

# Project 1

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## Load Packages and Functions

In this section, we load all necessary libraries.

```
library(tidyverse)
library(readr)
library(ggplot2)
```

## Task 1: Data Processing

### Question 1: Read in the dataset

We want to read in some of this Census data set, but not all of it. Here we specify which columns we want to read in and we named this data set: `df_selected`. We then slice the first 5 lines to display them to confirm we read the data in correctly.

```
#Read in the data while selecting specific columns
df_selected <- read_csv(
  "https://www4.stat.ncsu.edu/~online/datasets/EDU01a.csv",
  show_col_types = FALSE) %>%
  select(Area_name, STCOU, ends_with("D")) %>% #select specified columns
  rename(area_name = Area_name) #rename "Area_name" as directed

#Display the first 5 lines
df_selected %>%
  slice(1:5)
```

```
# A tibble: 5 x 12
  area_name      STCOU EDU010187D EDU010188D EDU010189D EDU010190D EDU010191D
  <chr>          <chr>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>
```

```

1 UNITED STATES 00000 40024299 39967624 40317775 40737600 41385442
2 ALABAMA 01000 733735 728234 730048 728252 725541
3 Autauga, AL 01001 6829 6900 6920 6847 7008
4 Baldwin, AL 01003 16417 16465 16799 17054 17479
5 Barbour, AL 01005 5071 5098 5068 5156 5173
# i 5 more variables: EDU010192D <dbl>, EDU010193D <dbl>, EDU010194D <dbl>,
# EDU010195D <dbl>, EDU010196D <dbl>

```

## Question 2: Convert to long format

Now we want to convert the data into long format where each row has only one enrollment value for `area_name`. This converted data will be called `df_long`. We then display the first 5 rows to make sure everything looks as expected.

```

df_long <- pivot_longer(
  df_selected,
  cols = ends_with("D"),
  names_to = "Survey",
  values_to = "Enrollment Value"
)

#Display the first 5 lines
df_long %>%
  slice(1:5)

```

```

# A tibble: 5 x 4
  area_name      STCOU Survey      `Enrollment Value`
  <chr>          <chr> <chr>          <dbl>
1 UNITED STATES 00000 EDU010187D      40024299
2 UNITED STATES 00000 EDU010188D      39967624
3 UNITED STATES 00000 EDU010189D      40317775
4 UNITED STATES 00000 EDU010190D      40737600
5 UNITED STATES 00000 EDU010191D      41385442

```

## Question 3: Split a variable into 2 variables

Now we need to separate some values that are currently combined in `Survey`. The first 7 digits of `Survey` are currently a `Item_ID` (public school enrollment) and the last 2 digits followed by D are the school year. We want to separate these values to create 2 corresponding variables and turn the year into a 4 digit format. Since we will not be working with any data that was

before the year 1925 or after the year 2025, we can do some simple math. The Year 1987 will be referring to the Fall 1986-1987 school year.

```
#Separate and create variables from Survey
long_updated <- df_long %>%
  mutate(
    Year = as.numeric(substr(Survey, 8, 9)),
    Year = ifelse(Year > 25, Year + 1900, Year + 2000),
    Item_ID = substr(Survey, 1, 7)
  )
#Display the first 5 lines
long_updated %>%
  slice(1:5)
```

```
# A tibble: 5 x 6
  area_name      STCOU Survey      `Enrollment Value`   Year Item_ID
  <chr>          <chr> <chr>                <dbl> <dbl> <chr>
1 UNITED STATES 00000 EDU010187D          40024299  1987 EDU0101
2 UNITED STATES 00000 EDU010188D          39967624  1988 EDU0101
3 UNITED STATES 00000 EDU010189D          40317775  1989 EDU0101
4 UNITED STATES 00000 EDU010190D          40737600  1990 EDU0101
5 UNITED STATES 00000 EDU010191D          41385442  1991 EDU0101
```

#### Question 4: Create county and state data sets

Now we want to create a data set for non-county data and a data set for only county level data. As directed, we will add a class to the county level data tibble that's called **county** and we will create a class for the non-county data called **state**. Then we will print the first 10 rows of each tibble to make sure they look correct.

```
#Create the county and state data sets
county_idx <- grep(" ", "\\w\\w", long_updated$area_name)
county_tibble <- long_updated[county_idx, ]
state_tibble <- long_updated[-county_idx, ]

#add class accordingly
class(county_tibble) <- c("county", class(county_tibble))
class(state_tibble) <- c("state", class(state_tibble))

#display first 10 lines of county data
county_tibble %>%
  slice(1:10)
```

```
# A tibble: 10 x 6
```

	area_name	STCOU	Survey	`Enrollment Value`	Year	Item_ID	
	<chr>	<chr>	<chr>	<dbl>	<dbl>	<chr>	
1	Autauga,	AL	01001	EDU010187D	6829	1987	EDU0101
2	Autauga,	AL	01001	EDU010188D	6900	1988	EDU0101
3	Autauga,	AL	01001	EDU010189D	6920	1989	EDU0101
4	Autauga,	AL	01001	EDU010190D	6847	1990	EDU0101
5	Autauga,	AL	01001	EDU010191D	7008	1991	EDU0101
6	Autauga,	AL	01001	EDU010192D	7137	1992	EDU0101
7	Autauga,	AL	01001	EDU010193D	7152	1993	EDU0101
8	Autauga,	AL	01001	EDU010194D	7381	1994	EDU0101
9	Autauga,	AL	01001	EDU010195D	7568	1995	EDU0101
10	Autauga,	AL	01001	EDU010196D	7834	1996	EDU0101

```
#display first 10 lines of state data
state_tibble %>%
slice(1:10)
```

```
# A tibble: 10 x 6
```

	area_name	STCOU	Survey	`Enrollment Value`	Year	Item_ID
	<chr>	<chr>	<chr>	<dbl>	<dbl>	<chr>
1	UNITED STATES	00000	EDU010187D	40024299	1987	EDU0101
2	UNITED STATES	00000	EDU010188D	39967624	1988	EDU0101
3	UNITED STATES	00000	EDU010189D	40317775	1989	EDU0101
4	UNITED STATES	00000	EDU010190D	40737600	1990	EDU0101
5	UNITED STATES	00000	EDU010191D	41385442	1991	EDU0101
6	UNITED STATES	00000	EDU010192D	42088151	1992	EDU0101
7	UNITED STATES	00000	EDU010193D	42724710	1993	EDU0101
8	UNITED STATES	00000	EDU010194D	43369917	1994	EDU0101
9	UNITED STATES	00000	EDU010195D	43993459	1995	EDU0101
10	UNITED STATES	00000	EDU010196D	44715737	1996	EDU0101

### Question 5: Add the state to each county

Now we want to add the state that each county corresponds to as a new variable in our tibble. This new variable will be called `State`.

```
# Add State to county
county_tibble <- county_tibble %>%
  mutate(State = substr(area_name, nchar(area_name) - 1, nchar(area_name)))
```

```
#display first 10 lines of county data
county_tibble %>%
slice(1:10)
```

```
# A tibble: 10 x 7
```

	area_name	STCOU	Survey	`Enrollment Value`	Year	Item_ID	State
	<chr>	<chr>	<chr>	<dbl>	<dbl>	<chr>	<chr>
1	Autauga, AL	01001	EDU010187D	6829	1987	EDU0101	AL
2	Autauga, AL	01001	EDU010188D	6900	1988	EDU0101	AL
3	Autauga, AL	01001	EDU010189D	6920	1989	EDU0101	AL
4	Autauga, AL	01001	EDU010190D	6847	1990	EDU0101	AL
5	Autauga, AL	01001	EDU010191D	7008	1991	EDU0101	AL
6	Autauga, AL	01001	EDU010192D	7137	1992	EDU0101	AL
7	Autauga, AL	01001	EDU010193D	7152	1993	EDU0101	AL
8	Autauga, AL	01001	EDU010194D	7381	1994	EDU0101	AL
9	Autauga, AL	01001	EDU010195D	7568	1995	EDU0101	AL
10	Autauga, AL	01001	EDU010196D	7834	1996	EDU0101	AL

### Question 6: Add Division to the state tibble

Now for our non-county level tibble, we want to create a new variable called `Division` that corresponds to the state's classification of division. Since we will have many conditions, we will use `case_when`. For rows that correspond to a non-state (such as UNITED STATES) we will return `ERROR`.

```
# Add Division to non-county tibble
state_tibble <- state_tibble %>%
mutate(area_name = toupper(area_name)) %>%
mutate(
  Division = case_when(
    area_name %in% c(
      "CONNECTICUT", "MAINE", "MASSACHUSETTS",
      "NEW HAMPSHIRE", "RHODE ISLAND", "VERMONT"
    ) ~ "New England",
    area_name %in% c(
      "NEW JERSEY", "NEW YORK", "PENNSYLVANIA"
    ) ~ "Middle Atlantic",
    area_name %in% c(
      "ILLINOIS", "INDIANA", "MICHIGAN", "OHIO", "WISCONSIN"
    ) ~ "East North Central",
    area_name %in% c(
```

```

    "IOWA", "KANSAS", "MINNESOTA", "MISSOURI", "NEBRASKA",
    "NORTH DAKOTA", "SOUTH DAKOTA"
  ) ~ "West North Central",
  area_name %in% c(
    "DELAWARE", "MARYLAND", "DISTRICT OF COLUMBIA",
    "VIRGINIA", "WEST VIRGINIA", "NORTH CAROLINA",
    "SOUTH CAROLINA", "GEORGIA", "FLORIDA"
  ) ~ "South Atlantic",
  area_name %in% c(
    "ALABAMA", "KENTUCKY", "MISSISSIPPI", "TENNESSEE"
  ) ~ "East South Central",
  area_name %in% c(
    "ARKANSAS", "LOUISIANA", "OKLAHOMA", "TEXAS"
  ) ~ "West South Central",
  area_name %in% c(
    "ARIZONA", "COLORADO", "IDAHO", "MONTANA",
    "NEVADA", "NEW MEXICO", "UTAH", "WYOMING"
  ) ~ "Mountain",
  area_name %in% c(
    "ALASKA", "CALIFORNIA", "HAWAII", "OREGON", "WASHINGTON"
  ) ~ "Pacific",
  TRUE ~ "ERROR"
)
)

# Check Division table (no zeros!)
state_tibble %>% count(Division, area_name) %>% print(n = 52)

```

```

# A tibble: 52 x 3
  Division      area_name      n
  <chr>         <chr>      <int>
1 ERROR        UNITED STATES    10
2 East North Central ILLINOIS    10
3 East North Central INDIANA    10
4 East North Central MICHIGAN    10
5 East North Central OHIO    10
6 East North Central WISCONSIN    10
7 East South Central ALABAMA    10
8 East South Central KENTUCKY    10
9 East South Central MISSISSIPPI    10
10 East South Central TENNESSEE    10
11 Middle Atlantic NEW JERSEY    10

```

12	Middle Atlantic	NEW YORK	10
13	Middle Atlantic	PENNSYLVANIA	10
14	Mountain	ARIZONA	10
15	Mountain	COLORADO	10
16	Mountain	IDAHO	10
17	Mountain	MONTANA	10
18	Mountain	NEVADA	10
19	Mountain	NEW MEXICO	10
20	Mountain	UTAH	10
21	Mountain	WYOMING	10
22	New England	CONNECTICUT	10
23	New England	MAINE	10
24	New England	MASSACHUSETTS	10
25	New England	NEW HAMPSHIRE	10
26	New England	RHODE ISLAND	10
27	New England	VERMONT	10
28	Pacific	ALASKA	10
29	Pacific	CALIFORNIA	10
30	Pacific	HAWAII	10
31	Pacific	OREGON	10
32	Pacific	WASHINGTON	10
33	South Atlantic	DELAWARE	10
34	South Atlantic	DISTRICT OF COLUMBIA	20
35	South Atlantic	FLORIDA	10
36	South Atlantic	GEORGIA	10
37	South Atlantic	MARYLAND	10
38	South Atlantic	NORTH CAROLINA	10
39	South Atlantic	SOUTH CAROLINA	10
40	South Atlantic	VIRGINIA	10
41	South Atlantic	WEST VIRGINIA	10
42	West North Central	IOWA	10
43	West North Central	KANSAS	10
44	West North Central	MINNESOTA	10
45	West North Central	MISSOURI	10
46	West North Central	NEBRASKA	10
47	West North Central	NORTH DAKOTA	10
48	West North Central	SOUTH DAKOTA	10
49	West South Central	ARKANSAS	10
50	West South Central	LOUISIANA	10
51	West South Central	OKLAHOMA	10
52	West South Central	TEXAS	10

## Task 2: Simplify Task 1 through 6 with functions

### Questions 1-2 with a function

We created a function that has an optional argument to allow the user to specify the name of the column representing the value (which is `Enrollment Value` in our case)

```
#Create the function
long_format_conversion <- function(df, value = "Enrollment Value") {
  selected_data <- df %>%
    # Select required columns
    select(Area_name, STCOU, ends_with("D")) %>%
    # Rename Area_name to area_name
    rename(area_name = Area_name)

  # Convert to long format
  long_updated <- pivot_longer(
    selected_data,
    cols = ends_with("D"),
    names_to = "Survey",
    values_to = value
  )
  return(long_updated)
}
```

### Question 3 with a function

Now we want to break survey into 2 variables just like we did in question 3, however, we are going to make a function this time.

```
survey_function <- function(long_updated) {
  long_data_updated <- long_updated %>%
    # Extract the last two digits as Year and convert to numeric
    mutate(Year = as.numeric(substr(Survey, start = 8, stop = 9))) %>%
    # Convert two-digit Year to four-digit Year
    mutate(Year = ifelse(Year > 25, Year + 1900, Year + 2000)) %>%
    # Extract the first 7 characters as Item_ID
    mutate(Item_ID = substr(Survey, start = 1, stop = 7))

  return(long_data_updated)
}
```



### Question 5 with a function

Now we are creating a function to do step 5.

```
state_function <- function(county_tibble) {  
  new_county_tibble <- county_tibble %>%  
    mutate(State = substr(area_name, start = nchar(area_name) - 1,  
                          stop = nchar(area_name)))  
  return(new_county_tibble)  
}
```

### Question 6 with a function

Now we are completing step 6, but with a function

```
division_function <- function(state_tibble) {  
  state_tibble_updated <- state_tibble %>%  
    mutate(area_name = toupper(area_name)) %>% # Ensure uppercase for matching  
    mutate(  
      Division = case_when(  
        area_name %in% c(  
          "CONNECTICUT", "MAINE", "MASSACHUSETTS",  
          "NEW HAMPSHIRE", "RHODE ISLAND", "VERMONT"  
        ) ~ "New England",  
        area_name %in% c(  
          "NEW JERSEY", "NEW YORK", "PENNSYLVANIA"  
        ) ~ "Middle Atlantic",  
        area_name %in% c(  
          "ILLINOIS", "INDIANA", "MICHIGAN", "OHIO", "WISCONSIN"  
        ) ~ "East North Central",  
        area_name %in% c(  
          "IOWA", "KANSAS", "MINNESOTA", "MISSOURI",  
          "NEBRASKA", "NORTH DAKOTA", "SOUTH DAKOTA"  
        ) ~ "West North Central",  
        area_name %in% c(  
          "DELAWARE", "MARYLAND", "DISTRICT OF COLUMBIA",  
          "VIRGINIA", "WEST VIRGINIA", "NORTH CAROLINA",  
          "SOUTH CAROLINA", "GEORGIA", "FLORIDA"  
        ) ~ "South Atlantic",  
        area_name %in% c(  
          "ALABAMA", "KENTUCKY", "MISSISSIPPI", "TENNESSEE"  
        ) ~ "East South Central",  
      )  
    )  
}
```

```

    area_name %in% c(
      "ARKANSAS", "LOUISIANA", "OKLAHOMA", "TEXAS"
    ) ~ "West South Central",
    area_name %in% c(
      "ARIZONA", "COLORADO", "IDAHO", "MONTANA",
      "NEVADA", "NEW MEXICO", "UTAH", "WYOMING"
    ) ~ "Mountain",
    area_name %in% c(
      "ALASKA", "CALIFORNIA", "HAWAII", "OREGON", "WASHINGTON"
    ) ~ "Pacific",
    TRUE ~ "ERROR"
  )
)
return(state_tibble_updated)
}

```

### Questions 4-6 with 1 function

In order to create two new tibbles from the output from step 3 and apply steps 5-6, we need to create a new function to perform the following: take in the output from step 3, Creates 2 tibbles like in step 4, then calls the functions from steps 5 and 6 to return two final tibbles.

```

create_datasets <- function(long_updated) {
  # Split by presence of ", XX" at end of area_name
  county_indices <- grep(pattern = ", \\w\\w", long_updated$area_name)
  state_tibble <- long_updated[-county_indices, ]
  class(state_tibble) <- c("state", class(state_tibble))

  county_tibble <- long_updated[county_indices, ]
  class(county_tibble) <- c("county", class(county_tibble))

  final_county_tibble <- state_function(county_tibble)
  final_state_tibble <- division_function(state_tibble)

  return(list(county = final_county_tibble, state = final_state_tibble))
}

```

### Create a wrapper function

We can combine all of these functions into one function call. We want this function to take in a URL of a .csv file and apply all of the functions listed above.

```
my_wrapper <- function(url, value = "Enrollment Value") {
  result <- read_csv(url, show_col_types = FALSE) %>%
    long_format_conversion(value = value) %>%
    survey_function() %>%
    create_datasets()
  return(result)
}
```

## Call and combine Our data

We want to combine our data from two different URLs. We will create a single short function that takes in the results of the two calls in our wrapper function. At the end of this report, we will apply our wrapper functions to a few different data sets.

```
combine_wrapper_results <- function(wrapper1, wrapper2) {
  combined_county <- bind_rows(wrapper1$county, wrapper2$county)
  combined_state <- bind_rows(wrapper1$state, wrapper2$state)
  list(county = combined_county, state = combined_state)
}
```

## Task 3: Writing a Generic Function for Summarizing

### Plot function for state

Now we want to summarize some of our data. We want to do this efficiently, so we will create some functions. Here, we do not want to include any of the Divisions that are listed as **ERROR** in our data sets. This function will be used to create our state plots.

```
plot.state <- function(df, var_name = "Enrollment Value") {
  df %>%
    filter(Division != "ERROR") %>% #exclude error divisions
    group_by(Division, Year) %>%
    summarize(mean_value = mean(get(var_name), na.rm = TRUE), .groups = "drop") %>%
    ggplot(aes(x = Year, y = mean_value, color = Division)) +
    geom_line() +
    labs(
      title = "Mean Value per Division over Years",
      y = paste("Mean of", var_name),
      x = "Year"
    ) +
}
```

```

    scale_y_continuous(labels = scales::comma) + # normal number format
    guides(color = guide_legend(ncol = 2))      # multi-column legend
}

```

## Plot function for county

We will now create a similar function for our county data set, however we will add some calculations to this function. We want to be able to display the top or bottom mean enrollment values with our function, so we have to do some sorting. It's important to note that we are sorting based on mean values, but we are plotting the actual statistic values.

```

plot.county <- function(df, var_name = "Enrollment Value",
                        state = "NC", top_or_bottom = "top", n = 5) {

  # Filter data for selected state
  df_state <- df %>% filter(State == state)

  # Calculate mean for each county
  mean_df <- df_state %>%
    group_by(area_name) %>%
    summarize(mean_value = mean(get(var_name), na.rm = TRUE), .groups = "drop")

  # Sort by mean and pick top or bottom n
  if (top_or_bottom == "top") {
    mean_df <- mean_df %>% arrange(desc(mean_value))
  } else {
    mean_df <- mean_df %>% arrange(mean_value)
  }

  selected_areas <- mean_df %>% slice_head(n = n) %>% pull(area_name)

  # Filter original data for selected counties
  df_selected <- df_state %>% filter(area_name %in% selected_areas)

  # Plot actual values (not means)
  ggplot(df_selected, aes(x = Year, y = get(var_name), color = area_name)) +
    geom_line() +
    labs(
      title = paste(top_or_bottom, n, "counties in", state),
      y = var_name,
      x = "Year"
    )
}

```

```
)+
  scale_y_continuous(labels = scales::comma) + # normal number format
  guides(color = guide_legend(ncol = 2))      # multi-column legend
}
```

## Task 4: Putting it Together

### Process the EDU Data Sets

We run our wrapper function on the two EDU data sets and inspect the results. Since we are looking at enrollment values, all of our plots and code use `value = Enrollment value`. This is referring to the census observed enrollment counts at given locations and given years. Again, With our current data sets, we are looking at the Public school enrollment for school years.

```
edu1 <- my_wrapper(
  "https://www4.stat.ncsu.edu/~online/datasets/EDU01a.csv",
  value = "Enrollment Value"
)
edu2 <- my_wrapper(
  "https://www4.stat.ncsu.edu/~online/datasets/EDU01b.csv",
  value = "Enrollment Value"
)
```

```
# Inspect to ensure correctness
head(edu1$county)
```

```
# A tibble: 6 x 7
  area_name STCOU Survey `Enrollment Value` Year Item_ID State
  <chr>      <chr> <chr>          <dbl> <dbl> <chr>   <chr>
1 Autauga, AL 01001 EDU010187D      6829  1987 EDU0101 AL
2 Autauga, AL 01001 EDU010188D      6900  1988 EDU0101 AL
3 Autauga, AL 01001 EDU010189D      6920  1989 EDU0101 AL
4 Autauga, AL 01001 EDU010190D      6847  1990 EDU0101 AL
5 Autauga, AL 01001 EDU010191D      7008  1991 EDU0101 AL
6 Autauga, AL 01001 EDU010192D      7137  1992 EDU0101 AL
```

```
head(edu1$state)
```

```
# A tibble: 6 x 7
  area_name STCOU Survey `Enrollment Value` Year Item_ID Division
  <chr>      <chr> <chr>          <dbl> <dbl> <chr>   <chr>
```

	<chr>	<chr>	<chr>	<dbl>	<dbl>	<chr>	<chr>
1	UNITED STATES	00000	EDU010187D	40024299	1987	EDU0101	ERROR
2	UNITED STATES	00000	EDU010188D	39967624	1988	EDU0101	ERROR
3	UNITED STATES	00000	EDU010189D	40317775	1989	EDU0101	ERROR
4	UNITED STATES	00000	EDU010190D	40737600	1990	EDU0101	ERROR
5	UNITED STATES	00000	EDU010191D	41385442	1991	EDU0101	ERROR
6	UNITED STATES	00000	EDU010192D	42088151	1992	EDU0101	ERROR

## Combine EDU Data Sets

Here we use our combining function to merge the two processed data sets. We then display the first few rows to make sure this worked as intended.

```
edu_combined <- combine_wrapper_results(edu1, edu2)
head(edu_combined$county)
```

```
# A tibble: 6 x 7
  area_name STCOU Survey `Enrollment Value` Year Item_ID State
  <chr>      <chr> <chr>          <dbl> <dbl> <chr> <chr>
1 Autauga, AL 01001 EDU010187D      6829  1987 EDU0101 AL
2 Autauga, AL 01001 EDU010188D      6900  1988 EDU0101 AL
3 Autauga, AL 01001 EDU010189D      6920  1989 EDU0101 AL
4 Autauga, AL 01001 EDU010190D      6847  1990 EDU0101 AL
5 Autauga, AL 01001 EDU010191D      7008  1991 EDU0101 AL
6 Autauga, AL 01001 EDU010192D      7137  1992 EDU0101 AL
```

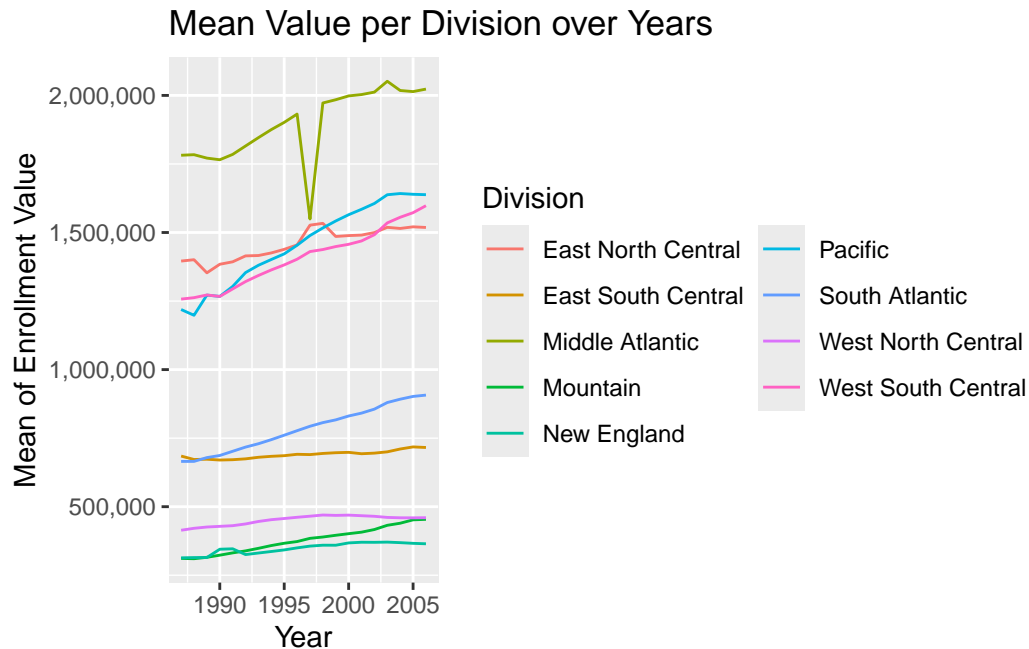
```
head(edu_combined$state)
```

```
# A tibble: 6 x 7
  area_name STCOU Survey `Enrollment Value` Year Item_ID Division
  <chr>      <chr> <chr>          <dbl> <dbl> <chr> <chr>
1 UNITED STATES 00000 EDU010187D      40024299 1987 EDU0101 ERROR
2 UNITED STATES 00000 EDU010188D      39967624 1988 EDU0101 ERROR
3 UNITED STATES 00000 EDU010189D      40317775 1989 EDU0101 ERROR
4 UNITED STATES 00000 EDU010190D      40737600 1990 EDU0101 ERROR
5 UNITED STATES 00000 EDU010191D      41385442 1991 EDU0101 ERROR
6 UNITED STATES 00000 EDU010192D      42088151 1992 EDU0101 ERROR
```

## State Plot for EDU Data

This plot shows the mean enrollment by Division across years. We are excluding Error divisions.

```
plot(edu_combined$state, var_name = "Enrollment Value")
```

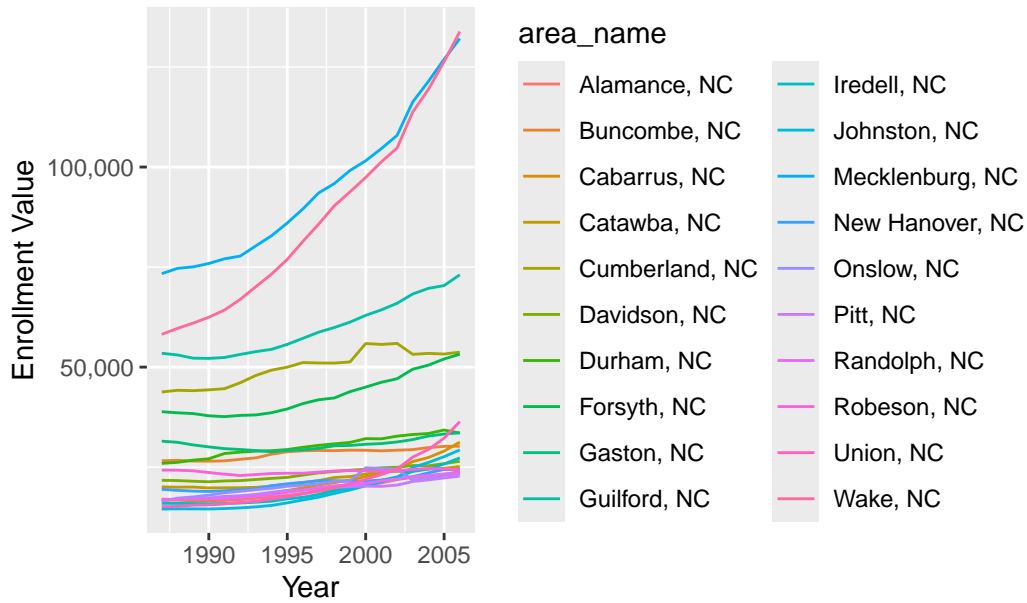


## County Plots for EDU Data

Below are various plots for county data, demonstrating flexibility in selecting state, top/bottom, and count.

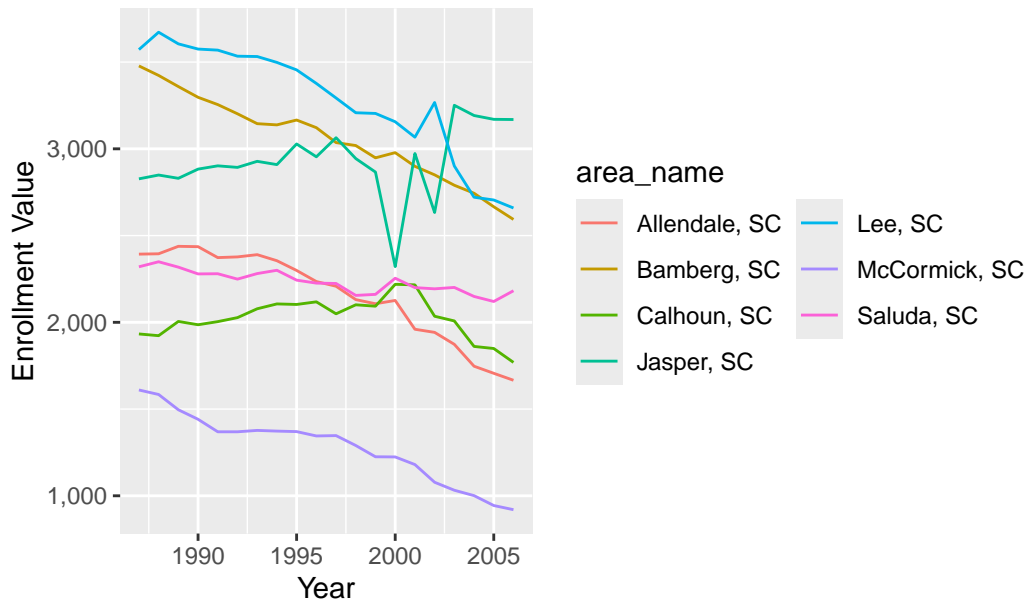
```
# NC, top 20
plot(edu_combined$county, var_name = "Enrollment Value",
     state = "NC", top_or_bottom = "top", n = 20)
```

top 20 counties in NC



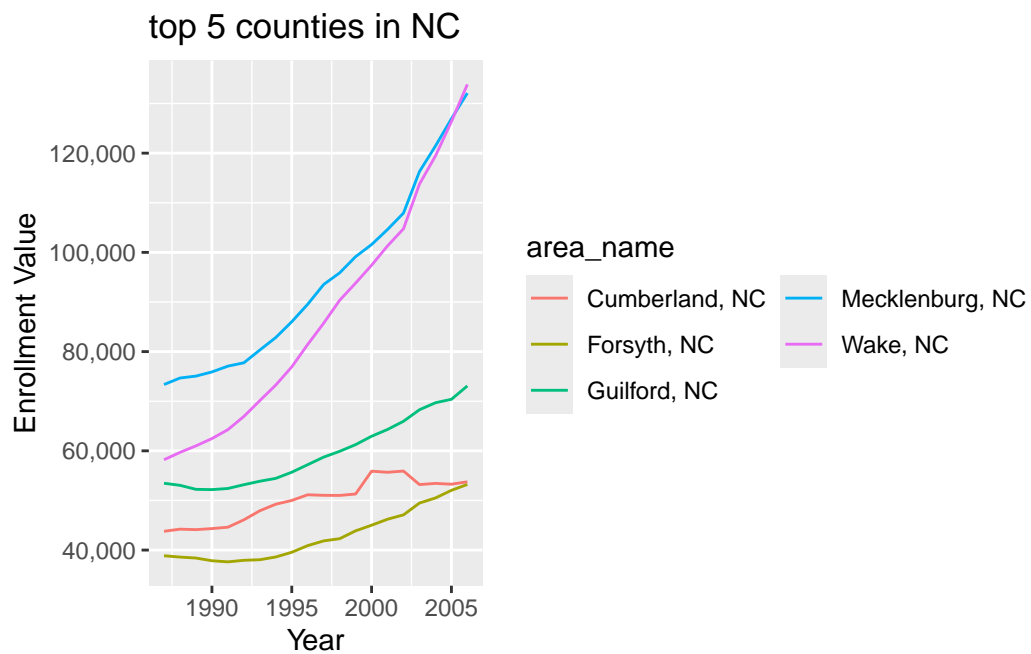
```
# SC, bottom 7
plot(edu_combined$county, var_name = "Enrollment Value",
     state = "SC", top_or_bottom = "bottom", n = 7)
```

bottom 7 counties in SC

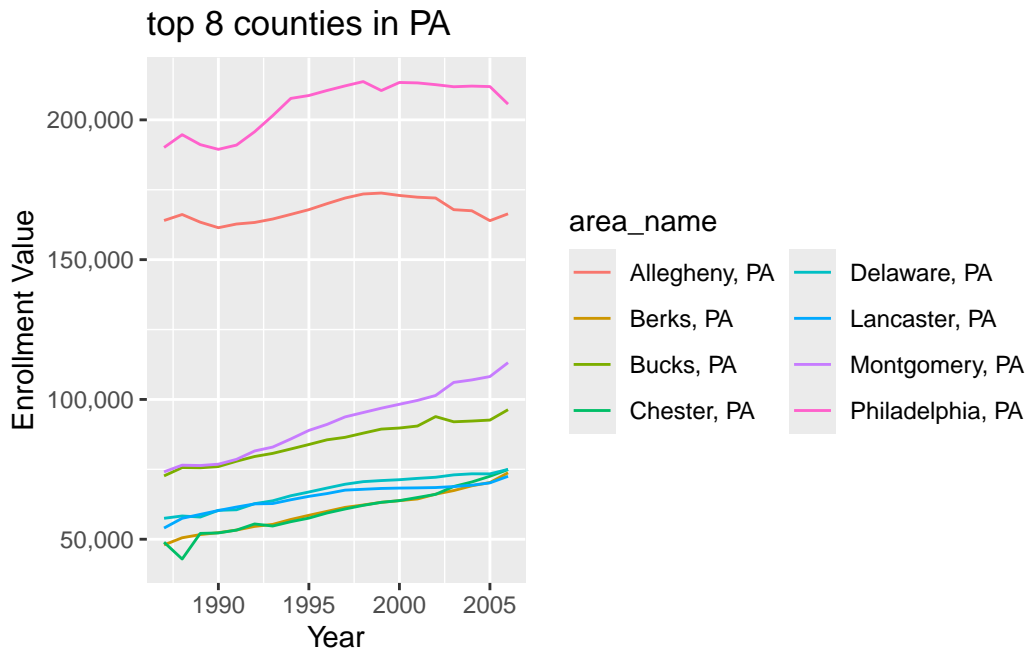




```
# Default (uses NC top 5)
plot(edu_combined$county, var_name = "Enrollment Value")
```



```
# PA, top 8
plot(edu_combined$county, var_name = "Enrollment Value",
     state = "PA", top_or_bottom = "top", n = 8)
```



##Task 4.5: Apply workflow to PST data sets

### Process PST Data Sets

We repeat the same workflow for the four PST data sets. We start by processing that data then combining the data sets. For the PST data, we are looking at Resident total population estimates at given locations on given years. **Year** 1971 is referencing July 1st 1971.

```
pst1 <- my_wrapper(
  "https://www4.stat.ncsu.edu/~online/datasets/PST01a.csv",
  value = "Resident Population"
)
pst2 <- my_wrapper(
  "https://www4.stat.ncsu.edu/~online/datasets/PST01b.csv",
  value = "Resident Population"
)
pst3 <- my_wrapper(
  "https://www4.stat.ncsu.edu/~online/datasets/PST01c.csv",
  value = "Resident Population"
)
pst4 <- my_wrapper(
  "https://www4.stat.ncsu.edu/~online/datasets/PST01d.csv",
```

```

    value = "Resident Population"
  )

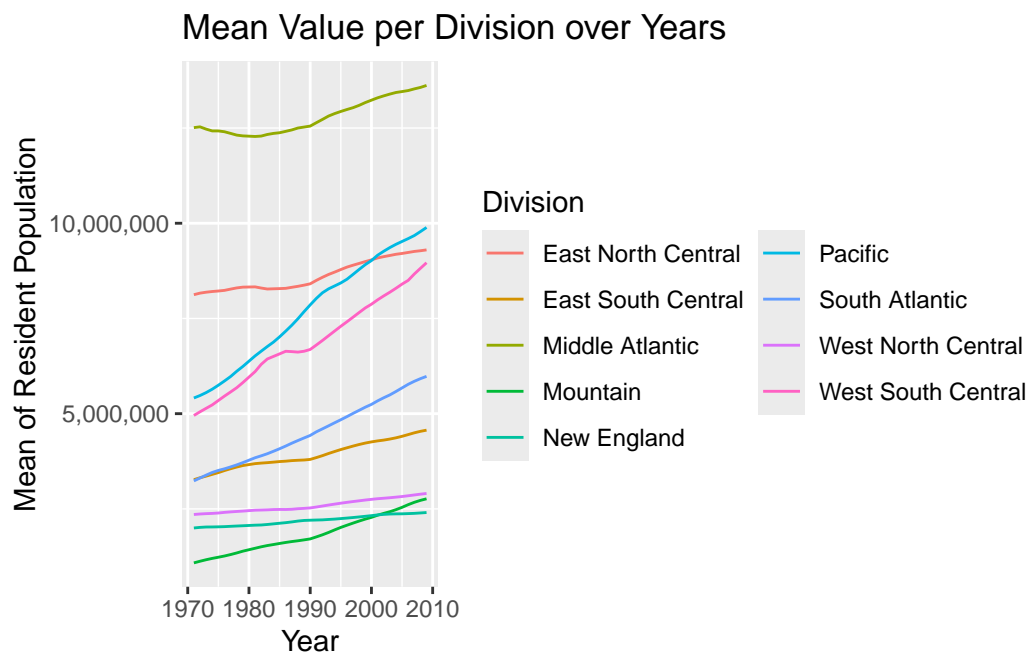
# Combine step by step
pst12 <- combine_wrapper_results(pst1, pst2)
pst34 <- combine_wrapper_results(pst3, pst4)
pst_combined <- combine_wrapper_results(pst12, pst34)

```

## State Plot for PST Data

Now that the PST data is processed, we will make our state plot.

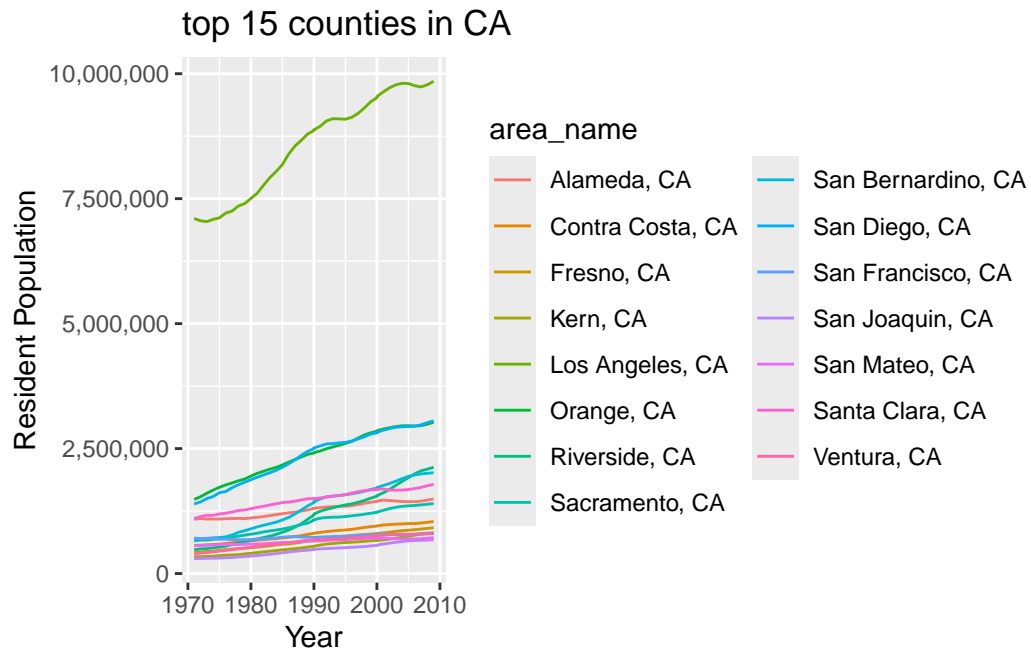
```
plot(pst_combined$state, var_name = "Resident Population")
```



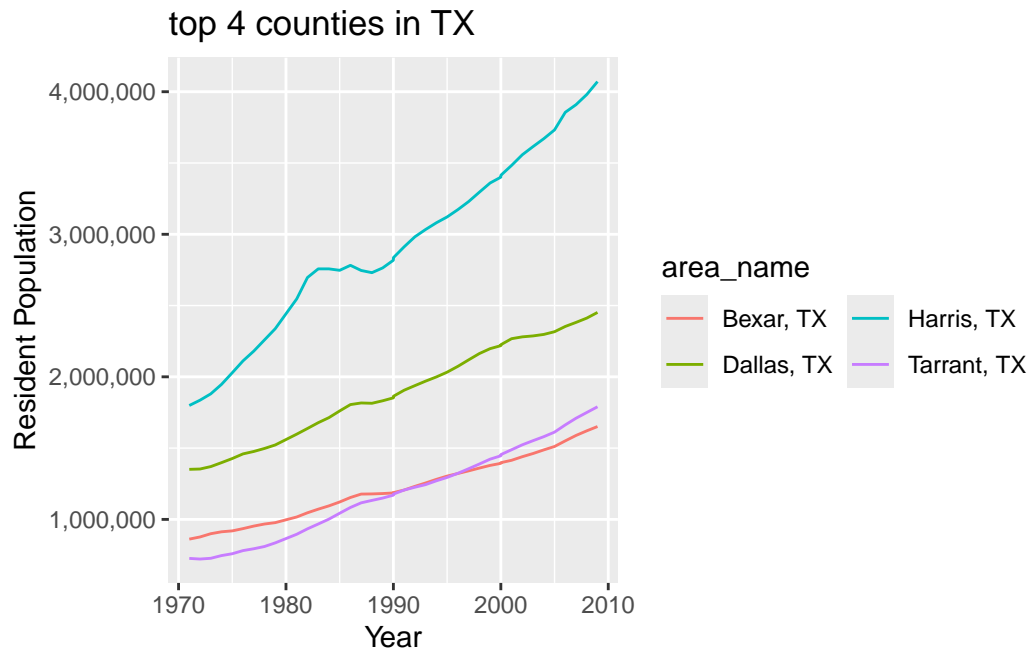
## County Plots for PST Data

Finally, we want to demonstrate the flexibility of our summary functions with the PST Data. As noted in the code chunk below, we created plots that show: CA top 15, TX top 4, the default plot function, and NY top 10.

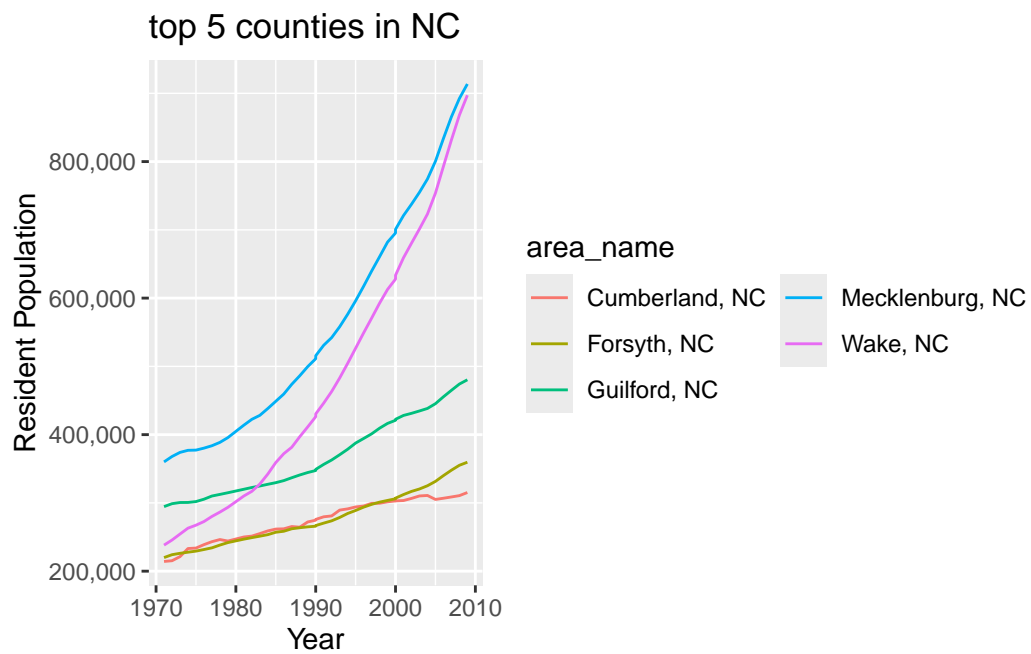
```
# CA, top 15
plot(pst_combined$county, var_name = "Resident Population",
     state = "CA", top_or_bottom = "top", n = 15)
```



```
# TX, top 4
plot(pst_combined$county, var_name = "Resident Population",
     state = "TX", top_or_bottom = "top", n = 4)
```



```
# Default
plot(pst_combined$county, var_name = "Resident Population")
```



```
# NY, top 10
plot(pst_combined$county, var_name = "Resident Population",
     state = "NY", top_or_bottom = "top", n = 10)
```

