Week 7 challenge

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2023-10-04

Question 1: Palmer Penguins

Solutions:

```
# Enter code here
library(tidyverse)
```

```
## — Attaching core tidyverse packages -
                                                            – tidyverse 2.0.0 —
## √ dplyr 1.1.2
                        √ readr
                                    2.1.4
## √ forcats
              1.0.0

√ stringr

                                    1.5.0
## √ ggplot2 3.4.3
                        √ tibble
                                    3.2.1
## ✓ lubridate 1.9.2
                        √ tidyr
                                    1.3.0
## √ purrr
              1.0.2
## — Conflicts —
                                                    —— tidyverse_conflicts() —
## X dplyr::filter() masks stats::filter()
## X dplyr::lag()
                 masks stats::lag()
### i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to be
come errors
```

```
library(palmerpenguins)
glimpse(penguins)
```

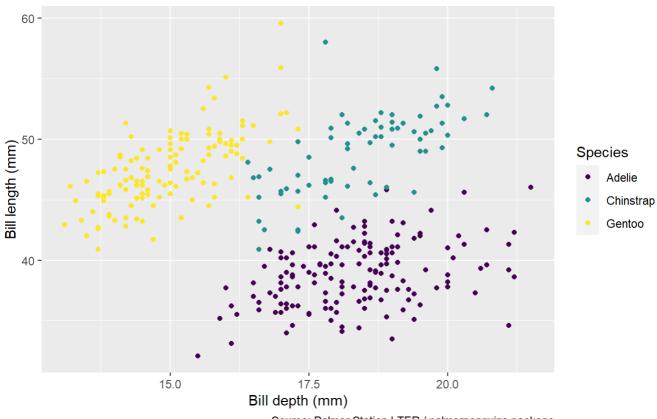
```
## Rows: 344
## Columns: 8
## $ species
                       <fct> Adelie, Adelie, Adelie, Adelie, Adelie, Adelie, Adeli...
## $ island
                       <fct> Torgersen, Torgersen, Torgersen, Torgerse...
                       <dbl> 39.1, 39.5, 40.3, NA, 36.7, 39.3, 38.9, 39.2, 34.1, ...
## $ bill length mm
## $ bill depth mm
                       <dbl> 18.7, 17.4, 18.0, NA, 19.3, 20.6, 17.8, 19.6, 18.1, ...
## $ flipper_length_mm <int> 181, 186, 195, NA, 193, 190, 181, 195, 193, 190, 186...
## $ body_mass_g
                       <int> 3750, 3800, 3250, NA, 3450, 3650, 3625, 4675, 3475, ...
                       <fct> male, female, female, NA, female, male, female, male...
## $ sex
## $ year
                       <int> 2007, 2007, 2007, 2007, 2007, 2007, 2007, 2007, 2007...
```

Question 1.2:Plot Palmer Penguins

Warning: Removed 2 rows containing missing values (`geom_point()`).

Bill depth and length

Dimensions for Adelie, Chinstrap, and Gentoo Penguins

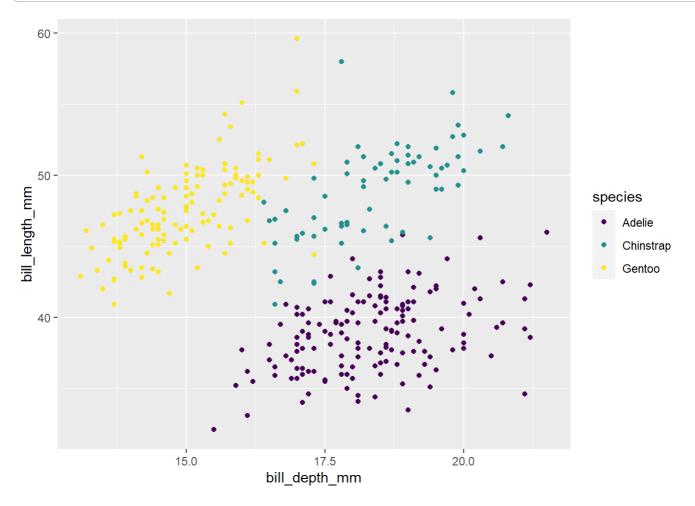


Source: Palmer Station LTER / palmerpenguins package

####Question 1.3:Palmer Penguins: colour

```
# Enter code here
ggplot(penguins) + aes(x = bill_depth_mm, y = bill_length_mm,
colour = species) + # specifying x and y
geom_point() + #represent each observation with a point
    scale_colour_viridis_d() #discrete colour scale that is designed to be perceived by viewers
with common forms of colour blindness
```

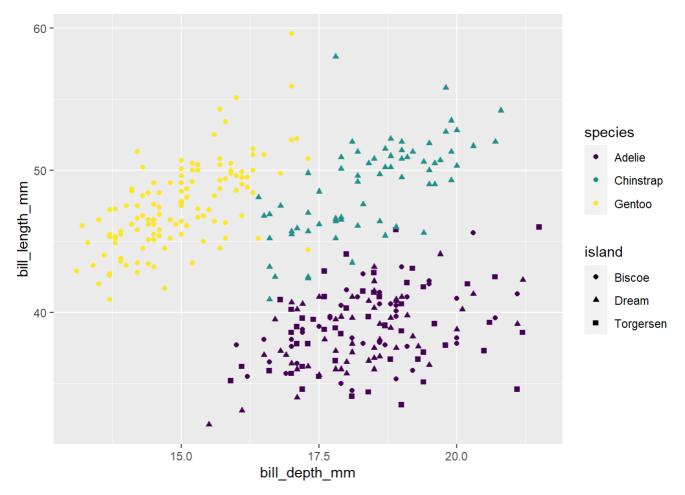
Warning: Removed 2 rows containing missing values (`geom_point()`).



####Question 1.4:Palmer Penguins: shape

Solutions:

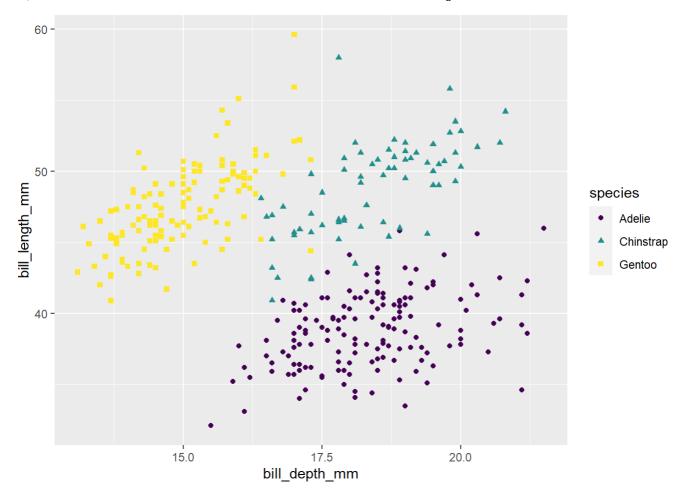
```
# Enter code here
ggplot(penguins, aes(x = bill_depth_mm, y = bill_length_mm, colour = species, #specifying x an
d y
shape = island)) + #shaoe of points of islands
geom_point() + #represent each observation with a point
scale_colour_viridis_d() #discrete colour scale that is designed to be perceived by viewers
with common forms of colour blindness
```



####Question 1.5:Palmer Penguins: shape

Solutions:

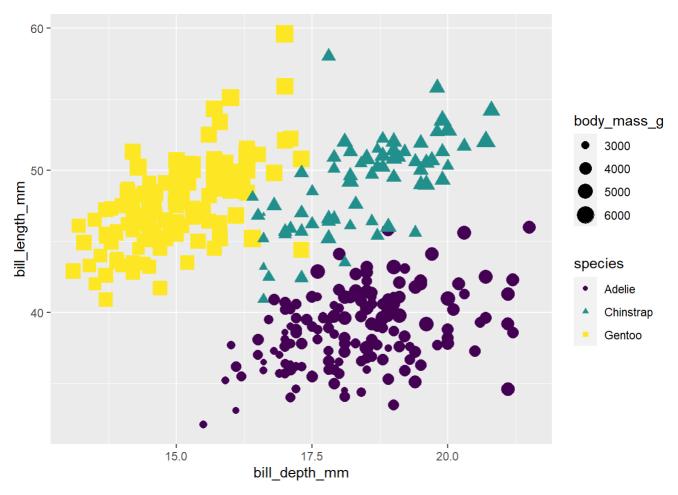
```
# Enter code here
ggplot(penguins, aes(x = bill_depth_mm, y = bill_length_mm, colour = species, #specifying x an
d y
shape = species)) + #shape of points of species
geom_point() + #represent each observation with a point
scale_colour_viridis_d() #discrete colour scale that is designed to be perceived by viewers
with common forms of colour blindness
```



####Question 1.6:Palmer Penguins: size

Solutions:

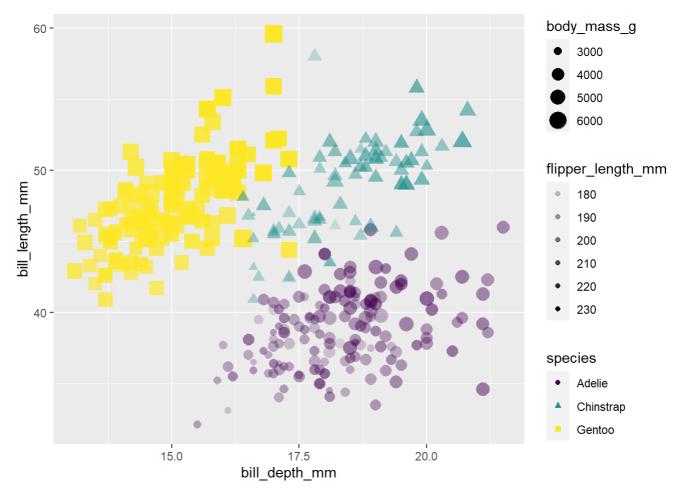
```
# Enter code here
ggplot(penguins, aes(x = bill_depth_mm, y = bill_length_mm, colour = species, shape = specie
s,
    size = body_mass_g)) + #size of points by body mass
geom_point() + #represent each observation with a point
    scale_colour_viridis_d() #discrete colour scale that is designed to be perceived by viewers
with common forms of colour blindness
```



####Question 1.7:Palmer Penguins: Alpha

Solutions:

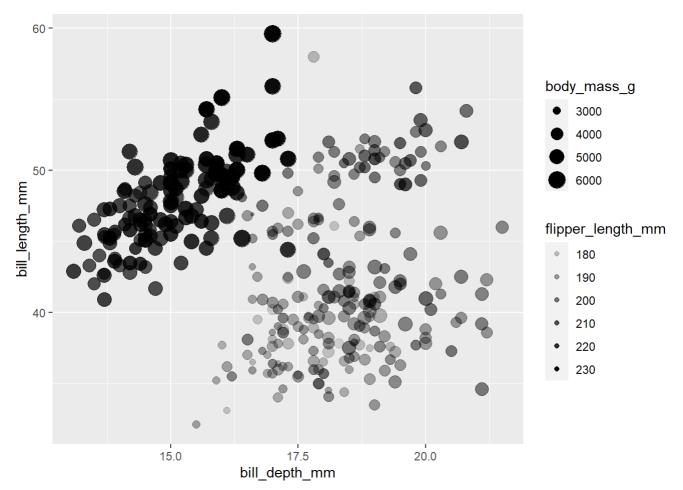
```
# Enter code here
ggplot(penguins, aes(x = bill_depth_mm, y = bill_length_mm, colour = species,
    shape = species, size = body_mass_g, alpha = flipper_length_mm)) + #alpha of points by flipp
erlength
    geom_point() + #represent each observation with a point
    scale_colour_viridis_d()#discrete colour scale that is designed to be perceived by viewers
with common forms of colour blindness
```



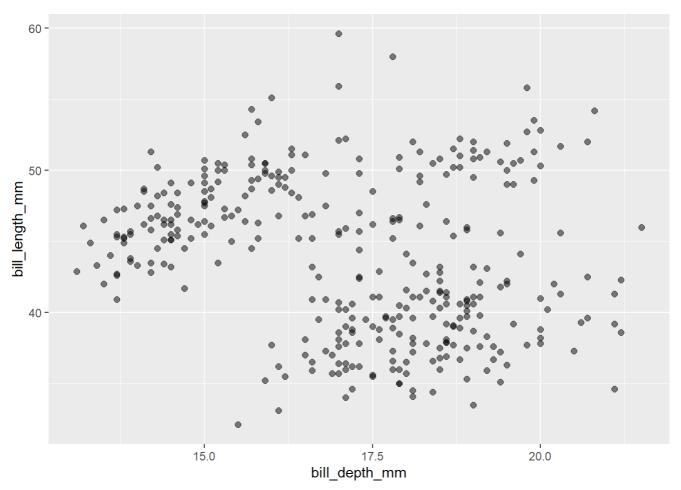
####Question 1.8:Palmer Penguins: Mapping vs Setting

```
# Enter code here
ggplot(penguins) +
aes(x = bill_depth_mm,
y = bill_length_mm,
size = body_mass_g,
alpha = flipper_length_mm) +
geom_point()
```

```
## Warning: Removed 2 rows containing missing values (`geom_point()`).
```



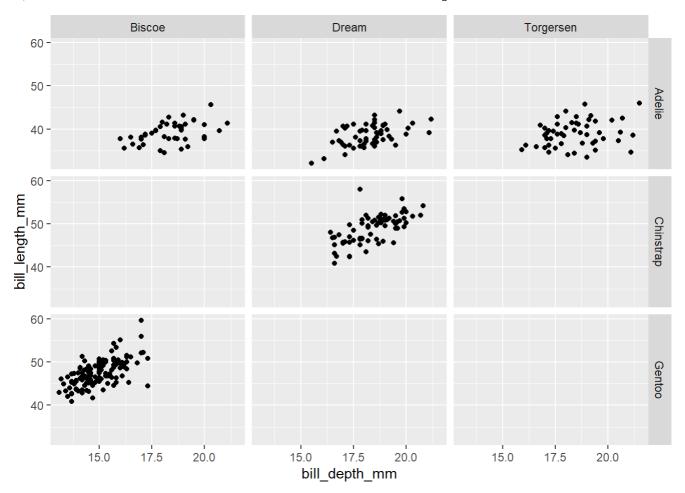
```
ggplot(penguins) +
aes(x = bill_depth_mm,
y = bill_length_mm) +
geom_point(size = 2, alpha = 0.5)
```



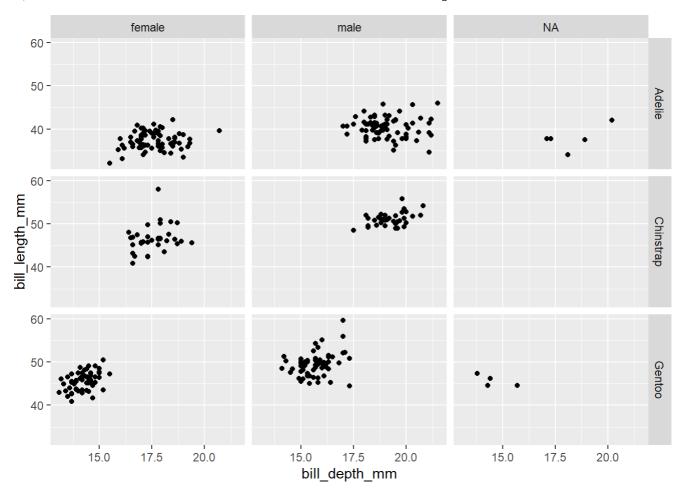
####Question 1.9:Palmer Penguins: Faceting

```
# Enter code here
ggplot(penguins) +
aes(x = bill_depth_mm,
y = bill_length_mm) +
geom_point() + #represent each observation with a point
facet_grid(species ~ island) #create smaller plots that display different subsets of the dat
a
```

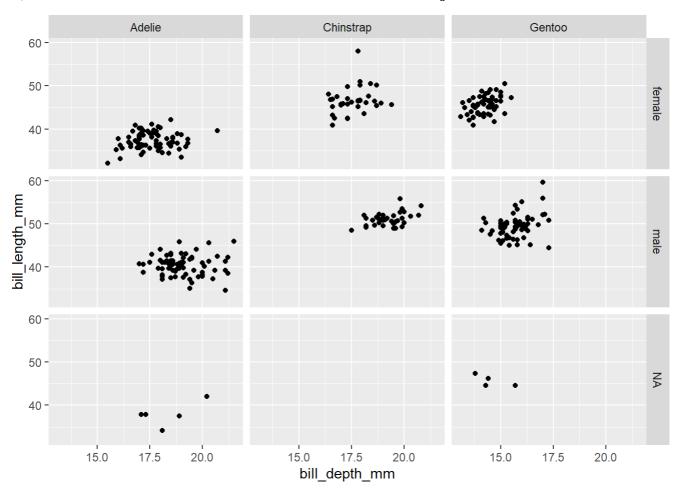
```
## Warning: Removed 2 rows containing missing values (`geom_point()`).
```



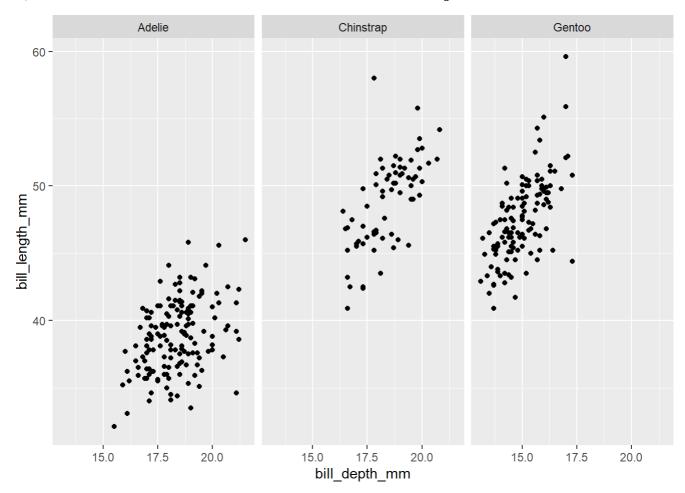
Enter code here
ggplot(penguins, aes(x = bill_depth_mm, y = bill_length_mm)) + geom_point() + #represent each
observation with a point
facet_grid(species ~ sex)#create smaller plots that display different subsets of the data



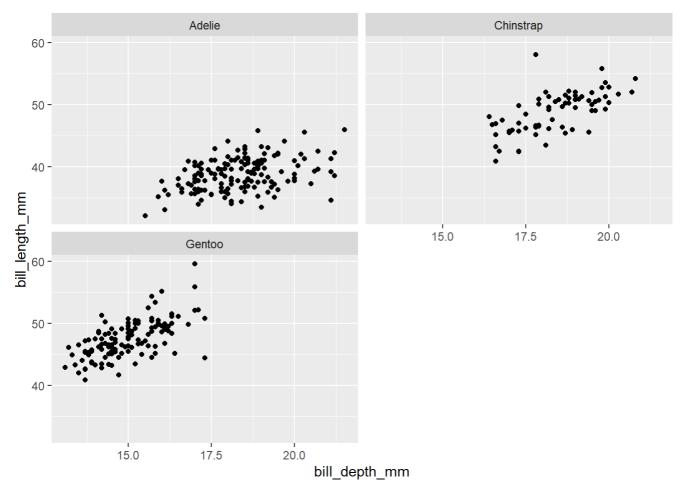
Enter code here
ggplot(penguins, aes(x = bill_depth_mm, y = bill_length_mm)) + geom_point() + #represent each
observation with a point
facet_grid(sex ~ species)#create smaller plots that display different subsets of the data bas
e on sex and species



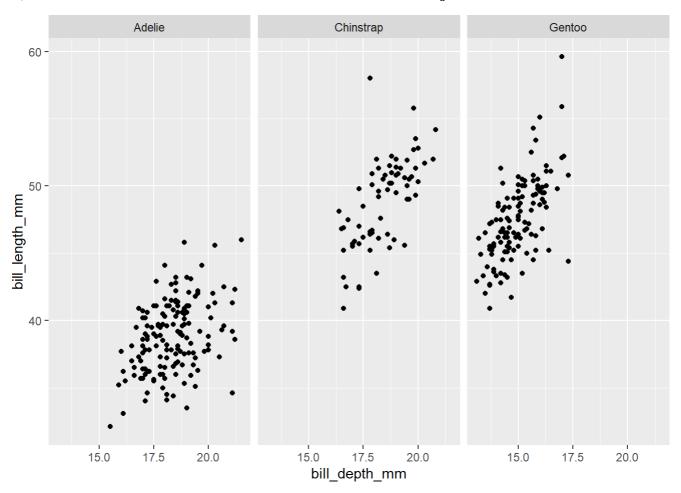
Enter code here
ggplot(penguins, aes(x = bill_depth_mm, y = bill_length_mm)) + geom_point() + #represent each
observation with a point
facet_wrap(~ species)#create smaller plots that display different subsets of the data base on
species not sex



Enter code here
ggplot(penguins, aes(x = bill_depth_mm, y = bill_length_mm)) + geom_point() + #represent each
observation with a point
facet_wrap(~ species, ncol = 2)#create smaller plots that display different subsets of the da
ta base on species, arranged in 2 columns



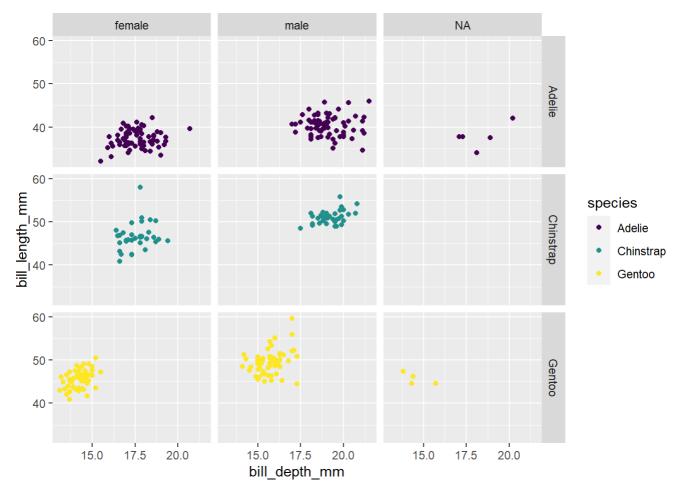
Enter code here
ggplot(penguins, aes(x = bill_depth_mm, y = bill_length_mm)) + geom_point() + #represent each
observation with a point
facet_wrap(. ~ species)#create smaller plots that display different subsets of the data base
on species



Enter code here
ggplot(penguins, aes(x = bill_depth_mm, y = bill_length_mm, color = species)) +
geom_point() + #represent each observation with a point

 $facet_grid(species \sim sex) + \#create smaller plots that display different subsets of the dat a base on sex and species$

scale_color_viridis_d()#discrete colour scale that is designed to be perceived by viewers
with common forms of colour blindness

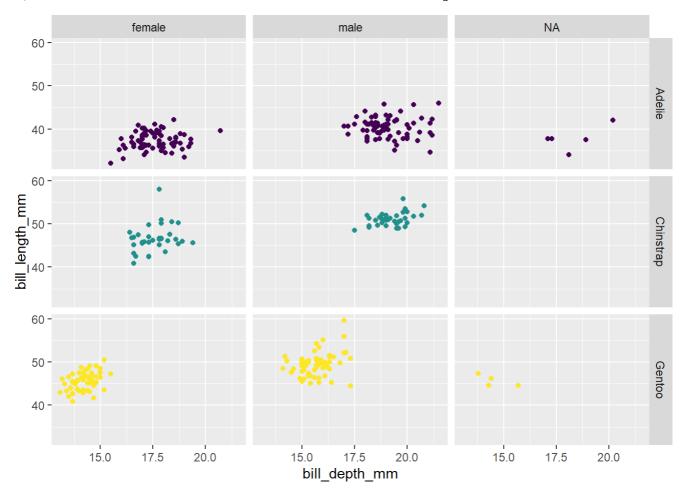


Enter code here
ggplot(penguins, aes(x = bill_depth_mm, y = bill_length_mm, color = species)) +
geom_point() + #represent each observation with a point

 $facet_grid(species \sim sex) + \#create smaller plots that display different subsets of the dat a base on sex and species$

 $scale_color_viridis_d() + \#discrete\ colour\ scale\ that\ is\ designed\ to\ be\ perceived\ by\ view\ ers\ with\ common\ forms\ of\ colour\ blindness$

guides(color = "none") # Remove the color legend.



####Question 2: Lending set

Solutions:

```
# Enter code here
library(openintro)

## Loading required package: airports
```

Loading required package: cherryblossom

Loading required package: usdata

glimpse(loans_full_schema)#peek at data

Rows: 10,000 ## Columns: 55 ## \$ emp_title <chr> "global config engineer ", "warehouse... ## \$ emp_length <dbl> 3, 10, 3, 1, 10, NA, 10, 10, 10, 3, 1... ## \$ state <fct> NJ, HI, WI, PA, CA, KY, MI, AZ, NV, I... ## \$ homeownership <fct> MORTGAGE, RENT, RENT, RENT, RENT, OWN... <dbl> 90000, 40000, 40000, 30000, 35000, 34... ## \$ annual_income ## \$ verified income <fct> Verified, Not Verified, Source Verifi... ## \$ debt_to_income <dbl> 18.01, 5.04, 21.15, 10.16, 57.96, 6.4... ## \$ annual_income_joint <dbl> NA, NA, NA, NA, 57000, NA, 155000, NA... ## \$ verification_income_joint <fct> , , , Verified, , Not Verified, , ,... ## \$ debt_to_income_joint <dbl> NA, NA, NA, NA, 37.66, NA, 13.12, NA,... ## \$ delinq_2y <int> 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 0... <int> 38, NA, 28, NA, NA, 3, NA, 19, 18, NA... ## \$ months_since_last_deling <dbl> 2001, 1996, 2006, 2007, 2008, 1990, 2... ## \$ earliest_credit_line <int> 6, 1, 4, 0, 7, 6, 1, 1, 3, 0, 4, 4, 8... ## \$ inquiries_last_12m ## \$ total_credit_lines <int> 28, 30, 31, 4, 22, 32, 12, 30, 35, 9,... <int> 10, 14, 10, 4, 16, 12, 10, 15, 21, 6,... ## \$ open_credit_lines <int> 70795, 28800, 24193, 25400, 69839, 42... ## \$ total_credit_limit ## \$ total_credit_utilized <int> 38767, 4321, 16000, 4997, 52722, 3898... <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0. ## \$ num_collections_last_12m ## \$ num_historical_failed_to_pay <int> 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0... ## \$ months_since_90d_late <int> 38, NA, 28, NA, NA, 60, NA, 71, 18, N... ## \$ current_accounts_deling <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0... ## \$ total_collection_amount_ever <int> 1250, 0, 432, 0, 0, 0, 0, 0, 0, 0, 0, ... <int> 2, 0, 1, 1, 1, 0, 2, 2, 6, 1, 2, 1, 2... ## \$ current_installment_accounts ## \$ accounts_opened_24m <int> 5, 11, 13, 1, 6, 2, 1, 4, 10, 5, 6, 7... ## \$ months_since_last_credit_inquiry <int> 5, 8, 7, 15, 4, 5, 9, 7, 4, 17, 3, 4,... ## \$ num_satisfactory_accounts <int> 10, 14, 10, 4, 16, 12, 10, 15, 21, 6,... ## \$ num_accounts_120d_past_due <int> 0, 0, 0, 0, 0, 0, NA, 0, 0, 0, ... ## \$ num_accounts_30d_past_due <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0... ## \$ num_active_debit_accounts <int> 2, 3, 3, 2, 10, 1, 3, 5, 11, 3, 2, 2,... ## \$ total debit limit <int> 11100, 16500, 4300, 19400, 32700, 272... ## \$ num_total_cc_accounts <int> 14, 24, 14, 3, 20, 27, 8, 16, 19, 7, ... ## \$ num_open_cc_accounts <int> 8, 14, 8, 3, 15, 12, 7, 12, 14, 5, 8,... ## \$ num_cc_carrying_balance <int> 6, 4, 6, 2, 13, 5, 6, 10, 14, 3, 5, 3... ## \$ num_mort_accounts <int> 1, 0, 0, 0, 0, 3, 2, 7, 2, 0, 2, 3, 3... ## \$ account_never_delinq_percent <dbl> 92.9, 100.0, 93.5, 100.0, 100.0, 78.1... ## \$ tax liens <int> 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0... ## \$ public record bankrupt <int> 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0... ## \$ loan purpose <fct> moving, debt_consolidation, other, de... ## \$ application_type <fct> individual, individual, individual, i... <int> 28000, 5000, 2000, 21600, 23000, 5000... ## \$ loan_amount ## \$ term <dbl> 60, 36, 36, 36, 36, 60, 60, 36, 3... <dbl> 14.07, 12.61, 17.09, 6.72, 14.07, 6.7... ## \$ interest_rate ## \$ installment <dbl> 652.53, 167.54, 71.40, 664.19, 786.87... <fct> C, C, D, A, C, A, C, B, C, A, C, B, C... ## \$ grade ## \$ sub_grade <fct> C3, C1, D1, A3, C3, A3, C2, B5, C2, A... ## \$ issue month <fct> Mar-2018, Feb-2018, Feb-2018, Jan-201... ## \$ loan_status <fct> Current, Current, Current, C... <fct> whole, whole, fractional, whole, whol... ## \$ initial listing status ## \$ disbursement method <fct> Cash, Cash, Cash, Cash, Cash, Cash, C... <dbl> 27015.86, 4651.37, 1824.63, 18853.26,... ## \$ balance ## \$ paid_total <dbl> 1999.330, 499.120, 281.800, 3312.890,... <dbl> 984.14, 348.63, 175.37, 2746.74, 1569... ## \$ paid_principal

####Question 2.1: Lending set: Selected variables

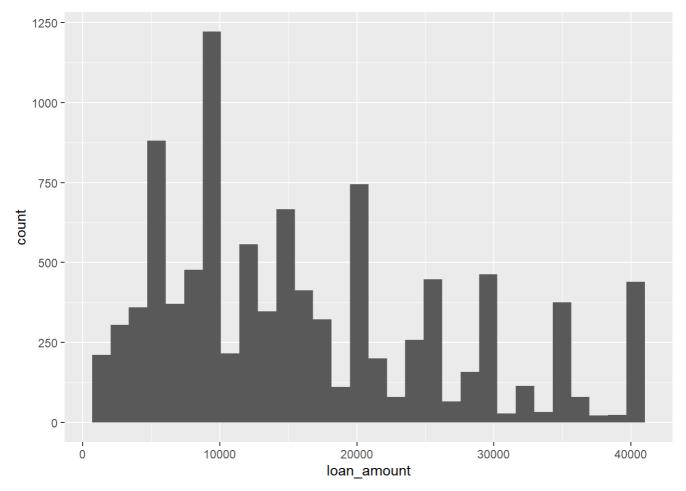
Solutions:

```
# Enter code here
loans <- loans_full_schema %>%
select(loan_amount, interest_rate, term, grade,
state, annual_income, homeownership, debt_to_income)
glimpse(loans)
```

####Question 2.2: Lending set: Histogram

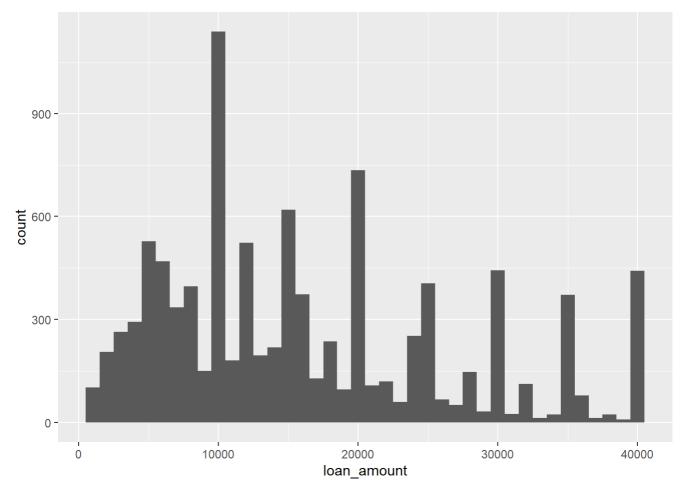
```
# Enter code here
ggplot(loans) + aes(x = loan_amount) +
geom_histogram() # Create a histogram plot of loan amounts from the 'loans' dataset with 'lo
an_amount' on the x-axis.
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

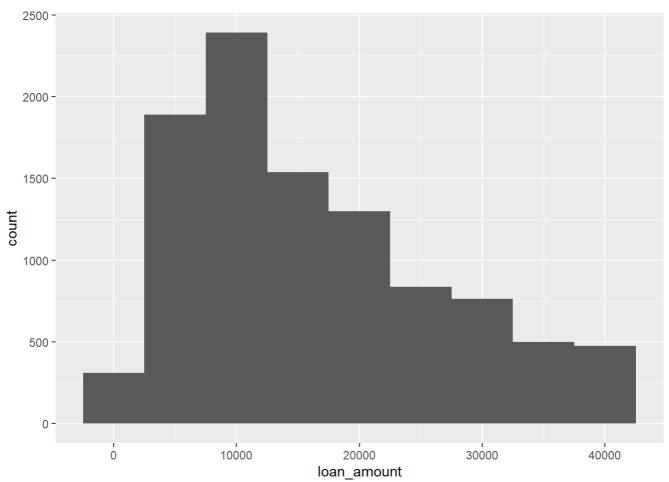


####Question 2.3: Lending set: Histograms and binwidth=100

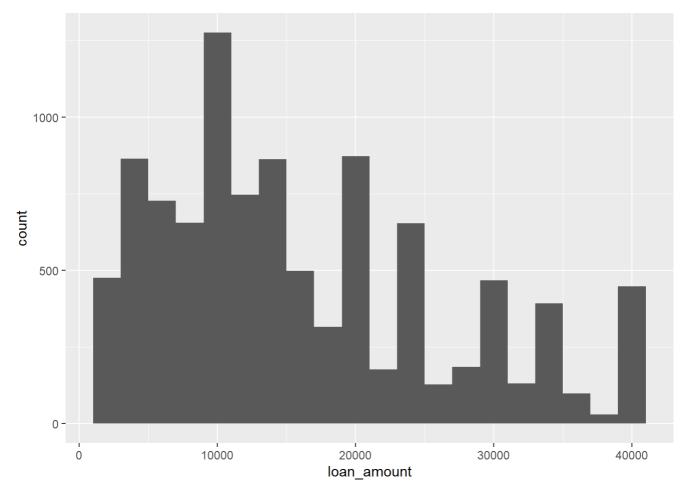
```
# Enter code here
# binwidth = 1000
ggplot(loans, aes(x = loan_amount)) +
  geom_histogram(binwidth = 1000) # Create a histogram plot of loan amounts from the 'loans' d
  ataset with 'loan_amount' on the x-axis and specified binwidth of 1000.
```



```
# Enter code here
# binwidth = 1000
ggplot(loans, aes(x = loan_amount)) +
  geom_histogram(binwidth = 5000) # Create a histogram plot of loan amounts from the 'loans' d
ataset with 'loan_amount' on the x-axis and specified binwidth of 5000.
```



```
# Enter code here
# binwidth = 1000
ggplot(loans, aes(x = loan_amount)) +
  geom_histogram(binwidth = 2000) # Create a histogram plot of loan amounts from the 'loans' d
ataset with 'loan_amount' on the x-axis and specified binwidth of 2000.
```

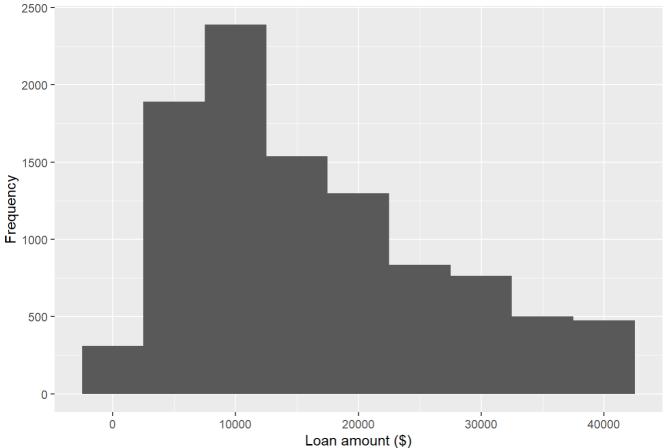


####Question 2.4: Lending set: Customizing histograms

Solutions:

Enter code here
ggplot(loans, aes(x = loan_amount)) + geom_histogram(binwidth = 5000) + # Create a histogram
plot of loan amounts from the 'loans' dataset with 'loan_amount' on the x-axis and a specifie
d binwidth of 5000.
labs(x = "Loan amount (\$)", y = "Frequency", title = "Amounts of Lending Club loans") # Add
labels to the x-axis, y-axis, and title of the plot for clarity.



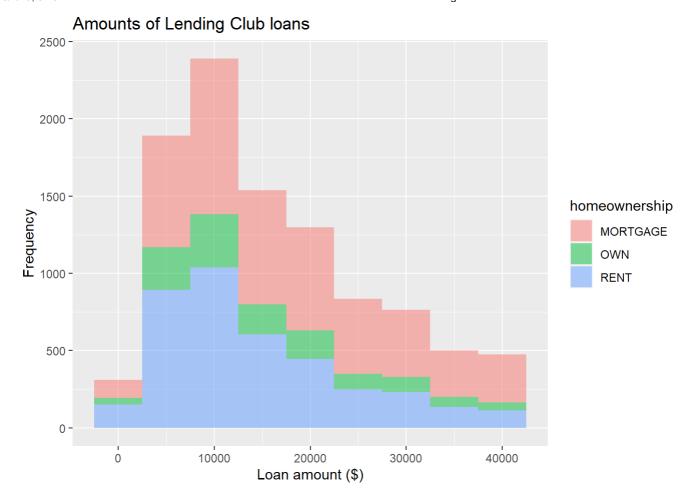


####Question 2.5: Lending set: Fill with a categorical variable

Solutions:

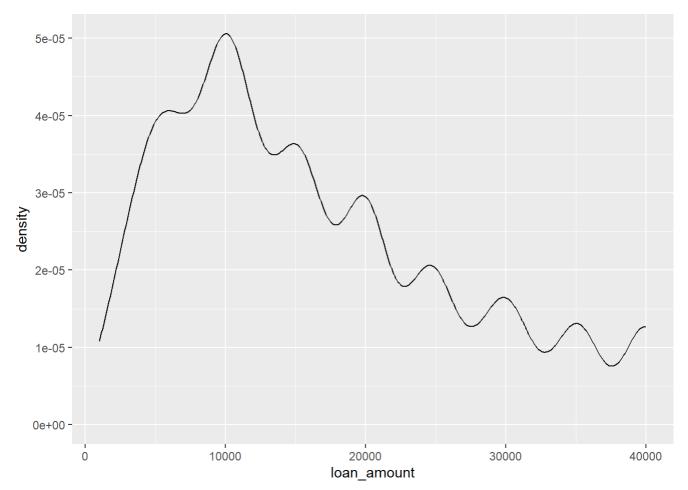
10/4/23, 8:20 PM

```
# Enter code here
ggplot(loans, aes(x = loan_amount, fill = homeownership)) +
 geom_histogram(binwidth = 5000, alpha = 0.5) + #specifies binwidth and setsalpha transparenc
y to 0.5 for the bars
 labs(x = "Loan amount (\$)", y = "Frequency", title = "Amounts of Lending Club loans")
```



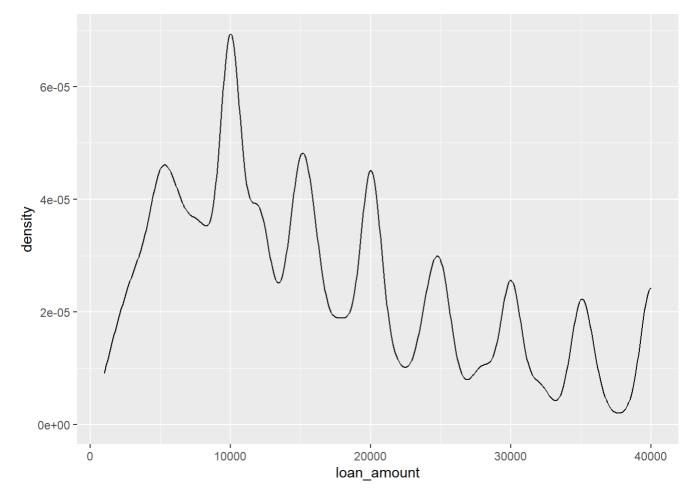
####Question 2.6: Lending set: Density plot

```
# Enter code here
ggplot(loans, aes(x = loan_amount)) +
  geom_density() # Creates a kernel density plot of loan amounts from the 'loans' dataset with
  'loan_amount' on the x-axis.
```

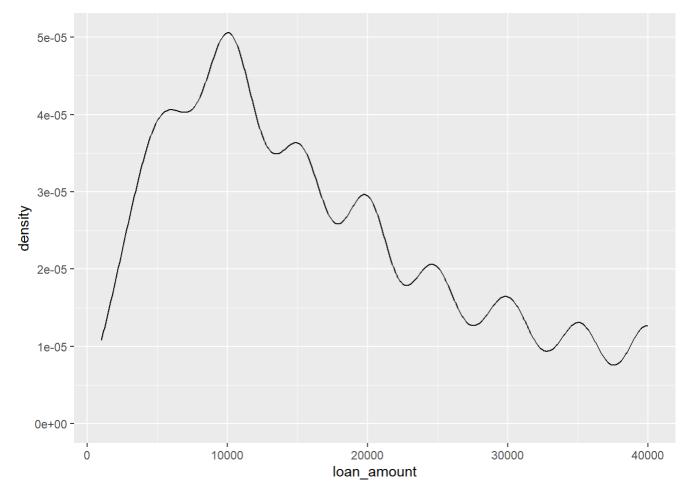


####Question 2.7: Lending set: Density plots and adjusting bandwidth

```
# Enter code here
ggplot(loans, aes(x = loan_amount)) +
geom_density(adjust = 0.5) # Creates a kernel density plot of loan amounts from the 'loans'
dataset with 'loan_amount' on the x-axis, and adjust the smoothing bandwidth to 0.5.
```

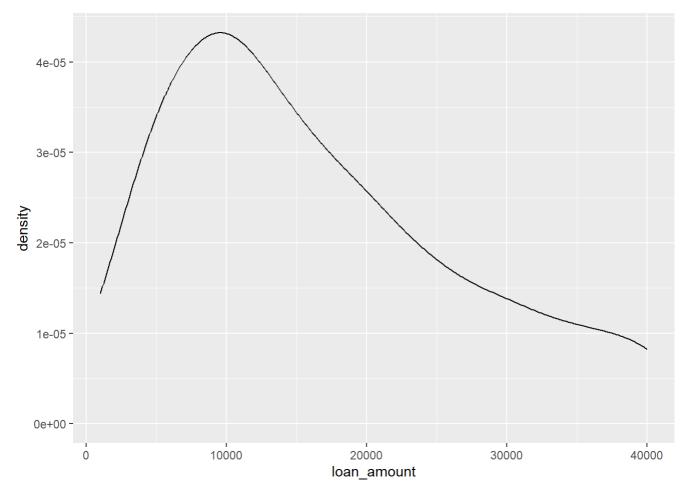


```
# Enter code here
ggplot(loans, aes(x = loan_amount)) +
geom_density(adjust = 1) # default bandwidth
```



Creates a kernel density plot of loan amounts from the 'loans' dataset with 'loan_amount' o n the x-axis, and adjust the smoothing bandwidth to 1

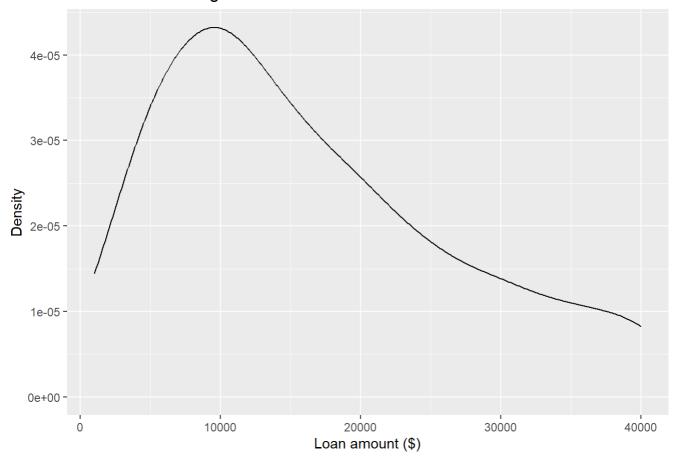
```
# Enter code here
ggplot(loans, aes(x = loan_amount)) +
  geom_density(adjust = 2) # Creates a kernel density plot of loan amounts from the 'loans' da
taset with 'loan_amount' on the x-axis, and adjust the smoothing bandwidth to 2.
```



####Question 2.7: Lending set: Customizing density plots

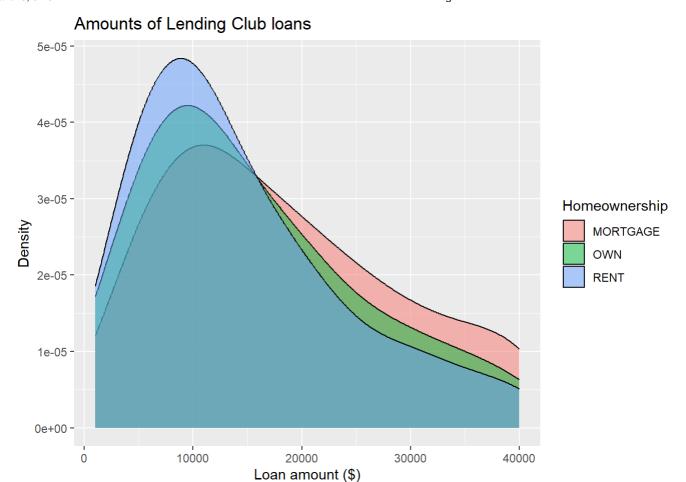
```
# Enter code here
ggplot(loans, aes(x = loan_amount)) +
geom_density(adjust = 2) + # Creates a kernel density plot of loan amounts from the 'loans'
dataset with 'loan_amount' on the x-axis, and adjust the smoothing bandwidth to 2.
labs( x = "Loan amount ($)", y = "Density", title = "Amounts of Lending Club loans" ) # Add
labels to the x-axis, y-axis, and title of the plot for clarity.
```

Amounts of Lending Club loans



####Question 2.8: Lending set:Adding a categorical variable

```
# Enter code here
ggplot(loans, aes(x = loan_amount, fill = homeownership)) +
geom_density(adjust = 2, alpha = 0.5) + # Add a density plot layer to the plot with a smooth
ing bandwidth adjustment of 2 and set the transparency of the density curves to 0.5.
  labs(x = "Loan amount ($)",y = "Density",title = "Amounts of Lending Club loans", fill = "H
omeownership") #Add labels to the x-axis, y-axis, and title of the plot for clarity, and labe
l the legend as "Homeownership" for the fill colors.
```

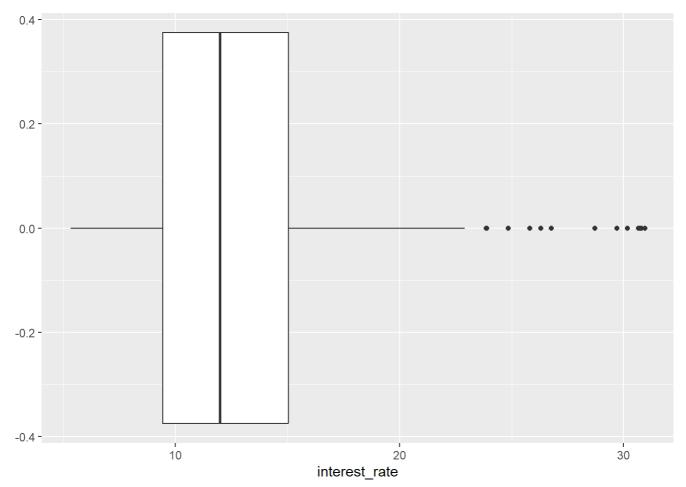


####Question 2.9: Lending set:Box plot

Solutions:

Enter code here
ggplot(loans, aes(x = interest_rate)) + # Create a ggplot object using the 'loans' dataset, s
pecifying 'interest_rate' for the x-axis.

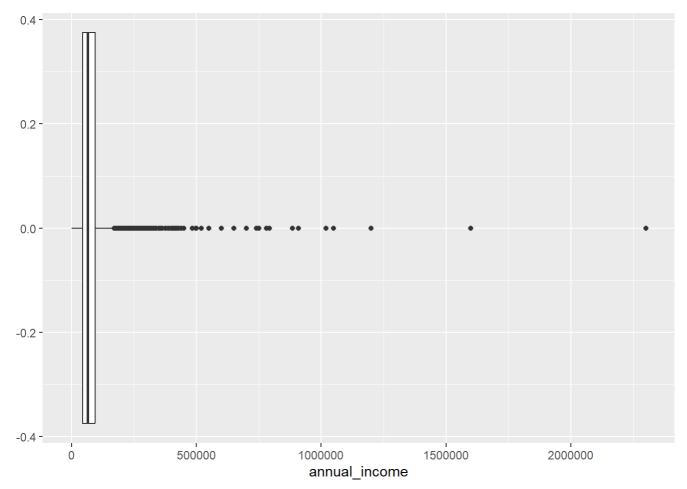
 $geom_boxplot()$ # Add a boxplot layer to the plot to visualize the distribution of interest r ates.



###Question 2.10: Lending set:Box plot and outliers

Solutions:

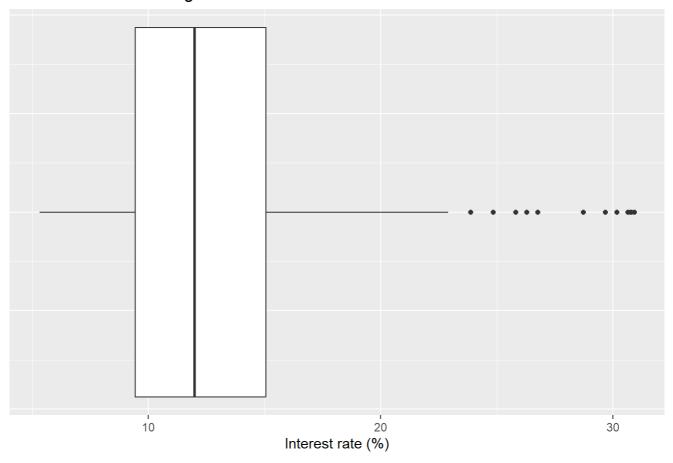
Enter code here
ggplot(loans, aes(x = annual_income)) + # Create a ggplot object using the 'loans' dataset, s
pecifying 'annual_income' for the x-axis
geom_boxplot() #Add a boxplot layer to the plot to visualize the distribution of annual inco
me.



###Question 2.11: Lending set:Customizing box plots

```
# Enter code here
ggplot(loans, aes(x = interest_rate)) +
    geom_boxplot() + # Add a boxplot layer to the plot to visualize the distribution of interes
t rates.
    labs(x = "Interest rate (%)",y = NULL,
    title = "Interest rates of Lending Club loans") + # Add a label to the x-axis, remove the la
bel for the y-axis, and set the title of the plot.
    theme( axis.ticks.y = element_blank(), axis.text.y = element_blank() ) # Customize the appea
rance of the y-axis by removing ticks and text.
```

Interest rates of Lending Club loans

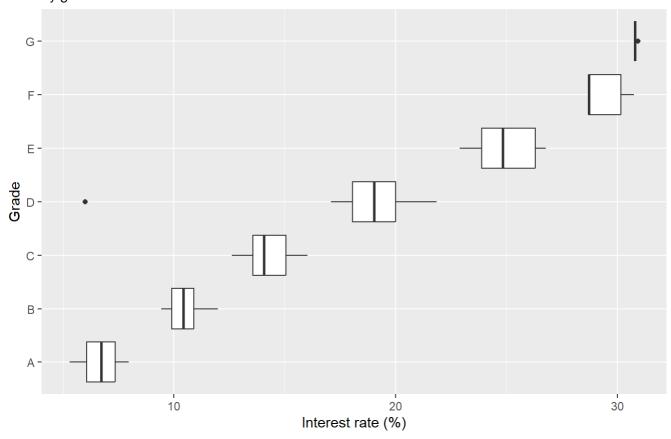


###Question 2.12: Lending set:Adding a categoric variable

```
# Enter code here
ggplot(loans, aes(x = interest_rate,
  y = grade)) + # Create a ggplot object using the 'loans' dataset, specifying 'interest_rate'
for the x-axis and 'grade' for the y-axis.
  geom_boxplot() + # Add a boxplot layer to the plot to visualize the distribution of interest
rates for different grades.
geom_boxplot() +
  labs(x = "Interest rate (%)",y = "Grade",title = "Interest rates of Lending Club loans",subt
itle ="by grade of loan") # Add labels to the x-axis and y-axis, as well as a title and a sub
title to the plot for clarity.
```

Interest rates of Lending Club loans

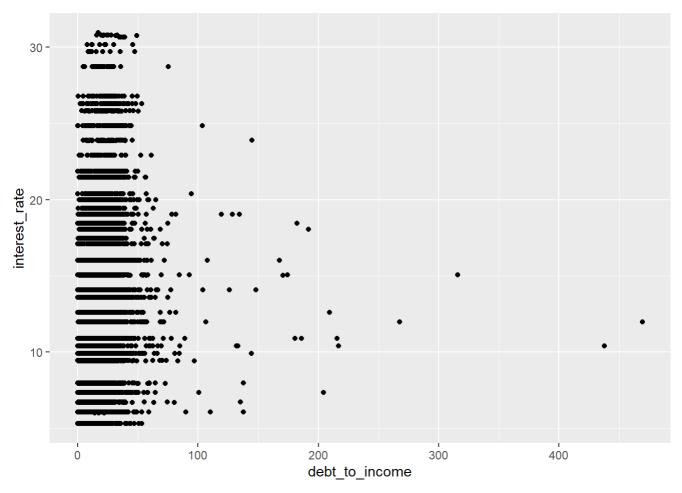
by grade of loan



###Question 2.13: Lending set:Scatterplot

Solutions:

```
# Enter code here
ggplot(loans, aes(x = debt_to_income, y = interest_rate)) + # Add a scatterplot layer to the
plot to visualize the relationship between debt-to-income ratio and interest rates.
geom_point()
```



###Question 2.14: Lending set:Hex Plot

```
# Enter code here
ggplot(loans, aes(x = debt_to_income, y = interest_rate)) + # Add a hexbin plot layer to the
plot to visualize the relationship between debt-to-income ratio and interest rates using hexa
gonal bins.
geom_hex()
```

```
## Warning: Removed 24 rows containing non-finite values (`stat_binhex()`).
```

```
## Warning: Computation failed in `stat_binhex()`
## Caused by error in `compute_group()`:
## ! The package "hexbin" is required for `stat_binhex()`
```

```
interest_rate
```

debt_to_income

```
# Enter code here
ggplot(loans %>% filter(debt_to_income < 100),
aes(x = debt_to_income, y = interest_rate)) +
geom_hex()</pre>
```

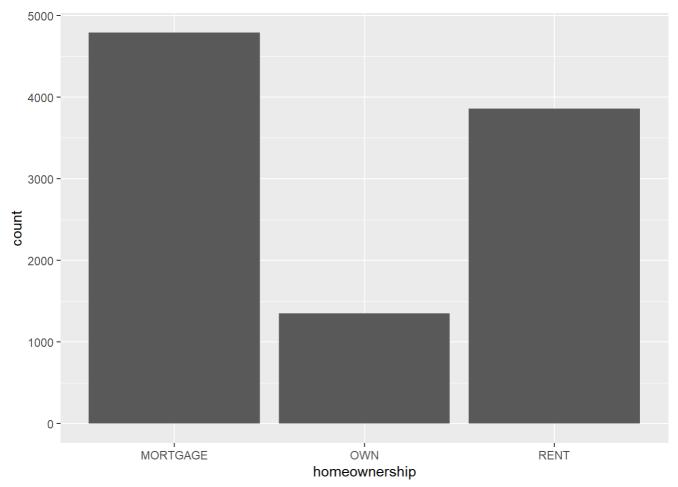
```
## Warning: Computation failed in `stat_binhex()`
## Caused by error in `compute_group()`:
## ! The package "hexbin" is required for `stat_binhex()`
```

interest_rate

debt_to_income

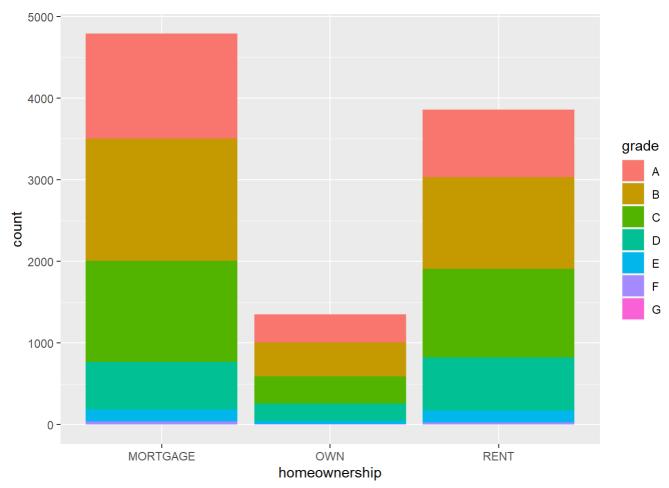
###Question 2.15: Lending set:Bar Plot

```
# Enter code here
ggplot(loans, aes(x = homeownership)) +
geom_bar()
```

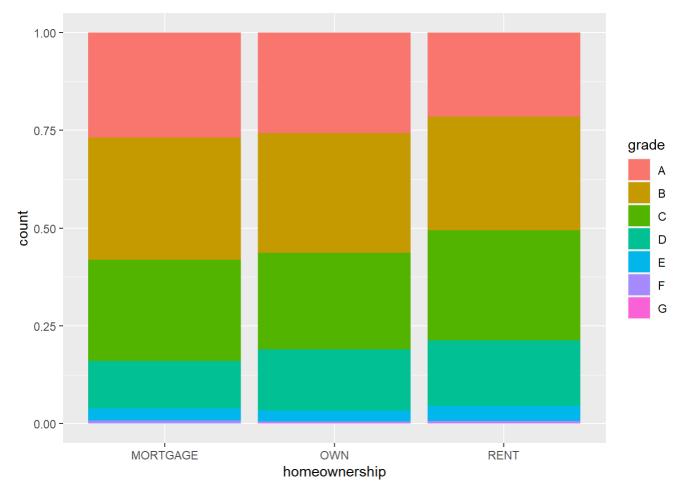


###Question 2.16: Lending set:Segmented bar plot

```
# Enter code here
ggplot(loans, aes(x = homeownership,
fill = grade)) +
geom_bar()
```



```
# Enter code here
ggplot(loans, aes(x = homeownership, fill = grade)) +
geom_bar(position = "fill")
```

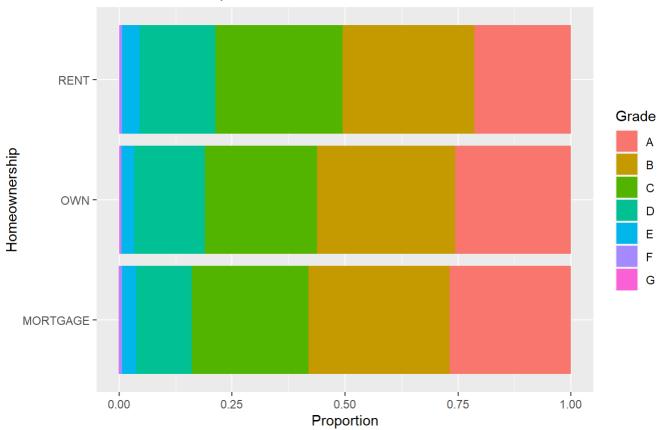


###Question 2.17: Lending set: Customiszing bar plot

```
# Enter code here
ggplot(loans, aes(y = homeownership, fill = grade)) + geom_bar(position = "fill") +
labs( x = "Proportion", y = "Homeownership", fill = "Grade", title = "Grades of Lending Club
loans", subtitle = "and homeownership of lendee")
```

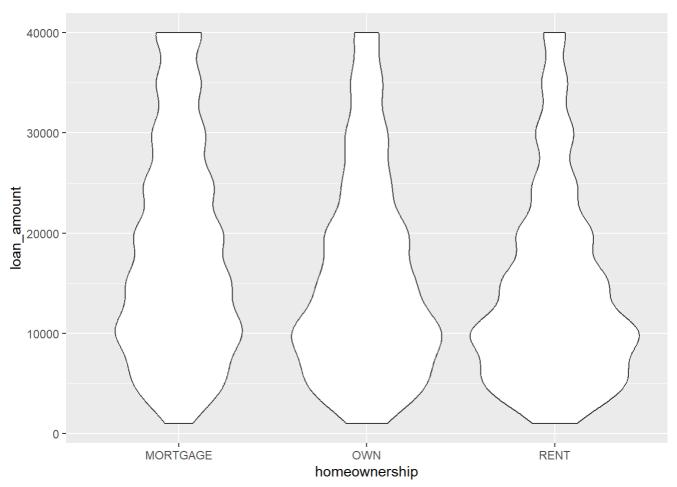
Grades of Lending Club loans

and homeownership of lendee



###Question 2.18: Lending set: Violin plot

```
# Enter code here
ggplot(loans, aes(x = homeownership, y = loan_amount)) +
geom_violin()
```



###Question 2.19: Lending set: Ridge plot

Solutions:

```
# Enter code here
library(ggridges)
ggplot(loans, aes(x = loan_amount, y = grade, fill = grade, color = grade)) +
geom_density_ridges(alpha = 0.5) # Add a density ridge plot layer to visualize the distribut
ion of loan amounts by grade, with transparency set to 0.5.
```

Picking joint bandwidth of 2360

