one categorical columns; 'total'

In [5]: `#Let's have a look at the descriptive statistics of the numerical columns`
`df.describe()`

```
Out[5]:
```

	_id	total
count	100.000000	100.000000
mean	50.500000	73.615000
std	29.011492	17.403404
min	1.000000	36.100000
25%	25.750000	59.275000
50%	50.500000	79.700000
75%	75.250000	87.875000
max	100.000000	98.500000

From the numerical analysis of the numerical dtypes above;

- The minimum and maximum proportion of vaccinated children is 36 and 98.
- The mean and the median values are almost the same.

```
In [6]: ▶ #Checking the statistics of the columns with the dtype `object`
df.describe(include='O')
```

```
Out[6]:
```

	state	vaccine
count	100	100
unique	20	5
top	Abia	Any
freq	5	20

We see that there are 20 unique values in the `state` column and 5 unique values in the `vaccine` column. The value "Abia" appears 5 times, making it the most frequent value in the `state` column. The vaccine "Any" was the most frequently administered counted 20 times.

2.2 Data Preparation

This is the actual preparation of the data to allow data analysis, it involves, data cleaning and formatting to ensure the Validity, Accuracy, Completeness, Consistency and Uniformity of the Data.

This will include checking for missing and duplicated values and values.

```
In [16]: ▶ # Checking for missing value
df.isnull().sum()
```

```
Out[16]: _id      0
state      0
vaccine    0
total      0
dtype: int64
```

```
In [17]: ▶ #Checking for duplicated values in the rows
df.duplicated().sum()
```

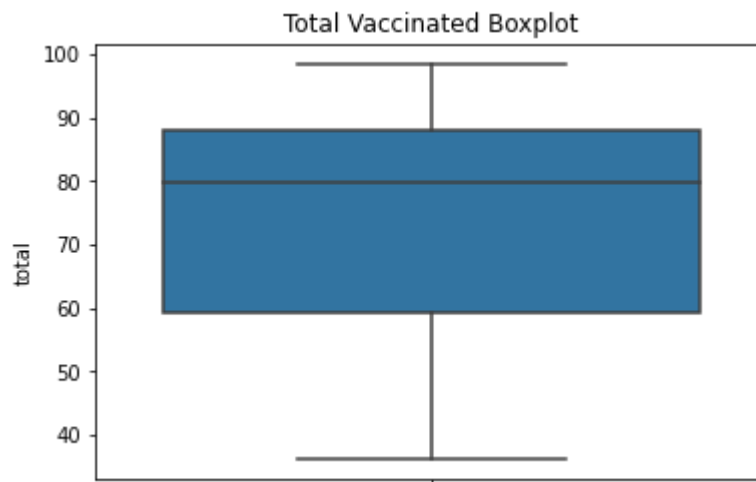
```
Out[17]: 0
```

There are no missing values or any duplicated values in the dataset

Now we can go ahead and check if there are any outliers in the 'Total' column, to see if there are any abnormalities which might influence how we explore and analyse our data.

```
In [22]: #Defining a function to check for outliers  
def find_outliers(data):  
    q1=data.quantile(0.25)  
    q3=data.quantile(0.75)  
    IQR=q3-q1  
    outliers = data[((data<(q1-1.5*IQR)) | (data>(q3+1.5*IQR)))]  
    return outliers  
  
## Finding outliers in the `total` column  
outliers = find_outliers(df['total'])  
print("number of outliers:" + str(len(outliers)))  
  
number of outliers:0
```

```
In [30]: #Using a boxplot to find outliers  
fig, ax = plt.subplots()  
sns.boxplot(y = df['total'])  
ax.set_title('Total Vaccinated Boxplot')  
plt.show()
```



Phewks! There are no outliers in our data, great Now we can begining exploring our data to get insights and achieve our objective.

2.3 Exploratory Data Analysis

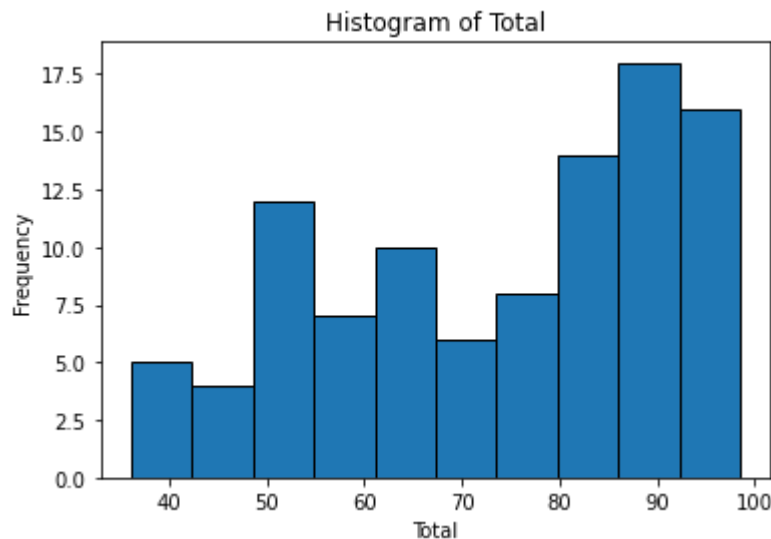
Here we are going to do three tasks which include;

- Univariate Analysis
- Bivariate Analysis
- Multivariate Analysis

2.3.1 Univariate Analysis

The purpose of the univariate analysis is to understand the distribution of values for a single variable.

```
In [38]: ▶ ##Visualising the `total` column  
  
n_bins= 10  
plt.hist(df['total'], bins=n_bins, edgecolor='black')  
  
plt.title('Histogram of Total')  
plt.xlabel('Total')  
plt.ylabel('Frequency')  
  
plt.show()
```



From the histogram above, it means more than 90% vaccines were administered to children in Nigeria

```
In [48]: ▶ # Analysis of the `vaccine` column  
vaccine_proportions = df['vaccine'].value_counts(normalize=True)  
print(vaccine_proportions)  
  
Any          0.2  
Measles      0.2  
Penta 1      0.2  
Penta 2      0.2  
Penta 3      0.2  
Name: vaccine, dtype: float64
```

This indicates that the distribution of vaccines was evenly spread, and each vaccine type represents approximately 20% of the total

```
In [50]: # Analysis of the `state` column  
state_proportions = df['state'].value_counts(normalize=True)  
print(state_proportions)
```

Abia	0.05
Adamawa	0.05
Kaduna	0.05
Jigawa	0.05
Imo	0.05
Gombe	0.05
Federal Capital Territory	0.05
Enugu	0.05
Ekiti	0.05
Edo	0.05
Ebonyi	0.05
Delta	0.05
Cross River	0.05
Borno	0.05
Benue	0.05
Bayelsa	0.05
Bauchi	0.05
Anambra	0.05
Akwa Ibom	0.05
Kano	0.05

Name: state, dtype: float64

All the vaccines were distributed evenly among all the states, from the above analysis in proportion.

2.3.2 Bivariate Analysis

Bivariate analysis is the analysis of exactly two variables. We will use bivariate analysis to find relationships between two variables.

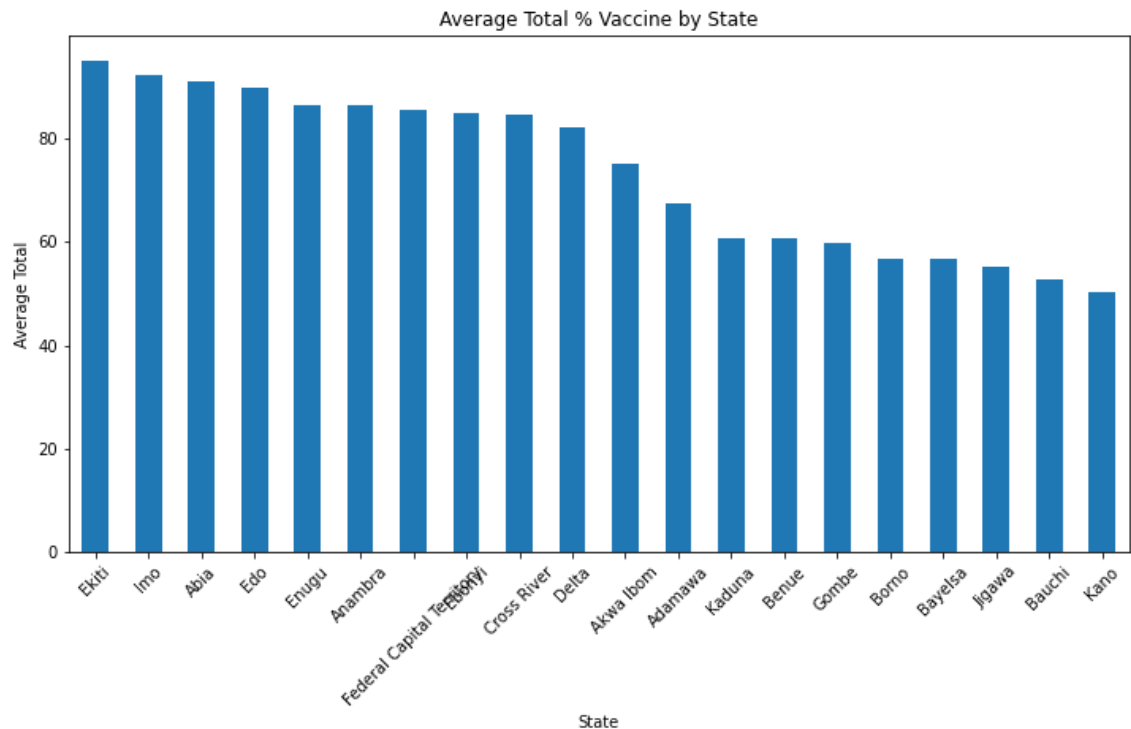
```
In [70]: # Group the DataFrame by 'state' and calculate the mean of 'total' for each
state_totals = df.groupby('state')['total'].mean().sort_values(ascending=False)

# Create a bar plot using the 'state_totals' Series
state_totals.plot(kind='bar', figsize=(12,6))

plt.title('Average Total % Vaccine by State')
plt.xlabel('State')
plt.ylabel('Average Total')

plt.xticks(rotation=45) # Rotate x-axis labels for better readability

plt.show()
```



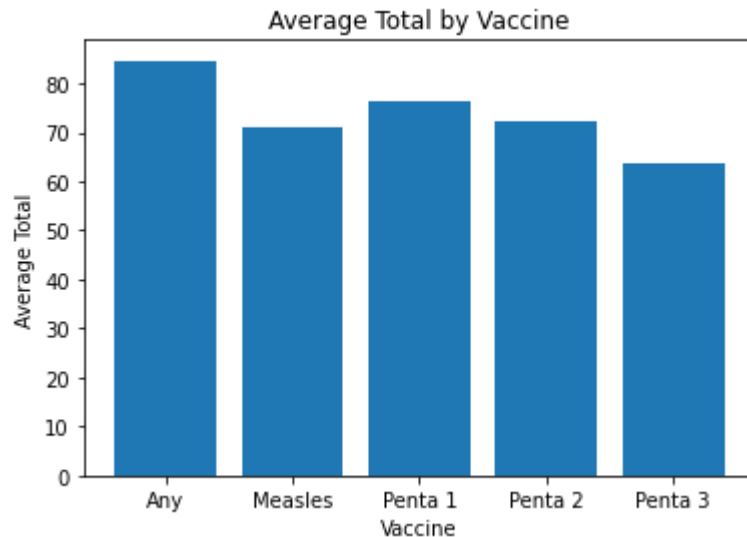
From the bar plot, we observe that the state that received the highest number of vaccines is 'Ekiti' and the least is 'Kano'

In [58]: `#We are going to do the same with the `vaccine` column and the `total` column`

```
vaccine_totals = df.groupby('vaccine')['total'].mean()

plt.bar(vaccine_totals.index, vaccine_totals.values)
plt.title('Average Total by Vaccine')
plt.xlabel('Vaccine')
plt.ylabel('Average Total')

plt.show()
```



The vaccine 'Any' was the most frequent vaccine administered to the children in total and 'Penta 3' was the least administered. We can go the extra mile and find out, if it is the same case for every single state, let's take a look

```
In [96]: ▶ # Most common vaccine by state
most_common_vaccine = df.groupby('state')['vaccine'].agg(lambda x: x.value_
print("Most Common Vaccine by State:")
print(most_common_vaccine)
```

Most Common Vaccine by State:

state	
Abia	Any
Adamawa	Any
Akwa Ibom	Any
Anambra	Any
Bauchi	Any
Bayelsa	Any
Benue	Any
Borno	Any
Cross River	Any
Delta	Any
Ebonyi	Any
Edo	Any
Ekiti	Any
Enugu	Any
Federal Capital Territory	Any
Gombe	Any
Imo	Any
Jigawa	Any
Kaduna	Any
Kano	Any

Name: vaccine, dtype: object

```
In [97]: # Least common vaccine by state
least_common_vaccine = df.groupby('state')['vaccine'].agg(lambda x: x.value)
print("Least Common Vaccine by State:")
print(least_common_vaccine)
```

Least Common Vaccine by State:

state	
Abia	Penta 3
Adamawa	Penta 3
Akwa Ibom	Penta 3
Anambra	Penta 3
Bauchi	Penta 3
Bayelsa	Penta 3
Benue	Penta 3
Borno	Penta 3
Cross River	Penta 3
Delta	Penta 3
Ebonyi	Penta 3
Edo	Penta 3
Ekiti	Penta 3
Enugu	Penta 3
Federal Capital Territory	Penta 3
Gombe	Penta 3
Imo	Penta 3
Jigawa	Penta 3
Kaduna	Penta 3
Kano	Penta 3

Name: vaccine, dtype: object

Wow! It is exactly the same output for all the states, we will have to do research and find out why this is the case

Let us then go ahead to do a statistical analysis of the highest and lowest vaccination rates per state, we are going to group the dataframe by the 'state' column and find the maximum and minimum values of the 'total' columns of each unique state

```
In [82]: # The highest vaccination rate by finding the maximum value in the total column
# We are going to group them by state and total columns

highest_vaccination_rate = df.groupby('state')['total'].max()
highest_vaccination_rate_sorted = highest_vaccination_rate.sort_values(ascending=False)
print(highest_vaccination_rate_sorted)
```

```
state
Ekiti          98.5
Imo            96.2
Anambra        96.2
Abia           95.5
Edo            95.3
Enugu          94.9
Ebonyi         94.0
Delta          93.8
Cross River    89.4
Federal Capital Territory 89.0
Benue          84.4
Kaduna         83.6
Adamawa        83.5
Akwa Ibom      83.5
Bayelsa        76.7
Jigawa         72.7
Gombe         69.7
Borno         66.7
Bauchi         65.9
Kano           63.3
Name: total, dtype: float64
```

For the highest vaccination rates, the analysis identified the states with the highest rates and provided their corresponding vaccination rates. The highest rates was 98.5 in Ekiti

```
In [90]: # The Lowest vaccination rate by finding the minimum value in the total column
# We are going to group them by state and total columns

lowest_vaccination_rate = df.groupby('state')['total'].min()
#lowest_vaccination_rate
lowest_vaccination_rate_sorted = lowest_vaccination_rate.sort_values(ascending=False)
print(lowest_vaccination_rate_sorted[:5])
```

```
state
Kano           36.1
Jigawa         38.0
Benue          40.6
Bauchi         41.5
Bayelsa        41.7
Name: total, dtype: float64
```

Similarly, for the lowest vaccination rates, the analysis identified the states with the lowest rates and provided their corresponding vaccination rates. The lowest rates ranged from 36.1 in Kano to 41.5 in Bauchi. This information helps us identify the states that need more focus and support to improve their vaccination rates.

2.3.3 Multivariate Analysis

This is the analysis of more than 2 variables at the same time and finding out if they are correlated

A grouped bar chart to compare the total values for each vaccine category across different states. Each bar represents a state, and within each bar, there are grouped bars representing the total values for each vaccine category.

In [98]:

```
plt.figure(figsize=(12, 6)) # Set the size of the plot

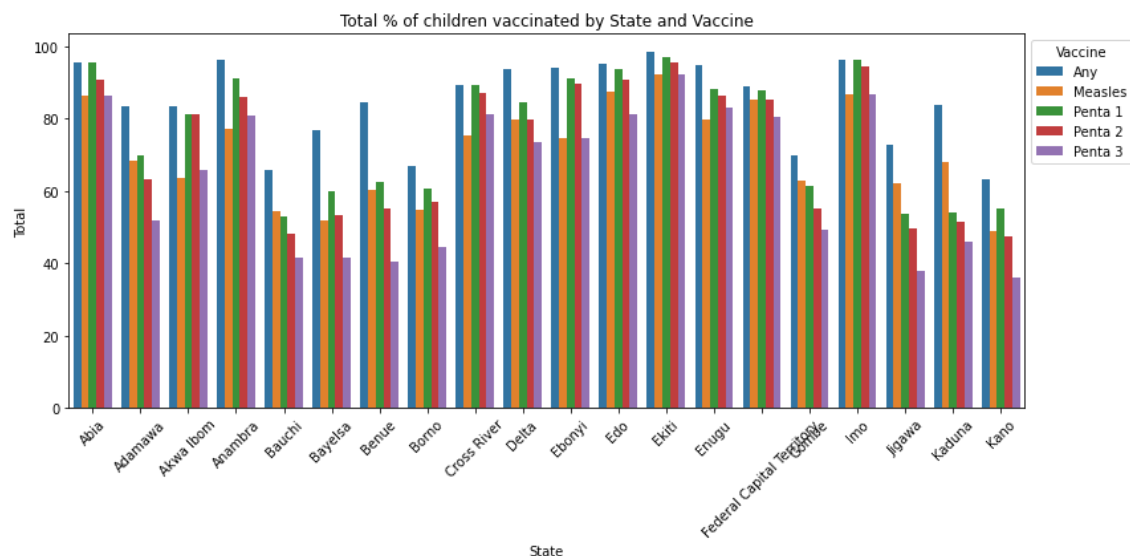
sns.barplot(data=df, x='state', y='total', hue='vaccine')

plt.title('Total % of children vaccinated by State and Vaccine')
plt.xlabel('State')
plt.ylabel('Total')

plt.xticks(rotation=45)

plt.legend(title='Vaccine', bbox_to_anchor=(1, 1))
plt.tight_layout()

plt.show()
```



From the grouped bar chart, we see that Ekiti is still in the lead of the all the states that has the highest percentage of children who have been vaccinated

3. Data Report

The following is a summary of the data exploration and analysis;

1. The 'Any' was the modal vaccine administered in Nigeria in 2018 in all of the states
2. Similarly, the least administered vaccine was 'Penta 3' for all the states.
3. The state that had the highest vaccination rates was "Ekiti" with 98.5 rate
4. The lowest vaccination rate was from "Kano" with 36.1
5. The vaccines were all distributed equally and evenly spread among the states.

Now let's go ahead and do some research gain insights why.