

What is your major mal-function?

PROGRAMMING WITH DPLYR



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Recall Uruguay's CPI

```
imf_data %>%  
  select(iso, country, year,  
         consumer_price_index) %>%  
  filter(country == "Uruguay",  
         year > 2010)
```

```
# A tibble: 9 × 4  
  iso    country    year consumer_price_index  
  <chr> <chr>    <int>         <dbl>  
1 URY    Uruguay    2011          105.  
2 URY    Uruguay    2012          114.  
3 URY    Uruguay    2013          123.  
4 URY    Uruguay    2014          134.  
5 URY    Uruguay    2015          146.  
6 URY    Uruguay    2016          160.  
7 URY    Uruguay    2017          170.  
8 URY    Uruguay    2018          183.  
9 URY    Uruguay    2019          197.
```

Another country's CPI

```
imf_data %>%  
  select(iso, country, year,  
         consumer_price_index) %>%  
  filter(country == "Belize",  
         year > 2010)
```

```
# A tibble: 9 × 4  
  iso    country    year consumer_price_index  
  <chr> <chr>    <int>         <dbl>  
1 BLZ    Belize    2011         130.  
2 BLZ    Belize    2012         132.  
3 BLZ    Belize    2013         133.  
4 BLZ    Belize    2014         134.  
5 BLZ    Belize    2015         133.  
6 BLZ    Belize    2016         134.  
7 BLZ    Belize    2017         136.  
8 BLZ    Belize    2018         136.  
9 BLZ    Belize    2019         136.
```

Creating a function instead

```
cpi_by_country <- function(country_name) {  
  imf_data %>%  
    select(iso, country, year,  
           consumer_price_index) %>%  
    filter(country == country_name,  
           year > 2010)  
}
```

Use the function

```
cpi_by_country <- function(country_name) {  
  imf_data %>%  
    select(iso, country, year,  
           consumer_price_index) %>%  
    filter(country == country_name,  
           year > 2010)  
}
```

```
cpi_by_country(country_name = "Samoa")
```

```
# A tibble: 9 × 4  
  iso    country    year consumer_price_index  
  <chr> <chr>    <int>         <dbl>  
1 WSM    Samoa    2011         102.  
2 WSM    Samoa    2012         109.  
3 WSM    Samoa    2013         108.  
4 WSM    Samoa    2014         107.  
5 WSM    Samoa    2015         109.  
6 WSM    Samoa    2016         109.  
7 WSM    Samoa    2017         111.  
8 WSM    Samoa    2018         115.  
9 WSM    Samoa    2019         117.
```

Joining IMF and World Bank data

```
joined <- imf_data %>%  
  inner_join(world_bank_data,  
            by = c("iso", "year")) %>%  
  relocate(continent, region, .after = year)
```

Results of the join

joined

```
# A tibble: 299 × 22
  iso   country.x  year continent region          gdp_in_billions...
<chr> <chr>      <dbl> <fct>     <fct>          <dbl>
1 ALB   Albania     2011 Europe   Southern Europe  12.9
2 ALB   Albania     2012 Europe   Southern Europe  12.3
3 AGO   Angola      2014 Africa   Middle Africa    146.
# ... with 296 more rows, and 16 more variables:
#   usd_conversion_rate <dbl>, total_investment_as_perc_gdp <dbl>,
#   consumer_price_index <dbl>, imports_perc_change <dbl>,
#   exports_perc_change <dbl>, population_in_millions <dbl>,
#   gov_revenue_as_perc_gdp <dbl>, gov_net_debt_as_perc_gdp <dbl>,
#   country.y <chr>, infant_mortality_rate <dbl>, fertility_rate <dbl>,
#   perc_electric_access <dbl>, perc_college_complete <dbl>, ...
```

Mean Government Revenue as GDP % by Continent

```
joined %>%  
  group_by(continent) %>%  
  summarize(mean_gov_revenue = mean(  
    gov_revenue_as_perc_gdp)  
  )
```

```
# A tibble: 5 × 2  
  continent mean_gov_revenue  
  <fct>      <dbl>  
1 Africa      18.2  
2 Americas   25.6  
3 Asia       26.7  
4 Europe     41.5  
5 Oceania    35.6
```

A function for grouping?

```
grouped_mean_gov_revenue <- function(group_col) {  
  joined %>%  
    group_by(group_col) %>%  
    summarize(mean_gov_revenue = mean(gov_revenue_as_perc_gdp))  
}  
grouped_mean_gov_revenue(group_col = year)
```

```
Error: Must group by variables found in `.data`.  
* Column `group_col` is not found.
```

The curly-curly `{{ }}` operator

```
library(rlang)
grouped_mean_gov_revenue <- function(group_col) {
  joined %>%
    group_by({{ group_col }}) %>%
    summarize(mean_gov_revenue = mean(gov_revenue_as_perc_gdp))
}
grouped_mean_gov_revenue(group_col = year)
```

```
# A tibble: 17 × 2
  year mean_gov_revenue
  <dbl>         <dbl>
1  2000             39.4
2  2001             30.8
...
16 2015             34.7
17 2016             32.3
```

Let's practice!

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Bang-bang!!

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Unpacking curly-curly

```
grouped_mean_gov_revenue <- function(group_col) {  
  joined %>%  
    group_by({{ group_col }}) %>%  
    summarize(mean_gov_revenue = mean(gov_revenue_as_perc_gdp))  
}  
grouped_mean_gov_revenue(group_col = continent)
```

`{{ }}`

- Forces function argument
- Defuses function argument

Bang-bang-enquo forces and defuses

`!!enquo()` is the same as `{{ }}`

```
grouped_mean_gov_revenue <- function(group_col) {  
  joined %>%  
    group_by(!!enquo(group_col)) %>%  
    summarize(  
      mean_gov_revenue = mean(  
        gov_revenue_as_perc_gdp  
      )  
    )  
  }  
}  
grouped_mean_gov_revenue(group_col = continent)
```

```
# A tibble: 17 × 2  
  year mean_gov_revenue  
  <dbl> <dbl>  
1  2000 39.4  
2  2001 30.8  
...  
16 2015 34.7  
17 2016 32.3
```

Multiple function arguments

```
grouped_mean_gov_revenue <- function(group_col) {  
  joined %>%  
    group_by(!!enquo(group_col)) %>%  
    summarize(mean_gov_revenue = mean(gov_revenue_as_perc_gdp))  
}
```

```
grouped_mean_for_column <- function(group_col, col_to_mean) {  
  joined %>%  
    group_by(!!enquo(group_col)) %>%  
    summarize(mean(!!enquo(col_to_mean)))  
}
```

Calling `grouped_mean_for_column()`

```
grouped_mean_for_column(group_col = continent,  
                        col_to_mean = perc_cvd_crd_70)
```

```
# A tibble: 5 × 2  
  continent `mean(perc_cvd_crd_70)`  
  <fct>          <dbl>  
1 Africa          23.8  
2 Americas        14.7  
3 Asia            19.8  
4 Europe          16.3  
5 Oceania          9.91
```

Let's practice!

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Rlang-ing in your rocking chair

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Strange summarize column name

```
grouped_mean_by_column <- function(group_col, col_to_mean) {  
  joined %>%  
    group_by(!!enquo(group_col)) %>%  
    summarize(mean(!!enquo(col_to_mean)))  
}  
grouped_mean_by_column(group_col = year, col_to_mean = perc_cvd_crd_70)
```

```
# A tibble: 17 × 2  
  year `mean(perc_cvd_crd_70)`  
  <dbl> <dbl>  
1  2000 15.8  
2  2001 23.3  
...  
17 2016 15.8
```

What we'd like it to be

```
# A tibble: 17 × 2
  year mean_of_perc_cvd_crd_70
  <dbl> <dbl>
1 2000 15.8
2 2001 23.3
3 2002 15
4 2003 14.8
5 2004 21.1
...
15 2014 16.7
16 2015 16.0
17 2016 15.8
```

as_name() and the walrus operator :=

`as_name()`

- Converts unquoted, defused column name to be a string

`:=`

- Allows for a variable in R to be on the left-hand side in `mutate()`

Wrapping up the function

```
grouped_mean_by_column <- function(.data, group_col, col_to_mean) {  
  name_of_col_to_mean <- as_name(enquo(col_to_mean))  
  new_col_name <- paste0("mean_of_", name_of_col_to_mean)  
  .data %>%  
    group_by(!!enquo(group_col)) %>%  
    summarize(!!new_col_name := mean(!!enquo(col_to_mean)))  
}
```

Applying the wrapped up function

```
grouped_mean_by_column(.data = joined,  
                        group_col = continent,  
                        col_to_mean = perc_rural_pop)
```

```
# A tibble: 5 × 2  
  continent mean_of_perc_rural_pop  
  <fct>          <dbl>  
1 Africa          65.1  
2 Americas        28.1  
3 Asia            31.9  
4 Europe          30.1  
5 Oceania         14.1
```

Cleaning things up a bit

```
grouped_mean_by_column <- function(.data, group_col, col_to_mean) {  
  name_of_col_to_mean <- as_name(enquo(col_to_mean))  
  new_col_name <- paste0("mean_of_", name_of_col_to_mean)  
  .data %>%  
    group_by({{ group_col }}) %>%  
    summarize(!!new_col_name := mean({{ col_to_mean }}))  
}  
joined %>%  
  grouped_mean_by_column(group_col = continent,  
                          col_to_mean = perc_rural_pop)
```

Let's practice!

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A great ggplot twist

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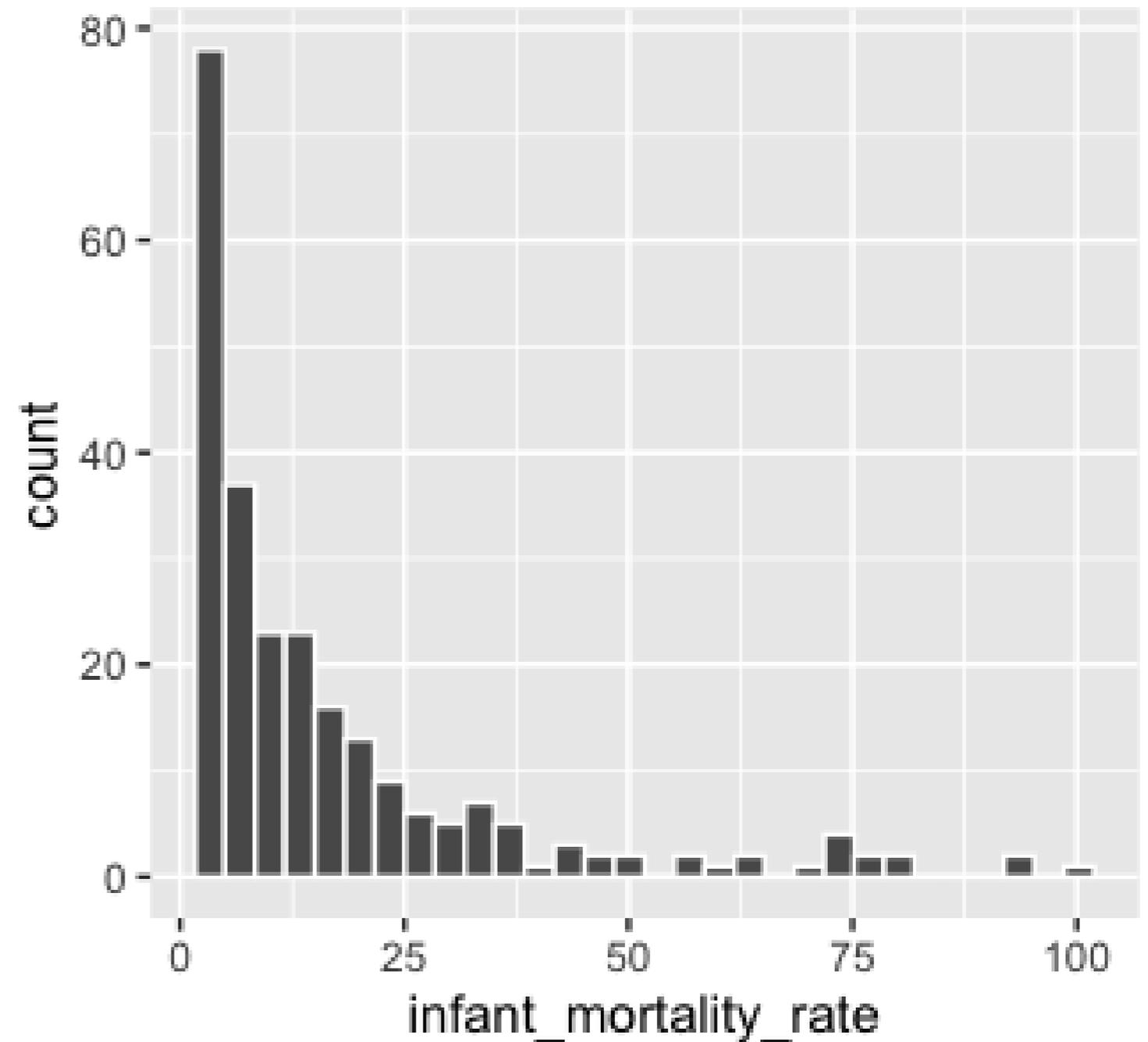


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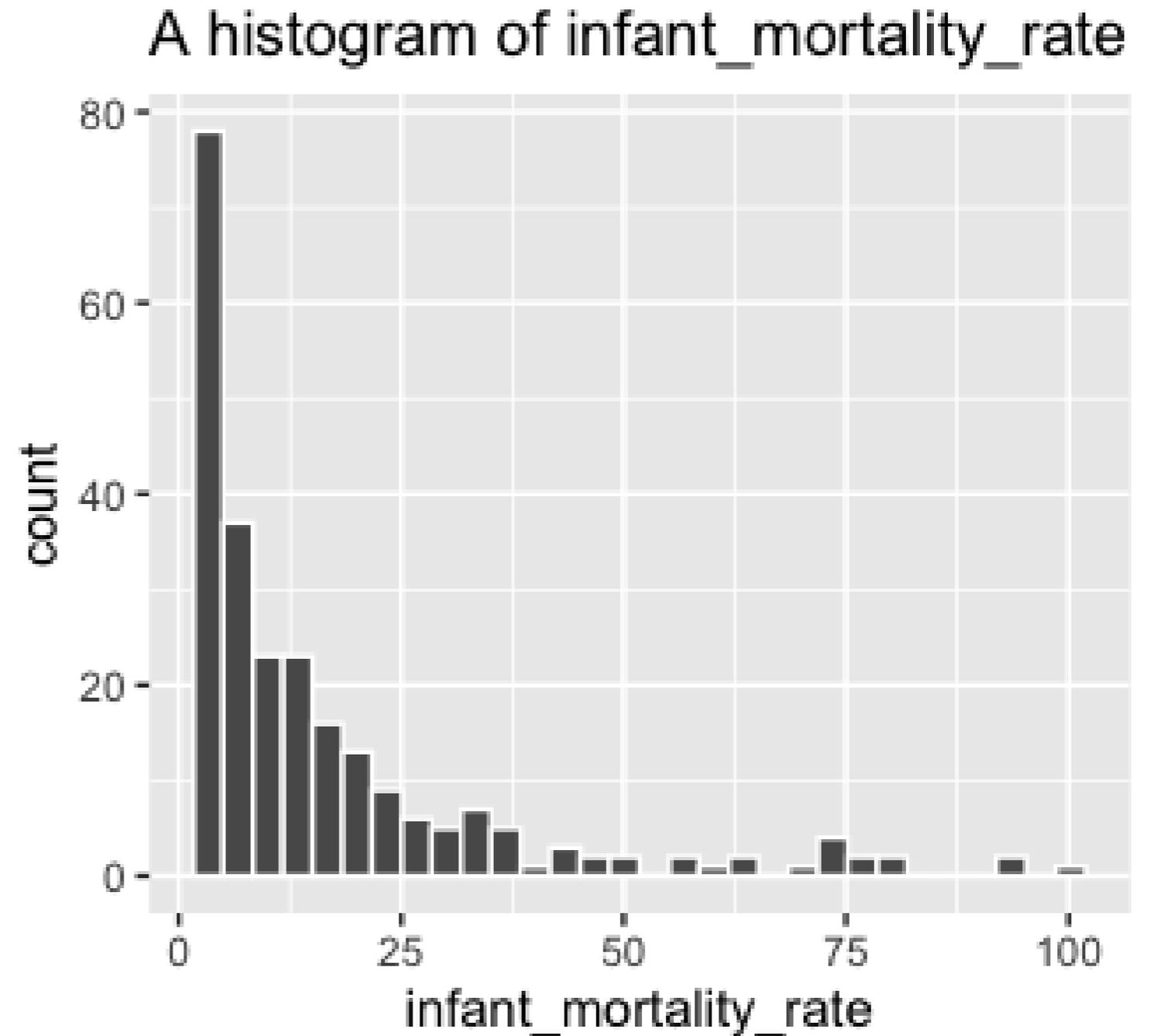
ggplot2 basics

```
library(ggplot2)
ggplot(
  world_bank_data,
  aes(x = infant_mortality_rate)
) +
  geom_histogram(color = "white")
```



Adding a title

```
ggplot(  
  world_bank_data,  
  aes(x = infant_mortality_rate)  
) +  
  geom_histogram(color = "white") +  
  ggtitle("A histogram of infant_mortality_rate")
```



Wrapping into a function

```
# Previous plot code
ggplot(world_bank_data, aes(x = infant_mortality_rate)) +
  geom_histogram(color = "white") +
  ggtitle("A histogram of infant_mortality_rate")
```

```
# Define a function
my_histogram <- function(
  df, x_var) {
  ggplot(df, aes(x = x_var)) +
    geom_histogram(color = "white") +
    ggtitle(paste("A histogram of", x_var))
}
```

Working on our function

```
# First attempt at a function
my_histogram <- function(df, x_var) {
  ggplot(df, aes(x = x_var)) +
    geom_histogram(color = "white") +
    ggtitle(paste("A histogram of", x_var))
}

# Call the function
my_histogram(df = world_bank_data,
             x_var = infant_mortality_rate)
```

```
Error in paste("A histogram of", x_var) :
  object 'infant_mortality_rate' not found
```

Adding in rlang

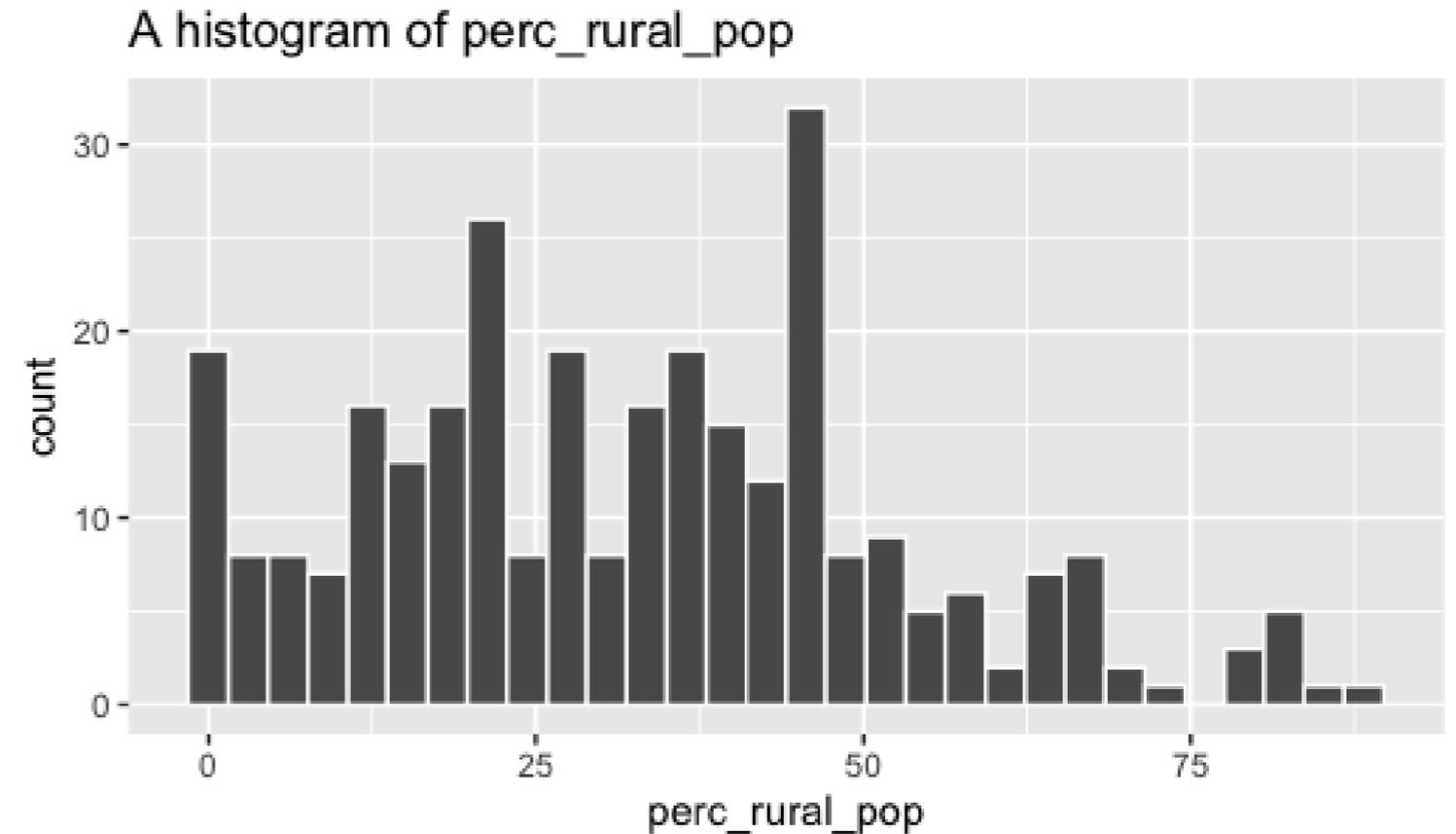
```
# First attempt at a function
my_histogram <- function(df, x_var) {
  ggplot(df, aes(x = x_var)) +
    geom_histogram(color = "white") +
    ggtitle(paste("A histogram of", x_var))
}
```

```
# Making needed tweaks to stop error
my_histogram <- function(df, x_var) {
  ggplot(df, aes(x = {{ x_var }})) +
    geom_histogram(color = "white") +
    ggtitle(paste("A histogram of",
                  as_label(enquo(x_var))))
}
```

Using our function

```
# Defined function with rlang operators
my_histogram <- function(df, x_var) {
  ggplot(df, aes(x = {{ x_var }})) +
    geom_histogram(color = "white") +
    ggtitle(paste(
      "A histogram of",
      as_label(enquo(x_var))
    ))
}

# Call the function with the pipe!
world_bank_data %>%
  my_histogram(x_var = perc_rural_pop)
```



Let's practice!

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Congratulations!

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Chapter 1 - `select()` and its helper functions

- Pipelines with the main `dplyr` verbs of `select()`, `filter()`, `summarize()`, `group_by()`, and `mutate()`
- `starts_with()`, `ends_with()`, and `contains()` helper functions
- `matches()` to look for regular expression patterns

Chapter 2 - Column transformations

- Move columns around with `select()`, `everything()`, and `relocate()`
- Use `across()` to do the same transformation on many columns
- Choose rows that match certain column criteria with `if_any()` and `if_all()`
- Calculate across each row with `rowwise()` and `c_across()`

Chapter 3 - Joins and set theory

- `dplyr` joins
 - Review of `inner_join()`, `left_join()`, and `anti_join()`
- Set theory clauses
 - `intersect()`
 - `union()` and `union_all()`
 - `setequal()` and `setdiff()`

Chapter 4 - Getting your feet wet with rlang

- Writing functions using curly-curly `{{ }}`
- Using the forcing "bang-bang" operator `!!` and the defusing operator `enquo()`
- Creating variable names using the walrus operator `:=`, `as_name()`, and `!!`
- Wrapping `ggplot2` plotting code in a function with `as_label()`

Other relevant courses/tracks

Courses

- Visualization Best Practices in R
- Introduction to Data Visualization with ggplot2
- Intermediate Data Visualization with ggplot2

Tracks

- Tidyverse Fundamentals with R
- Intermediate Tidyverse Toolbox
- R Programmer

Woohoo for you!

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