EDA on COVID-19 Clinical Trials

1. Introduction

The COVID-19 pandemic initiated an unprecedented surge in clinical research activities around the world. Thousands of clinical trials were launched to discover treatments, vaccines, and diagnostic methods.

This project performs **Exploratory Data Analysis** (**EDA**) on a dataset of **COVID-19 related clinical trials** to understand trends, patterns, and important insights.

2. Objectives

- To explore the **status** of clinical trials.
- To analyze **study types** (Interventional, Observational).
- To identify the most common **medical conditions** studied.
- To find the major **sponsors** behind the studies.
- To study the **geographic distribution** of trials.
- To investigate the **phases** of Interventional studies.

3. Dataset Overview

The data is loaded from a CSV file using Python's pandas library.

The dataset contains columns such as:

- **Status**: Current state of the trial (e.g., Completed, Recruiting).
- **Conditions**: Disease(s) being studied.
- **Sponsor**: Organization leading or funding the study.
- **Study Type**: Type of research (Interventional, Observational).
- **Phase**: Phase of the trial (Phase 1–4).
- Locations: Where the trial is being conducted

4. Code, Explanation

4.1 Import Visualization Libraries

```
[126] import pandas as pd
    import numpy as np
    from matplotlib import pyplot as plt
    %matplotlib inline
    import matplotlib
    import seaborn as sns
    from plotly.subplots import make_subplots
    import plotly.graph_objects as go
```

- pandas: For data handling.
- **numpy:** For numerical computations.
- matplotlib.pyplot and seaborn: For creating visualizations.

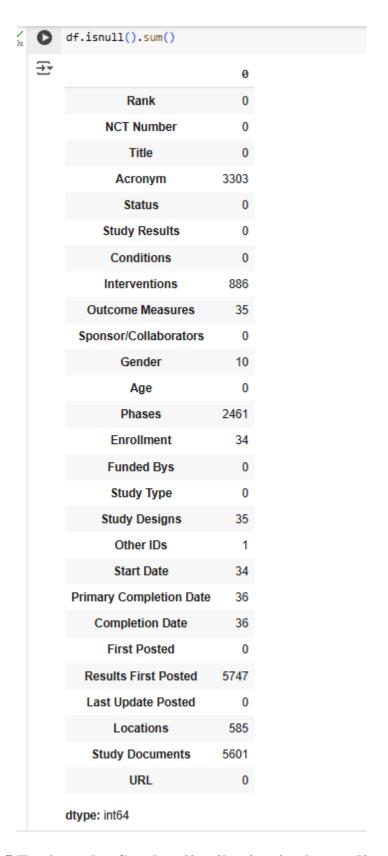
4.2 Reading the Dataset

importing the csv file as a dataframe

```
[127] df = pd.read_csv('/content/COVID clinical trials.csv')
```

4.3 Checking Basic Information

4.4 Data Cleaning



4.5 Explore the Gender distribution in the studies

```
df['Gender'].unique()
        array(['All', 'Female', 'Male', nan], dtype=object)
[ [219] df['Gender'].value_counts()
   ₹
                    count
          Gender
            AII
                     5567
          Female
                      162
           Male
                        44
         dtype: int64
   sns.barplot(x=df['Gender'].value_counts().index,
                 y=df['Gender'].value_counts().values)
    plt.xlabel('Gender')
    plt.ylabel('Frequency')
    plt.title('Show of gender Bar Plot')
    plt.show()
<del>∑</del>*
                               Show of gender Bar Plot
       5000
       4000
     Frequency
       3000
       2000
       1000
           0
```

Observation: The majority of studies are categorized as "All" genders, with few specifically labeled as "Female" or "Male."

Female Gender Male

ΑII

4.6 Exploring Study Status Distribution

```
df.Status.unique()
array(['Active, not recruiting', 'Not yet recruiting', 'Recruiting',
        'Enrolling by invitation', 'Suspended', 'Completed', 'Withdrawn',
        'Terminated', 'No longer available', 'Available',
        'Approved for marketing', 'Temporarily not available'],
      dtype=object)
df['Status'].value_counts()
                          count
                  Status
        Recruiting
                           2805
        Completed
                           1025
     Not yet recruiting
                           1004
   Active, not recruiting
                            526
  Enrolling by invitation
                            181
        Withdrawn
                            107
```

sns.countplot(y="Status", data=df, color="red")

74

27

19

12

2

<Axes: xlabel='count', ylabel='Status'>

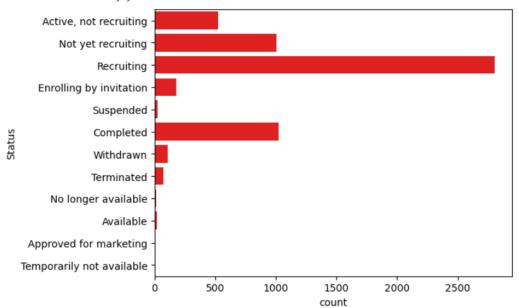
Terminated

Suspended

Available

No longer available

Approved for marketing



Observation: The study status distribution reveals that most studies are still in the process of recruiting participants, indicating that the research is ongoing.

4.7 Cleaning Age Column

Cleaninig Age Column

df.Age.unique()

df[['Age']].head()

```
→ array(['18 Years and older \xa0 (Adult, Older Adult)',
           'Child, Adult, Older Adult', '18 Years to 48 Years \xa0 (Adult)',
           '18 Years to 75 Years \xa0 (Adult, Older Adult)',
           '18 Years to 45 Years \xa0 (Adult)',
           '18 Years to 99 Years \xa0 (Adult, Older Adult)',
           '18 Years to 55 Years \xa0 (Adult)',
           '15 Years and older \xa0 (Child, Adult, Older Adult)',
           '18 Years to 80 Years \xa0 (Adult, Older Adult)',
           '45 Years and older \xa0 (Adult, Older Adult)'
           '20 Years to 100 Years \xa0 (Adult, Older Adult)',
           '8 Years to 88 Years \xa0 (Child, Adult, Older Adult)',
           '5 Years to 65 Years \xa0 (Child, Adult, Older Adult)',
           'up to 99 Years \xa0 (Child, Adult, Older Adult)',
           '18 Years to 85 Years \xa0 (Adult, Older Adult)',
           '18 Years to 65 Years \xa0 (Adult, Older Adult)',
           'up to 29 Days \xa0 (Child)',
           '18 Years to 70 Years \xa0 (Adult, Older Adult)',
           '18 Years to 59 Years \xa0 (Adult)',
from string import digits
def remove digits(text):
      return text.translate(str.maketrans('', '', digits))
df["Age"] = df["Age"].apply(lambda text: remove_digits(text))
```

- **from string import digits**: Imports the digits constant, which contains all digits ('0123456789').
- **remove_digits(text)**: A function that removes all digit characters from a given string text by using str.translate() to replace digits (from digits) with nothing (i.e., removes them).
- **df["Age"].apply(lambda text: remove_digits(text))**: For each value in the "Age" column of the df DataFrame, the remove_digits function is applied to remove any digits.
- **df**[['**Age**']].**head**(): Displays the first 5 rows of the "Age" column after removing digits.

```
Age
Years and older (Adult, Older Adult)
Years and older (Adult, Older Adult)
Years and older (Adult, Older Adult)
Child, Adult, Older Adult
```

```
from nltk.corpus import stopwords
stopwords = stopwords.words('english')
def remove_stopwords(text):
    return " " .join([word for word in str(text).split() if word not in stopwords])

df["Age"] = df["Age"].apply(lambda text: remove_stopwords(text))
df[['Age']].head()
```

Years to Years (Adult)

- **from nltk.corpus import stopwords**: Imports common English stopwords (like "the", "is", "and") from NLTK.
- **remove_stopwords(text)**: Defines a function that removes these stopwords from a given text by keeping only the words not in the stopwords list.
- **df["Age"].apply(lambda text: remove_stopwords(text))**: Applies the remove_stopwords function to each entry in the "Age" column.
- **df**[['**Age**']].head(): Displays the first 5 rows of the cleaned "Age" column.

	Age
0	Years older (Adult, Older Adult)
1	Years older (Adult, Older Adult)
2	Years older (Adult, Older Adult)
3	Child, Adult, Older Adult
4	Years Years (Adult)

```
df.Age.unique()
array(['Years older (Adult, Older Adult)', 'Child, Adult, Older Adult',
        'Years Years (Adult)', 'Years Years (Adult, Older Adult)',
       'Years older (Child, Adult, Older Adult)',
       'Years Years (Child, Adult, Older Adult)',
       'Years (Child, Adult, Older Adult)', 'Days (Child)',
       'Years (Child, Adult)', 'Years older (Older Adult)',
       'Years Years (Child, Adult)', 'Years (Child)',
       'Months older (Child, Adult, Older Adult)',
       'Year Years (Child, Adult, Older Adult)', 'Years Years (Child)',
       'Months Years (Child, Adult, Older Adult)', 'Minutes (Child)'
       'Weeks Weeks (Child)', 'Year older (Child, Adult, Older Adult)'
       'Month Years (Child, Adult, Older Adult)', 'Year Years (Child)',
       'Year Years (Child, Adult)', 'Month Years (Child, Adult)',
       'Month Years (Child)', 'Hours (Child)', 'Months (Child)',
       'Months Years (Child, Adult)', 'Years Years (Older Adult)'
       'Months older (Adult, Older Adult)', 'Months Years (Child)',
       'Days Years (Child, Adult)', 'Month (Child)',
       'Month older (Child, Adult, Older Adult)',
       'Weeks Years (Child, Adult)', 'Months Months (Child)',
       'Days older (Child, Adult, Older Adult)', 'Year (Child)'],
      dtype=object)
df["Age"]=df["Age"].apply(lambda x:str(x).replace('Years','')if 'Years' in str(x) else str(x))
df["Age"]=df["Age"].apply(lambda x:str(x).replace('Year','')if 'Year' in str(x) else str(x))
```

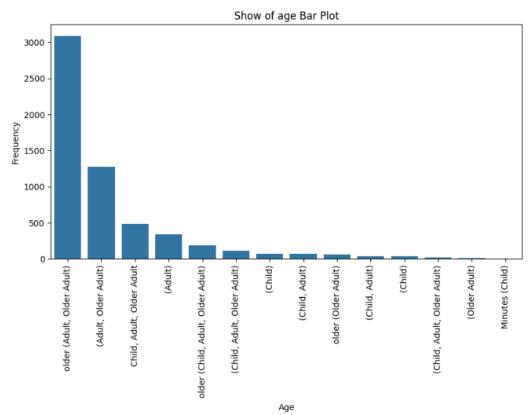
- This code removes the words "Years" and "Year" from each value in the "Age" column.
- It checks if "Years" or "Year" exists in the text, and if yes, replaces them with an empty string.

```
df.Age.unique()
array([' older (Adult, Older Adult)', 'Child, Adult, Older Adult',
           (Adult)', ' (Adult, Older Adult)',
         ' older (Child, Adult, Older Adult)',
         ' (Child, Adult, Older Adult)', ' (Child, Adult, Older Adult)', 'Days (Child)', ' (Child, Adult)', ' older (Older Adult)',
         ' (Child, Adult)', ' (Child)',
         'Months older (Child, Adult, Older Adult)', ' (Child)',
         'Months (Child, Adult, Older Adult)', 'Minutes (Child)'
         'Weeks Weeks (Child)', 'Month (Child, Adult, Older Adult)',
         'Month (Child, Adult)', 'Month (Child)', 'Hours (Child)',
         'Months (Child)', 'Months (Child, Adult)', ' (Older Adult)',
         'Months older (Adult, Older Adult)', 'Months (Child)',
         'Days (Child, Adult)', 'Month (Child)',
         'Month older (Child, Adult, Older Adult)', 'Weeks (Child, Adult)',
         'Months Months (Child)', 'Days older (Child, Adult, Older Adult)'],
       dtype=object)
```

```
df["Age"]=df["Age"].apply(lambda x:str(x).replace('Months','')if 'Months' in str(x) else str(x))
df["Age"]=df["Age"].apply(lambda x:str(x).replace('Month','')if 'Month' in str(x) else str(x))
df["Age"]=df["Age"].apply(lambda x:str(x).replace('Days','')if 'Days' in str(x) else str(x))
df["Age"]=df["Age"].apply(lambda x:str(x).replace('Weeks','')if 'Weeks' in str(x) else str(x))
df["Age"]=df["Age"].apply(lambda x:str(x).replace('Hours','')if 'Hours' in str(x) else str(x))
```

- This code **removes** the words "Months", "Month", "Days", "Weeks", and "Hours" from each value in the "Age" column.
- For each word, it checks if it exists in the text and replaces it with an empty string.

df.Age.unique()



Observation: Adult, Older Adult age bracket are mostly studied --Child and older Adult age bracket has the lowest studies.

```
i = 0
fig = make_subplots(rows=3, cols=2, subplot_titles=list(pd.DataFrame(df.groupby(['Age'])['Gender'].value_counts()).unstack().index))
for row in range(1,4):
    for col in range(1,3):
        dt = pd.DataFrame(df.groupby(['Age'])['Gender'].value_counts()).unstack().iloc[i]
        # Check if dt is a Series and convert it to DataFrame if necessary
        if isinstance(dt, pd.Series):
            dt = dt.to_frame[name='Gender']]
            #This converts it to a DataFrame with 'Gender' as column name.
        fig.add_trace(go.Bar(x=dt.index, y=dt.Gender.values), row=row, col=col) #Use dt.index instead of dt.Gender.index
        i+=1
fig.show()
```



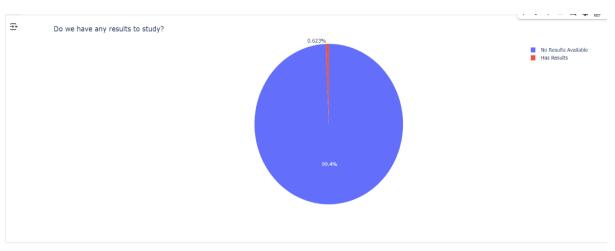
Observations:

Most studies have taken data from All Genders;

In (Adult) and (Child, Adult) Category there is significant number of Female patients considered for the studies

4.8 Exploring study results

```
import plotly.express as px
fig = px.pie(df,'Study Results')
fig.update_layout(title='Do we have any results to study?')
fig.show()
```

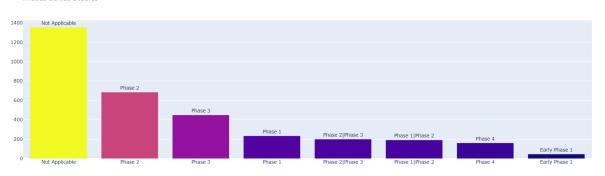


0.038% studies have attained some results; Remaining 99.97% results have NO RESULTS

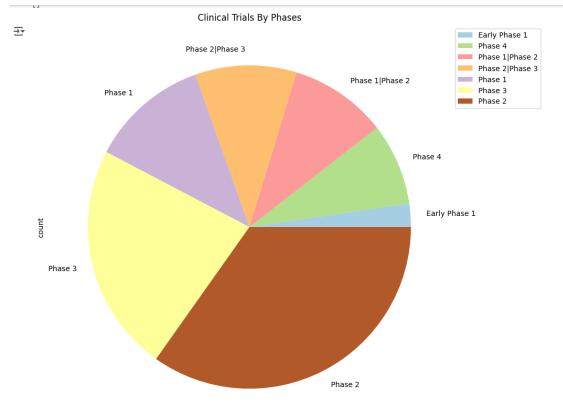
4.9 Exploring Study Phases

```
fig = go.Figure(go.Bar(
    x= df.groupby('Phases').agg('count')['Rank'].sort_values(ascending=False).index,
    y= df.groupby('Phases').agg('count')['Rank'].sort_values(ascending=False).values,
    text=df.groupby('Phases').agg('count')['Rank'].sort_values(ascending=False).index,
    textposition='outside',
    marker_color=df.groupby('Phases').agg('count')['Rank'].sort_values(ascending=False).values
))
fig.update_layout(title='Phases across Studies')
fig.show()
```

Phases across Studies







Observation: Most Study where applicable are in Phase 2 and Phase 3

```
df.Interventions.unique()
 array(['Drug: Drug COVID19-0001-USR|Drug: normal saline',
           'Other: Lung CT scan analysis in COVID-19 patients',
           'Diagnostic Test: COVID 19 Diagnostic Test', ...,
           'Biological: FluBlok|Other: Placebo',
           'Biological: ASP2390|Biological: Placebo',
           'Other: Antibiotic treatment Other: No antibiotic treatment'],
         dtype=object)
/ [239] interventions = df[df['Study Type']=='Interventional']
     interventions['Interventions'].value_counts().head(6).sort_values().plot(kind='barh', color='g', title='Top 5 Drugs tested')
  Axes: title={'center': 'Top 5 Drugs tested'}, ylabel='Interventions'>
                                                               Top 5 Drugs tested
                      Biological: Convalescent plasma
                          Drug: Hydroxychloroquine
                                Drug: Tocilizumab
                Drug: Hydroxychloroquine|Drug: Placebo
                     Drug: Nitazoxanide|Drug: Placebo
        Drug: Hydroxychloroquine|Drug: Placebo oral tablet
                                                                               10
                                                                                      12
                                                                                             14
```

"For each row, if a cell is missing, fill it with the value from the left."

df.head()

	Rank	NCT Number	Title	Acronym	Status	Study Results	Conditions	Interventions	Outcome Measures	Sponsor/Collaborators	 Other IDs	Start Date	Primary Completion Date	Completion Date	First Posted	Results First Postec
1	2	NCT04595136	Study to Evaluate the Efficacy of COVID19- 0001	COVID- 19	Not yet recruiting	No Results Available	SARS-CoV-2 Infection	Drug: Drug COVID19- 0001- USR Drug: normal saline	Change on viral load results from baseline aft	United Medical Specialties	COVID19- 0001-USR	November 2, 2020	December 15, 2020	January 29, 2021	October 20, 2020	October 20 2020
2	3	NCT04395482	Lung CT Scan Analysis of SARS-CoV2 Induced Lun	TAC- COVID19	Recruiting	No Results Available	covid19	Other: Lung CT scan analysis in COVID-19 patients	A qualitative analysis of parenchymal lung dam	University of Milano Bicocca	TAC- COVID19	May 7, 2020	June 15, 2021	June 15, 2021	May 20, 2020	May 20 2020
3	4	NCT04416061	The Role of a Private Hospital in Hong Kong Am	COVID- 19	Active, not recruiting	No Results Available	COVID	Diagnostic Test COVID 19 Diagnostic Test	Proportion of asymptomatic subjects Proportion	Hong Kong Sanatorium & Hospital	RC-2020- 08	May 25, 2020	July 31, 2020	August 31, 2020	June 4, 2020	June 4 2020
4	5	NCT04395924	Maternal- foetal Transmission of SARS- Cov-2	TMF- COVID- 19	Recruiting	No Results Available	Maternal Fetal Infection Transmission COVID- 19	Diagnostic Test Diagnosis of SARS-Cov2 by RT	COVID-19 by positive PCR in cord blood and / o	Centre Hospitalier Régional d'Orléans Centre d	CHRO- 2020-10	May 5, 2020	May 2021	May 2021	May 20, 2020	May 20 2020
5	6	NCT04516954	Convalescent Plasma for COVID-19 Patients	CPCP	Enrolling by invitation	No Results Available	COVID 19	Biological: Convalescent COVID 19 Plasma	Evaluate the safety Change in requirement for	Vinmec Research Institute of Stem Cell and Gen	ISC.20.11.1	August 1, 2020	November 30, 2020	December 30, 2020	August 18, 2020	Augusi 18 2020

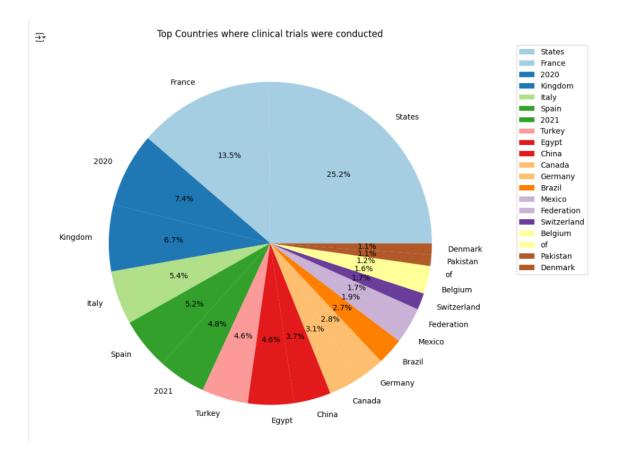
[414] df.isnull().sum()



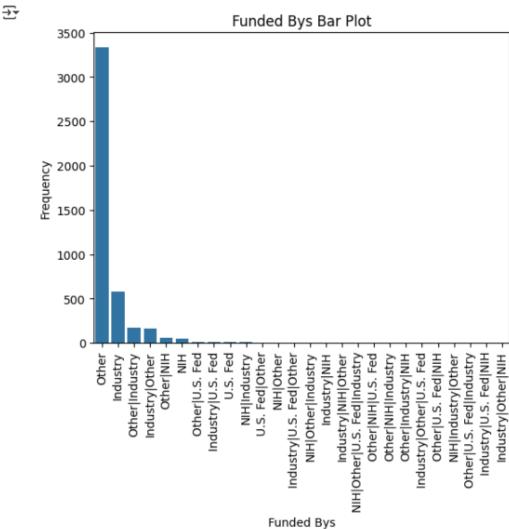
	0
Rank	0
NCT Number	0
Title	0
Acronym	0
Status	0
Study Results	0
Conditions	0
Interventions	0
Outcome Measures	0
Sponsor/Collaborators	0
Gender	0
Age	0
Phases	0
Enrollment	0
Funded Bys	0
Study Type	0
Study Designs	0
Other IDs	0
Start Date	0
Primary Completion Date	0
Completion Date	0
First Posted	0
Results First Posted	0
Last Update Posted	0
Locations	0
Study Documents	0
URL	0

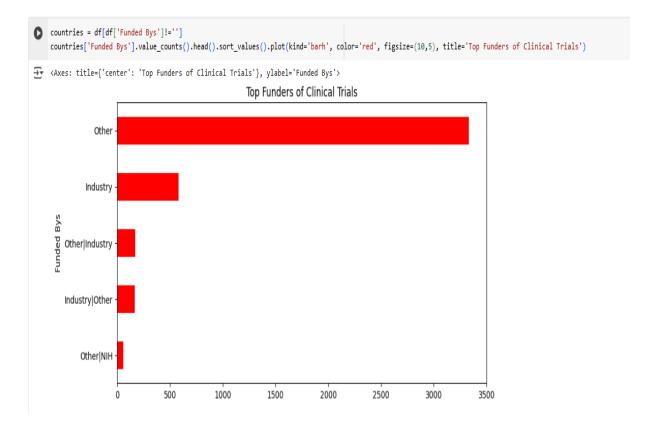
```
[415] df['country'] = [coutry.split()[-1] for coutry in df.Locations]
countries = df[df['country']!='']
     countries['country'].value_counts().head(20).sort_values().plot(kind='barh', color='red', figsize=(10,5), title='Top Countries where clinical triala where conducted')
→ <Axes: title={'center': 'Top Countries where clinical triala where conducted'}, ylabel='country'>
                                          Top Countries where clinical triala where conducted
              States -
             France
               2020
           Kingdom
               Italy
               Spain
               2021
             Turkey
              Egypt
              China
            Canada
           Germany
             Mexico
          Federation
         Switzerland -
            Belgium
           Denmark
            Pakistan
                                                                                                    800
```

```
# Filter top 20 countries by number of clinical trials
    top_countries = df['country'].value_counts().head(20)
    # Plot
    ax = top_countries.plot(
        kind='pie',
        figsize=(10, 10),
        colormap='Paired',
        autopct='%1.1f%%',
                           # Show percentage
        title='Top Countries where clinical trials were conducted',
        ylabel='',
                            # Hide the y-label
        legend=False
    # Display the plot
    plt.legend(loc='upper left', bbox_to_anchor=(1.1, 1.0))
    plt.tight_layout()
    plt.show()
```



Observation: Most studies have taken data from State and France



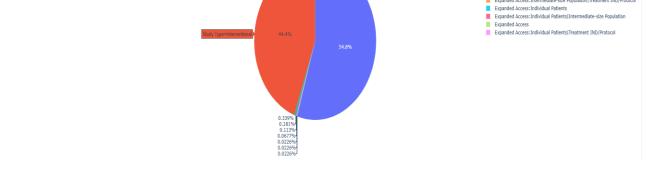


Observation: Max Funding is by Industry

```
import plotly.express as px
fig = px.pie(df,'Study Type')
fig.update_layout(title='Do we have any results to study?')
fig.show()

Do we have any results to study?

Do we have any results to study?
```



4.10 States Explore the Gender distribution in the studies

```
[637] dfc = df.groupby('country')
```

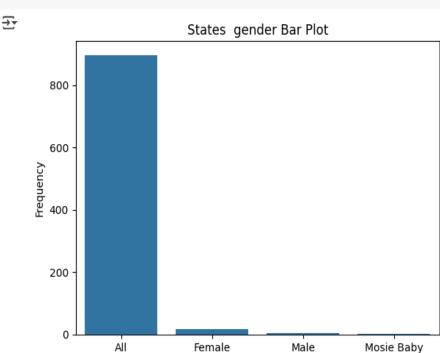


• Shows the first row for each country group

• dfcc is a new dataframe that contains only studies from "States".

dfcc.head()																			
•	Rank	NCT Number	Title	Acronym	Status	Study Results	Conditions	Interventions	Outcome Measures	Sponsor/Collaborators		Start Date	Primary Completion Date	Completion Date	First Posted	Results First Posted	Last Update Posted	Locations	Study Documents
10	11	NCT04355897	CoVID-19 Plasma in Treatment of COVID-19 Patients	CoVID-19 Plasma in Treatment of COVID-19 Patients	Recruiting	No Results Available	COVID 19	Biological: Convalescent COVID 19 Plasma	Reduce mortality Reduce requirement for mechan	The Christ Hospital		April 28, 2020	July 2020	August 2020	April 21, 2020	April 21, 2020	May 20, 2020	The Christ Hospital, Cincinnati, Ohio, United	The Christ Hospital, Cincinnati, Ohio, United
12	13	NCT04659759	COVID-19 Pregnancy Related Immunological, Clin	COVID- PRICE	Recruiting	No Results Available	Covid19	Other: COVID-19 exposure Biological: COVID-19	Maternal COVID- 19 serology (IgG and IgM) Mater	Thomas Jefferson University[Nemours	***	November 17, 2020	December 31, 2021	June 30, 2022	December 9, 2020	December 9, 2020	March 5, 2021	Thomas Jefferson University Hospital, Philadel	Thomas Jefferson University Hospital, Philadel
27	28	NCT04424004	MURDOCK Cabarrus County COVID-19 Prevalence an	C3PI	Active, not recruiting	No Results Available	COVID 19	Other: COVID-19 PCR and serology testing	Estimate the prevalence of COVID-19 infection	Duke University North Carolina Department of H		June 9, 2020	June 30, 2021	June 30, 2021	June 9, 2020	June 9, 2020	November 17, 2020	Duke CTSI Translational Population Health Offi	Duke CTSI Translational Population Health Offi
37	38	NCT04372004	Comparison of the Efficacy of Rapid Tests to I	CATCH COVID-19	Recruiting	No Results Available	COVID-19	Diagnostic Test: diagnostic tests for COVID-19	detection of viral infection using serology an	Texas Cardiac Arrhythmia Research Foundation	***	May 8, 2020	May 2021	June 2021	May 1, 2020	May 1, 2020	August 11, 2020	Texas Cardiac Arrhythmia Institute, Austin, Te	Texas Cardiac Arrhythmia Institute, Austin, Te
40	41	NCT04412486	COVID-19 Convalescent Plasma (CCP) Transfusion	COVID-19 Convalescent Plasma (CCP) Transfusion	Recruiting	No Results Available	COVID-19	Biological: COVID Convalescent Plasma	Change in PaO2/FiO2 after CCP transfusion. Cha	Gailen D. Marshall Jr., MD PhD University of M		June 1, 2020	May 31, 2022	May 31, 2022	June 2, 2020	June 2, 2020	July 2, 2020	University of Mississippi Medical Center, Jack	University of Mississippi Medical Center, Jack

```
sns.barplot(x=dfcc['Gender'].value_counts().index,
              y=dfcc['Gender'].value_counts().values)
plt.xlabel('Gender')
plt.ylabel('Frequency')
plt.title('States gender Bar Plot')
plt.show()
```



5. Conclusion

1. Top Countries for Clinical Trials

The United States (States) conducted the maximum number of COVID-19 clinical trials.

Male

Gender

- France, United Kingdom, Italy, and Spain were also major contributors.
- This shows that developed countries were leading clinical research during the pandemic.

2. Funding Sources

- Most clinical trials were funded under the category "Other".
- The second largest funding source was "Industry" (pharmaceutical companies, biotech firms).
- Combination funding like **Industry**|Other also appeared but was much less frequent.

• Conclusion: Clinical trials were majorly privately or independently funded rather than solely by governments or official health bodies.

3. Gender Distribution in Studies (for 'States')

- A majority of clinical trials were open to **All Genders** (both male and female participants).
- Some studies focused specifically on **Male** or **Female** participants, but these were much fewer.
- **Conclusion**: Researchers mostly designed studies to include **both genders**, aiming for a broader understanding of COVID-19's impact.

4. Timeline Observations (if any)

• A lot of trials spiked around **2020-2021**, which aligns perfectly with the global emergency and vaccine development phases.