

Seminar_1

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Install and load data

```
library(tidyverse)
```

```
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr     1.1.4     v readr     2.1.6
v forcats   1.0.1     v stringr   1.6.0
v ggplot2   4.0.1     v tibble    3.3.1
v lubridate  1.9.4     v tidyr    1.3.2
v purrr     1.2.1
-- 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 becom
```

```
library(datasauRus)
```

Have a look at the data

```
datasaurus_dozen_wide
```

```
# A tibble: 142 x 26
  away_x away_y bullseye_x bullseye_y circle_x circle_y dino_x dino_y dots_x
  <dbl>  <dbl>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>
 1    32.3    61.4      51.2      83.3      56.0      79.3      55.4      97.2
 2    53.4    26.2      59.0      85.5      50.0      79.0      51.5      96.0
 3    63.9    30.8      51.9      85.8      51.3      82.4      46.2      94.5
```

```

4   70.3   82.5      48.2      85.0      51.2      79.2      42.8      91.4      50.1
5   34.1   45.7      41.7      84.0      44.4      78.2      40.8      88.3      50.6
6   67.7   37.1      37.9      82.6      45.0      77.9      38.7      84.9      50.3
7   53.3   97.5      39.5      80.8      48.6      78.8      35.6      79.9      25.6
8   63.5   25.1      39.6      82.7      42.1      76.9      33.1      77.6      25.5
9   68.0   81.0      34.8      80.0      41.0      76.4      29.0      74.5      25.4
10  67.4   29.7      27.6      72.8      34.6      72.7      26.2      71.4      25.6
# i 132 more rows
# i 17 more variables: dots_y <dbl>, h_lines_x <dbl>, h_lines_y <dbl>,
#   high_lines_x <dbl>, high_lines_y <dbl>, slant_down_x <dbl>,
#   slant_down_y <dbl>, slant_up_x <dbl>, slant_up_y <dbl>, star_x <dbl>,
#   star_y <dbl>, v_lines_x <dbl>, v_lines_y <dbl>, wide_lines_x <dbl>,
#   wide_lines_y <dbl>, x_shape_x <dbl>, x_shape_y <dbl>

data("datasaurus_dozen_wide")
datasauRus::twelve_from_slant_wide

# A tibble: 182 x 24
  bullseye_x bullseye_y circle_x circle_y dots_x dots_y h_lines_x h_lines_y
    <dbl>     <dbl>     <dbl>     <dbl>     <dbl>     <dbl>     <dbl>     <dbl>
1     66.3     84.2     57.6     76.2     73.5     82.5     66.1     89.8
2     73.5     47.4     82.4     43.2     72.5     50.0     85.9     49.6
3     34.9     49.2     27.2     37.1     27.6     48.6     26.0     49.6
4     56.9     28.8     51.7     19.4     49.5     17.2     49.2     29.6
5     68.4     60.8     62.9     75.2     52.2     48.7     51.0     49.6
6     57.0     86.1     61.7     75.5     51.3     82.2     65.4     89.7
7     39.9     12.1     42.1     21.6     49.9     17.3     38.7     9.56
8     70.9     13.1     64.9     21.3     72.4     19.5     74.2     29.9
9     42.8     11.2     44.1     20.9     26.4     17.7     35.4     9.63
10    61.5     30.0     60.5     20.0     50.0     17.0     55.9     29.6
# i 172 more rows
# i 16 more variables: high_lines_x <dbl>, high_lines_y <dbl>, slant_x <dbl>,
#   slant_y <dbl>, slant_down_x <dbl>, slant_down_y <dbl>, slant_up_x <dbl>,
#   slant_up_y <dbl>, star_x <dbl>, star_y <dbl>, v_lines_x <dbl>,
#   v_lines_y <dbl>, wide_lines_x <dbl>, wide_lines_y <dbl>, x_shape_x <dbl>,
#   x_shape_y <dbl>
```

Obtain info

```
?datasaurus_dozen_wide
```

Part 1

```
summary(datasaurus_dozen_wide)
```

	away_x	away_y	bullseye_x	bullseye_y
Min.	:15.56	: 0.01512	:19.29	: 9.692
1st Qu.	:39.72	:24.62589	:41.63	:26.245
Median	:53.34	:47.53527	:53.84	:47.383
Mean	:54.27	:47.83472	:54.27	:47.831
3rd Qu.	:69.15	:71.80315	:64.80	:72.533
Max.	:91.64	:97.47577	:91.74	:85.876
	circle_x	circle_y	dino_x	dino_y
Min.	:21.86	:16.33	:22.31	: 2.949
1st Qu.	:43.38	:18.35	:44.10	:25.288
Median	:54.02	:51.03	:53.33	:46.026
Mean	:54.27	:47.84	:54.26	:47.832
3rd Qu.	:64.97	:77.78	:64.74	:68.526
Max.	:85.66	:85.58	:98.21	:99.487
	dots_x	dots_y	h_lines_x	h_lines_y
Min.	:25.44	:15.77	:22.00	:10.46
1st Qu.	:50.36	:17.11	:42.29	:30.48
Median	:50.98	:51.30	:53.07	:50.47
Mean	:54.26	:47.84	:54.26	:47.83
3rd Qu.	:75.20	:82.88	:66.77	:70.35
Max.	:77.95	:94.25	:98.29	:90.46
	high_lines_x	high_lines_y	slant_down_x	slant_down_y
Min.	:17.89	:14.91	:18.11	: 0.3039
1st Qu.	:41.54	:22.92	:42.89	:27.8409
Median	:54.17	:32.50	:53.14	:46.4013
Mean	:54.27	:47.84	:54.27	:47.8359
3rd Qu.	:63.95	:75.94	:64.47	:68.4394
Max.	:96.08	:87.15	:95.59	:99.6442
	slant_up_x	slant_up_y	star_x	star_y
Min.	:20.21	: 5.646	:27.02	:14.37
1st Qu.	:42.81	:24.756	:41.03	:20.37
Median	:54.26	:45.292	:56.53	:50.11
Mean	:54.27	:47.831	:54.27	:47.84

3rd Qu.:64.49	3rd Qu.:70.856	3rd Qu.:68.71	3rd Qu.:63.55
Max. :95.26	Max. :99.580	Max. :86.44	Max. :92.21
v_lines_x	v_lines_y	wide_lines_x	wide_lines_y
Min. :30.45	Min. : 2.735	Min. :27.44	Min. : 0.217
1st Qu.:49.96	1st Qu.:22.753	1st Qu.:35.52	1st Qu.:24.347
Median :50.36	Median :47.114	Median :64.55	Median :46.279
Mean :54.27	Mean :47.837	Mean :54.27	Mean :47.832
3rd Qu.:69.50	3rd Qu.:65.845	3rd Qu.:67.45	3rd Qu.:67.568
Max. :89.50	Max. :99.695	Max. :77.92	Max. :99.284
x_shape_x	x_shape_y		
Min. :31.11	Min. : 4.578		
1st Qu.:40.09	1st Qu.:23.471		
Median :47.14	Median :39.876		
Mean :54.26	Mean :47.840		
3rd Qu.:71.86	3rd Qu.:73.610		
Max. :85.45	Max. :97.838		

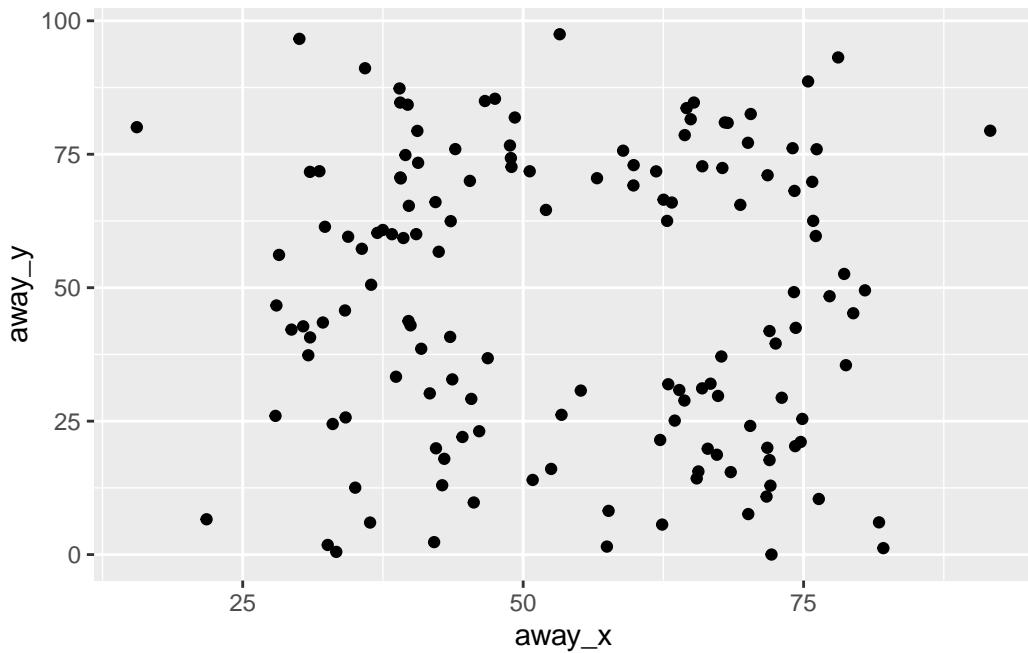
```
glimpse(datasaurus_dozen_wide)
```

Rows: 142
 Columns: 26

\$ away_x	<dbl> 32.33111, 53.42146, 63.92020, 70.28951, 34.11883, 67.6707~
\$ away_y	<dbl> 61.411101, 26.186880, 30.832194, 82.533649, 45.734551, 37~
\$ bullseye_x	<dbl> 51.20389, 58.97447, 51.87207, 48.17993, 41.68320, 37.8904~
\$ bullseye_y	<dbl> 83.33978, 85.49982, 85.82974, 85.04512, 84.01794, 82.5674~
\$ circle_x	<dbl> 55.99303, 50.03225, 51.28846, 51.17054, 44.37791, 45.0102~
\$ circle_y	<dbl> 79.27726, 79.01307, 82.43594, 79.16529, 78.16463, 77.8808~
\$ dino_x	<dbl> 55.3846, 51.5385, 46.1538, 42.8205, 40.7692, 38.7179, 35.~
\$ dino_y	<dbl> 97.1795, 96.0256, 94.4872, 91.4103, 88.3333, 84.8718, 79.~
\$ dots_x	<dbl> 51.14792, 50.51713, 50.20748, 50.06948, 50.56285, 50.2885~
\$ dots_y	<dbl> 90.86741, 89.10239, 85.46005, 83.05767, 82.93782, 82.9752~
\$ h_lines_x	<dbl> 53.36657, 52.80198, 47.05413, 42.44843, 42.70404, 32.3789~
\$ h_lines_y	<dbl> 90.20803, 90.08806, 90.45894, 89.50770, 90.44263, 90.1441~
\$ high_lines_x	<dbl> 57.61323, 51.27439, 50.75390, 37.02118, 42.88176, 37.1557~
\$ high_lines_y	<dbl> 83.90517, 82.81798, 76.75413, 81.95447, 80.18477, 84.9541~
\$ slant_down_x	<dbl> 52.87202, 59.01414, 56.37511, 37.83920, 39.88537, 44.0774~
\$ slant_down_y	<dbl> 97.34322, 93.57487, 96.30515, 94.35944, 90.63466, 84.1258~
\$ slant_up_x	<dbl> 47.69520, 44.60998, 43.85638, 41.57893, 49.17742, 42.6522~
\$ slant_up_y	<dbl> 95.24119, 93.07584, 94.08587, 90.30357, 96.61053, 90.5606~
\$ star_x	<dbl> 58.21361, 58.19605, 58.71823, 57.27837, 58.08202, 57.4894~
\$ star_y	<dbl> 91.88189, 92.21499, 90.31053, 89.90761, 92.00815, 88.0852~
\$ v_lines_x	<dbl> 50.48151, 50.28241, 50.18670, 50.32691, 50.45621, 30.4648~

```
$ v_lines_y      <dbl> 93.22270, 97.60998, 99.69468, 90.02205, 89.98741, 82.0892~  
$ wide_lines_x <dbl> 65.81554, 65.67227, 39.00272, 37.79530, 35.51390, 39.2194~  
$ wide_lines_y <dbl> 95.58837, 91.93340, 92.26184, 93.53246, 89.59919, 83.5434~  
$ x_shape_x     <dbl> 38.33776, 35.75187, 32.76722, 33.72961, 37.23825, 36.0272~  
$ x_shape_y     <dbl> 92.47272, 94.11677, 88.51829, 88.62227, 83.72493, 82.0407~
```

```
# Command + shift + m: %>% --> pipe operator  
datasaurus_dozen_wide %>%  
  ggplot() +  
    geom_point(mapping = aes(x = away_x, y = away_y))
```



Tidying Data

column = variable and row = observation

```
df <- datasaurus_dozen_wide %>%  
  mutate(id = row_number()) %>%  
  pivot_longer(cols = -id,  
               names_to = c("dataset", "var_name"),  
               values_to = "value",  
               names_pattern = "^(.*)(.*)$") %>%
```

```
pivot_wider(names_from = var_name, values_from = value) %>%
  select(-id)
```

Summary statistics

```
# names of the dataset
df %>%
  distinct(dataset)
```

```
# A tibble: 13 x 1
  dataset
  <chr>
  1 away
  2 bullseye
  3 circle
  4 dino
  5 dots
  6 h_lines
  7 high_lines
  8 slant_down
  9 slant_up
 10 star
 11 v_lines
 12 wide_lines
 13 x_shape
```

```
dataset_list <- c("dino", "star", "bullseye", "away")
```

```
df %>%
  filter(dataset == "dino") %>%
  summarise(mean_x = mean(x),
            sd_x = sd(x))
```

```
# A tibble: 1 x 2
  mean_x   sd_x
  <dbl> <dbl>
1     54.3  16.8
```

```
df %>%
  filter(dataset %in% dataset_list) %>%
  group_by(dataset) %>%
  summarise(across(c(x,y), list(mean = mean, sd = sd)))
```

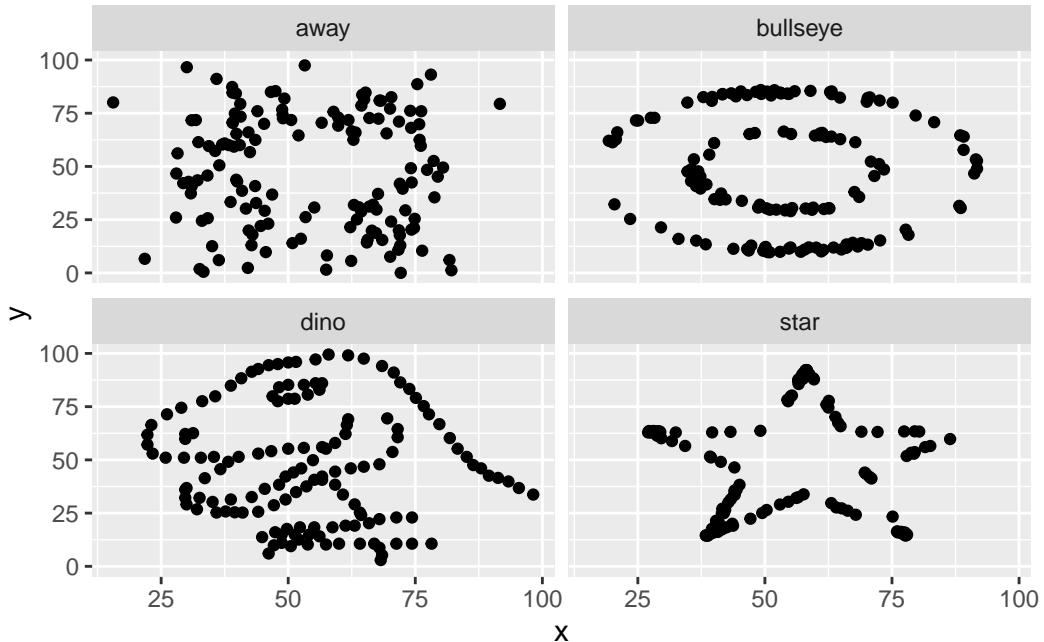
```
# A tibble: 4 x 5
  dataset   x_mean   x_sd   y_mean   y_sd
  <chr>     <dbl>   <dbl>   <dbl>   <dbl>
1 away      54.3    16.8    47.8    26.9
2 bullseye  54.3    16.8    47.8    26.9
3 dino      54.3    16.8    47.8    26.9
4 star      54.3    16.8    47.8    26.9
```

```
df %>%
  filter(dataset %in% dataset_list) %>%
  group_by(dataset) %>%
  summarise(across(c(x, y), list(mean = mean, sd = sd)),
            slope = coef(lm(y ~ x))[2])
```

```
# A tibble: 4 x 6
  dataset   x_mean   x_sd   y_mean   y_sd   slope
  <chr>     <dbl>   <dbl>   <dbl>   <dbl>   <dbl>
1 away      54.3    16.8    47.8    26.9 -0.103
2 bullseye  54.3    16.8    47.8    26.9 -0.110
3 dino      54.3    16.8    47.8    26.9 -0.104
4 star      54.3    16.8    47.8    26.9 -0.101
```

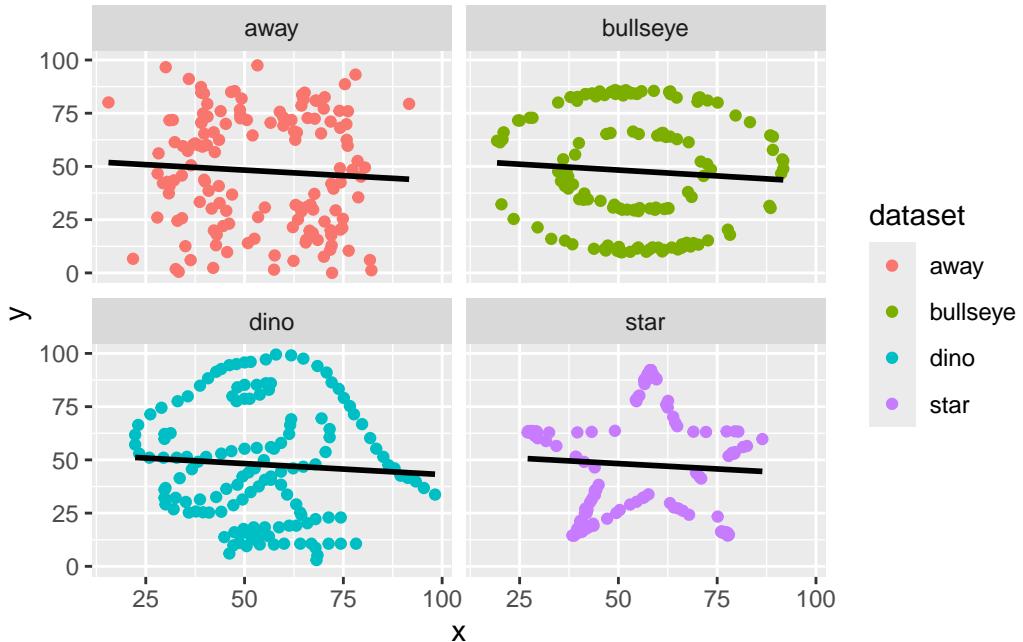
Plotting

```
df %>%
  filter(dataset %in% dataset_list) %>%
  ggplot() +
  geom_point(aes(x, y)) +
  facet_wrap(vars(dataset))
```



```
df %>%
  filter(dataset %in% dataset_list) %>%
  ggplot(aes(x, y, color = dataset)) +
  geom_point() +
  geom_smooth(method = "lm", se = FALSE, color = "black") +
  facet_wrap(vars(dataset))

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



Part 2: The Titanic Dataset

```
# use assignment side
titanic <- read_csv("/Users/ingridcanelles/Documents/GitHub/statcomp/datasets/titanic.csv")
```

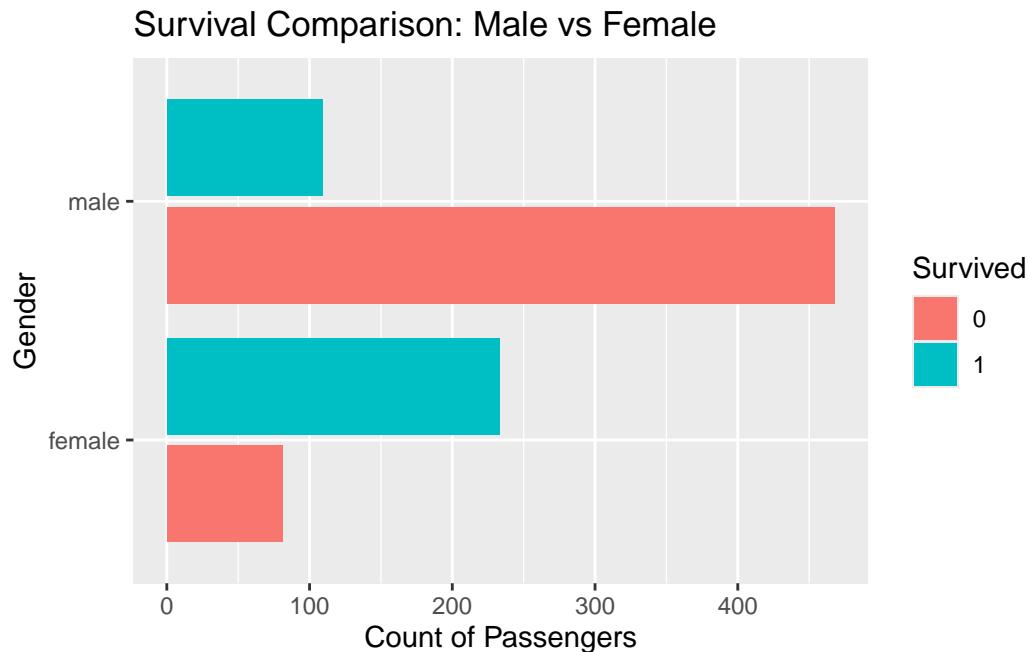
```
Rows: 891 Columns: 12
-- Column specification -----
Delimiter: ","
chr (5): Name, Sex, Ticket, Cabin, Embarked
dbl (7): PassengerId, Survived, Pclass, Age, SibSp, Parch, Fare

i Use `spec()` to retrieve the full column specification for this data.
i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

Gender vs survival

```
titanic = titanic %>%
  mutate(across(c(Survived, Pclass, Sex), as.factor))
```

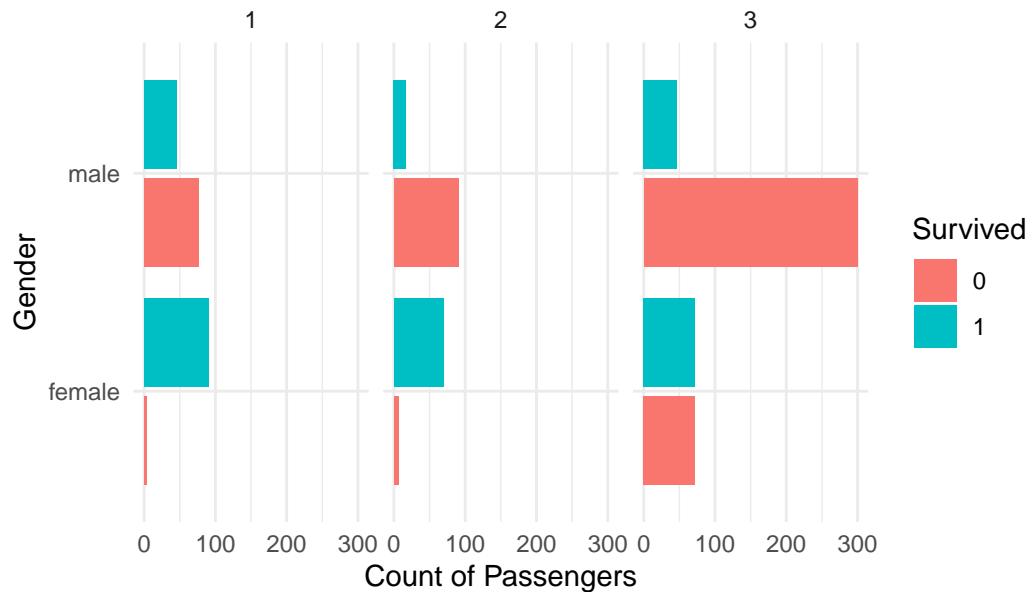
```
titanic %>%
  ggplot() +
  geom_bar(aes(y = Sex, fill = Survived), position = "dodge2") +
  labs(title = "Survival Comparison: Male vs Female",
       x = "Count of Passengers",
       y = "Gender")
```



Survival by gender and class

```
titanic %>%
  ggplot() +
  geom_bar(aes(y = Sex, fill = Survived), position = "dodge2") +
  facet_wrap(~Pclass) +
  theme_minimal() +
  labs(title = "Survival by Gender across Passenger Classes",
       x = "Count of Passengers",
       y = "Gender")
```

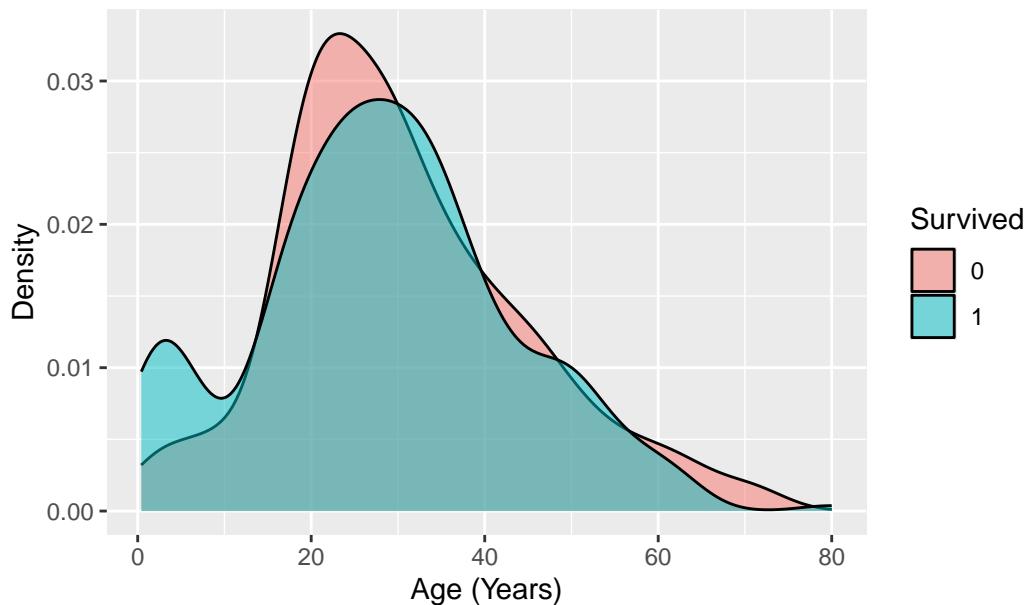
Survival by Gender across Passenger Classes



Age distribution

```
titanic %>%
  filter(!is.na(Age)) %>%
  ggplot(aes(x = Age, fill = Survived)) +
  geom_density(alpha = 0.5) +
  labs(title = "Age Distribution of Survivors vs Non-Survivors",
       x = "Age (Years)",
       y = "Density")
```

Age Distribution of Survivors vs Non-Survivors



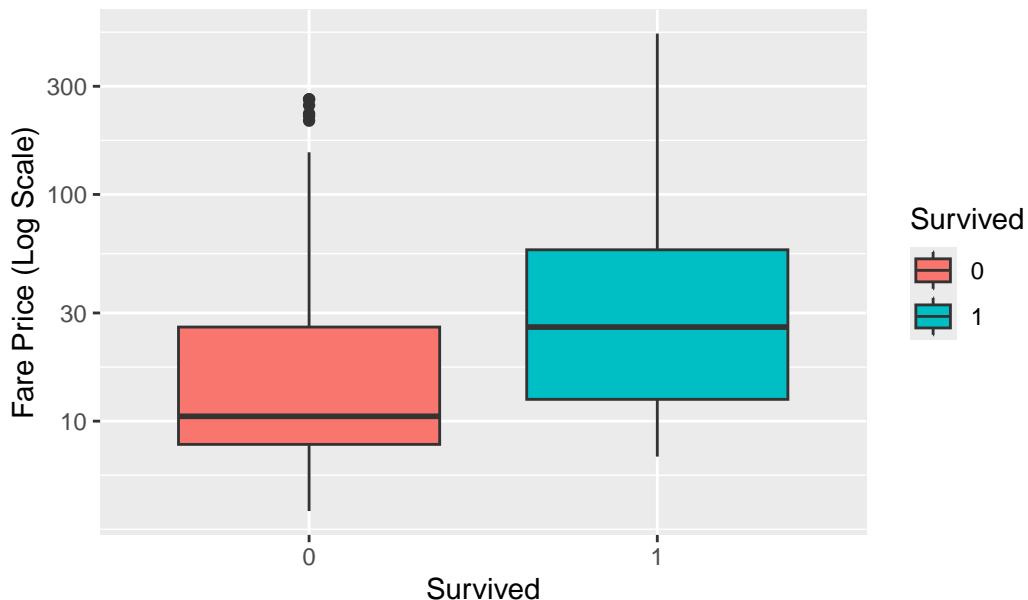
Fare and survival

```
titanic %>%
  ggplot(aes(x = Survived, y = Fare, fill = Survived)) +
  geom_boxplot() +
  scale_y_log10() +
  labs(title = "Survival Outcome based on Ticket Fare",
       y = "Fare Price (Log Scale)",
       x = "Survived")
```

Warning in scale_y_log10(): log-10 transformation introduced infinite values.

Warning: Removed 15 rows containing non-finite outside the scale range
(`stat_boxplot()`).

Survival Outcome based on Ticket Fare



Interpretation

- Women were much more likely to survive than men.
- Passengers in higher classes showed higher survival rates.
- Higher ticket fares were associated with a greater chance of survival.
- Third class passengers, especially men, had the lowest survival probability.
- Overall, gender and social class strongly influenced survival.