

Exam 1 review

Questions

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Note

Suggested answers can be found [here](#), but resist the urge to peek before you go through it yourself.

Blizzard salaries

In 2020, employees of Blizzard Entertainment circulated a spreadsheet to anonymously share salaries and recent pay increases amidst rising tension in the video game industry over wage disparities and executive compensation. (Source: [Blizzard Workers Share Salaries in Revolt Over Pay](#))

The name of the data frame used for this analysis is `blizzard_salary` and the variables are:

- `percent_incr`: Raise given in July 2020, as percent increase with values ranging from 1 (1% increase) to 21.5 (21.5% increase)
- `salary_type`: Type of salary, with levels Hourly and Salaried
- `annual_salary`: Annual salary, in USD, with values ranging from \$50,939 to \$216,856.
- `performance_rating`: Most recent review performance rating, with levels Poor, Successful, High, and Top. The Poor level is the lowest rating and the Top level is the highest rating.

The first ten rows of `blizzard_salary` are shown below:

```
# A tibble: 409 x 4
  percent_incr salary_type annual_salary performance_rating
      <dbl>   <chr>          <dbl>   <chr>
1           1 Salaried             1 High
2           1 Salaried             1 Successful
3           1 Salaried             1 High
4           1 Hourly        33987. Successful
5          NA Hourly        34798. High
6          NA Hourly        35360 <NA>
7          NA Hourly        37440 <NA>
```

```

8          0   Hourly          37814. <NA>
9          4   Hourly          41101. Top
10         1.2 Hourly          42328  <NA>
# i 399 more rows

```

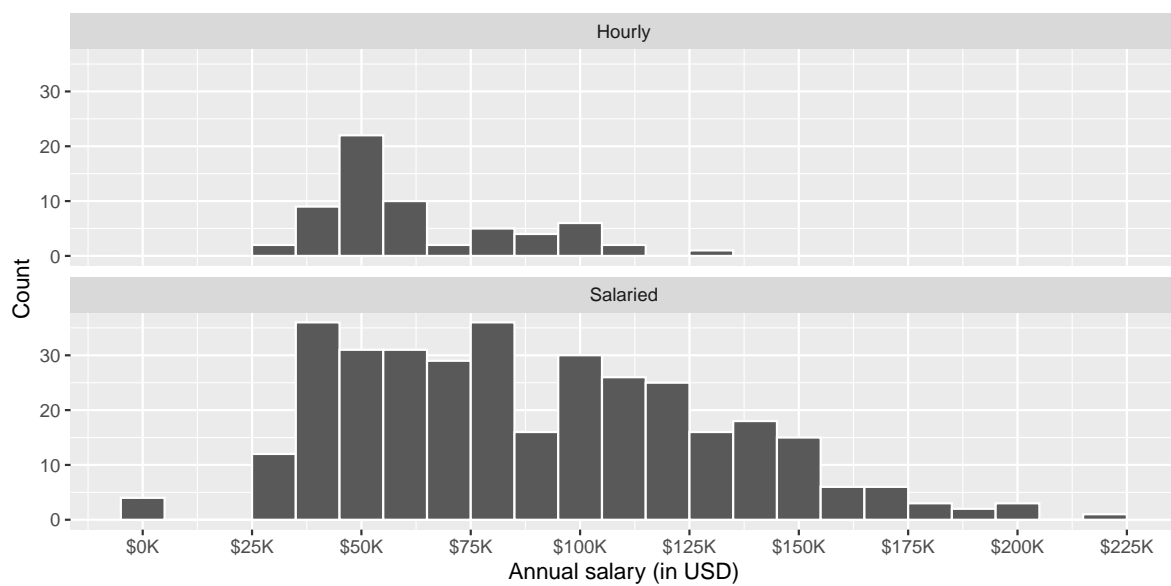
Question 1

Which of the following is correct? Choose all that apply.

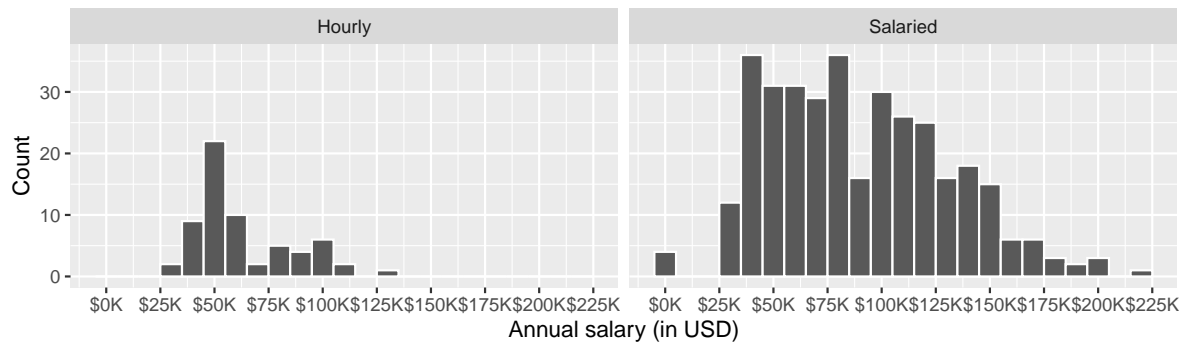
- a. The `blizzard_salary` dataset has 399 rows.
- b. The `blizzard_salary` dataset has 4 columns.
- c. Each row represents a Blizzard Entertainment worker who filled out the spreadsheet.
- d. The `percent_incr` variable is numerical and discrete.
- e. The `salary_type` variable is numerical.
- f. The `annual_salary` variable is numerical.
- g. The `performance_rating` variable is categorical and ordinal.

Question 2

Figure 1a and Figure 1b show the distributions of annual salaries of hourly and salaried workers. The two figures show the same data, with the facets organized across rows and across columns. Which of the two figures is better for comparing the median annual salaries of hourly and salaried workers. Explain your reasoning.



(a) Option 1



(b) Option 2

Figure 1: Distribution of annual salaries of Blizzard employees

Question 3

Suppose your teammate wrote the following code as part of their analysis of the data.

```
blizzard_salary |>
  group_by(salary_type) |>
  summarize(
    mean_annual_salary = mean(annual_salary),
    median_annual_salary = median(annual_salary)
  )
```

They then printed out the results shown below. Unfortunately one of the numbers got erased from the printout. It's indicated with _____ below.

```
# A tibble: 2 × 3
  salary_type mean_annual_salary median_annual_salary
  <chr>          <dbl>          <dbl>
1 Hourly          63003.          54246.
2 Salaried        90183.          _____
```

Which of the following is the best estimate for that erased value?

- a. 30,000
- b. 50,000
- c. 80,000
- d. 100,000

Question 4

Which distribution of annual salaries has a higher standard deviation?

- a. Hourly workers
- b. Salaried workers
- c. Roughly the same

Question 5

Which of the following alternate plots would also be useful for visualizing the distributions of annual salaries of hourly and salaried workers? Choose all that apply.

- a. Box plots
- b. Density plots
- c. Pie charts
- d. Waffle charts
- e. Histograms
- f. Scatterplots

Questions 6 and 7

Suppose you made the bar plot shown in Figure 2a to visualize the distribution of `performance_rating` and your teammate made the bar plot shown in Figure 2b.

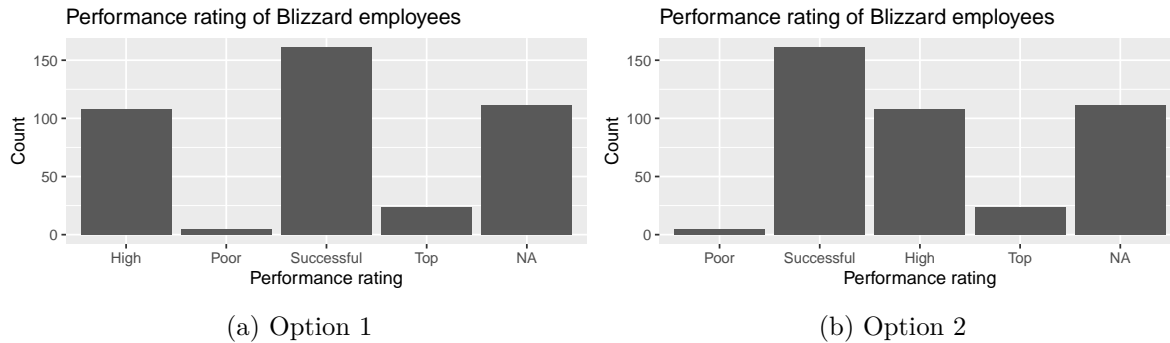


Figure 2: Distribution of performance rating

You made your bar plot without transforming the data in any way, while your friend did first transform the data with code like the following:

```
blizzard_salary <- blizzard_salary |>
  _(1)_(performance_rating = fct_relevel(performance_rating, _(2)_))
```

Question 6: What goes in the blank (1)?

- a. `arrange()`
- b. `filter()`
- c. `mutate()`
- d. `summarize()`

Question 7: What goes in the blank (2)?

- a. "Poor", "Successful", "High", "Top"
- b. "Successful", "High", "Top"
- c. "Top", "High", "Successful", "Poor"
- d. Poor, Successful, High, Top

Questions 8 - 10

Finally, another teammate creates the following two plots.

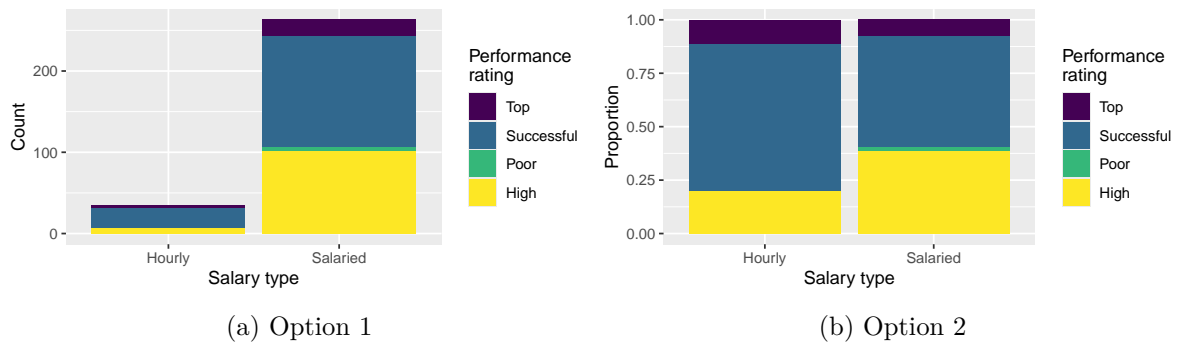


Figure 3: Distribution of salary type by performance rating

Question 8: Your teammate asks you for help deciding which one to use in the final report for visualizing the relationship between performance rating and salary type. In 1-3 sentences, can you help them make a decision, justify your choice, and write the narrative that should go with the plot?

Question 9: A friend with a keen eye points out that the number of observations in Figure 3a seems lower than the total number of observations in `blizzard_salary`. What might be going on here? Explain your reasoning.

Question 10: Below are the proportions of performance ratings for hourly and salaried workers. Place these values in the corresponding segments in Figure 3b.

```
# A tibble: 4 x 3
  performance_rating Hourly Salaried
  <chr>             <dbl>   <dbl>
1 High              0.2     0.384
2 Successful        0.686   0.521
3 Top               0.114   0.0760
4 Poor              0      0.0190
```

Questions 11 and 12

The table below shows the distribution of `salary_type` and `performance_rating`.

```
# A tibble: 2 x 6
  salary_type Poor Successful High Top `NA`
  <chr>      <int>      <int> <int> <int> <int>
1 Hourly      NA         24     7     4    28
2 Salaried     5        137    101    20    83
```

The pipeline below produces a data frame with a fewer number of rows than `blizzard_salary`.

```
blizzard_salary |>
  filter(salary_type _(1)_ "Hourly" _(2)_ performance_rating == "Poor") |>
  _(3)_ (annual_salary)
```

```
# A tibble: 5 x 4
  percent_incr salary_type annual_salary performance_rating
      <dbl> <chr>          <dbl> <chr>
1           0 Salaried      80000 Poor
2           3 Salaried      83000 Poor
3           0 Salaried     116000 Poor
4           0 Salaried     135219 Poor
5           0 Salaried     147500 Poor
```

Question 11: Which of the following goes in blanks (1) and (2)?

	(1)	(2)
a.	!=	
b.	==	&
c.	!=	&
d.	==	

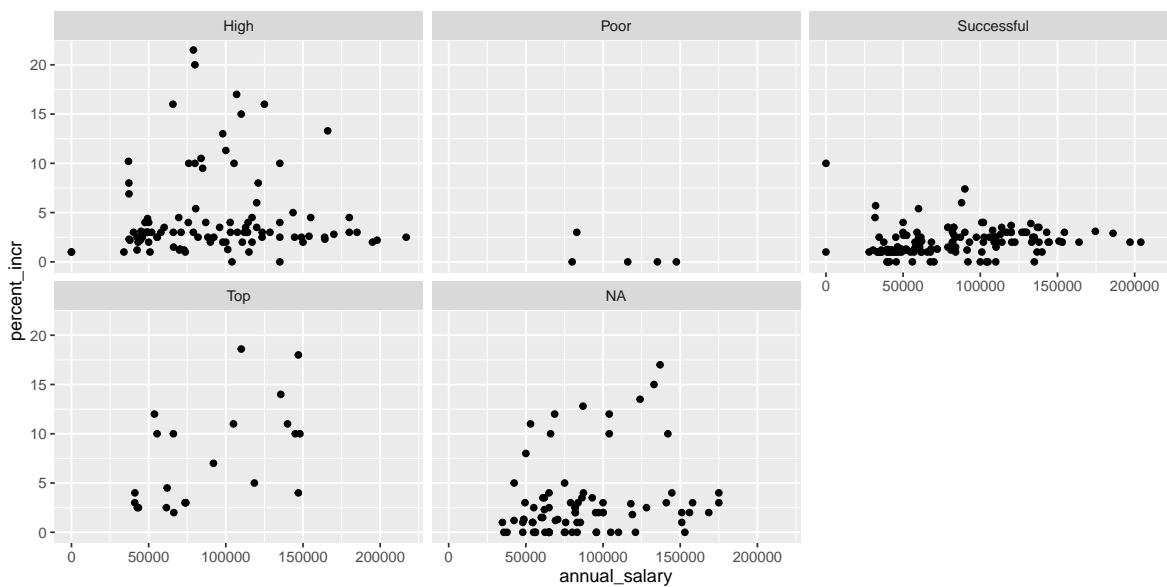
Question 12: Which function or functions go into blank (3)?

- a. `arrange()`

- b. `mutate()`
- c. `order()`
- d. `sort()`

Question 13

You're reviewing another team's work and they made the following visualization:



And they wrote the following interpretation for the relationship between annual salary and percent increase for Top performers:

The relationship is positive, having a higher salary results in a higher percent increase. There is one clear outlier.

Which of the following is/are the most accurate and helpful) peer review note for this interpretation. Choose all that apply.

- a. The interpretation is complete and perfect, no changes needed!
- b. The interpretation doesn't mention the direction of the relationship.
- c. The interpretation doesn't mention the form of the relationship, which is linear.

- d. The interpretation doesn't mention the strength of the relationship, which is somewhat strong.
- e. There isn't a clear outlier in the plot. If any points stand out as potential outliers, more guidance should be given to the reader to identify them (e.g., salary and/or percent increase amount).
- f. The interpretation is causal – we don't know if the cause of the high percent increase is higher annual salary based on observational data. The causal direction might be the other way around, or there may be other factors contributing to the apparent relationship.

Question 14

Below is some code and its output.

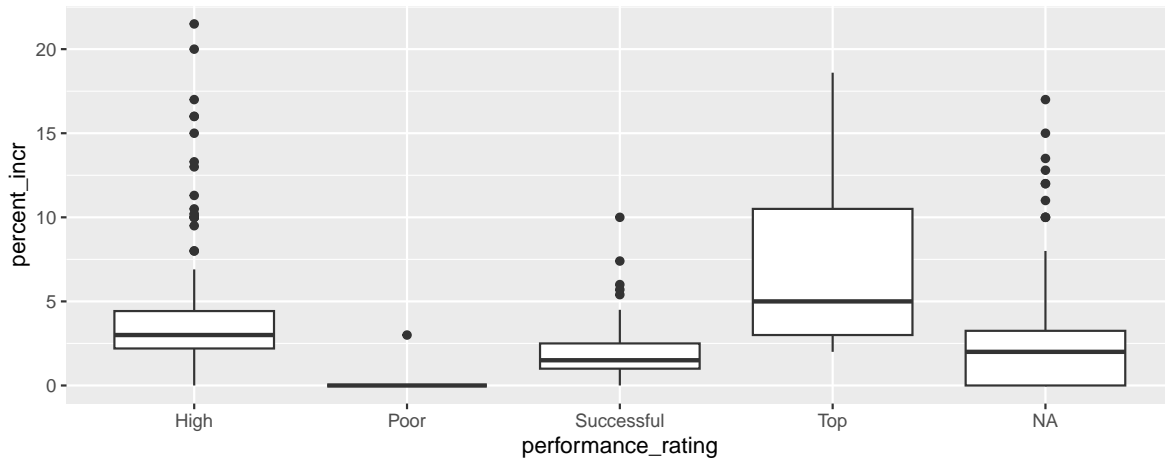
```

```{r}
label=plot blizzard

ggplot(blizzard_salary,aes(x=performance_rating,y=percent_incr))+geom_boxplot()
labs(x="Performance rating", y = "Percent increase")
```

```

```
Warning: Removed 39 rows containing non-finite outside the scale range
(`stat_boxplot()`).
```



```
<ggplot2::labels> List of 2
 $ x: chr "Performance rating"
 $ y: chr "Percent increase"
```

Part 1: List at least 5 things that should be fixed or improved in the code.

Part 2: What is the cause of the warning and what does it mean?

Question 15

You're working on a data analysis on salaries of Blizzard employees in a Quarto document in a project version controlled by Git. You create a plot and write up a paragraph describing any patterns in it. Then, your teammate says "render, commit, and push".

Part 1: What do they mean by each of these three steps. In 1-2 sentences for each, explain in your own words what they mean.

1. Render:

2. Commit:

3. Push:

Part 2: Your teammate is getting impatient and they interrupt you after you rendered and committed and say "I still can't see your changes in our shared GitHub repo when I look at it in my web browser." Which of the following answers is the most accurate?

- a. I rendered my document, you should be seeing my changes on GitHub when you look at it in your web browser.
- b. I committed my changes, you should be seeing my changes on GitHub when you look at it in your web browser.
- c. I didn't yet push my changes, it's expected that you are not seeing them on GitHub when you look at it in your web browser. Wait until I push, and check again.
- d. You need to pull to see my changes on GitHub in the web browser.

| Island | Adelie | Gentoo | Chinstrap | Total |
|-----------|--------|--------|-----------|-------|
| Biscoe | 44 | 124 | 0 | 168 |
| Dream | 56 | 0 | 68 | 124 |
| Torgersen | 52 | 0 | 0 | 52 |
| Total | 152 | 124 | 68 | 344 |

Penguins

The `penguins` data set includes measurements for penguin species, including: flipper length, body mass, bill dimensions, and sex. The following table summarizes information on which species of penguins (Adelie, Gentoo, and Chinstrap) live on which islands (Biscoe, Dream, or Torgersen).

Question 16

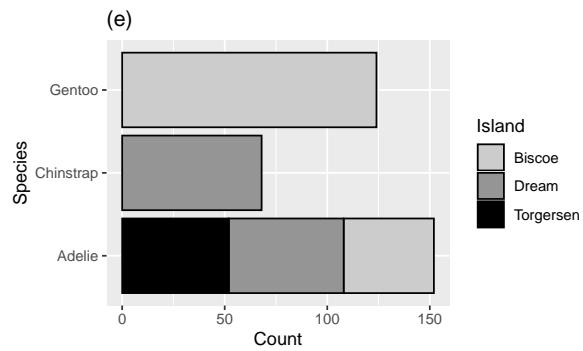
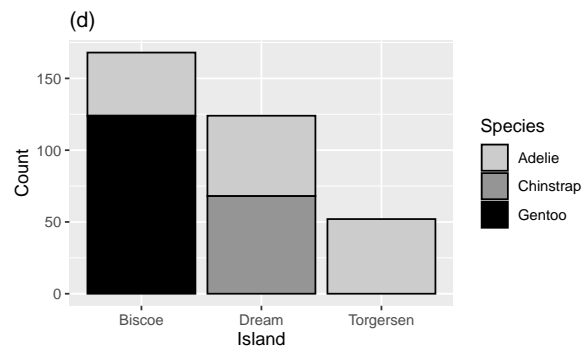
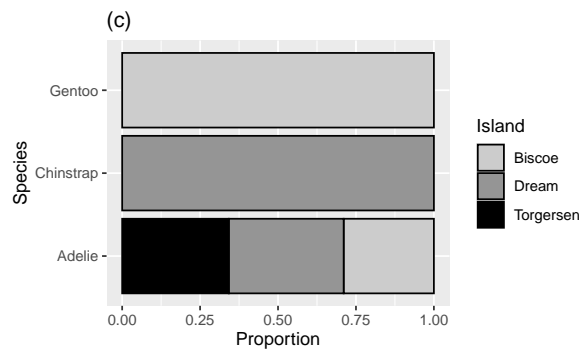
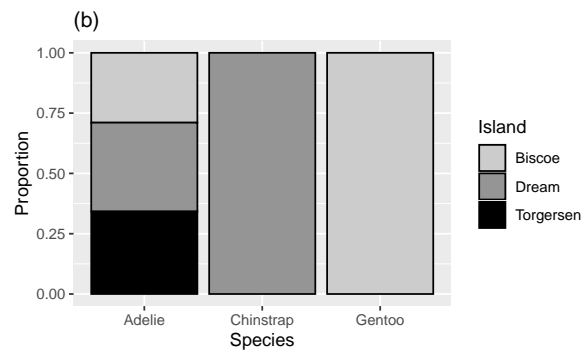
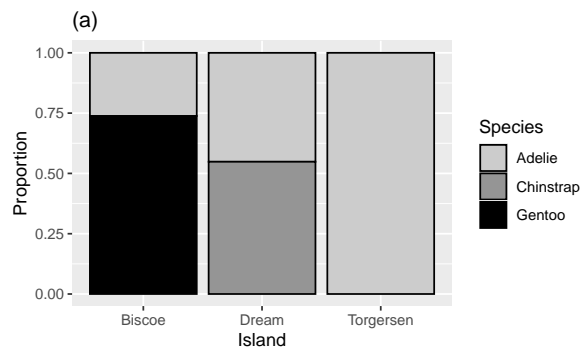
Which of the following plots is the result of the following code?

```
ggplot(penguins, aes(x = island, fill = species)) +
  geom_bar()
```

NYC Flights

The `flights` dataset includes characteristics of all flights departing from New York City airports (JFK, LGA, EWR) in 2013. Below is a peek at the first ten rows of the `flights` data.

```
# A tibble: 336,776 x 19
   year month   day arr_delay carrier dep_time sched_dep_time dep_delay
   <int> <int> <int>     <dbl> <chr>      <int>         <int>         <dbl>
1  2013     1     1      11 UA          517           515           2
2  2013     1     1      20 UA          533           529           4
3  2013     1     1      33 AA          542           540           2
4  2013     1     1     -18 B6          544           545          -1
5  2013     1     1     -25 DL          554           600          -6
6  2013     1     1      12 UA          554           558          -4
7  2013     1     1      19 B6          555           600          -5
8  2013     1     1     -14 EV          557           600          -3
9  2013     1     1      -8 B6          557           600          -3
```



```

10 2013      1      1          8 AA          558          600          -2
# i 336,766 more rows
# i 11 more variables: arr_time <int>, sched_arr_time <int>, flight <int>,
#   tailnum <chr>, origin <chr>, dest <chr>, air_time <dbl>, distance <dbl>,
#   hour <dbl>, minute <dbl>, time_hour <dtm>

```

Question 17

Based on this output, which of the following must be true about the `flights` data frame? **Select all that are true.**

- a. The `flights` data frame is a tibble.
- b. The `flights` data frame has 10 rows.
- c. The `flights` data frame has 8 columns.
- d. The `carrier` variable in the `flights` data frame is a character variable.
- e. There are no missing data in the `flights` data frame.

Question 18

Which of the following pipelines produce(s) the output shown below? **Select all that apply.**

```

# A tibble: 336,776 x 5
  arr_delay carrier  year month   day
    <dbl>   <chr>   <int> <int> <int>
1     1272 HA      2013     1     9
2     1127 MQ      2013     6    15
3     1109 MQ      2013     1    10
4     1007 AA      2013     9    20
5      989 MQ      2013     7    22
6      931 DL      2013     4    10
7      915 DL      2013     3    17
8      895 DL      2013     7    22
9      878 AA      2013    12     5
10     875 MQ      2013     5     3
# i 336,766 more rows

```

- a.

```
flights |>
  select(arr_delay, carrier, year, month, day) |>
  arrange(desc(arr_delay))
```

b.

```
flights |>
  select(arr_delay, carrier, year, month, day) |>
  arrange(arr_delay)
```

c.

```
flights |>
  select(arr_delay, carrier, year, month, day) |>
  arrange(year)
```

d.

```
flights |>
  arrange(desc(arr_delay)) |>
  select(arr_delay, carrier, year, month, day)
```

e.

```
flights |>
  arrange(desc(arr_delay)) |>
  select(day, month, year, arr_delay, carrier)
```

Countries and populations

We have a small dataset of six countries and their populations:

```
# A tibble: 6 x 2
  country      population
  <chr>         <dbl>
1 Curacao         150
2 Ecuador       18001
3 Iraq          44496.
4 New Zealand     5124.
5 Palau           18.0
6 United States  333288.
```

And another small dataset of five countries and the continent they're in:

```
# A tibble: 5 x 3
  entity      code continent
  <chr>      <chr> <chr>
1 Angola     AGO   Africa
2 Curacao    CUW   North America
3 Ecuador    ECU   South America
4 Iraq       IRQ   Asia
5 New Zealand NZL   Oceania
```

You join the two datasets with the following:

```
population |>
  left_join(continent, by = join_by(country == entity))
```

Question 19

How many rows will the resulting data frame have?

- a. 4
- b. 5
- c. 6
- d. 7
- e. 8

Question 20

What will be the columns of the resulting data frame?

- a. country, population
- b. country, population, code, continent
- c. entity, code, continent
- d. entity, population, code, continent
- e. country, entity, population, code, continent

| | Built earlier than 1950 | Built in 1950 or later |
|-----------|-------------------------|------------------------|
| Garage | 5 | 33 |
| No garage | 3 | 57 |

Duke Forest houses

The `duke_forest` dataset includes information on prices and various other features (number of bedrooms, bathrooms, area, year built, type of cooling, type of heating, etc.) of houses in the Duke Forest neighborhood of Durham, NC.

Rows: 98

Columns: 13

```
$ address    <chr> "1 Learned Pl, Durham, NC 27705", "1616 Pinecrest Rd, Durha~
$ price      <dbl> 1520000, 1030000, 420000, 680000, 428500, 456000, 1270000, ~
$ bed        <dbl> 3, 5, 2, 4, 4, 3, 5, 4, 4, 3, 4, 4, 3, 5, 4, 5, 3, 4, 4, 3,~
$ bath       <dbl> 4.0, 4.0, 3.0, 3.0, 3.0, 3.0, 5.0, 3.0, 5.0, 2.0, 3.0, 3.0,~
$ area       <dbl> 6040, 4475, 1745, 2091, 1772, 1950, 3909, 2841, 3924, 2173,~
$ type       <chr> "Single Family", "Single Family", "Single Family", "Single ~
$ year_built <dbl> 1972, 1969, 1959, 1961, 2020, 2014, 1968, 1973, 1972, 1964,~
$ heating    <chr> "Other, Gas", "Forced air, Gas", "Forced air, Gas", "Heat p~
$ cooling     <fct> central, central, central, central, central, central, centr~
$ parking    <chr> "0 spaces", "Carport, Covered", "Garage - Attached, Covered~
$ lot        <dbl> 0.97, 1.38, 0.51, 0.84, 0.16, 0.45, 0.94, 0.79, 0.53, 0.73,~
$ hoa        <chr> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA,~
$ url        <chr> "https://www.zillow.com/homedetails/1-Learned-Pl-Durham-NC-~
```

The following summary table gives us some information about whether homes in this data set have garages and when they were built.

The pipeline below produces a data frame with a fewer number of rows than `duke_forest`.

```
duke_forest |>
  filter(parking == "Garage" |_(1)_ year_built |_(2)_ 1950) |>
  select(parking, year_built, price, area) |>
  |_(3)_ (price_per_sqfeet = price / area)
```

A tibble: 5 x 5

```
  parking year_built  price  area price_per_sqfeet
  <chr>      <dbl>  <dbl> <dbl>          <dbl>
1 Garage    1945  900000  2933          307.
```

| | | | | | |
|---|--------|------|--------|------|------|
| 2 | Garage | 1938 | 265000 | 1300 | 204. |
| 3 | Garage | 1934 | 600000 | 2514 | 239. |
| 4 | Garage | 1941 | 412500 | 1661 | 248. |
| 5 | Garage | 1940 | 105000 | 1094 | 96.0 |

Question 21

Which of the following goes in blanks (1) and (2)?

| | (1) | (2) |
|----|-----|-----|
| a. | & | < |
| b. | | < |
| c. | & | >= |
| d. | | >= |
| e. | & | != |

Question 22

Which function or functions go into blank (3)? **Select all that apply.**

- a. `arrange()`
- b. `mutate()`
- c. `filter()`
- d. `summarize()`
- e. `slice()`

Law & Order

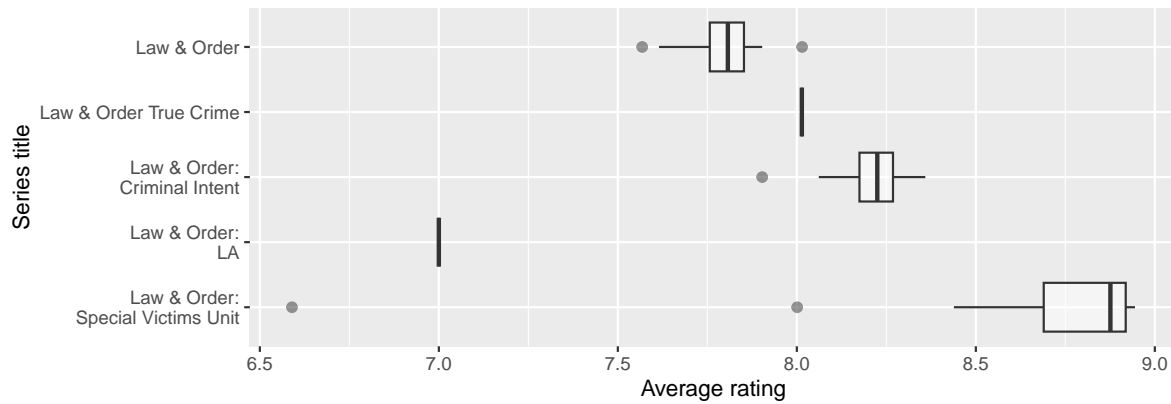
You've heard of the tidyverse, now let's visit the Law & Order-verse. Doink doink!¹

Law & Order is a police procedural and legal drama television series that has been running since the 1990s. The Law & Order franchise includes a number of series such as Law & Order, Law & Order: SVU, Law & Order: Criminal Intent, etc.

¹“Doink doink” is the scene and episode introductory sound on the Law & Order series. If you've never heard it, you're not at any disadvantage for the exam. If you've ever heard it, good luck getting it out of your head!

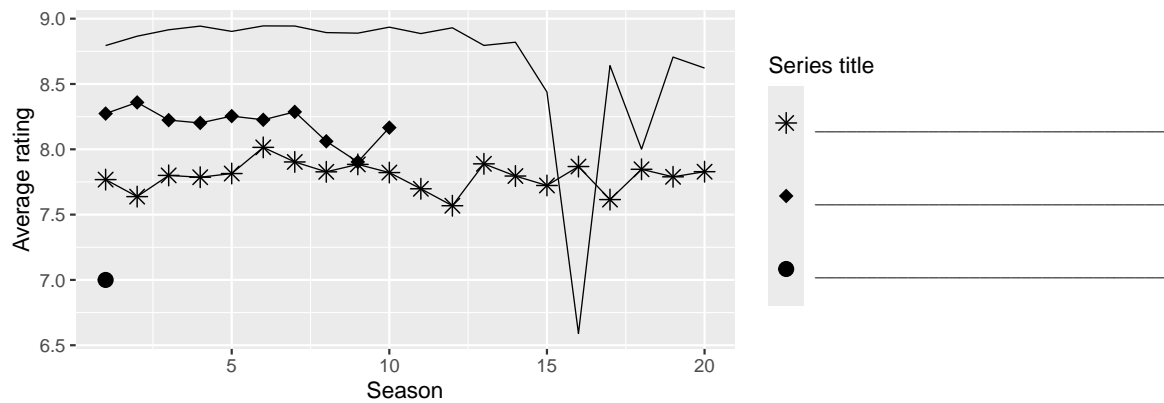
You will work with data on average ratings for each season of three series from the Law & Order-verse – a subset of the data from the previous questions. Below is a peek at the first ten rows of the Law & Order data.

The plot below shows the distributions of average ratings of various Law & Order series across seasons.



Question 23

Based on the information from the side-by-side box plots, fill in the legend of the plot below with Law & Order series titles.



Question 24

The following code calculates the standard deviations of average season ratings of the five Law & Order series. Unfortunately, the output is partially erased and replaced with blanks.

```
lo_titles <- c("Law & Order", "Law & Order: Criminal Intent", "Law & Order: SVU")
```

```
law_and_order |>
  filter(title %in% lo_titles) |>
  group_by(title) |>
  summarize(mean_av_rating = mean(av_rating), sd_av_rating = sd(av_rating))
```

```
# A tibble: 5 × 3
```

| | title | mean_av_rating | sd_av_rating |
|---|------------------------------|----------------|--------------|
| | <chr> | <dbl> | <dbl> |
| 1 | Law & Order | _(1)_ | 0.106 |
| 2 | Law & Order: Criminal Intent | 8.20 | 0.129 |
| 4 | Law & Order: SVU | 8.67 | _(2)_ |

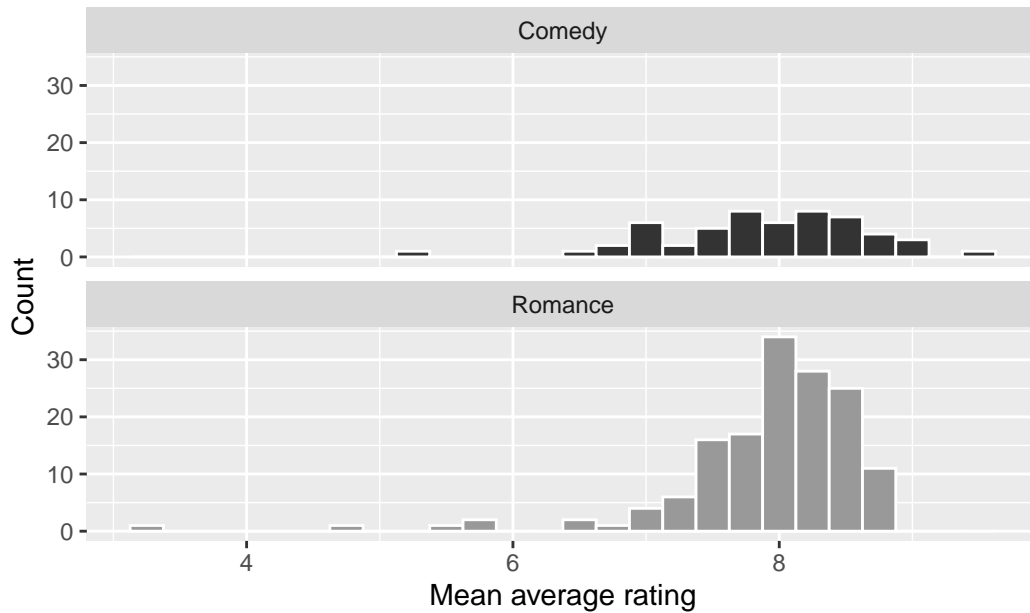
Based on the visualizations you’ve seen of these data so far, which of the following is true about the blanks in the output? **Select all that are true.**

- The **mean** of average ratings (Blank 1) of Law & Order seasons is **lower** than the other two means.
- The **mean** of average ratings (Blank 1) of Law & Order seasons is **higher** than the other two means.
- The **standard deviation** of average ratings of Law & Order: SVU seasons (Blank 2) is **lower** than the other two standard deviations.
- The **standard deviation** of average ratings of Law & Order: SVU seasons (Blank 2) is **higher** than the other two standard deviations.
- The **standard deviation** of average ratings of Law & Order: SVU seasons (Blank 2) is **between** the other two standard deviations.

Romance and comedy

Finally, we focus on romance and comedy shows. We first filter the dataset for any shows that have romance or comedy as their genre (**genre_1**, **genre_2**, or **genre_3**) and then remove shows that have both of these genre labels. For the next two questions, we focus on these shows that we identify as either romance or comedy. We then calculate the mean of the average season ratings for each show, to obtain a single “mean average rating” value per show.

The plot below shows the distributions of mean average ratings of seasons of comedy and romance shows.



Question 25

Which of the following statements is **true** about these distributions? **Select all that are true.**

- a. Mean average ratings of romance shows are bimodal.
- b. Mean average ratings of comedy are unimodal.
- c. Mean average ratings of romance shows is left skewed.
- d. Mean average ratings of comedy shows is right skewed.
- e. There are more romance shows than comedy shows.

IMDB

The data for the next few questions come from the Internet Movie Database (IMDB). Specifically, the data are a random sample of movies released between 1980 and 2020.

The name of the data frame used for this analysis is **movies**, and it contains the variables shown in Table 1.

Table 1: Data dictionary for **movies**

| Variable | Description |
|-------------|-------------------|
| name | name of the movie |

| Variable | Description |
|-----------------|---|
| rating | rating of the movie (R, PG, etc.) |
| genre | main genre of the movie. |
| runtime | duration of the movie |
| year | year of release |
| release_date | release date (YYYY-MM-DD) |
| release_country | release country |
| score | IMDB user rating |
| votes | number of user votes |
| director | the director |
| writer | writer of the movie |
| star | main actor/actress |
| country | country of origin |
| budget | the budget of a movie (some movies don't have this, so it appears as 0) |
| gross | revenue of the movie |
| company | the production company |

The first thirty rows of the `movies` data frame are shown in Table 2, with variable types suppressed (since we'll ask about them later).

Question 26

The `name` and `runtime` variables are shown below, with the variable types suppressed.

| # A tibble: 500 x 2 | | What is the type of the <code>runtime</code> variable? |
|-----------------------------|---------|--|
| name | runtime | |
| 1 Blue City | 83 mins | a. Character |
| 2 Winter Sleep | 196 | b. Double |
| 3 Rang De Basanti | 167 | c. Factor |
| 4 Pokémon Detective Pikachu | 104 | d. Integer |
| 5 A Bad Moms Christmas | 104 | e. Logical |
| 6 Replicas | 107 | |
| # i 494 more rows | | |

Table 2

First 30 rows of movies, with variable types suppressed.

A tibble: 500 x 16

| | name | score | runtime | genre | rating | release_country | release_date |
|----|----------------|-----------|---------|-----------|------------------|-----------------|---------------|
| 1 | Blue City | 4.4 | 83 mins | Action | R | United States | 1986-05-02 |
| 2 | Winter Sleep | 8.1 | 196 | Drama | Not Rated | Turkey | 2014-06-12 |
| 3 | Rang De Basan~ | 8.1 | 167 | Comedy | Not Rated | United States | 2006-01-26 |
| 4 | Pokémon Detec~ | 6.6 | 104 | Action | PG | United States | 2019-05-10 |
| 5 | A Bad Moms Ch~ | 5.6 | 104 | Comedy | R | United States | 2017-11-01 |
| 6 | Replicas | 5.5 | 107 | Drama | PG-13 | United States | 2019-01-11 |
| 7 | Windy City | 5.8 | 103 | Drama | R | Uruguay | 1986-01-01 |
| 8 | War for the P~ | 7.4 | 140 | Action | PG-13 | United States | 2017-07-14 |
| 9 | Tales from th~ | 6.4 | 98 | Crime | R | United States | 1995-05-24 |
| 10 | Fire with Fire | 6.5 | 103 | Drama | PG-13 | United States | 1986-05-09 |
| 11 | Raising Helen | 6 | 119 | Comedy | PG-13 | United States | 2004-05-28 |
| 12 | Feeling Minne~ | 5.4 | 99 | Comedy | R | United States | 1996-09-13 |
| 13 | The Babe | 5.9 | 115 | Biography | PG | United States | 1992-04-17 |
| 14 | The Real Blon~ | 6 | 105 | Comedy | R | United States | 1998-02-27 |
| 15 | To vlemma tou~ | 7.6 | 176 | Drama | Not Rated | United States | 1997-11-01 |
| 16 | Going the Dis~ | 6.3 | 102 | Comedy | R | United States | 2010-09-03 |
| 17 | Jung on zo | 6.8 | 103 | Action | R | Hong Kong | 1993-06-24 |
| 18 | Rita, Sue and~ | 6.5 | 93 | Comedy | R | United Kingdom | 1987-05-29 |
| 19 | Phone Booth | 7 | 81 | Crime | R | United States | 2003-04-04 |
| 20 | Happy Death D~ | 6.6 | 96 | Comedy | PG-13 | United States | 2017-10-13 |
| 21 | Barely Legal | 4.7 | 90 | Comedy | R | Thailand | 2006-05-25 |
| 22 | Three Kings | 7.1 | 114 | Action | R | United States | 1999-10-01 |
| 23 | Menace II Soc~ | 7.5 | 97 | Crime | R | United States | 1993-05-26 |
| 24 | Four Rooms | 6.8 | 98 | Comedy | R | United States | 1995-12-25 |
| 25 | Quartet | 6.8 | 98 | Comedy | PG-13 | United States | 2013-03-01 |
| 26 | Tape | 7.2 | 86 | Drama | R | Denmark | 2002-07-12 |
| 27 | Marked for De~ | 6 | 93 | Action | R | United States | 1990-10-05 |
| 28 | Congo | 5.2 | 109 | Action | PG-13 | United States | 1995-06-09 |
| 29 | Stop-Loss | 6.4 | 112 | Drama | R | United States | 2008-03-28 |
| 30 | Con Air | 6.9 | 115 | Action | R | United States | 1997-06-06 |
| | budget | gross | votes | year | director | writer | star |
| 1 | 10000000 | 6947787 | 1100 | 1986 | Michelle Manning | Ross Macdona~ | Judd Nelson |
| 2 | NA | 4018705 | 48000 | 2014 | Nuri Bilge Ceyl~ | Ebru Ceylan | Haluk Bilgin~ |
| 3 | NA | 10800778 | 115000 | 2006 | Rakeysh Ompraka~ | Renzil D'Sil~ | Aamir Khan |
| 4 | 150000000 | 433921300 | 146000 | 2019 | Rob Letterman | Dan Hernandez | Ryan Reynolds |
| 5 | 28000000 | 130560428 | 46000 | 2017 | Jon Lucas | Jon Lucas | Mila Kunis |
| 6 | 30000000 | 9330075 | 34000 | 2018 | Jeffrey Nachman~ | Chad St. John | Keanu Reeves |
| 7 | NA | 343890 | 262 | 1984 | Armyan Bernstein | Armyan Berns~ | John Shea |
| 8 | 150000000 | 490719763 | 235000 | 2017 | Matt Reeves | Mark Bombback | Andy Serkis |
| 9 | 6000000 | 11837928 | 7400 | 1995 | Rusty Cundieff | Rusty Cundie~ | Clarence Wil~ |
| 10 | NA | 4636169 | 1500 | 1986 | Duncan Gibbins | Bill Phillips | Craig Sheffer |
| 11 | 50000000 | 49718611 | 36000 | 2004 | Garry Marshall | Patrick J. C~ | Kate Hudson |
| 12 | NA | 3124440 | 11000 | 1996 | Steven Baigelman | Steven Baige~ | Keanu Reeves |
| 13 | NA | 19930973 | 9300 | 1992 | Arthur Hiller | John Fusco | John Goodman |
| 14 | NA | 83488 | 3900 | 1997 | Tom DiCillo | Tom DiCillo | Matthew Modi~ |
| 15 | NA | NA | 6400 | 1995 | Theodoros Angel~ | Theodoros An~ | Harvey Keitel |
| 16 | 32000000 | 42059111 | 57000 | 2010 | Nanette Burstein | Geoff LaTuli~ | Drew Barrymo~ |

Question 27

The code below summarizes the data in a certain way.

```
movies |>
  summarize(sum(release_country == "United States"))
```

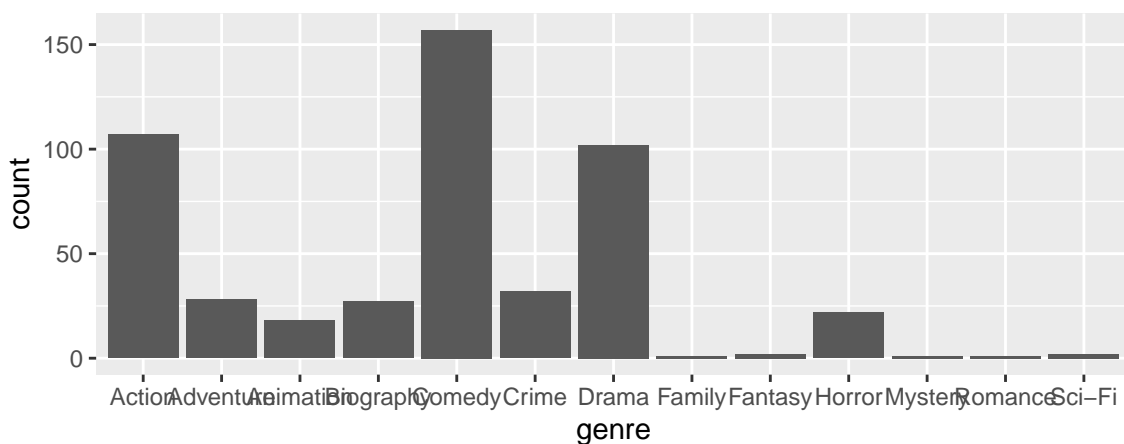
```
# A tibble: 1 x 1
  `sum(release_country == "United States")`
    <int>
1                435
```

Which of the following is **TRUE** about the code and its result? **Select all that are true.**

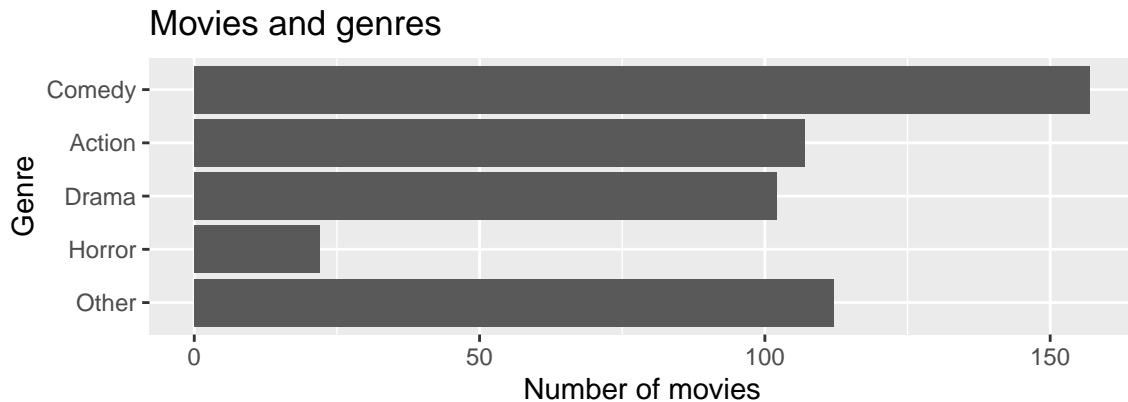
- a. Evaluates whether each `release_country` is equal to "United States" or not, which results in a logical variable.
- b. Filters out rows where `release_country` is not equal to "United States" and counts the remaining rows.
- c. Sums the logical values, where each TRUE is considered a 1 and each FALSE is considered a 0.
- d. Results in a character vector.
- e. The result shows there are 435 movies released in the United States.

Question 28

Suppose you want a visualization that shows the number of movies in the sample in each genre. Your first attempt is as follows.



A friend of yours says that the visualization is difficult to read and they suggest using the following visualization instead.



Source: IMDB.

Which of the following modifications would your friend have made to your code to create their version? **Select all that apply.**

- Combine movies in genres other than Comedy, Drama, Action, and Horror into a new level called "Other".
- Reorder the levels in descending order of numbers of observations, except for the "Other" level.
- Map `genre` to the y aesthetic.
- Add a title, x and y-axis labels, and a caption.
- Filter out all moves in genres other than Comedy, Drama, Action, and Horror before plotting.

Question 29

Which of the following is **TRUE** about the code and its result? **Select all that are true.**

```
movies |>
  count(rating, genre) |>
  pivot_wider(names_from = genre, values_from = n, values_fill = 0)
```

```
# A tibble: 6 x 6
  rating Other Drama Action Comedy Horror
<fct>   <int> <int>  <int>  <int>  <int>
1 G           5     1     1     1     0
```

| | | | | | | |
|---|-----------|----|----|----|----|----|
| 2 | PG | 38 | 13 | 10 | 18 | 0 |
| 3 | PG-13 | 19 | 25 | 35 | 35 | 0 |
| 4 | R | 45 | 50 | 57 | 96 | 21 |
| 5 | NC-17 | 1 | 2 | 0 | 1 | 0 |
| 6 | Not Rated | 4 | 11 | 4 | 6 | 1 |

- The code counts how many movies are in each rating and genre combination.
- The code sorts the results in descending order.
- Each row of the output is a movie.
- The output shows that there are six distinct ratings in the dataset.
- The code reduces the number of variables and observations in the `movies` data frame to six.

Bonus

Pick a concept we introduced in class so far that you've been struggling with and explain it in your own words.