In the following section, I load the COVID-19 dataset and filter it to include only the columns related to cases and deaths. The dataset is in wide format, so I use pd.melt() to reshape it into a long format, making it easier to perform time-based analysis. Then I converted the 'date' column to datetime format.

```
In [21]: import pandas as pd
           # Load the dataset
           file_path = '/Users/amnasohail/Desktop/405Stage2indv/final_merged_data.csv'
           state_data = pd.read_csv(file_path)
           # Filter columns for cases and deaths
           # Only keep columns that contain '_cases' or '_deaths' in their names, and exclude non-relevant columns
           cases_columns = [col for col in state_data.columns if '_cases' in col] deaths_columns = [col for col in state_data.columns if '_deaths' in col]
           # Filter for case and death columns and necessary identifiers
           filtered_cases = ['County Name', 'State', 'population'] + cases_columns
filtered_deaths = ['County Name', 'State', 'population'] + deaths_columns
           filtered cases data = state data[filtered cases]
           filtered_deaths_data = state_data[filtered_deaths]
           # Melt the data to long format for cases
           data_cases_long = pd.melt(
                filtered_cases_data,
                id_vars=['County Name', 'State', 'population'],
                value_vars=cases_columns, # Only melt the case columns
                var_name='date',
                value_name='cases'
           # Melt the data to long format for deaths
           data_deaths_long = pd.melt(
                filtered_deaths_data,
                id_vars=['County Name', 'State', 'population'],
                value_vars=deaths_columns, # Only melt the death columns
                var_name='date',
                value_name='deaths'
           # Clean the 'date' column by removing '_cases' or '_deaths' and convert to datetime
data_cases_long['date'] = pd.to_datetime(data_cases_long['date'].str.replace('_cases', ''), format='%Y-%m-%d
data_deaths_long['date'] = pd.to_datetime(data_deaths_long['date'].str.replace('_deaths', ''), format='%Y-%m
           # Group by State and Date, and calculate weekly sums
           weekly_cases = data_cases_long.groupby(['State', pd.Grouper(key='date', freq='W')]).sum().reset_index()
weekly_deaths = data_deaths_long.groupby(['State', pd.Grouper(key='date', freq='W')]).sum().reset_index()
           # List of states to filter
           selected_states = ['AL', 'CA', 'WA', 'SC', 'MI', 'MA']
           # Filter weekly cases for the selected states
           weekly_cases_filtered = weekly_cases[weekly_cases['State'].isin(selected_states)]
           # Filter weekly deaths for the selected states
           weekly_deaths_filtered = weekly_deaths[weekly_deaths['State'] isin(selected_states)]
           # Display the filtered weekly data for cases and deaths
           print(weekly_cases_filtered.head())
           print(weekly_deaths_filtered.head())
```

```
State
                          date
                                                                        County Name \
         182
                 AL 2020-01-26 Autauga County Baldwin County Barbour County B...
         183
                 AL 2020-02-02
                                Autauga County Baldwin County Barbour County B...
         184
                 AL 2020-02-09
                                Autauga County Baldwin County Barbour County B...
         185
                 AL 2020-02-16 Autauga County Baldwin County Barbour County B...
                 AL 2020-02-23 Autauga County Baldwin County Barbour County B...
         186
              population cases
         182
                 24515925
         183
                 34322295
                               0
         184
                 34322295
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                 34322295
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             State
                          date
                                                                        County Name \
                                Autauga County Baldwin County Barbour County B...
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                 AL 2020-01-26
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                 AL 2020-02-09 Autauga County Baldwin County Barbour County B...
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                 AL 2020-02-16 Autauga County Baldwin County Barbour County B...
         186
                 AL 2020-02-23
                                Autauga County Baldwin County Barbour County B...
              population deaths
         182
                 24515925
                                a
         183
                 34322295
                                0
         184
                 34322295
                                0
         185
                 34322295
                                0
         186
                 34322295
                                0
In [35]: # Filter the data for Alabama
         alabama_cases = weekly_cases_filtered[weekly_cases_filtered['State'] == 'AL'].copy()
         alabama_deaths = weekly_deaths_filtered[weekly_deaths_filtered['State'] == 'AL'].copy()
         # Normalize cases and deaths per 100,000 people for Alabama using .loc to avoid SettingWithCopyWarning
         alabama_cases.loc[:, 'cases_per_100k'] = (alabama_cases['cases'] / alabama_cases['population']) * 100000
alabama_deaths.loc[:, 'deaths_per_100k'] = (alabama_deaths['deaths'] / alabama_deaths['population']) * 10000
         # Calculate weekly mean, median, and mode for cases and deaths in Alabama
         weekly_mean_cases_al = alabama_cases_stats['mean_cases'].mean()
         weekly_median_cases_al = alabama_cases_stats['median_cases'].median()
         weekly_mode_cases_al = alabama_cases_stats['mode_cases'].mode()[0]
         weekly_mean_deaths_al = alabama_deaths_stats['mean_deaths'].mean()
         weekly_median_deaths_al = alabama_deaths_stats['median_deaths'].median()
         weekly_mode_deaths_al = alabama_deaths_stats['mode_deaths'].mode()[0]
         # Display the weekly statistics for Alabama in a formatted manner
         print(f"Weekly Mean of Cases in Alabama: {weekly_mean_cases_al}")
         print(f"Weekly Median of Cases in Alabama: {weekly_median_cases_al}")
         print(f"Weekly Mode of Cases in Alabama: {weekly_mode_cases_al}")
         print()
         print(f"Weekly Mean of Deaths in Alabama: {weekly_mean_deaths_al}")
         print(f"Weekly Median of Deaths in Alabama: {weekly_median_deaths_al}")
         print(f"Weekly Mode of Deaths in Alabama: {weekly_mode_deaths_al}")
         print()
         Weekly Mean of Cases in Alabama: 17380.16780405117
         Weekly Median of Cases in Alabama: 16602.770881143
         Weekly Mode of Cases in Alabama: 33854.23964219176
         Weekly Mean of Deaths in Alabama: 258.0357283298914
         Weekly Median of Deaths in Alabama: 309.48105305895194
         Weekly Mode of Deaths in Alabama: 0.0
```

```
In [50]: # List of states
            other_states = ['CA', 'WA', 'SC', 'MI', 'MA']
            for state in other_states:
                 print(f"Weekly statistics for {state}:\n")
                 # Filter data for each state
                 state_cases = weekly_cases_filtered[weekly_cases_filtered['State'] == state].copy()
                 state_deaths = weekly_deaths_filtered[weekly_deaths_filtered['State'] == state].copy()
                 # Normalize cases and deaths per 100,000 people for the state
                 state_cases['cases_per_100k'] = (state_cases['cases'] / state_cases['population']) * 100000
state_deaths['deaths_per_100k'] = (state_deaths['deaths'] / state_deaths['population']) * 100000
                 # Calculate weekly mean, median, and mode for cases
                weekly_mean_cases = state_cases.groupby(pd.Grouper(key='date', freq='W'))['cases_per_100k'].mean().mean(
weekly_median_cases = state_cases.groupby(pd.Grouper(key='date', freq='W'))['cases_per_100k'].median().m
weekly_mode_cases = state_cases.groupby(pd.Grouper(key='date', freq='W'))['cases_per_100k'].apply(lambda
                 # Calculate weekly mean, median, and mode for deaths
                 weekly_mean_deaths = state_deaths.groupby(pd.Grouper(key='date', freq='W'))['deaths_per_100k'].mean().me
                 weekly_median_deaths = state_deaths.groupby(pd.Grouper(key='date', freq='W'))['deaths_per_100k'].median(
weekly_mode_deaths = state_deaths.groupby(pd.Grouper(key='date', freq='W'))['deaths_per_100k'].apply(lam
                 # Display the weekly statistics for the states
                 print(f"Weekly Mean of Cases in {state}: {weekly_mean_cases}")
                 print(f"Weekly Median of Cases in {state}: {weekly_median_cases}")
                 print(f"Weekly Mode of Cases in {state}: {weekly_mode_cases}\n")
                 print(f"Weekly Mean of Deaths in {state}: {weekly_mean_deaths}")
                 print(f"Weekly Median of Deaths in {state}: {weekly_median_deaths}")
                 print(f"Weekly Mode of Deaths in {state}: {weekly_mode_deaths}\n")
```

### Weekly statistics for CA:

- Weekly Mean of Cases in CA: 14396.782024823066
- Weekly Median of Cases in CA: 11624.479196873188
- Weekly Mode of Cases in CA: 0.0
- Weekly Mean of Deaths in CA: 157.67247795422975
- Weekly Median of Deaths in CA: 177.96301376701297
- Weekly Mode of Deaths in CA: 0.0

### Weekly statistics for WA:

- Weekly Mean of Cases in WA: 12098.968116695483
- Weekly Median of Cases in WA: 9151.312340923803
- Weekly Mode of Cases in WA: 0.013132160885254724
- Weekly Mean of Deaths in WA: 110.55408939904117
- Weekly Median of Deaths in WA: 108.64518103818776
- Weekly Mode of Deaths in WA: 0.0

#### Weekly statistics for SC:

- Weekly Mean of Cases in SC: 16654.692497998785
- Weekly Median of Cases in SC: 17205.989967314446
- Weekly Mode of Cases in SC: 28777.011113843186
- Weekly Mean of Deaths in SC: 218.0548464755134
- Weekly Median of Deaths in SC: 257.80223954952635
- Weekly Mode of Deaths in SC: 347.0575370859597

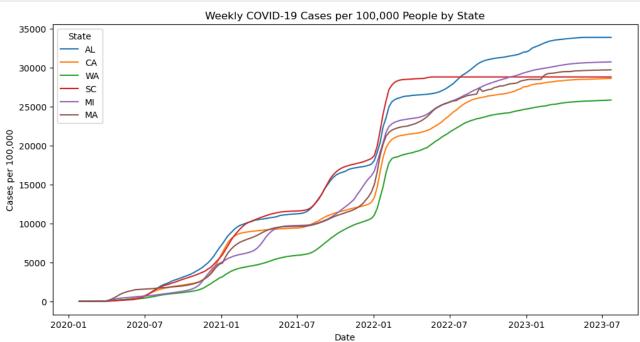
### Weekly statistics for MI:

- Weekly Mean of Cases in MI: 15102.57338205475
- Weekly Median of Cases in MI: 11967.98738296028
- Weekly Mode of Cases in MI: 0.0
- Weekly Mean of Deaths in MI: 242.4092866468423
- Weekly Median of Deaths in MI: 228.80858025989272
- Weekly Mode of Deaths in MI: 0.0

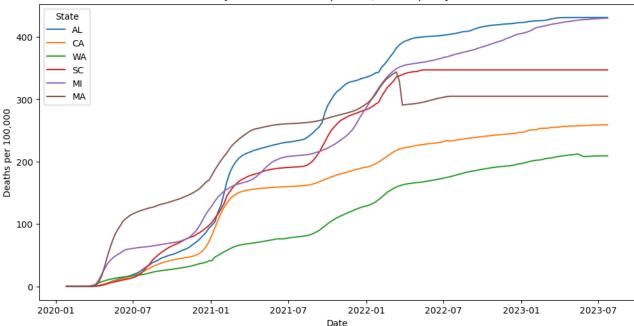
## Weekly statistics for MA:

- Weekly Mean of Cases in MA: 14902.080249095174
- Weekly Median of Cases in MA: 11291.142813327133
- Weekly Mode of Cases in MA: 28464.652100985666
- Weekly Mean of Deaths in MA: 233.3014360274787
- Weekly Median of Deaths in MA: 273.1943471251093
- Weekly Mode of Deaths in MA: 304.8384599905143

```
In [48]: import matplotlib.pyplot as plt
         import seaborn as sns
         # list of states including Alabama
         selected_states = ['AL', 'CA', 'WA', 'SC', 'MI', 'MA']
         # empty lists to store the data for cases and deaths
         consolidated_cases_data = []
         consolidated_deaths_data = []
         for state in selected_states:
             # Filter data for each state
             state_cases = weekly_cases_filtered[weekly_cases_filtered['State'] == state].copy()
             state_deaths = weekly_deaths_filtered[weekly_deaths_filtered['State'] == state].copy()
             # Normalize cases and deaths per 100,000 people for the state
state_cases['cases_per_100k'] = (state_cases['cases'] / state_cases['population']) * 100000
             state_deaths['deaths_per_100k'] = (state_deaths['deaths'] / state_deaths['population']) * 100000
             consolidated_cases_data.append(state_cases)
             consolidated_deaths_data.append(state_deaths)
         # Combine all the states' data into a single DataFrame for cases and deaths
         consolidated_cases_data = pd.concat(consolidated_cases_data)
         consolidated_deaths_data = pd.concat(consolidated_deaths_data)
         # Plot Weekly COVID-19 Cases per 100,000 People for Each State
         plt.figure(figsize=(12, 6))
         sns.lineplot(data=consolidated_cases_data, x='date', y='cases_per_100k', hue='State')
         plt.title('Weekly COVID-19 Cases per 100,000 People by State')
         plt.xlabel('Date')
         plt.ylabel('Cases per 100,000')
         plt.legend(title='State')
         plt.show()
         # Plot Weekly COVID-19 Deaths per 100,000 People for Each State
         plt.figure(figsize=(12, 6))
         sns.lineplot(data=consolidated_deaths_data, x='date', y='deaths_per_100k', hue='State')
         plt.title('Weekly COVID-19 Deaths per 100,000 People by State')
         plt.xlabel('Date')
         plt.ylabel('Deaths per 100,000')
         plt.legend(title='State')
         plt.show()
```



# Weekly COVID-19 Deaths per 100,000 People by State



Analysis of Weekly COVID-19 Trends Across States

Some states, such as South Carolina and Michigan, experienced higher peaks in cases, but Washington and Massachusetts had lower but more sustained case numbers. The timing and the magnitude of deaths peaks was different across states, which I think shows differences in healthcare systems, population density, and public health responses.

Why the Rates Differ Across States

factors that could have contributed to the differences in COVID-19 rates across these states:
-Population Density: States with larger urban populations, such as California and Massachusetts, may have seen more rapid transmission due to closer physical contact.

-Healthcare Capacity: States with stronger healthcare infrastructure, like Massachusetts, may have been better equipped to handle severe cases, leading to lower death rates even with high case numbers.
-Public Health Measures: Differences in state policies (e.g., mask mandates, lockdowns) affected the spread of the virus. For example, stricter measures in Washington likely helped control the spread earlier in the nandemic.

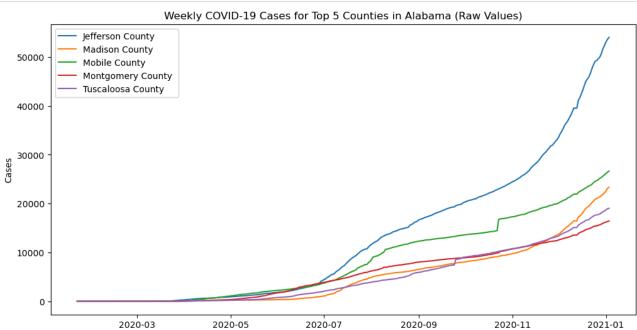
```
In [51]:
         # Find the peak week for cases for each state
         peak_cases_week = consolidated_cases_data.loc[consolidated_cases_data.groupby('State')['cases_per_100k'].idx
         # Find the peak week for deaths for each state
         peak_deaths_week = consolidated_deaths_data.loc[consolidated_deaths_data.groupby('State')['deaths_per_100k']
         # Display the peak weeks for cases and deaths
         print("Peak weeks for cases:")
         print(peak_cases_week[['State', 'date', 'cases_per_100k']])
         print("\nPeak weeks for deaths:")
         print(peak_deaths_week[['State', 'date', 'deaths_per_100k']])
         Peak weeks for cases:
              State
                           date
                                 cases_per_100k
         355
                 AL 2023-05-21
                                   33854.239642
         909
                 CA 2023-07-23
                                   28589.035058
         3639
                 MA 2023-07-23
                                   29695.115320
         4185
                 MI 2023-07-23
                                   30704.008564
                 SC 2022-05-22
                                   28777.011114
         7401
         8735
                 WA 2023-07-23
                                   25819.287846
         Peak weeks for deaths:
                                 deaths_per_100k
              State
                           date
                                      431.107535
         355
                 AL 2023-05-21
         909
                 CA 2023-07-23
                                      259.048953
         3569
                 MA 2022-03-13
                                      343.706778
         4185
                 MI 2023-07-23
                                      429.996573
         7401
                 SC 2022-05-22
                                      347.057537
         8726
                 WA 2023-05-14
                                      212.325155
```

The peaks in COVID-19 cases and deaths across the states are probably related to factors like population density, public health measures, and vaccination rates. States like California, Massachusetts, and Michigan had case peaks in July 2023, most likely because of summer travel and gatherings. South Carolina peaked earlier in May 2022, reflecting looser restrictions. Death peaks usually followed case peaks by a few weeks. States like Alabama and South Carolina, with earlier peaks, probably struggled with lower vaccination rates and healthcare capacity. Overall, these patterns match national trends, with spikes occurring during periods of increased travels and social interaction.

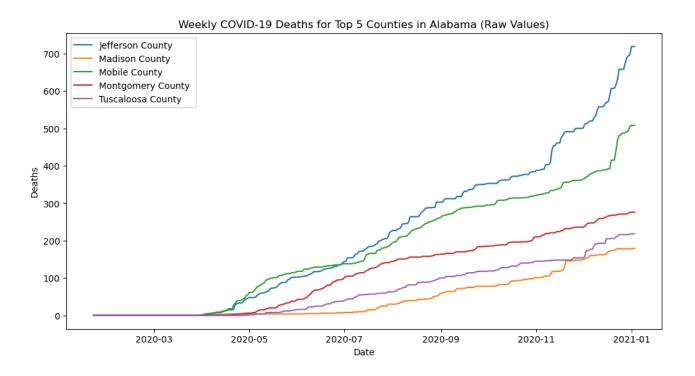
```
In [69]: # Define the end date
         end_date = '2021-01-03'
         # Filter the data for Alabama
         alabama_cases = data_cases_long[data_cases_long['State'] == 'AL']
         alabama_deaths = data_deaths_long[data_deaths_long['State'] == 'AL']
         # Filter the data for dates up to and including the end_date
         alabama_cases_until_date = alabama_cases[alabama_cases['date'] <= end_date]</pre>
         alabama_deaths_until_date = alabama_deaths[alabama_deaths['date'] <= end_date]</pre>
         # Group by county and sum up the cases and deaths up to the end date
         total_cases = alabama_cases_until_date.groupby('County Name').agg({
             'cases': 'sum',
'population': 'max'
         }).reset_index()
         total_deaths = alabama_deaths_until_date.groupby('County Name').agg({
              'deaths': 'sum',
             'population': 'max
         }).reset_index()
         # Sort and get the highest case and death rates by county
         top case counties = total cases.nlargest(5, 'cases')[['County Name', 'cases']]
         top_death_counties = total_deaths.nlargest(5, 'deaths')[['County Name', 'deaths']]
         # Print top counties by total cases and deaths until the end date
         print(f"Top 5 Counties by Number of Cases up to {end_date}")
         print(top_case_counties)
         print(f"\nTop 5 Counties by Number of Deaths up to {end_date}")
         print(top_death_counties)
         # Get the list of top 5 county names
         top_5_county_names = top_case_counties['County Name'].tolist()
         Top 5 Counties by Number of Cases up to 2021-01-03
                    County Name
                                   cases
              Jefferson County
                                 4399139
         12
                 Mobile County
                                 2857206
         50
             Montgomery County
                                 1884843
                                 1764606
         44
                Madison County
         62 Tuscaloosa County
                                 1720640
         Top 5 Counties by Number of Deaths up to 2021-01-03
                    County Name deaths
         36
              Jefferson County
                                  75338
                 Mobile County
                                  62330
                                  38269
         50 Montgomery County
         62 Tuscaloosa County
                                  23862
         61 Tallapoosa County
                                  19576
In [63]: # Filter data for the top 5 counties
         top 5 county cases = alabama cases until date[alabama cases until date['County Name'].isin(top 5 county name
         top_5_county_deaths = alabama_deaths_until_date[alabama_deaths_until_date['County Name'].isin(top_5_county_n
```

```
In [64]: # Plot weekly cases for the top 5 counties
plt.figure(figsize=(12, 6))
sns.lineplot(data=top_5_county_cases, x='date', y='cases', hue='County Name')
plt.title(f'Weekly COVID-19 Cases for Top 5 Counties in Alabama (Raw Values)')
plt.ylabel('Date')
plt.ylabel('Cases')
plt.legend()
plt.show()

# Plot weekly deaths for the top 5 counties
plt.figure(figsize=(12, 6))
sns.lineplot(data=top_5_county_deaths, x='date', y='deaths', hue='County Name')
plt.title(f'Weekly COVID-19 Deaths for Top 5 Counties in Alabama (Raw Values)')
plt.ylabel('Date')
plt.ylabel('Date')
plt.legend()
plt.show()
```

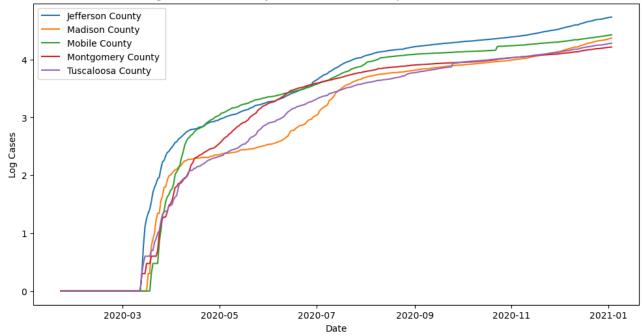


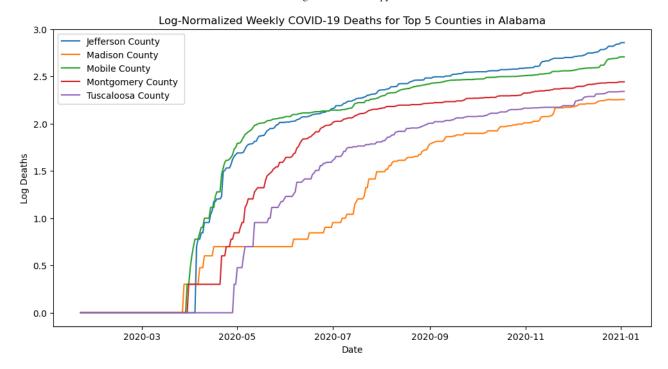
Date



```
In [67]: import numpy as np
           # Log-normalize the cases and deaths using .loc[] to avoid the SettingWithCopyWarning top_5_county_cases.loc[:, 'log_cases'] = np.log10(top_5_county_cases['cases'] + 1) top_5_county_deaths.loc[:, 'log_deaths'] = np.log10(top_5_county_deaths['deaths'] + 1)
           # Plot log-normalized weekly cases for the top 5 counties
           plt.figure(figsize=(12, 6))
           sns.lineplot(data=top_5_county_cases, x='date', y='log_cases', hue='County Name')
           plt.title(f'Log-Normalized Weekly COVID-19 Cases for Top 5 Counties in Alabama')
           plt.xlabel('Date')
           plt.ylabel('Log Cases')
           plt.legend()
           plt.show()
           # Plot log-normalized weekly deaths for the top 5 counties
           plt.figure(figsize=(12, 6))
sns.lineplot(data=top_5_county_deaths, x='date', y='log_deaths', hue='County_Name')
           plt title(f'Log-Normalized Weekly COVID-19 Deaths for Top 5 Counties in Alabama')
           plt.xlabel('Date')
           plt ylabel('Log Deaths')
           plt.legend()
           plt.show()
```







The COVID-19 peaks in Alabama's top counties, like Jefferson and Mobile, were mainly becayse of their larger populations and urban settings, which made it easier for the virus to spread. The spikes in cases and deaths around May 2023 suggest that public health measures and vaccines might have taken longer to have an effect in these areas. While all the counties followed the same general trend as the state, the number of cases and deaths varied depending on how big and crowded the counties are. Overall, these counties' trends fit in with what was happening across Alabama at the time.