

```

from google.cloud import bigquery
from google.colab import auth
import matplotlib.pyplot as plt
import pandas as pd

auth.authenticate_service_account()
client = bigquery.Client()
project_name='weatherlink-404323'

```

Successfully saved credentials for pipelineauth@weatherlink-404323.iam.gserviceaccount.com

```

dataset = 'weatherlink-master'

# Define your SQL query
query = """
SELECT
  a.number_of_fatalities, a.geoid, a.year, a.timestamp_of_crash, a.hour_of_arrival,
  c.income_per_capita, c.black_pop, c.total_pop, c.hispanic_pop, c.asian_pop, c.american_indian_pop
FROM
  `weatherlink-404323.weatherlink_master.census_master` c
JOIN
  `weatherlink-404323.weatherlink_master.accident_master` a
ON
  c.YEAR = a.YEAR AND c.geo_id = a.GEOID
"""

```

```

# Execute the query and store the results in a Pandas DataFrame
df = client.query(query).to_dataframe()

```

```
print(df)
```

122001	99	99	9
122002	99	99	0
122003	99	99	20
122004	99	99	15
122005	99	99	15

```
minute_of_crash  hour_of_ems_arrival_at_hospital  \
```

0	43	3
1	18	0
2	53	1
3	43	1
4	36	0
...
122001	13	88
122002	0	99
122003	30	99
122004	23	99
122005	0	99

	minute_of_ems_arrival_at_hospital	hour_of_notification	\
0	13	0	
1	42	0	
2	35	0	
3	7	0	
4	48	0	
...	
122001	88	99	
122002	99	99	
122003	99	99	
122004	99	99	
122005	99	99	

	minute_of_notification	income_per_capita	black_pop	total_pop	\
0	47	22926.0	2219.0	204737.0	
1	20	37894.0	NaN	65018.0	
2	45	27582.0	NaN	119980.0	
3	43	26971.0	NaN	253178.0	
4	37	28227.0	49780.0	633473.0	
...	
122001	99	29760.0	416126.0	2716940.0	
122002	99	34907.0	22202.0	1160437.0	
122003	99	32998.0	233317.0	1471968.0	
122004	99	16199.0	721.0	71887.0	
122005	99	34744.0	248228.0	4485414.0	

	hispanic_pop	asian_pop	amerindian_pop
0	33111.0	2479.0	5440.0
1	NaN	NaN	NaN
2	NaN	NaN	NaN
3	NaN	NaN	NaN
4	134895.0	26050.0	2223.0
...
122001	1886364.0	39907.0	2217.0
122002	218243.0	48128.0	7734.0
122003	436896.0	59534.0	2035.0
122004	4736.0	160.0	52536.0
122005	1408855.0	187233.0	76504.0

```

# Count the number of rows with null values in time-related columns
num_nulls = df[['hour_of_notification', 'minute_of_notification', 'hour_of_crash']]

print(f"Number of Rows with Null Values: {num_nulls}")

    Number of Rows with Null Values: 0

# Create a function to convert hour and minute to total minutes
def time_to_minutes(hour, minute):
    return hour * 60 + minute

# Calculate how long until the notification came in
df['NotificationResponseTime'] = df.apply(lambda row: (time_to_minutes(row['hour_of_notification'], row['minute_of_notification'])), axis=1)

# Calculate response time for emergency services (from when the crash happened to when they arrived)
df['EmsResponseTime'] = df.apply(lambda row: (time_to_minutes(row['hour_of_crash'], row['minute_of_crash']) - time_to_minutes(row['hour_of_arrival'], row['minute_of_arrival'])), axis=1)

# Calculate delivery time for emergency services (when they get to the hospital from when they arrived)
df['EmsDeliveryTime'] = df.apply(lambda row: (time_to_minutes(row['hour_of_crash'], row['minute_of_crash']) - time_to_minutes(row['hour_of_hospital'], row['minute_of_hospital'])), axis=1)

# Calculate total response time from when the crash happened to when they made it to the hospital
df['TotalResponseTime'] = df.apply(lambda row: (time_to_minutes(row['hour_of_crash'], row['minute_of_crash']) - time_to_minutes(row['hour_of_hospital'], row['minute_of_hospital'])), axis=1)

# Print a few rows to verify the calculations
print("Sample Rows:")
print(df[['NotificationResponseTime', 'EmsResponseTime', 'EmsDeliveryTime', 'TotalResponseTime']])

# Print averages
average_notification_response_time = df['NotificationResponseTime'].mean()
average_ems_response_time = df['EmsResponseTime'].mean()
average_ems_delivery_time = df['EmsDeliveryTime'].mean()
average_total_response_time = df['TotalResponseTime'].mean()

print("\nAverages:")
print(f"Notification Response Time: {average_notification_response_time:.2f} minutes")
print(f"EMS Response Time: {average_ems_response_time:.2f} minutes")
print(f"EMS Delivery Time: {average_ems_delivery_time:.2f} minutes")
print(f"Total Response Time: {average_total_response_time:.2f} minutes")

```

Sample Rows:

	NotificationResponseTime	EmsResponseTime	EmsDeliveryTime	\
0	4	17	133	
1	2	8	16	
2	1432	1432	50	
3	0	6	18	
4	1	6	6	

	TotalResponseTime
0	150
1	24
2	42
3	24
4	12

Averages:

Notification Response Time: 465.15 minutes
EMS Response Time: 398.20 minutes
EMS Delivery Time: 298.65 minutes
Total Response Time: 569.26 minutes

```
(time_to_minutes(0, 22) - time_to_minutes(23, 17)) % (24 * 60)
```

65

```
project_id = 'weatherlink-404323'  
dataset_id = 'weatherlink_master'  
table_id = 'accident_response_times'  
table_ref = f'{project_id}.{dataset_id}.{table_id}'
```

```
# Calculate average response times for each county (GE0ID)  
average_response_times = df.groupby('geoid').agg({  
    'TotalResponseTime': 'mean',  
    'EmsResponseTime': 'mean',  
    'EmsDeliveryTime': 'mean',  
    'NotificationResponseTime': 'mean',  
    # Add other columns as needed  
}).reset_index()
```

average_response_times

	geoid	TotalResponseTime	EmsResponseTime	EmsDeliveryTime	NotificationRe
0	1003	670.807453	129.571429	621.732919	
1	1015	592.980000	88.020000	548.160000	
2	1043	572.055046	110.844037	527.266055	
3	1049	544.628205	79.294872	502.256410	
4	1051	664.050000	34.033333	654.016667	
...	
826	55133	456.188119	262.574257	307.673267	
827	55139	665.777778	322.644444	407.133333	
828	55141	617.282051	219.410256	397.871795	
829	56021	385.966102	50.745763	335.220339	
830	56025	352.595745	56.744681	326.489362	

831 rows x 5 columns

```

from io import StringIO
from pandas_gbq import to_gbq
from pandas_gbq.schema import generate_bq_schema

# Define the schema for the new table
schema = [
    bigquery.SchemaField('geoid', 'STRING', mode='REQUIRED'),
    bigquery.SchemaField('TotalResponseTime', 'FLOAT', mode='REQUIRED'),
    bigquery.SchemaField('EmsResponseTime', 'FLOAT', mode='REQUIRED'),
    bigquery.SchemaField('EmsDeliveryTime', 'FLOAT', mode='REQUIRED'),
    bigquery.SchemaField('NotificationResponseTime', 'FLOAT', mode='REQUIRED'),
    # Add other columns as needed
]

# This is a weird work around to get the dataframe acceptable for upload
# temporarily store the dataframe as a csv in a string variable
temp_csv_string = average_response_times.to_csv(sep=";", index=False)
temp_csv_string_IO = StringIO(temp_csv_string)
# create new dataframe from string variable
new_df = pd.read_csv(temp_csv_string_IO, sep=";")

to_gbq(new_df, f"{dataset_id}.{table_id}", project_id=project_id, if_exists='replace')

print(f"New table created: {table_ref}")

100%|██████████| 1/1 [00:00<00:00, 5322.72it/s]New table created: weatherlink-

```

It looks like some of those are erroneously large, so we need to clean the data

```

# Count the number of rows with null values in time-related columns
num_nulls = df[['hour_of_notification', 'minute_of_notification', 'hour_of_crash']]

print(f"Number of Rows with Null Values: {num_nulls}")

Number of Rows with Null Values: 0

```

Wow it looks like there are no null values. So lets remove any huge outliers

```

# Count NaN values for each column
nan_counts = df[['NotificationResponseTime', 'EmsResponseTime', 'EmsDeliveryTime']]

print("NaN Counts:")
print(nan_counts)

# Count NaN values for each column
nan_counts = df[['NotificationResponseTime', 'EmsResponseTime', 'EmsDeliveryTime']]

print("Null Counts:")
print(nan_counts)

NaN Counts:
NotificationResponseTime    0
EmsResponseTime             0
EmsDeliveryTime             0
TotalResponseTime           0
dtype: int64
Null Counts:
NotificationResponseTime    0
EmsResponseTime             0
EmsDeliveryTime             0
TotalResponseTime           0
dtype: int64

import seaborn as sns
import matplotlib.pyplot as plt

# Convert relevant columns to numeric
df[['NotificationResponseTime', 'EmsResponseTime', 'EmsDeliveryTime', 'TotalResponseTime']] = df[['NotificationResponseTime', 'EmsResponseTime', 'EmsDeliveryTime', 'TotalResponseTime']].astype(float)

# Calculate z-scores for all rows
z_scores_all = stats.zscore(df[['NotificationResponseTime', 'EmsResponseTime', 'EmsDeliveryTime', 'TotalResponseTime']])

# Set a z-score threshold for outliers (e.g., 2 standard deviations)
z_threshold = 2

# Create mask for outliers
outliers_mask = (z_scores_all >= z_threshold) | (z_scores_all <= -z_threshold)

# Apply any() along the columns to create a 1D mask
any_outliers_mask = outliers_mask.any(axis=1)

# Filter the DataFrame based on the outliers mask
df_filtered = df.loc[~any_outliers_mask]

# Explicitly select the columns of interest after filtering
columns_of_interest = ['NotificationResponseTime', 'EmsResponseTime', 'EmsDeliveryTime', 'TotalResponseTime']
df_filtered = df_filtered[columns_of_interest]

```

```

# Create a box and whisker plot with the filtered data
plt.figure(figsize=(10, 6))
sns.boxplot(data=df_filtered[columns_of_interest])
plt.title('Box and Whisker Plot of Response Times without Outliers')
plt.show()

# Calculate the number of rows filtered out
num_filtered = df.shape[0] - df.loc[~any_outliers_mask].shape[0]

# Summary stats about the cleaning process
print(f"Using a z score of {z_threshold} you filtered out {num_filtered} values ")

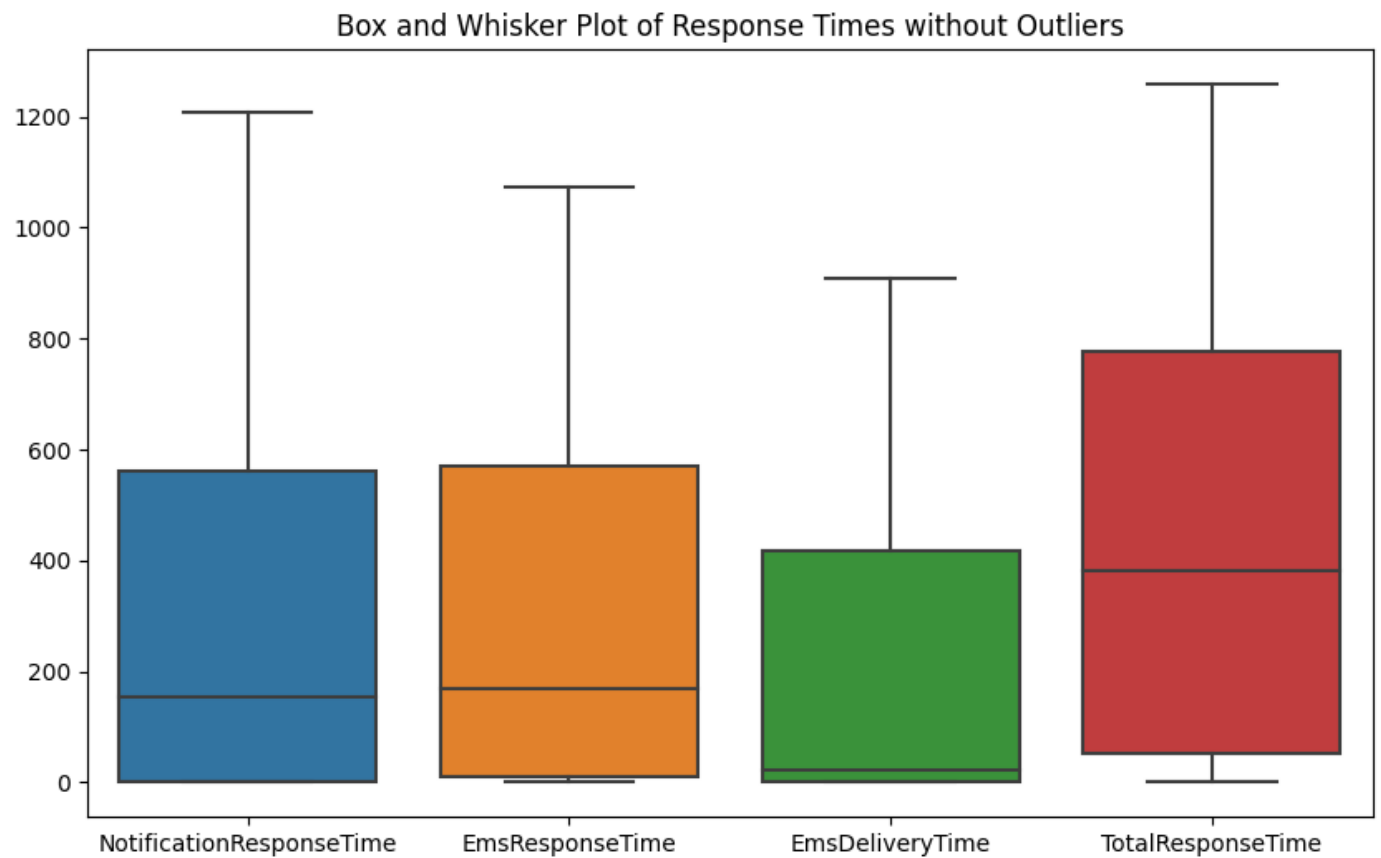
# Recalculate averages for the cleaned data
average_notification_response_time_cleaned = df['NotificationResponseTime'].mean()
average_ems_response_time_cleaned = df['EmsResponseTime'].mean()
average_ems_delivery_time_cleaned = df['EmsDeliveryTime'].mean()
average_total_response_time_cleaned = df['TotalResponseTime'].mean()

print("\nAverages (Uncleaned Data):")
print(f"Notification Response Time: {average_notification_response_time_cleaned:.2f} minutes")
print(f"EMS Response Time: {average_ems_response_time_cleaned:.2f} minutes")
print(f"EMS Delivery Time: {average_ems_delivery_time_cleaned:.2f} minutes")
print(f"Total Response Time: {average_total_response_time_cleaned:.2f} minutes")

average_notification_response_time_cleaned = df_filtered['NotificationResponseTime'].mean()
average_ems_response_time_cleaned = df_filtered['EmsResponseTime'].mean()
average_ems_delivery_time_cleaned = df_filtered['EmsDeliveryTime'].mean()
average_total_response_time_cleaned = df_filtered['TotalResponseTime'].mean()

print("\nAverages (Cleaned Data):")
print(f"Notification Response Time: {average_notification_response_time_cleaned:.2f} minutes")
print(f"EMS Response Time: {average_ems_response_time_cleaned:.2f} minutes")
print(f"EMS Delivery Time: {average_ems_delivery_time_cleaned:.2f} minutes")
print(f"Total Response Time: {average_total_response_time_cleaned:.2f} minutes")

```

Using a z score of 1.5 you filtered out 32576 values

Averages (UnCleaned Data):

Notification Response Time: 465.15 minutes

EMS Response Time: 398.20 minutes

EMS Delivery Time: 298.65 minutes

Total Response Time: 569.26 minutes

Averages (Cleaned Data):

Notification Response Time: 299.08 minutes

EMS Response Time: 307.20 minutes

EMS Delivery Time: 204.60 minutes

Total Response Time: 440.25 minutes

```
# Create new columns for percentage calculations
df_filtered['BlackPercentage'] = (df_filtered['black_pop'] / df_filtered['total_pop'])
df_filtered['HispanicPercentage'] = (df_filtered['hispanic_pop'] / df_filtered['total_pop'])
df_filtered['AsianPercentage'] = (df_filtered['asian_pop'] / df_filtered['total_pop'])
df_filtered['AmerindianPercentage'] = (df_filtered['amerindian_pop'] / df_filtered['total_pop'])

# If needed, you can drop the intermediate columns used for calculation
# df = df.drop(['black_pop', 'hispanic_pop', 'asian_pop', 'amerindian_pop'], axis=1)
```

```
<ipython-input-83-104f4fed2c84>:2: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
```

See the caveats in the documentation: <https://pandas.pydata.org/pandas-docs/stable/10min.html#copy-on-write>

```
df_filtered['BlackPercentage'] = (df_filtered['black_pop'] / df_filtered['total_pop'])
<ipython-input-83-104f4fed2c84>:3: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
```

See the caveats in the documentation: <https://pandas.pydata.org/pandas-docs/stable/10min.html#copy-on-write>

```
df_filtered['HispanicPercentage'] = (df_filtered['hispanic_pop'] / df_filtered['total_pop'])
<ipython-input-83-104f4fed2c84>:4: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
```

See the caveats in the documentation: <https://pandas.pydata.org/pandas-docs/stable/10min.html#copy-on-write>

```
df_filtered['AsianPercentage'] = (df_filtered['asian_pop'] / df_filtered['total_pop'])
<ipython-input-83-104f4fed2c84>:5: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
```

See the caveats in the documentation: <https://pandas.pydata.org/pandas-docs/stable/10min.html#copy-on-write>

```
df_filtered['AmerindianPercentage'] = (df_filtered['amerindian_pop'] / df_filtered['total_pop'])
```

```
# Print a few rows
print("Sample Rows:")
print(df_filtered[['BlackPercentage', 'HispanicPercentage', 'AsianPercentage', 'AmerindianPercentage']])

# Calculate and print the average
average_percentages = df_filtered[['BlackPercentage', 'HispanicPercentage', 'AsianPercentage', 'AmerindianPercentage']].mean()
print("\nAverage Percentages:")
print(average_percentages)
```

```
Sample Rows:
   BlackPercentage  HispanicPercentage  AsianPercentage  AmerindianPercentage
0         1.083829          16.172455          1.210822          2.657067
1              NaN              NaN              NaN              NaN
3              NaN              NaN              NaN              NaN
4         7.858267          21.294515          4.112251          0.350923
6        29.334279          55.118644          3.609689          0.204130
```

```
Average Percentages:
BlackPercentage      13.900822
HispanicPercentage   23.627632
AsianPercentage       6.019564
AmerindianPercentage  0.734391
dtype: float64
```

```
import pandas as pd
import matplotlib.pyplot as plt
```

```
# Assuming you have a DataFrame named 'df' with the relevant columns
# ...
```

```
# Scatterplot for Black Population
plt.scatter(df['BlackPercentage'], df['TotalResponseTime'], label='Black Population')
```

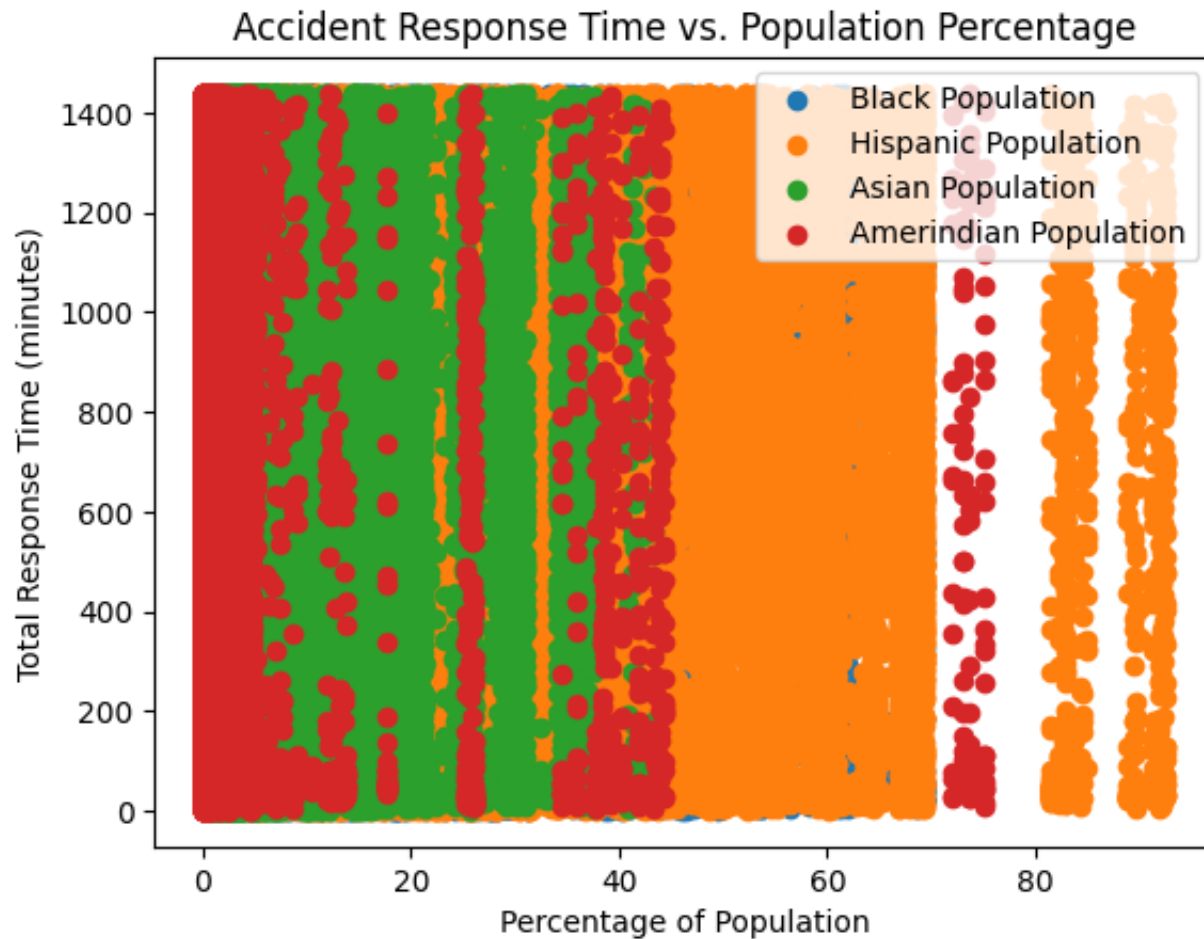
```
# Scatterplot for Hispanic Population
plt.scatter(df['HispanicPercentage'], df['TotalResponseTime'], label='Hispanic Population')
```

```
# Scatterplot for Asian Population
plt.scatter(df['AsianPercentage'], df['TotalResponseTime'], label='Asian Population')
```

```
# Scatterplot for Amerindian Population
plt.scatter(df['AmerindianPercentage'], df['TotalResponseTime'], label='Amerindian Population')
```

```
# Add labels and legend
plt.xlabel('Percentage of Population')
plt.ylabel('Total Response Time (minutes)')
plt.title('Accident Response Time vs. Population Percentage')
plt.legend()
plt.show()
```

```
/usr/local/lib/python3.10/dist-packages/IPython/core/pylabtools.py:151: UserWarning:
fig.canvas.print_figure(bytes_io, **kw)
```



```
# Assuming you have a DataFrame named 'df_filtered' with the relevant columns

# Define the races to plot
races = ['Black', 'Hispanic', 'Asian', 'Amerindian']

# Create a grid layout for the subplots
fig, axes = plt.subplots(nrows=2, ncols=2, figsize=(12, 10))
fig.suptitle('Accident Response Time vs. Population Percentage by Race', fontsize=14)

# Flatten the axes for easier indexing
axes = axes.flatten()

# Plot each race on a separate subplot
for i, race in enumerate(races):
    sns.scatterplot(data=df_filtered, x=f'{race}Percentage', y='TotalResponseTime')
    axes[i].set_title(f'{race} Population')

# Set common labels
for ax in axes:
    ax.set_xlabel('Percentage of Population')
```

```
ax.set_ylabel('Total Response Time (minutes)')
```

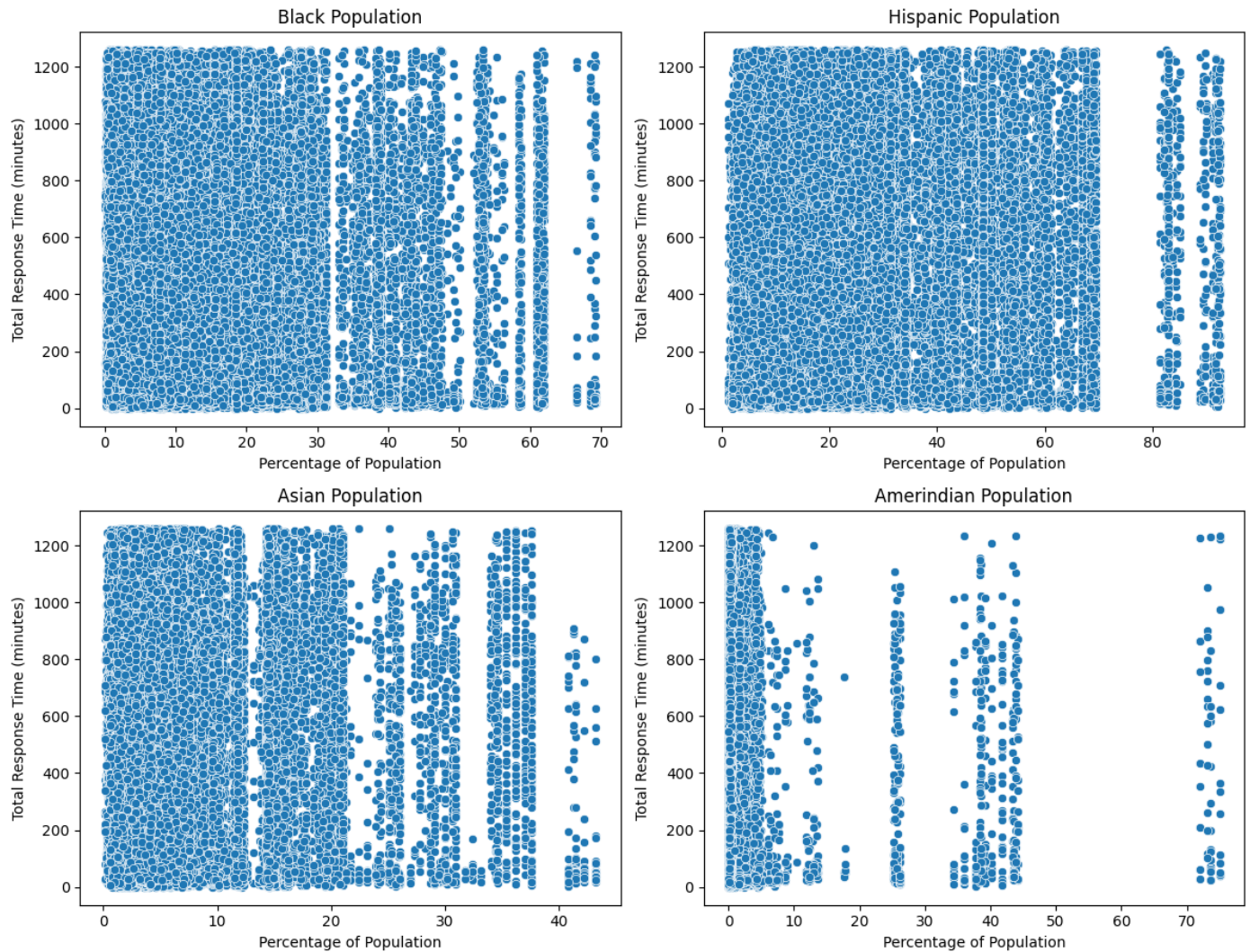
```
# Adjust layout
```

```
plt.tight_layout(rect=[0, 0, 1, 0.96]) # Add space for the title
```

```
plt.show()
```



Accident Response Time vs. Population Percentage by Race



```

import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

# Assuming you have a DataFrame named 'df_filtered' with the relevant columns

# Define the races to plot
races = ['Black', 'Hispanic', 'Asian', 'Amerindian']

# Create a grid layout for the subplots
fig, axes = plt.subplots(nrows=2, ncols=2, figsize=(12, 10))
fig.suptitle('Accident Response Time vs. Population Percentage by Race', fontsize=14)

# Flatten the axes for easier indexing
axes = axes.flatten()

# Define colors for each race
colors = ['blue', 'orange', 'green', 'red']

# Plot each race on a separate subplot with flipped axes and different colors
for i, (race, color) in enumerate(zip(races, colors)):
    sns.scatterplot(data=df_filtered, x='TotalResponseTime', y=f'{race}Percentage',
                    axes[i].set_title(f'{race} Population'))

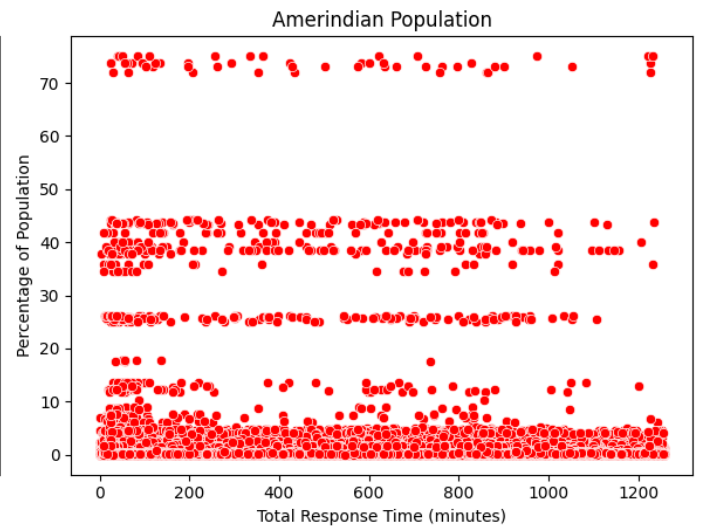
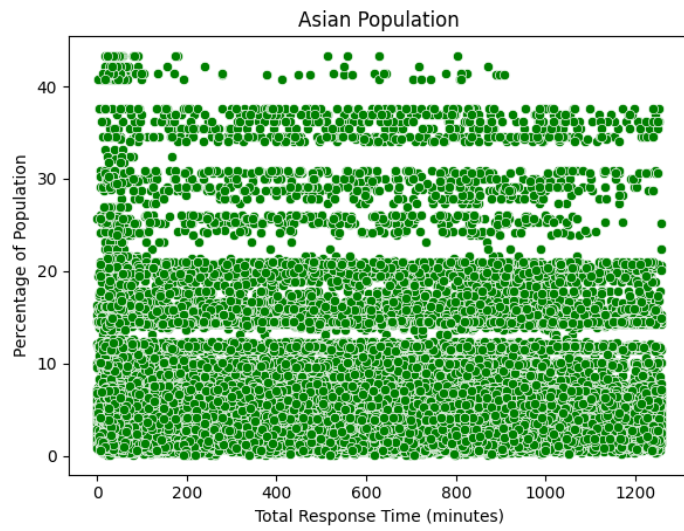
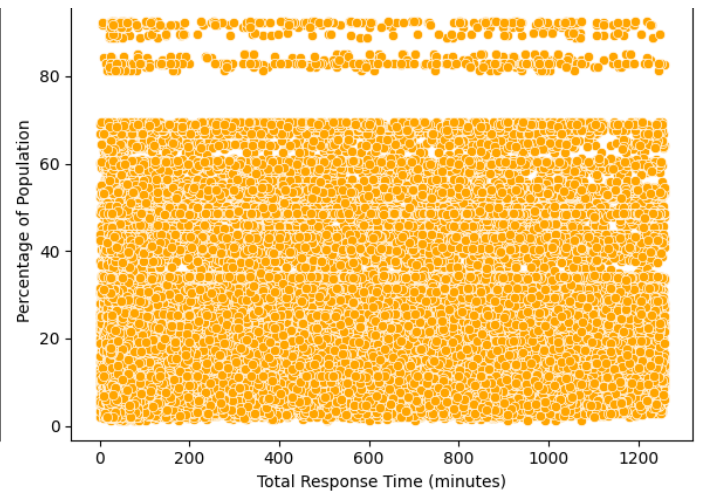
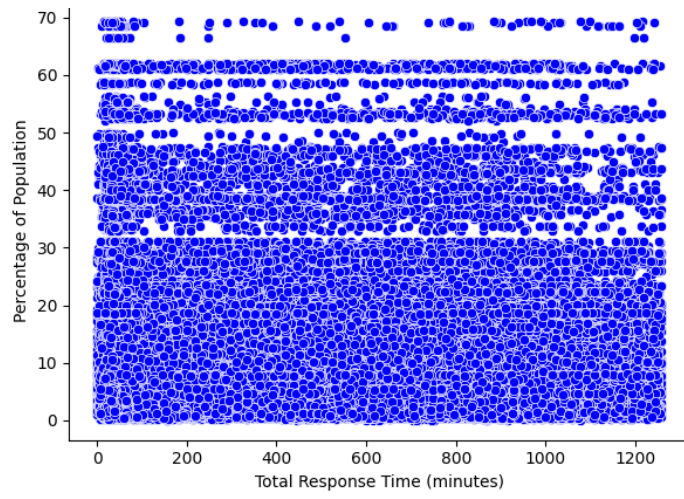
# Set common labels
for ax in axes:
    ax.set_xlabel('Total Response Time (minutes)')
    ax.set_ylabel('Percentage of Population')

# Adjust layout
plt.tight_layout(rect=[0, 0, 1, 0.96]) # Add space for the title
plt.show()

```

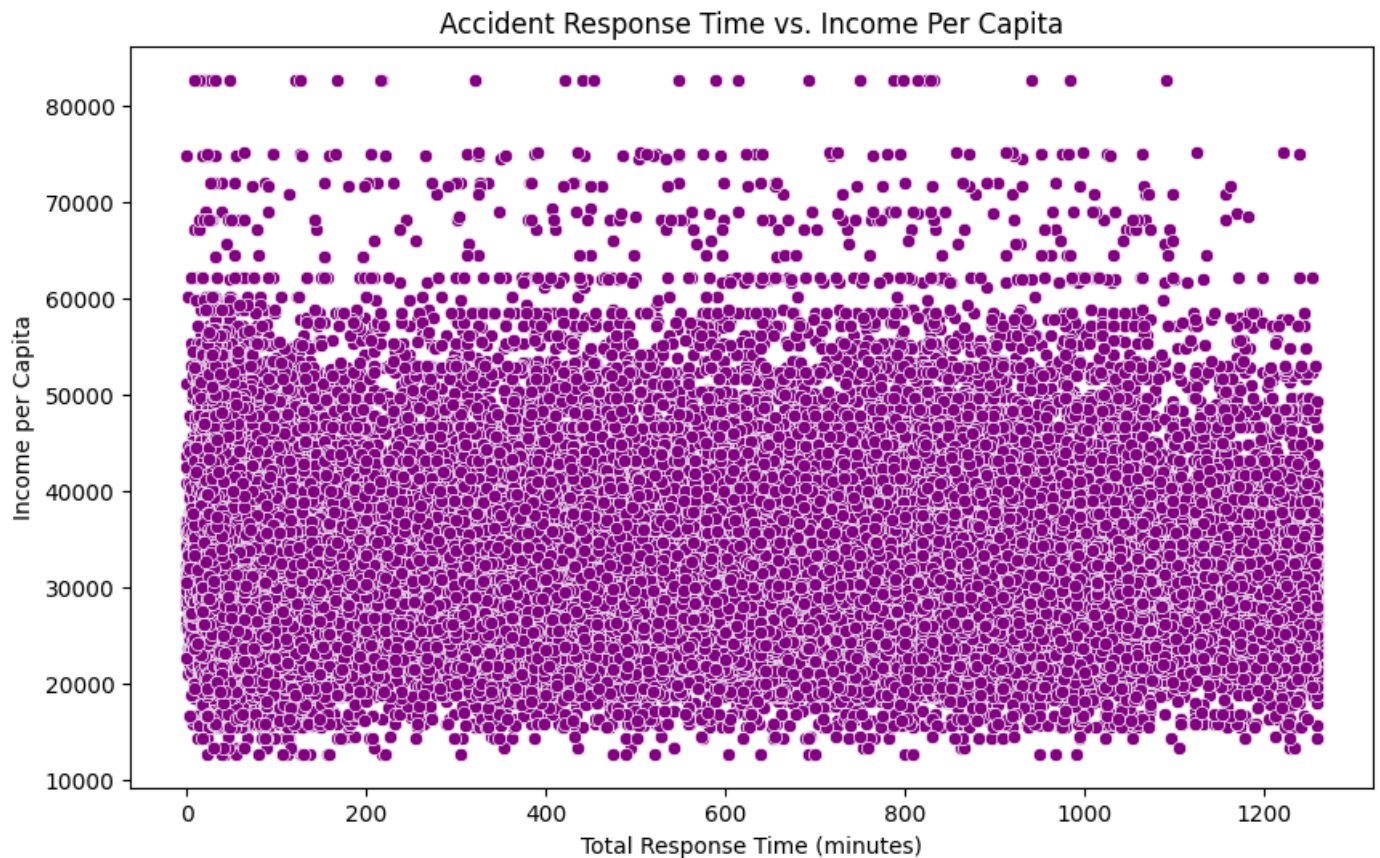
Accident Response Time vs. Population Percentage by Race





```
# Assuming you have a DataFrame named 'df_filtered' with the relevant columns

# Create a scatter plot for 'income_per_capita' against 'TotalResponseTime'
plt.figure(figsize=(10, 6))
sns.scatterplot(data=df_filtered, x='TotalResponseTime', y='income_per_capita', c=
plt.title('Accident Response Time vs. Income Per Capita')
plt.xlabel('Total Response Time (minutes)')
plt.ylabel('Income per Capita')
plt.show()
```



```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
# Assuming you have a DataFrame named 'df_filtered' with the relevant columns
```



```

# Create a grid layout for the subplots
fig, axes = plt.subplots(nrows=2, ncols=2, figsize=(12, 10))
fig.suptitle('Income Per Capita vs. Different Response Times', fontsize=16)

# Flatten the axes for easier indexing
axes = axes.flatten()

# Define response time columns
response_time_columns = ['NotificationResponseTime', 'EmsResponseTime', 'EmsDeliveryTime', 'TotalResponseTime']

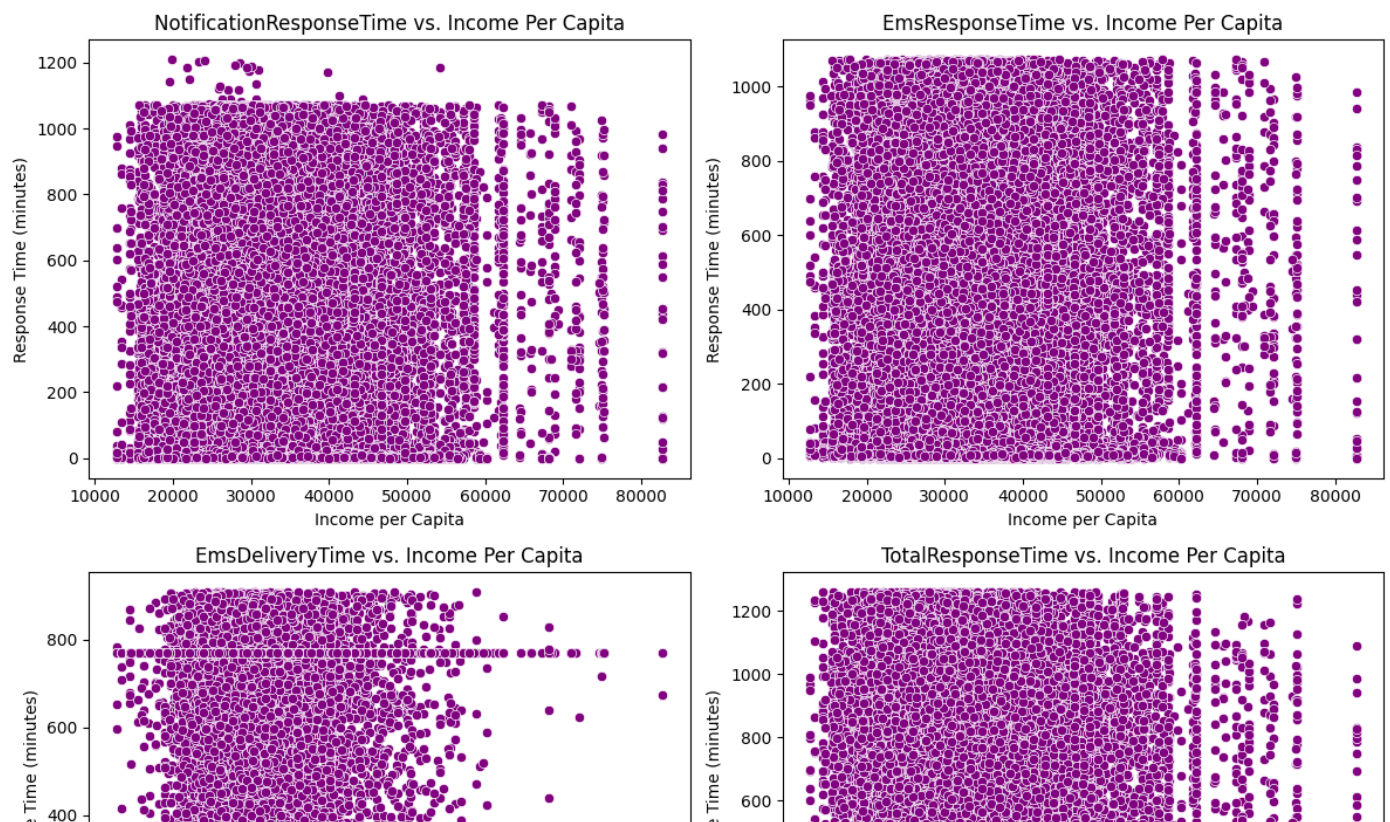
# Plot each response time on a separate subplot
for i, response_time in enumerate(response_time_columns):
    sns.scatterplot(data=df_filtered, x='income_per_capita', y=response_time, ax=axes[i])
    axes[i].set_title(f'{response_time} vs. Income Per Capita')

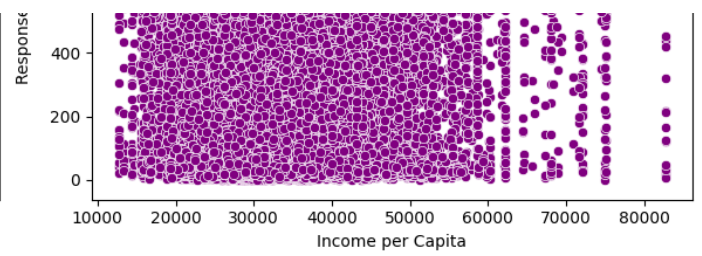
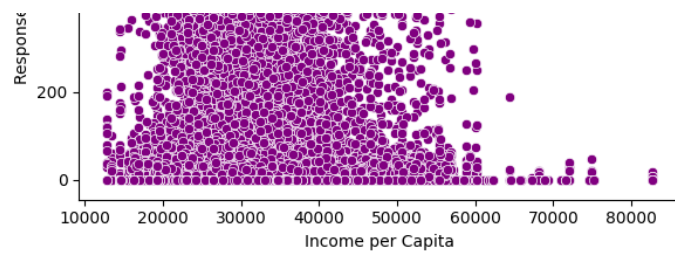
# Set common labels
for ax in axes:
    ax.set_xlabel('Income per Capita')
    ax.set_ylabel('Response Time (minutes)')

# Adjust layout
plt.tight_layout(rect=[0, 0, 1, 0.96]) # Add space for the title
plt.show()

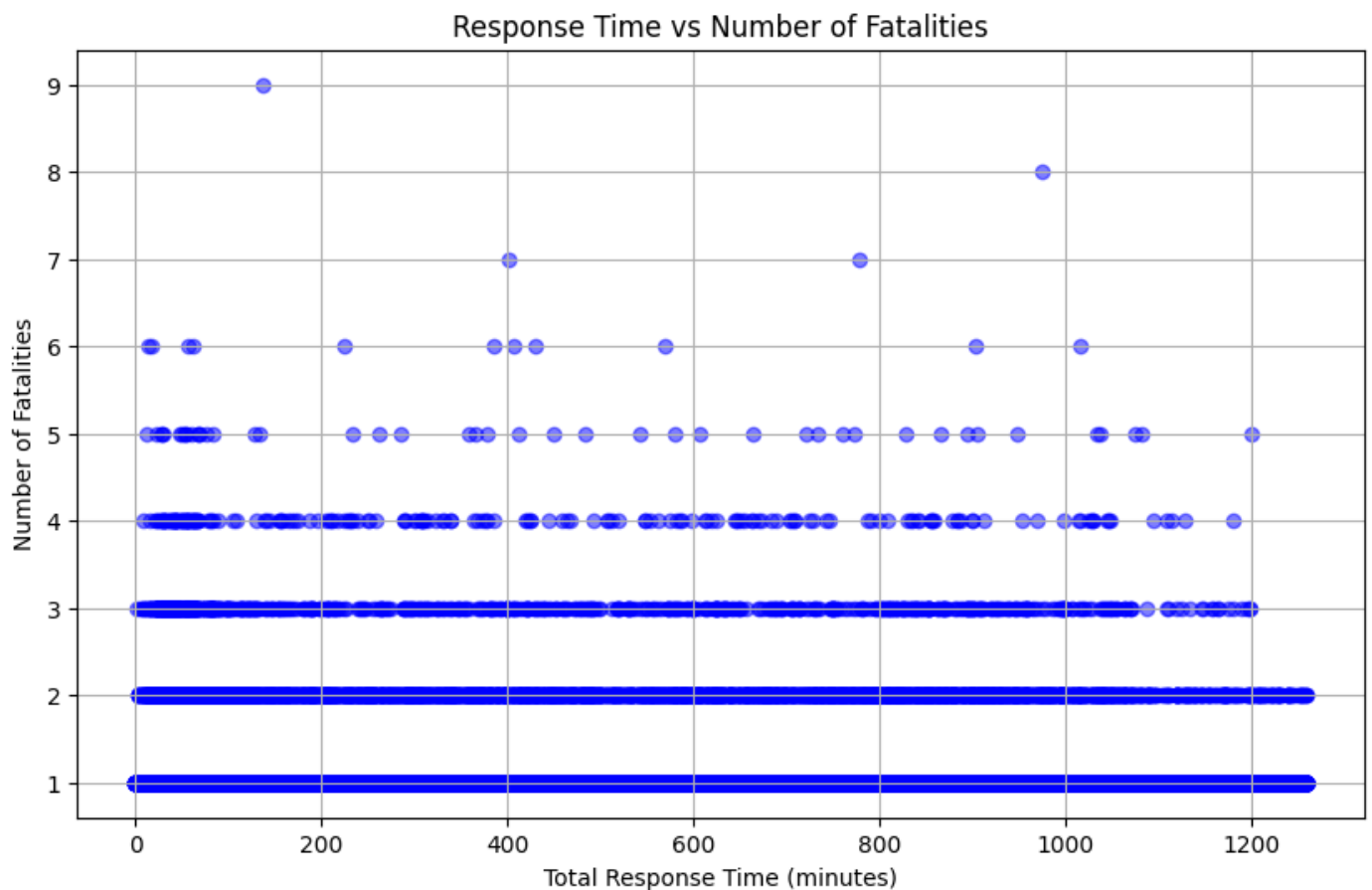
```

Income Per Capita vs. Different Response Times





```
# Create a scatter plot
plt.figure(figsize=(10, 6))
plt.scatter(df_filtered['TotalResponseTime'], df_filtered['number_of_fatalities'])
plt.title('Response Time vs Number of Fatalities')
plt.xlabel('Total Response Time (minutes)')
plt.ylabel('Number of Fatalities')
plt.grid(True)
plt.show()
```



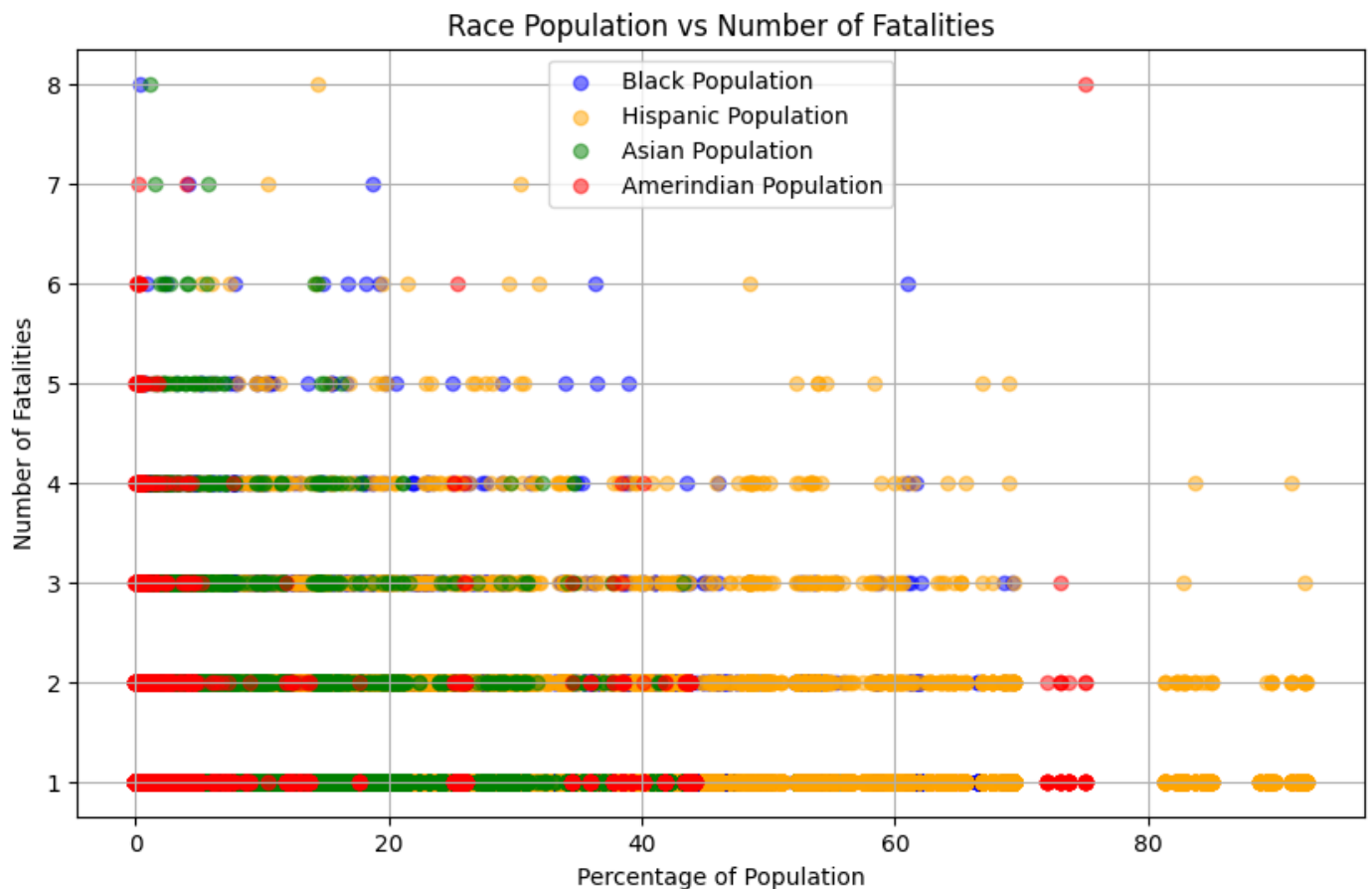
```
# Create a scatter plot for Black Population vs Number of Fatalities
plt.figure(figsize=(10, 6))
plt.scatter(df_filtered['BlackPercentage'], df_filtered['number_of_fatalities'],

# Create a scatter plot for Hispanic Population vs Number of Fatalities
plt.scatter(df_filtered['HispanicPercentage'], df_filtered['number_of_fatalities']
```

```
# Create a scatter plot for Asian Population vs Number of Fatalities
plt.scatter(df_filtered['AsianPercentage'], df_filtered['number_of_fatalities'],

# Create a scatter plot for Amerindian Population vs Number of Fatalities
plt.scatter(df_filtered['AmerindianPercentage'], df_filtered['number_of_fatalities']

# Add labels and legend
plt.title('Race Population vs Number of Fatalities')
plt.xlabel('Percentage of Population')
plt.ylabel('Number of Fatalities')
plt.legend()
plt.grid(True)
plt.show()
```



```
# Create a grid layout for the subplots
fig, axes = plt.subplots(nrows=2, ncols=2, figsize=(12, 10))
fig.suptitle('Race Percentage vs Total Number of Accidents', fontsize=16)
```

```

# Flatten the axes for easier indexing
axes = axes.flatten()

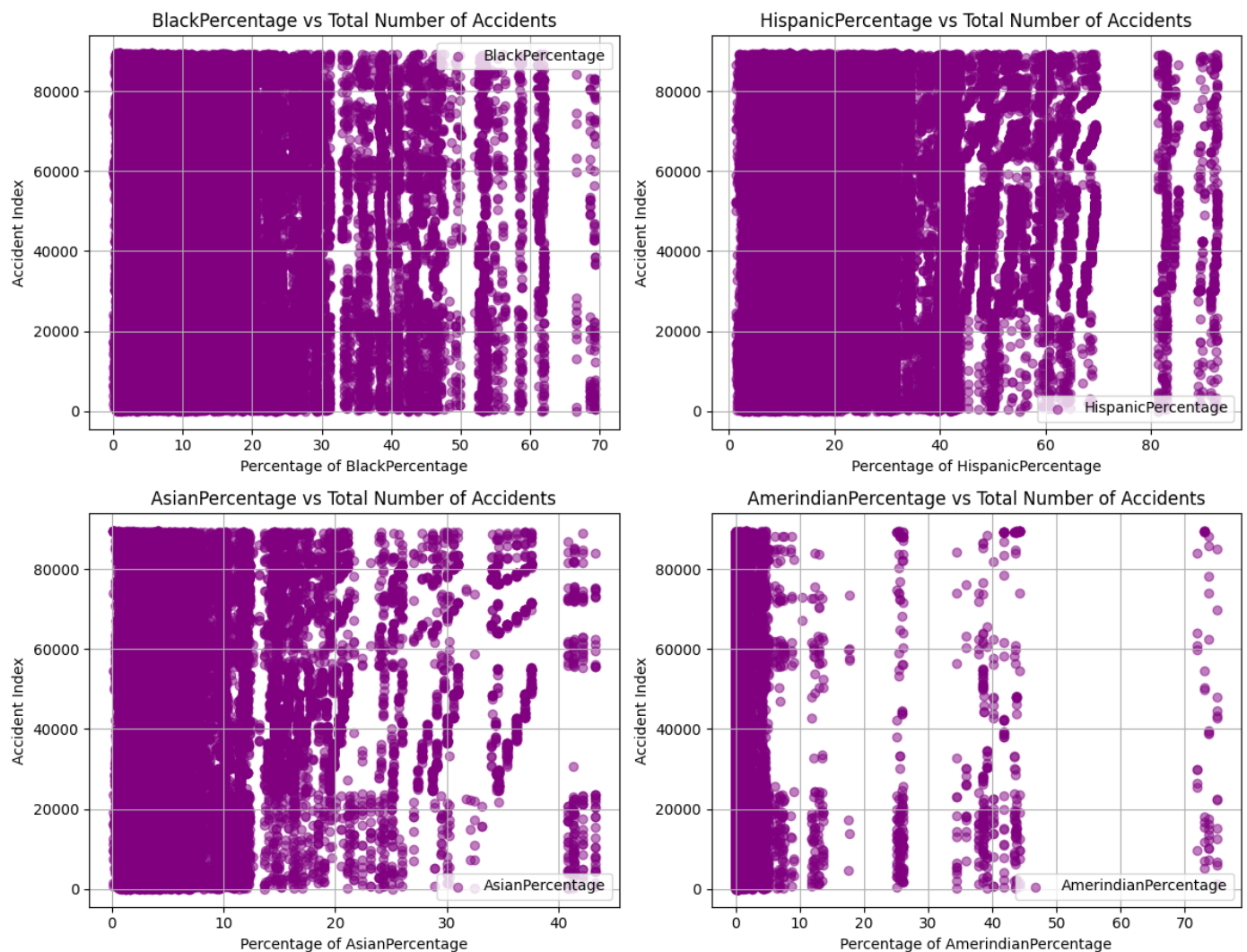
# Define race columns
race_columns = ['BlackPercentage', 'HispanicPercentage', 'AsianPercentage', 'Amer

# Plot each race percentage on a separate subplot
for i, race_column in enumerate(race_columns):
    axes[i].scatter(df_filtered[race_column], range(len(df_filtered)), label=race_
    axes[i].set_title(f'{race_column} vs Total Number of Accidents')
    axes[i].set_xlabel(f'Percentage of {race_column}')
    axes[i].set_ylabel('Accident Index')
    axes[i].grid(True)
    axes[i].legend()

# Adjust layout
plt.tight_layout(rect=[0, 0, 1, 0.96]) # Add space for the title
plt.show()

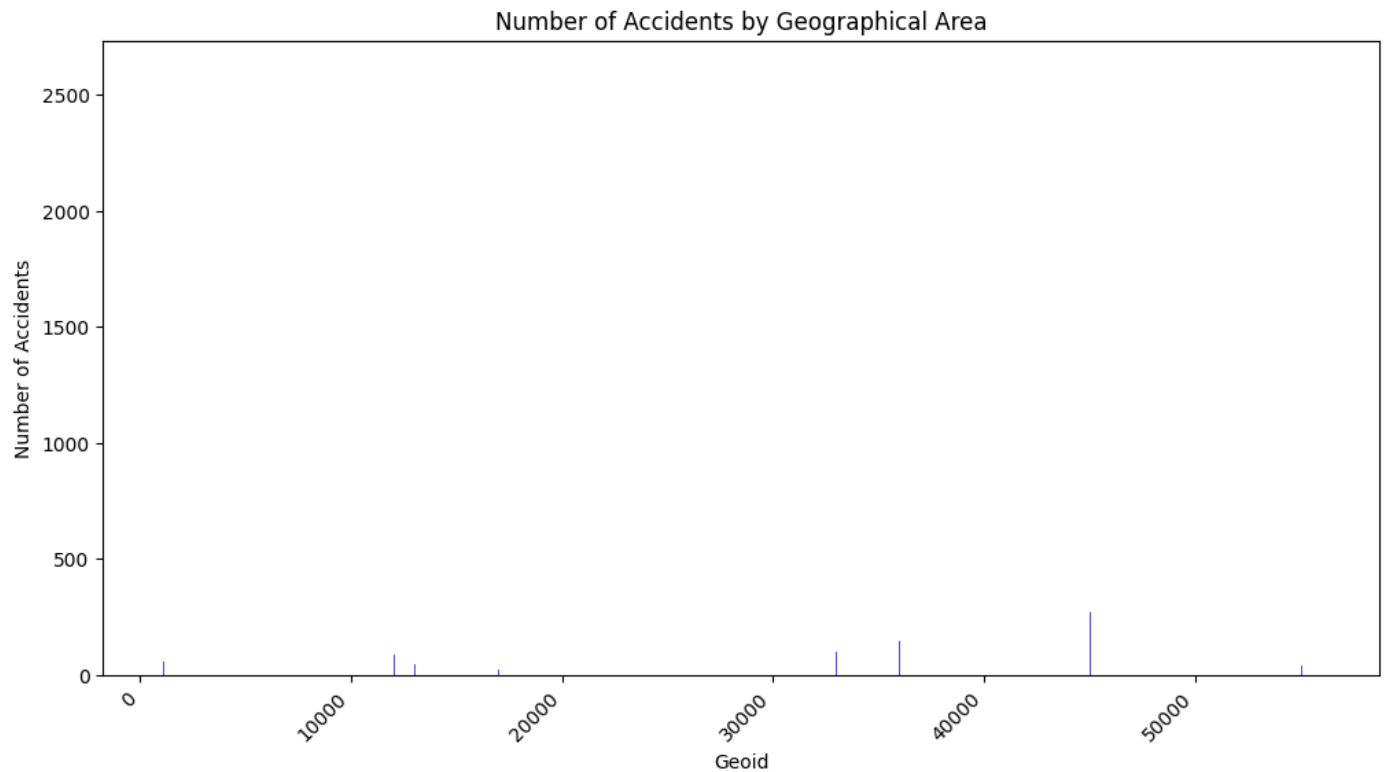
```

Race Percentage vs Total Number of Accidents




```
# Group by 'geoid' and count the number of accidents in each geographical area
accidents_by_geoid = df_filtered.groupby('geoid')['number_of_fatalities'].count()

# Create a bar chart
plt.figure(figsize=(12, 6))
plt.bar(accidents_by_geoid['geoid'], accidents_by_geoid['number_of_fatalities'],
plt.title('Number of Accidents by Geographical Area')
plt.xlabel('Geoid')
plt.ylabel('Number of Accidents')
plt.xticks(rotation=45, ha='right') # Rotate x-axis labels for better readability
plt.show()
```



```
distinct_geoIDs_count = df_filtered['geoid'].nunique()

print(f"Distinct Number of geoIDs: {distinct_geoIDs_count}")
```

Distinct Number of geoIDs: 831

```
!pip install pydeck
```

Collecting pydeck

Downloading pydeck-0.8.0-py2.py3-none-any.whl (4.7 MB)

4.7/4.7 MB 32.0 MB/s eta 0:00:00
Requirement already satisfied: jinja2>=2.10.1 in /usr/local/lib/python3.10/dist-packages (from pydeck==0.8.0)
Requirement already satisfied: numpy>=1.16.4 in /usr/local/lib/python3.10/dist-packages (from pydeck==0.8.0)
Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.10/dist-packages (from jinja2>=2.10.1->pydeck==0.8.0)
Installing collected packages: pydeck
Successfully installed pydeck-0.8.0

```
import pydeck as pdk
```

```
# Replace 'YOUR_MAPBOX_TOKEN' with your actual Mapbox API token
```

```
MAPBOX_TOKEN = 'pk.eyJ1IjoibXdpbmNoZXN0ZXIyIiwiaSI6ImNscHVteDVwbTBsZjgya282MDd6c2I'.
```

```
# Sample data (replace this with your data)
```

```
data = [
    {'latitude': 37.7749, 'longitude': -122.4194, 'weight': 10},
    {'latitude': 40.7128, 'longitude': -74.0060, 'weight': 5},
    # Add more data points as needed
]
```

```
# Create a PyDeck layer for the heatmap
```

```
heatmap_layer = pdk.Layer(
    'HeatmapLayer',
    data,
    get_position='[longitude, latitude]',
    get_weight='weight',
)
```

```
# Set the initial view state for the map
```

```
view_state = pdk.ViewState(
    latitude=37.7749,
    longitude=-122.4194,
    zoom=3,
    pitch=0,
)
```

```
# Create a PyDeck deck
```



```
deck = pdk.Deck(  
    layers=[heatmap_layer],  
    initial_view_state=view_state,  
    map_style='mapbox://styles/mapbox/dark-v10'  
)  
  
# Use the to_html() method to get the HTML code  
html_code = deck.to_html()  
  
# Display the HTML code in Colab  
from IPython.core.display import HTML  
  
HTML(html_code)
```

