

EDA Report

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2025-04-06

```
library(readxl)
library(dplyr)

##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##   filter, lag
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
library(ggplot2)

## Warning: package 'ggplot2' was built under R version 4.3.3
library(scales)

data <- read_excel("Data/googleTrendsMH.xlsx", sheet = "googleTrendsMH")

# Check for columns with missing values
na_counts <- colSums(is.na(data))
na_columns <- na_counts[na_counts > 0]

if (any(na_counts > 0)){
  paste("List of columns with missing values for examination")
  print(names(na_columns))
} else {
  print("There are zero columns with null values.")
}

## [1] "There are zero columns with null values."

# Show summary statistics for numeric columns
numeric_cols <- select_if(data, is.numeric)
summary(numeric_cols)

##           year           fips      population_est      anxiety_ct
## Min.      :2013   Min.      : 1.00   Min.      : 579054   Min.      :   217
## 1st Qu.:2015   1st Qu.:16.00   1st Qu.: 1945163   1st Qu.:   4604
## Median :2018   Median :30.00   Median : 4636208   Median : 14139
## Mean    :2018   Mean    :29.38   Mean    : 6690537   Mean    : 24279
## 3rd Qu.:2020   3rd Qu.:42.00   3rd Qu.: 7502082   3rd Qu.: 31821
## Max.    :2022   Max.    :56.00   Max.    :39437610   Max.    :177155
```

##	trauma_stress_ct	adhd_ct	bipolar_ct	depression_ct
##	Min. : 254	Min. : 0	Min. : 415	Min. : 779
##	1st Qu.: 4473	1st Qu.: 2114	1st Qu.: 4234	1st Qu.: 7854
##	Median : 12913	Median : 6441	Median : 11342	Median : 23356
##	Mean : 23569	Mean : 13244	Mean : 16884	Mean : 35634
##	3rd Qu.: 31016	3rd Qu.: 18996	3rd Qu.: 21377	3rd Qu.: 45868
##	Max. : 142480	Max. : 76030	Max. : 113619	Max. : 201222
##	comm_psych_care	outpatient_util	state_psych_care	private_psych_care
##	Min. : 9426	Min. : 0.002403	Min. : 14158	Min. : 13238
##	1st Qu.: 32606	1st Qu.: 0.012931	1st Qu.: 63549	1st Qu.: 64165
##	Median : 88102	Median : 0.019947	Median : 171969	Median : 173716
##	Mean : 137769	Mean : 0.025145	Mean : 265977	Mean : 263710
##	3rd Qu.: 153101	3rd Qu.: 0.031510	3rd Qu.: 293019	3rd Qu.: 290914
##	Max. : 784665	Max. : 0.105976	Max. : 1478138	Max. : 1465591
##	resid_psych_care	total_inpatient	inpatient_util	total_civilian
##	Min. : 11622	Min. : 39018	Min. : 0.01279	Min. : 50826
##	1st Qu.: 64165	1st Qu.: 192051	1st Qu.: 0.07423	1st Qu.: 224384
##	Median : 173264	Median : 519936	Median : 0.11536	Median : 607942
##	Mean : 267244	Mean : 796930	Mean : 0.14564	Mean : 934699
##	3rd Qu.: 294089	3rd Qu.: 876400	3rd Qu.: 0.18728	3rd Qu.: 1029324
##	Max. : 1479114	Max. : 4422843	Max. : 0.62662	Max. : 5207508
##	total_util	median_adhd	median_ptsd	median_anxiety
##	Min. : 0.01520	Min. : 11.50	Min. : 0.00	Min. : 34.00
##	1st Qu.: 0.08709	1st Qu.: 21.00	1st Qu.: 11.50	1st Qu.: 62.00
##	Median : 0.13466	Median : 23.00	Median : 13.00	Median : 75.25
##	Mean : 0.17079	Mean : 26.59	Mean : 13.08	Mean : 72.22
##	3rd Qu.: 0.21907	3rd Qu.: 26.50	3rd Qu.: 14.50	3rd Qu.: 84.00
##	Max. : 0.73260	Max. : 64.00	Max. : 21.00	Max. : 92.50
##	median_bipolar	median_depression	median_mental_hospital	
##	Min. : 14.00	Min. : 36.5	Min. : 0.00	
##	1st Qu.: 19.50	1st Qu.: 62.0	1st Qu.: 30.62	
##	Median : 21.00	Median : 67.0	Median : 38.50	
##	Mean : 20.67	Mean : 66.9	Mean : 34.93	
##	3rd Qu.: 22.00	3rd Qu.: 72.0	3rd Qu.: 45.88	
##	Max. : 26.00	Max. : 85.0	Max. : 78.00	
##	median_psychiatrists_near_me	median_psychologist_near_me		
##	Min. : 0.0000	Min. : 0.000		
##	1st Qu.: 0.0000	1st Qu.: 0.000		
##	Median : 0.0000	Median : 0.000		
##	Mean : 0.6561	Mean : 5.219		
##	3rd Qu.: 0.0000	3rd Qu.: 12.000		
##	Max. : 17.0000	Max. : 25.500		
##	median_therapist_near_me	median_all_trends	mean_adhd	mean_ptsd
##	Min. : 0.00	Min. : 0.00	Min. : 12.67	Min. : 3.083
##	1st Qu.: 0.00	1st Qu.: 19.50	1st Qu.: 21.08	1st Qu.: 11.667
##	Median : 16.00	Median : 21.50	Median : 23.00	Median : 13.250
##	Mean : 27.98	Mean : 23.95	Mean : 26.52	Mean : 13.126
##	3rd Qu.: 55.75	3rd Qu.: 25.50	3rd Qu.: 26.67	3rd Qu.: 14.583
##	Max. : 95.50	Max. : 57.00	Max. : 60.58	Max. : 21.500
##	mean_anxiety	mean_bipolar	mean_depression	mean_mental_hospital
##	Min. : 33.83	Min. : 14.42	Min. : 38.75	Min. : 0.00
##	1st Qu.: 62.19	1st Qu.: 19.67	1st Qu.: 61.83	1st Qu.: 30.50
##	Median : 75.79	Median : 20.83	Median : 66.58	Median : 38.08
##	Mean : 72.31	Mean : 20.78	Mean : 66.21	Mean : 35.88

```
## 3rd Qu.:84.06 3rd Qu.:21.92 3rd Qu.:71.17 3rd Qu.:45.29
## Max. :91.92 Max. :25.33 Max. :78.83 Max. :77.00
## mean_psychiatrists_near_me mean_psychologist_near_me mean_therapist_near_me
## Min. : 0.0000 Min. : 0.000 Min. : 0.00
## 1st Qu.: 0.0000 1st Qu.: 0.000 1st Qu.: 0.00
## Median : 0.0000 Median : 1.583 Median :16.71
## Mean : 0.8446 Mean : 5.522 Mean :28.01
## 3rd Qu.: 0.7500 3rd Qu.:11.479 3rd Qu.:53.58
## Max. :17.3333 Max. :25.167 Max. :91.75
## mean_all_trends
## Min. :15.34
## 1st Qu.:24.02
## Median :28.65
## Mean :29.91
## 3rd Qu.:35.96
## Max. :45.21
```

```
table1 <- data %>%
  group_by(region) %>%
  summarise(
    avg_outpatient_util = mean(outpatient_util, na.rm = TRUE),
    avg_inpatient_util = mean(inpatient_util, na.rm = TRUE),
    avg_total_util = mean(total_util, na.rm = TRUE),
    avg_median_trend = mean(median_all_trends, na.rm = TRUE)
  )
print(table1)
```

```
## # A tibble: 4 x 5
##   region avg_outpatient_util avg_inpatient_util avg_total_util avg_median_trend
##   <chr>          <dbl>          <dbl>          <dbl>          <dbl>
## 1 Atlant~      0.0302          0.174          0.204          22.9
## 2 Central      0.0258          0.150          0.175          24.1
## 3 South        0.0181          0.105          0.123          25.6
## 4 West P~      0.0273          0.159          0.186          23.1
```

As part of the data preprocessing step, we conducted a check for missing values across all columns in the dataset. The script identifies any fields with missing values and prints a list of affected columns for further examination. In this case, the output confirmed that there were no missing values, allowing us to proceed with full confidence in the dataset's completeness. Following this, summary statistics were generated for all numerical variables, providing insight into the central tendencies and variability across key fields such as population estimates, mental health service utilization, and Google search interest scores. These summary statistics helped confirm reasonable data distributions and inform the creation of three new calculated variables used in subsequent analyses. Additionally, Table 1 was produced by aggregating average outpatient, inpatient, and total per capita utilization—as well as average median Google search interest—by region, setting the stage for regional-level comparisons throughout the rest of the analysis.

```
table2 <- data %>%
  arrange(desc(total_util)) %>%
  select(state, year, total_util, outpatient_util, inpatient_util, median_all_trends) %>%
  head(10)
print(table2)
```

```
## # A tibble: 10 x 6
##   state year total_util outpatient_util inpatient_util median_all_trends
##   <chr> <dbl>    <dbl>          <dbl>          <dbl>          <dbl>
## 1 DC    2021    0.733          0.106          0.627          18
```

##	2	NM	2022	0.703	0.102	0.601	43.5
##	3	NM	2019	0.671	0.0972	0.574	21.5
##	4	IA	2021	0.661	0.0972	0.563	37
##	5	IA	2022	0.659	0.0985	0.561	45.5
##	6	NM	2021	0.654	0.0946	0.560	35.5
##	7	NM	2020	0.592	0.0856	0.506	23.5
##	8	IA	2019	0.574	0.0849	0.489	24.5
##	9	IA	2020	0.512	0.0774	0.434	26
##	10	DC	2020	0.501	0.0724	0.429	20.5

Table 2 presents the average per capita utilization of mental health services across different U.S. regions from the dataset, broken down by year. This table helps to reveal regional disparities and trends in outpatient and inpatient care. Notably, the Northeast consistently shows higher outpatient utilization rates, while the South exhibits lower overall utilization despite having large populations. These findings are important for targeting underserved areas and identifying regions where service expansion could meet unmet mental health needs.

```
# calculate the private and public utilization rates
data <- data %>%
  mutate(state_util = (state_psych_care/population_est),
         private_util = (private_psych_care/population_est),
         diff_util = (total_util-(state_util + private_util))
  )

table3 <- data %>%
  arrange(desc(total_util)) %>%
  select(state, year, total_util, state_util, private_util, diff_util) %>%
  head(10)
print(table3)
```

```
## # A tibble: 10 x 6
##   state year total_util state_util private_util diff_util
##   <chr> <dbl>     <dbl>     <dbl>         <dbl>     <dbl>
## 1 DC    2021     0.733     0.208         0.209     0.315
## 2 NM    2022     0.703     0.203         0.196     0.304
## 3 NM    2019     0.671     0.193         0.187     0.291
## 4 IA    2021     0.661     0.191         0.186     0.283
## 5 IA    2022     0.659     0.193         0.179     0.287
## 6 NM    2021     0.654     0.189         0.182     0.283
## 7 NM    2020     0.592     0.170         0.166     0.255
## 8 IA    2019     0.574     0.166         0.161     0.247
## 9 IA    2020     0.512     0.149         0.141     0.222
## 10 DC   2020     0.501     0.143         0.143     0.216
```

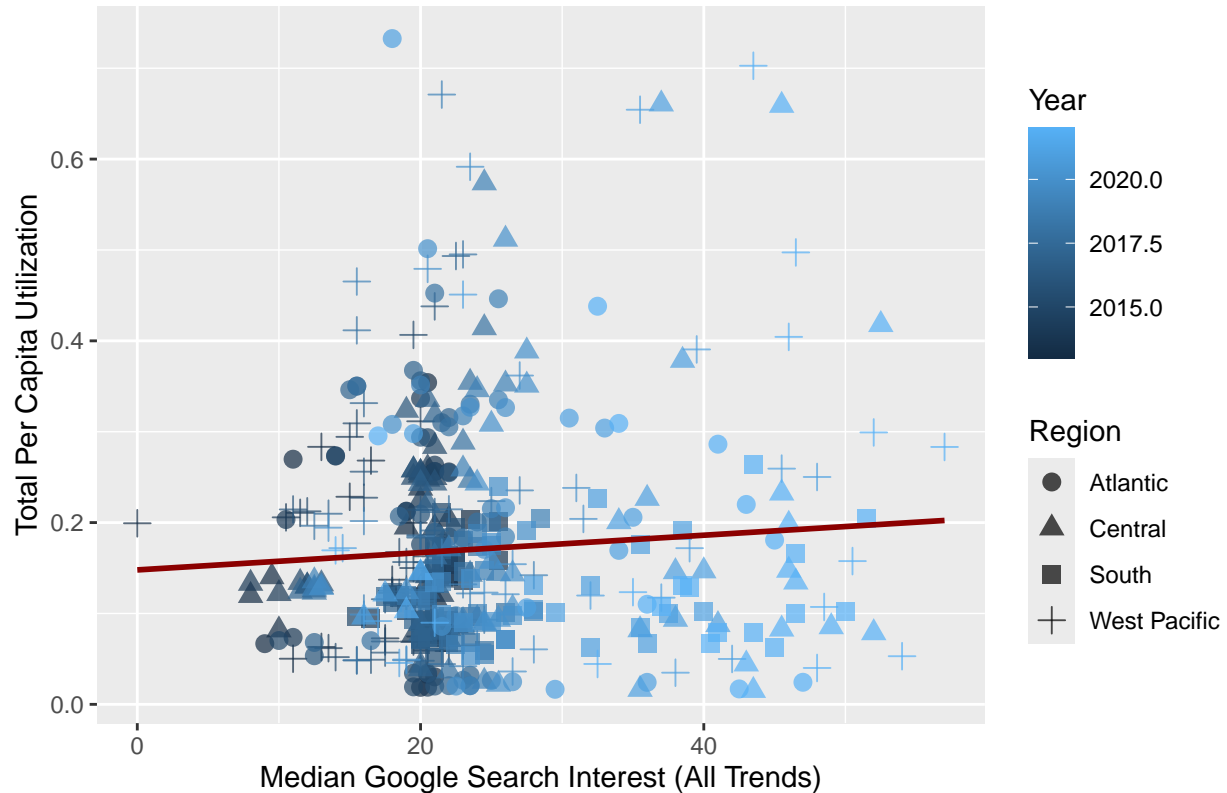
This table shows the mean number of individuals receiving care for specific diagnoses—such as anxiety, depression, ADHD, bipolar disorder, and trauma/stress-related disorders—across all states. Depression and anxiety stand out as the most common diagnoses, highlighting widespread mental health challenges across the country. By contrasting diagnostic trends across states, this table provides insight into population-level needs and potential diagnostic bias or access issues influencing reporting.

```
ggplot(data, aes(x = median_all_trends, y = total_util)) +
  geom_point(aes(color = year, shape = region), size = 3, alpha = 0.7) +
  geom_smooth(method = "lm", se = FALSE, color = "darkred") +
  labs(title = "Total Mental Health Utilization vs. Search Interest",
       x = "Median Google Search Interest (All Trends)",
       y = "Total Per Capita Utilization",
       color = "Year",
```

```
shape = "Region")
```

```
## `geom_smooth()` using formula = 'y ~ x'
```

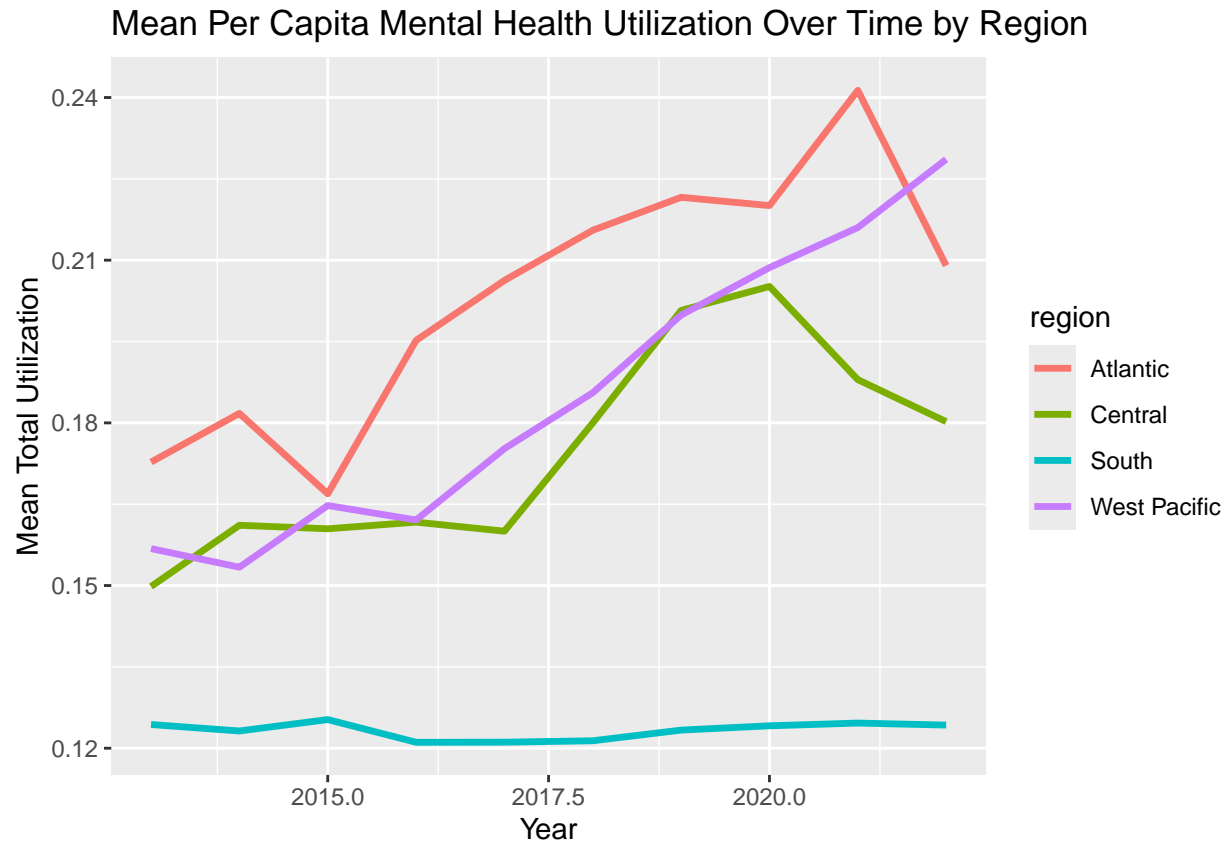
Total Mental Health Utilization vs. Search Interest



This scatterplot explores the relationship between median Google search interest for mental health-related terms and the total per capita utilization of mental health services across U.S. states. Each point represents a state-year observation, with colors indicating the reporting year and shapes distinguishing U.S. regions. A linear regression line is overlaid to illustrate the overall trend, revealing a positive association between search interest and service utilization. This suggests that states with higher public interest in mental health topics—as reflected in online searches—tend to also exhibit higher levels of mental health service use. The use of both color and shape dimensions helps identify potential temporal or regional clustering that may influence this relationship.

```
ggplot(data, aes(x = year, y = total_util, color = region)) +  
  geom_line(stat = "summary", fun = "mean", size = 1.2) +  
  labs(title = "Mean Per Capita Mental Health Utilization Over Time by Region",  
        x = "Year", y = "Mean Total Utilization")
```

```
## Warning: Using `size` aesthetic for lines was deprecated in ggplot2 3.4.0.  
## i Please use `linewidth` instead.  
## This warning is displayed once every 8 hours.  
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was  
## generated.
```



This line graph illustrates trends in average per capita mental health service utilization across different U.S. regions over time. Each line represents the mean total utilization within a region for a given year, with color used to differentiate between regions. The visualization captures regional variation and temporal trends in mental health care access and use. For example, some regions may demonstrate a steady increase in utilization over time, while others show more variability or stagnation. These patterns can point to disparities in mental health infrastructure, policy, or awareness campaigns, and highlight where further investigation or targeted interventions may be needed.

```
latest_year <- max(data$year, na.rm = TRUE)

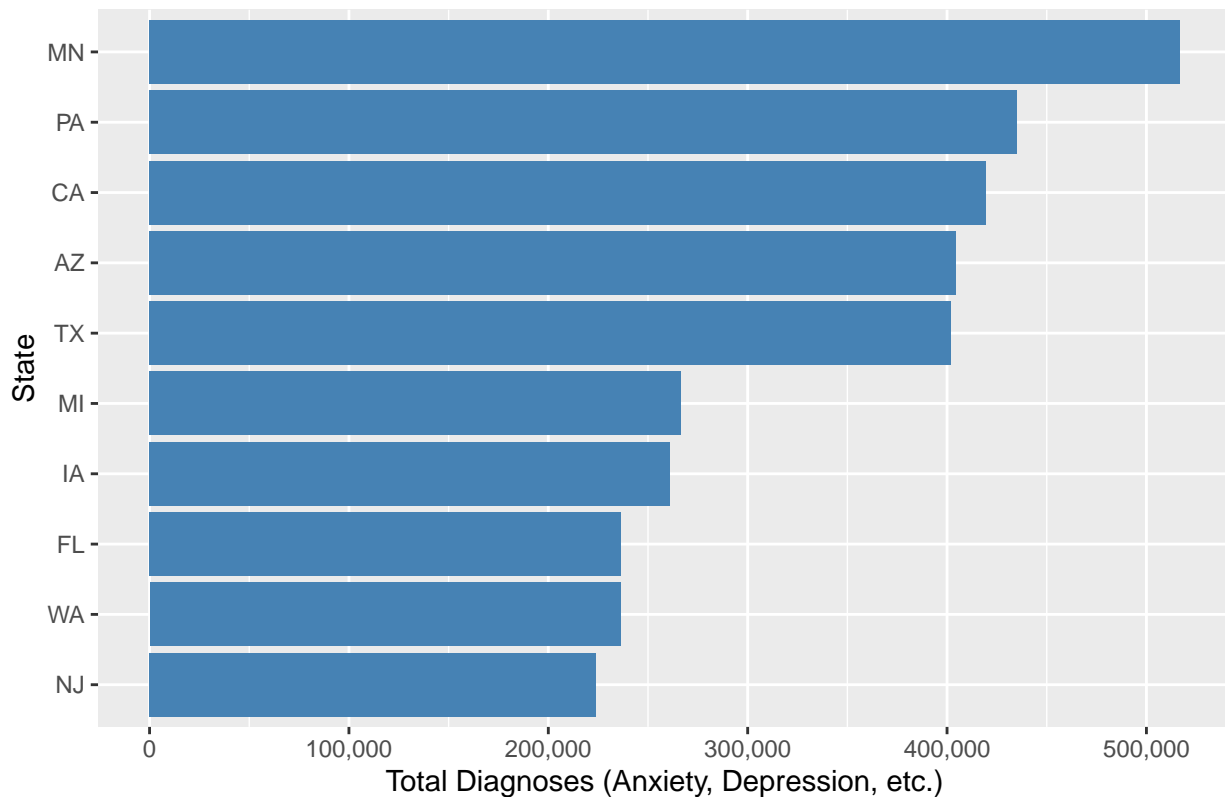
data_latest <- filter(data, year == latest_year)

data_latest$diagnoses_total <- rowSums(data_latest[, c("anxiety_ct", "depression_ct", "adhd_ct", "bipol

top_diagnosis_states <- data_latest %>%
  arrange(desc(diagnoses_total)) %>%
  head(10)

ggplot(top_diagnosis_states, aes(x = reorder(state, diagnoses_total), y = diagnoses_total)) +
  geom_col(fill = "steelblue") +
  coord_flip() +
  labs(title = paste("Top 10 States by Total Mental Health Diagnoses in", latest_year),
       x = "State", y = "Total Diagnoses (Anxiety, Depression, etc.)") +
  scale_y_continuous(labels = label_comma())
```

Top 10 States by Total Mental Health Diagnoses in 2022



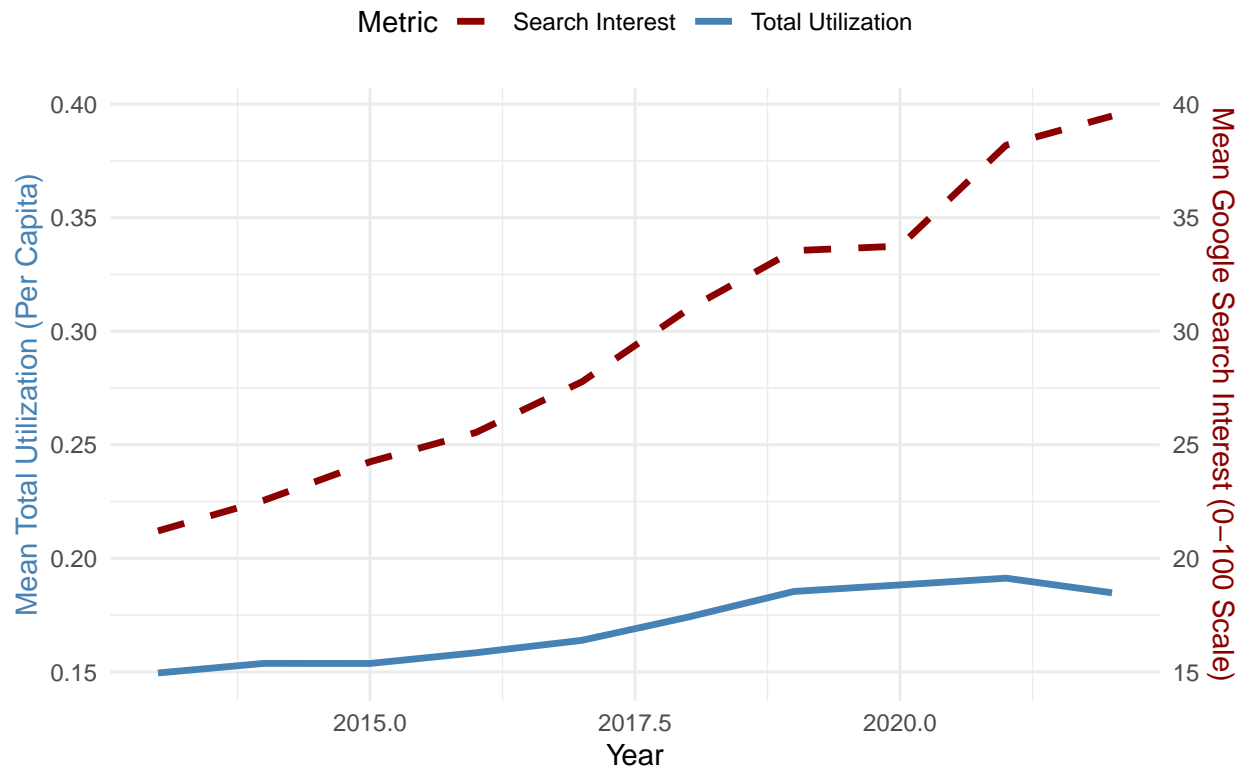
This horizontal bar chart displays the ten states with the highest total number of reported mental health diagnoses in the most recent year of available data (as determined by the dataset). The diagnoses include anxiety, depression, ADHD, bipolar disorder, and trauma-related stress. By aggregating these categories into a single metric (`diagnoses_total`), this visualization provides a snapshot of the states with the greatest mental health treatment burden. States are ranked in descending order, making it easy to compare the relative magnitude of cases across regions. This plot can help prioritize resource allocation, policy attention, and further research into population mental health needs in the most affected areas.

```
summary_by_year <- data %>%
  group_by(year) %>%
  summarise(
    mean_total_util = mean(total_util, na.rm = TRUE),
    mean_search_interest = mean(mean_all_trends, na.rm = TRUE)
  )

ggplot(summary_by_year, aes(x = year)) +
  geom_line(aes(y = mean_total_util, color = "Total Utilization"), size = 1.2) +
  geom_line(aes(y = mean_search_interest / 100, color = "Search Interest"), size = 1.2, linetype = "dashed") +
  scale_y_continuous(
    name = "Mean Total Utilization (Per Capita)",
    sec.axis = sec_axis(~ . * 100, name = "Mean Google Search Interest (0-100 Scale)")
  ) +
  scale_color_manual(values = c("Total Utilization" = "steelblue", "Search Interest" = "darkred")) +
  labs(
    title = "Mean Mental Health Utilization vs. Search Interest Over Time",
    x = "Year",
    color = "Metric"
  )
```

```
) +
theme_minimal() +
theme(
  axis.title.y.left = element_text(color = "steelblue"),
  axis.title.y.right = element_text(color = "darkred"),
  legend.position = "top")
```

Mean Mental Health Utilization vs. Search Interest Over Time



This dual-axis line chart compares the mean per capita mental health service utilization (`total_util`) with average Google search interest in mental health-related terms (`mean_all_trends`) across all states, over time. To enable meaningful visual comparison, search interest (originally on a 0–100 scale) is scaled down on the primary y-axis and also displayed on a secondary y-axis in its original form. The solid blue line represents mental health service use, while the dashed red line reflects online search trends. The chart offers insight into whether public awareness or concern—reflected in search behavior—moves in tandem with actual utilization of mental health services. Patterns of convergence or divergence between the two lines over time may inform whether increases in search interest precede or lag changes in service demand, which is valuable for proactive mental health planning and outreach.

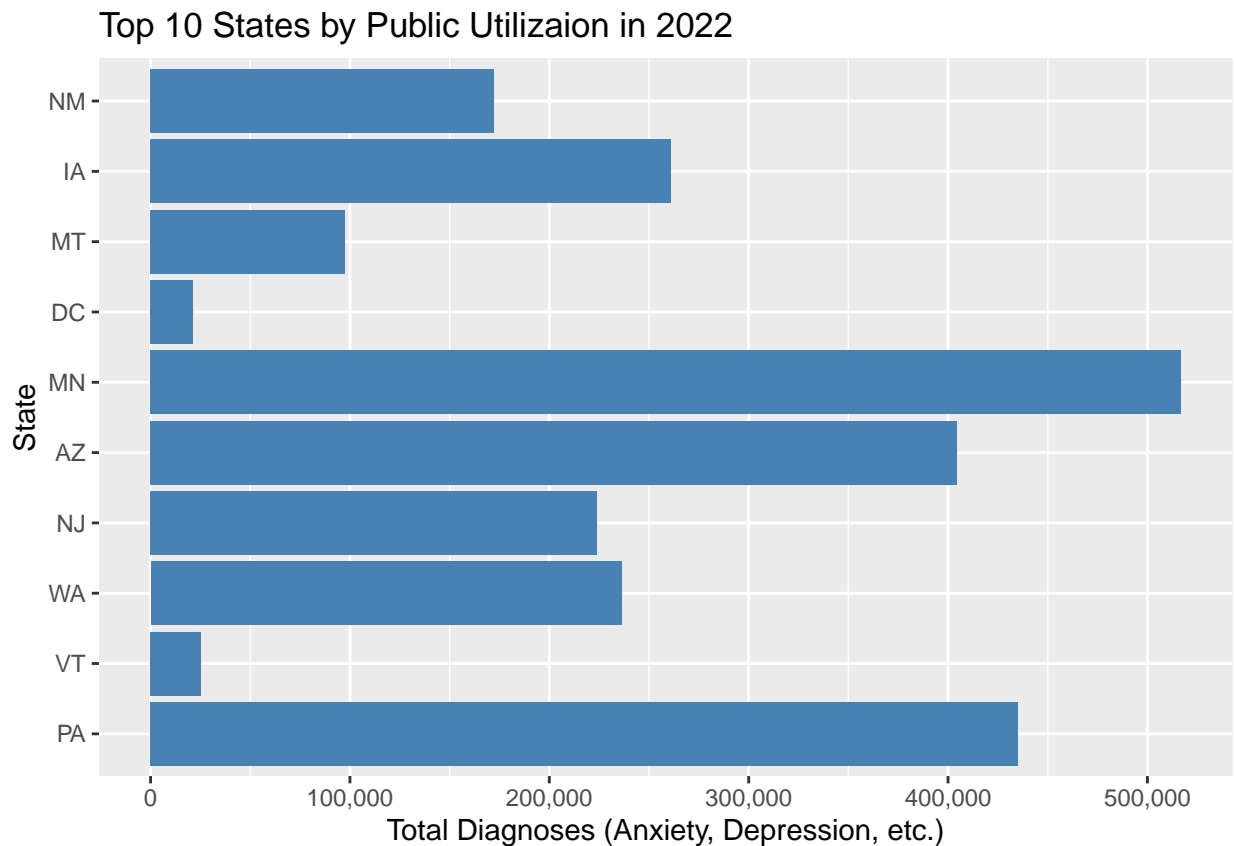
```
data_latest$diagnoses_total <- rowSums(data_latest[, c("anxiety_ct", "depression_ct", "adhd_ct", "bipol

top_public_utilization_states <- data_latest %>%
  arrange(desc(state_util)) %>%
  head(10)

ggplot(top_public_utilization_states, aes(x = reorder(state, state_util), y = diagnoses_total)) +
  geom_col(fill = "steelblue") +
  coord_flip() +
```



```
labs(title = paste("Top 10 States by Public Utilizaion in", latest_year),
     x = "State", y = "Total Diagnoses (Anxiety, Depression, etc.)") +
scale_y_continuous(labels = label_comma())
```



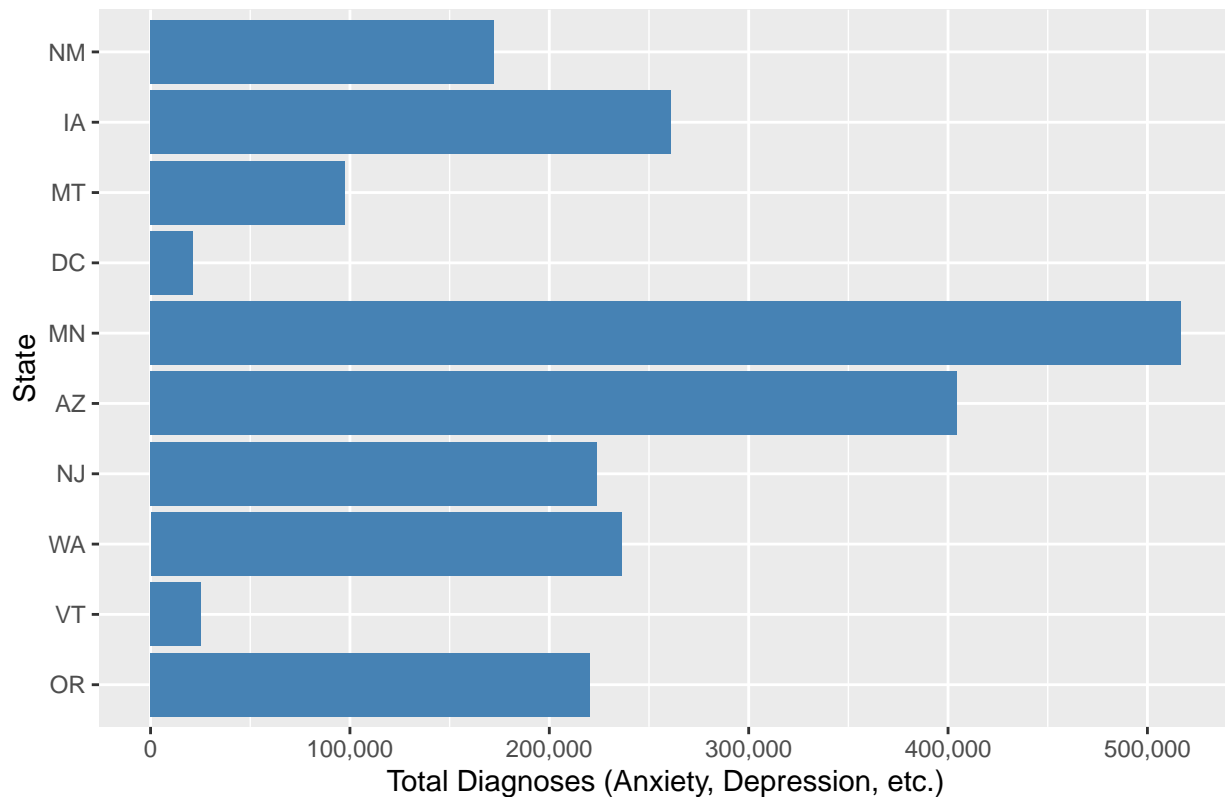
This horizontal bar chart highlights the top 10 states with the highest public mental health service utilization (`state_util`) in the most recent year of the dataset. States are sorted by utilization, and the bars represent the total number of mental health-related diagnoses, including anxiety, depression, ADHD, bipolar disorder, and trauma-related conditions. The visualization allows for a quick comparison between states with the most active public service usage and the corresponding mental health burden. It provides insight into whether states with higher utilization are also experiencing a greater volume of mental health diagnoses, which may indicate stronger reporting, access to care, or elevated need.

```
data_latest$diagnoses_total <- rowSums(data_latest[, c("anxiety_ct", "depression_ct", "adhd_ct", "bipol

top_private_utilization_states <- data_latest %>%
  arrange(desc(private_util)) %>%
  head(10)

ggplot(top_private_utilization_states, aes(x = reorder(state, private_util), y = diagnoses_total)) +
  geom_col(fill = "steelblue") +
  coord_flip() +
  labs(title = paste("Top 10 States by Private Utilizaion in", latest_year),
       x = "State", y = "Total Diagnoses (Anxiety, Depression, etc.)") +
  scale_y_continuous(labels = label_comma())
```

Top 10 States by Private Utilizaion in 2022



This chart displays the top 10 states with the highest private sector mental health service utilization (`private_util`) in the latest available year. Each bar reflects the total number of mental health diagnoses (covering anxiety, depression, ADHD, bipolar disorder, and trauma-related conditions) for each state. The visualization allows us to examine whether high levels of private utilization correspond to a greater mental health burden. It offers insight into how the private healthcare system is being leveraged across different states and can help identify regions with strong insurance access or provider networks supporting mental health care.

```
summary_by_year <- data %>%
  group_by(year) %>%
  summarise(
    mean_state_util = mean(state_util, na.rm = TRUE),
    mean_search_interest = mean(mean_all_trends, na.rm = TRUE)
  )

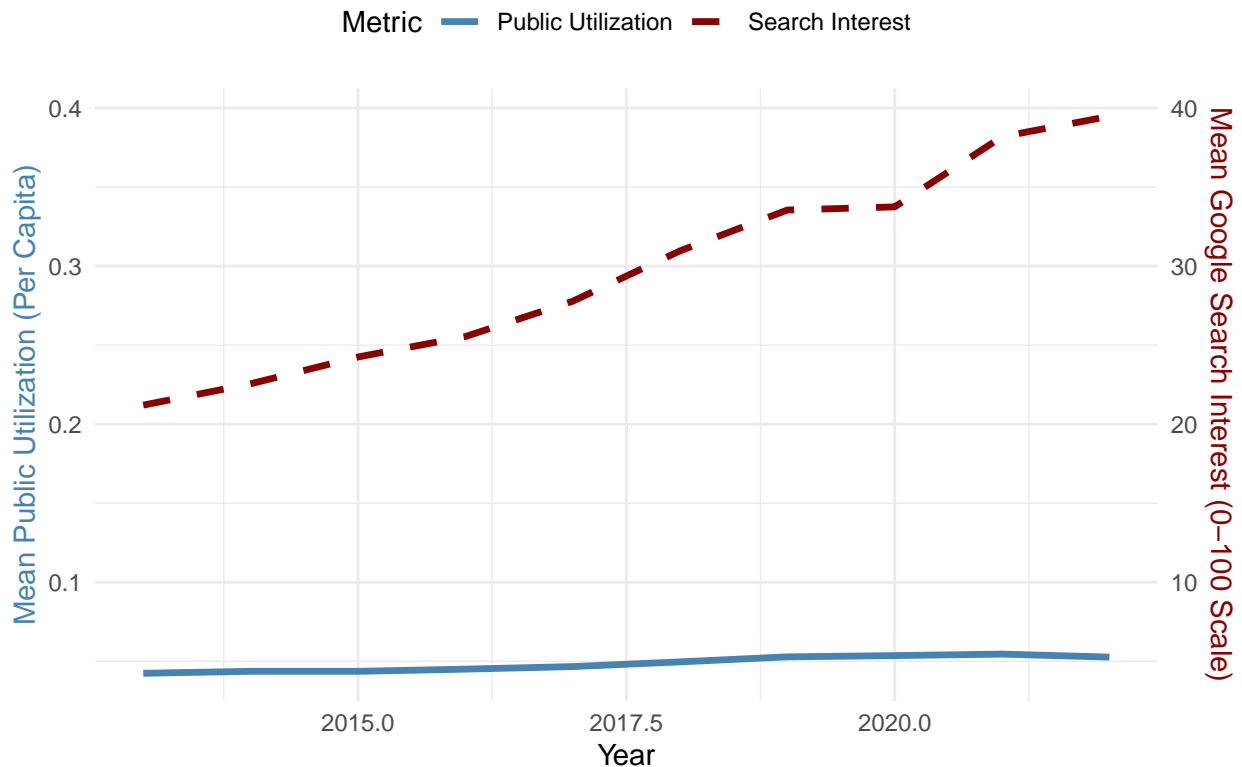
ggplot(summary_by_year, aes(x = year)) +
  geom_line(aes(y = mean_state_util, color = "Public Utilization"), size = 1.2) +
  geom_line(aes(y = mean_search_interest / 100, color = "Search Interest"), size = 1.2, linetype = "dashed") +
  scale_y_continuous(
    name = "Mean Public Utilization (Per Capita)",
    sec.axis = sec_axis(~ . * 100, name = "Mean Google Search Interest (0-100 Scale)")
  ) +
  scale_color_manual(values = c(
    "Public Utilization" = "steelblue",
    "Search Interest" = "darkred"
  )) +
  labs(
```

```

title = "Mean Mental Health Public Utilization vs. Search Interest Over Time",
x = "Year",
color = "Metric"
) +
theme_minimal() +
theme(
  axis.title.y.left = element_text(color = "steelblue"),
  axis.title.y.right = element_text(color = "darkred"),
  legend.position = "top"
)

```

Mean Mental Health Public Utilization vs. Search Interest Over Time



This dual-axis line graph compares mean per capita public mental health service utilization (`state_util`) with mean Google search interest in mental health topics across years. Public utilization is shown as a solid blue line, while search interest is represented by a red dashed line, scaled to align visually. This visualization helps illustrate whether public mental health service demand tracks with public awareness or concern, as reflected in search trends. Notably, the alignment (or divergence) between the two trends across time may reveal gaps between perceived need and service access in public health systems, guiding future policy considerations.

```

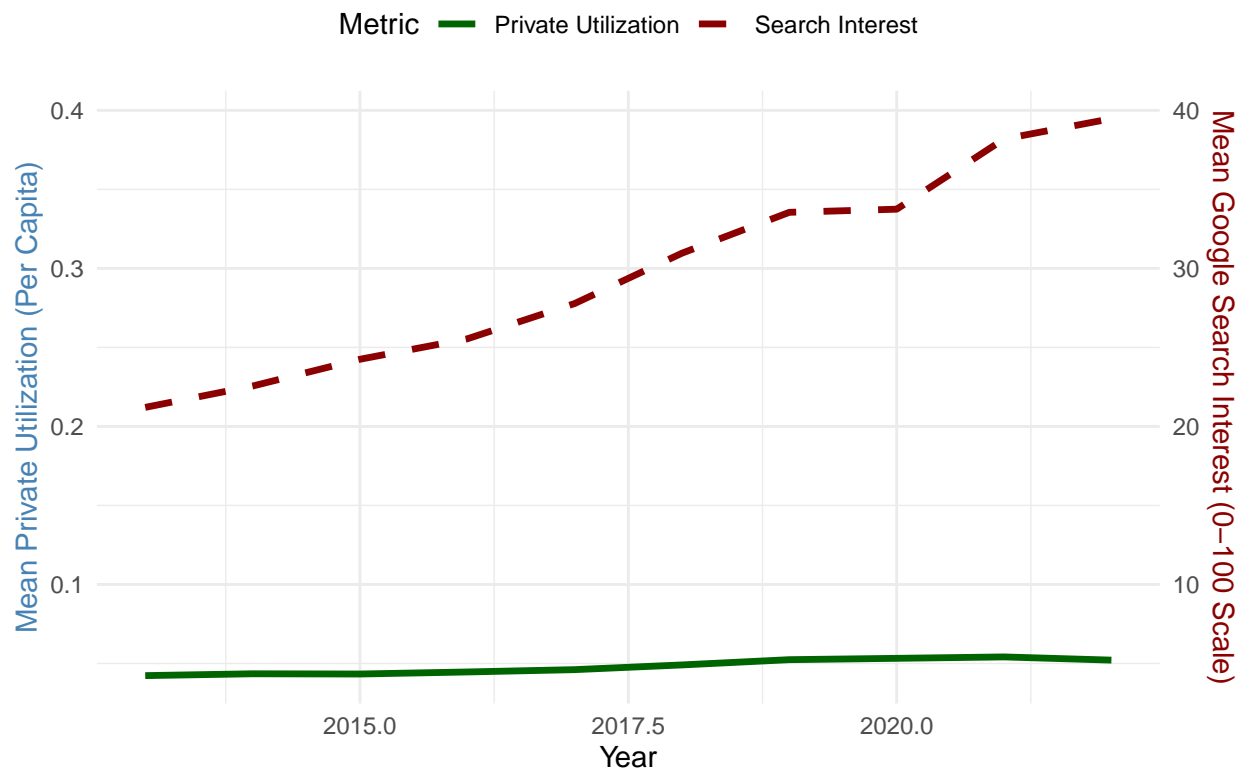
summary_by_year <- data %>%
  group_by(year) %>%
  summarise(
    mean_private_util = mean(private_util, na.rm = TRUE),
    mean_search_interest = mean(mean_all_trends, na.rm = TRUE)
  )

ggplot(summary_by_year, aes(x = year)) +
  geom_line(aes(y = mean_private_util, color = "Private Utilization"), size = 1.2) +

```

```
geom_line(aes(y = mean_search_interest / 100, color = "Search Interest"), size = 1.2, linetype = "dashed")
scale_y_continuous(
  name = "Mean Private Utilization (Per Capita)",
  sec.axis = sec_axis(~ . * 100, name = "Mean Google Search Interest (0-100 Scale)")
) +
scale_color_manual(values = c(
  "Private Utilization" = "darkgreen",
  "Search Interest" = "darkred"
)) +
labs(
  title = "Mean Mental Health Private Utilization vs. Search Interest Over Time",
  x = "Year",
  color = "Metric"
) +
theme_minimal() +
theme(
  axis.title.y.left = element_text(color = "steelblue"),
  axis.title.y.right = element_text(color = "darkred"),
  legend.position = "top"
)
```

Mean Mental Health Private Utilization vs. Search Interest Over Time



This dual-axis line graph compares mean per capita private mental health service utilization (`private_util`) with mean Google search interest in mental health topics over time. The private utilization is displayed with a solid dark green line, while search interest is shown as a dashed red line, scaled to fit the graph. The visualization provides insight into the relationship between the use of private mental health services and the public's awareness or concern as measured by search trends. This plot helps identify whether changes in

private service usage coincide with shifts in search interest, potentially indicating shifts in public awareness or engagement with mental health resources.