

# illnesses.qmd

```
library(tidyverse)
```

```
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr      1.1.3      v readr      2.1.4
v forcats    1.0.0      v stringr    1.5.0
v ggplot2    3.4.3      v tibble     3.2.1
v lubridate  1.9.2      v tidyr      1.3.0
v 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 become
```

```
library(tidymodels)
```

```
-- Attaching packages ----- tidymodels 1.1.1 --
v broom      1.0.5      v rsample     1.2.0
v dials      1.2.0      v tune        1.1.2
v infer      1.0.5      v workflows   1.1.3
v modeldata  1.2.0      v workflowsets 1.0.1
v parsnip    1.1.1      v yardstick   1.2.0
v recipes    1.0.8
-- Conflicts ----- tidymodels_conflicts() --
x scales::discard() masks purrr::discard()
x dplyr::filter()   masks stats::filter()
x recipes::fixed()  masks stringr::fixed()
x dplyr::lag()       masks stats::lag()
x yardstick::spec() masks readr::spec()
x recipes::step()    masks stats::step()
* Dig deeper into tidy modeling with R at https://www.tmw.org
```

```
covid <- read_csv("Covid Data.csv")
```

```
Rows: 1048575 Columns: 21
```

```
-- Column specification -----
```

```
Delimiter: ","
```

```
chr (1): DATE_DIED
```

```
dbl (20): USMER, MEDICAL_UNIT, SEX, PATIENT_TYPE, INTUBED, PNEUMONIA, AGE, P...
```

```
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.
```

```
# 0 = did not die
```

```
# 1 = died
```

```
covid_health <- covid |>
```

```
  mutate(
```

```
    died = if_else(DATE_DIED == "9999-99-99", 0, 1)
```

```
  ) |>
```

```
  filter(PNEUMONIA != 97 & PNEUMONIA != 99 & PNEUMONIA != 98) |>
```

```
  filter(DIABETES != 97 & DIABETES != 99 & DIABETES != 98) |>
```

```
  filter(INMSUPR != 97 & INMSUPR != 99 & INMSUPR != 98) |>
```

```
  filter(HIPERTENSION != 97 & HIPERTENSION != 99 & HIPERTENSION != 98) |>
```

```
  filter(OTHER_DISEASE != 97 & OTHER_DISEASE != 99 & OTHER_DISEASE != 98) |>
```

```
  filter(CARDIOVASCULAR != 97 & CARDIOVASCULAR != 99 & CARDIOVASCULAR != 98) |>
```

```
  filter(RENAL_CHRONIC != 97 & RENAL_CHRONIC != 99 & RENAL_CHRONIC != 98) |>
```

```
  filter(ASTHMA != 97 & ASTHMA != 99 & ASTHMA != 98) |>
```

```
  filter(AGE != 97 & AGE != 99 & AGE != 98) |>
```

```
  filter(SEX != 97 & SEX != 99 & SEX != 98)
```

```
covid_health
```

```
# A tibble: 1,025,722 x 22
```

	USMER	MEDICAL_UNIT	SEX	PATIENT_TYPE	DATE_DIED	INTUBED	PNEUMONIA	AGE
	<dbl>	<dbl>	<dbl>	<dbl>	<chr>	<dbl>	<dbl>	<dbl>
1	2	1	1	1	03/05/2020	97	1	65
2	2	1	2	1	03/06/2020	97	1	72
3	2	1	2	2	09/06/2020	1	2	55
4	2	1	1	1	12/06/2020	97	2	53
5	2	1	2	1	21/06/2020	97	2	68
6	2	1	1	2	9999-99-99	2	1	40
7	2	1	1	1	9999-99-99	97	2	64

```

8      2      1      1      1 9999-99-99      97      1      64
9      2      1      1      2 9999-99-99      2      2      37
10     2      1      1      2 9999-99-99      2      2      25
# i 1,025,712 more rows
# i 14 more variables: PREGNANT <dbl>, DIABETES <dbl>, COPD <dbl>,
#   ASTHMA <dbl>, INMSUPR <dbl>, HIPERTENSION <dbl>, OTHER_DISEASE <dbl>,
#   CARDIOVASCULAR <dbl>, OBESITY <dbl>, RENAL_CHRONIC <dbl>, TOBACCO <dbl>,
#   CLASIFFICATION_FINAL <dbl>, ICU <dbl>, died <dbl>

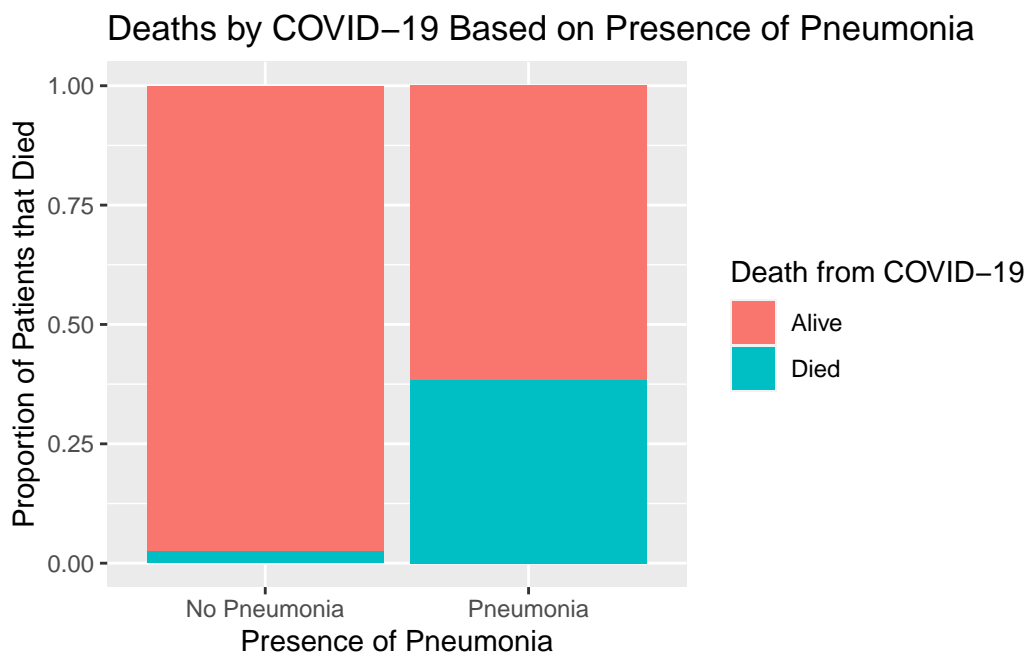
```

```
# pneumonia specifically influential bc it's a lung disease
```

```

covid_health |>
  mutate(PNEUMONIA = if_else(PNEUMONIA == 1, "Pneumonia", "No Pneumonia")) |>
  mutate(died = if_else(died == 1, "Died", "Alive")) |>
  ggplot(aes(x = PNEUMONIA, fill = died)) +
  geom_bar(position = "fill") +
  labs(title = "Deaths by COVID-19 Based on Presence of Pneumonia", y = "Proportion of Patients that Died")

```



```

covid_long <- covid_health |>
  select(PNEUMONIA, DIABETES, ASTHMA, INMSUPR, HIPERTENSION, OTHER_DISEASE, CARDIOVASCULAR, OBESITY, RENAL_CHRONIC, TOBACCO, CLASIFFICATION_FINAL, ICU, died)
  pivot_longer(

```

```

  cols = c('PNEUMONIA', 'DIABETES', 'ASTHMA', 'INMSUPR', 'HIPERTENSION', 'OTHER_DISEASE')
  names_to = "Health Condition",
  values_to = "Presence"
) |>
mutate(Presence = if_else(Presence == 1, "Yes", "No"))

```

covid\_long

```

# A tibble: 8,205,776 x 3
  died `Health Condition` Presence
  <dbl> <chr>             <chr>
1     1 1 PNEUMONIA          Yes
2     1 1 DIABETES           No
3     1 1 ASTHMA             No
4     1 1 INMSUPR            No
5     1 1 HIPERTENSION        Yes
6     1 1 OTHER_DISEASE       No
7     1 1 CARDIOVASCULAR      No
8     1 1 RENAL_CHRONIC       No
9     1 1 PNEUMONIA          Yes
10    1 1 DIABETES           No
# i 8,205,766 more rows

```

```

logit_mod_health <- glm(died ~ as.factor(PNEUMONIA) + as.factor(DIABETES) + as.factor(ASTHMA),
  data = covid_health,
  family = "binomial")

```

tidy(logit\_mod\_health)

```

# A tibble: 8 x 5
  term                estimate std.error statistic  p.value
  <chr>              <dbl>    <dbl>    <dbl>    <dbl>
1 (Intercept)        1.02     0.0464     22.0 2.13e-107
2 as.factor(PNEUMONIA)2 -2.94    0.00907   -324. 0
3 as.factor(DIABETES)2 -0.717   0.0107    -67.2 0
4 as.factor(ASTHMA)2    0.514   0.0308     16.7 1.33e- 62
5 as.factor(INMSUPR)2  -0.285   0.0277    -10.3 8.55e- 25
6 as.factor(HIPERTENSION)2 -0.749   0.0103    -72.7 0
7 as.factor(CARDIOVASCULAR)2 -0.230   0.0224    -10.3 8.32e- 25
8 as.factor(RENAL_CHRONIC)2 -0.555   0.0216    -25.7 1.22e-145

```

```
logit_health_aug <- augment(logit_mod_health)
```

```
logit_health_aug
```

```
# A tibble: 1,025,722 x 14
```

	died	`as.factor(PNEUMONIA)`	`as.factor(DIABETES)`	`as.factor(ASTHMA)`
	<dbl>	<fct>	<fct>	<fct>
1	1	1	2	2
2	1	1	2	2
3	1	2	1	2
4	1	2	2	2
5	1	2	1	2
6	0	1	2	2
7	0	2	2	2
8	0	1	1	2
9	0	2	1	2
10	0	2	2	2

```
# i 1,025,712 more rows
```

```
# i 10 more variables: `as.factor(INMSUPR)` <fct>,  
# `as.factor(HIPERTENSION)` <fct>, `as.factor(CARDIOVASCULAR)` <fct>,  
# `as.factor(RENAL_CHRONIC)` <fct>, .fitted <dbl>, .resid <dbl>, .hat <dbl>,  
# .sigma <dbl>, .cooksd <dbl>, .std.resid <dbl>
```

```
logit_health_aug <- logit_health_aug |>  
  mutate(prob = exp(.fitted)/(1 + exp(.fitted)),  
         pred_died = ifelse(prob > 0.5, "Died", "Did Not Die")) %>%  
  select(.fitted, prob, pred_died, died)
```

```
logit_health_aug
```

```
# A tibble: 1,025,722 x 4
```

	.fitted	prob	pred_died	died
	<dbl>	<dbl>	<chr>	<dbl>
1	-0.252	0.437	Did Not Die	1
2	0.303	0.575	Died	1
3	-3.22	0.0383	Did Not Die	1
4	-3.94	0.0191	Did Not Die	1
5	-2.47	0.0777	Did Not Die	1
6	-1.00	0.269	Did Not Die	0
7	-3.94	0.0191	Did Not Die	0

```

8    1.31  0.787  Died           0
9   -2.47  0.0777 Did Not Die    0
10  -3.94  0.0191 Did Not Die    0
# i 1,025,712 more rows

```

```
table(logit_health_aug$pred_died, logit_health_aug$died)
```

```

           0      1
Did Not Die 937687 61126
Died        13203 13706

```

```
##DEMOGRAPHICS:
```

```

covid_health$PREGNANT <- NULL
covid_health

```

```
# A tibble: 1,025,722 x 21
```

	USMER	MEDICAL_UNIT	SEX	PATIENT_TYPE	DATE_DIED	INTUBED	PNEUMONIA	AGE
	<dbl>	<dbl>	<dbl>	<dbl>	<chr>	<dbl>	<dbl>	<dbl>
1	2	1	1	1	03/05/2020	97	1	65
2	2	1	2	1	03/06/2020	97	1	72
3	2	1	2	2	09/06/2020	1	2	55
4	2	1	1	1	12/06/2020	97	2	53
5	2	1	2	1	21/06/2020	97	2	68
6	2	1	1	2	9999-99-99	2	1	40
7	2	1	1	1	9999-99-99	97	2	64
8	2	1	1	1	9999-99-99	97	1	64
9	2	1	1	2	9999-99-99	2	2	37
10	2	1	1	2	9999-99-99	2	2	25

```
# i 1,025,712 more rows
```

```

# i 13 more variables: DIABETES <dbl>, COPD <dbl>, ASTHMA <dbl>, INMSUPR <dbl>,
#   HIPERTENSION <dbl>, OTHER_DISEASE <dbl>, CARDIOVASCULAR <dbl>,
#   OBESITY <dbl>, RENAL_CHRONIC <dbl>, TOBACCO <dbl>,
#   CLASIFFICATION_FINAL <dbl>, ICU <dbl>, died <dbl>

```

```

covid_health<- covid_health |>
mutate(DIED = ifelse(DATE_DIED == '9999-99-99', 0, 1)) |>
mutate(SEX = as.factor(SEX)) |>

```

```
mutate(DIED = as.factor(DIED))
covid_health
```

```
# A tibble: 1,025,722 x 22
  USMER MEDICAL_UNIT SEX PATIENT_TYPE DATE_DIED INTUBED PNEUMONIA AGE
  <dbl> <dbl> <fct> <dbl> <chr> <dbl> <dbl> <dbl>
1 2 1 1 1 03/05/2020 97 1 65
2 2 1 2 1 03/06/2020 97 1 72
3 2 1 2 2 09/06/2020 1 2 55
4 2 1 1 1 12/06/2020 97 2 53
5 2 1 2 1 21/06/2020 97 2 68
6 2 1 1 2 9999-99-99 2 1 40
7 2 1 1 1 9999-99-99 97 2 64
8 2 1 1 1 9999-99-99 97 1 64
9 2 1 1 2 9999-99-99 2 2 37
10 2 1 1 2 9999-99-99 2 2 25
# i 1,025,712 more rows
# i 14 more variables: DIABETES <dbl>, COPD <dbl>, ASTHMA <dbl>, INMSUPR <dbl>,
# HIPERTENSION <dbl>, OTHER_DISEASE <dbl>, CARDIOVASCULAR <dbl>,
# OBESITY <dbl>, RENAL_CHRONIC <dbl>, TOBACCO <dbl>,
# CLASIFFICATION_FINAL <dbl>, ICU <dbl>, died <dbl>, DIED <fct>
```

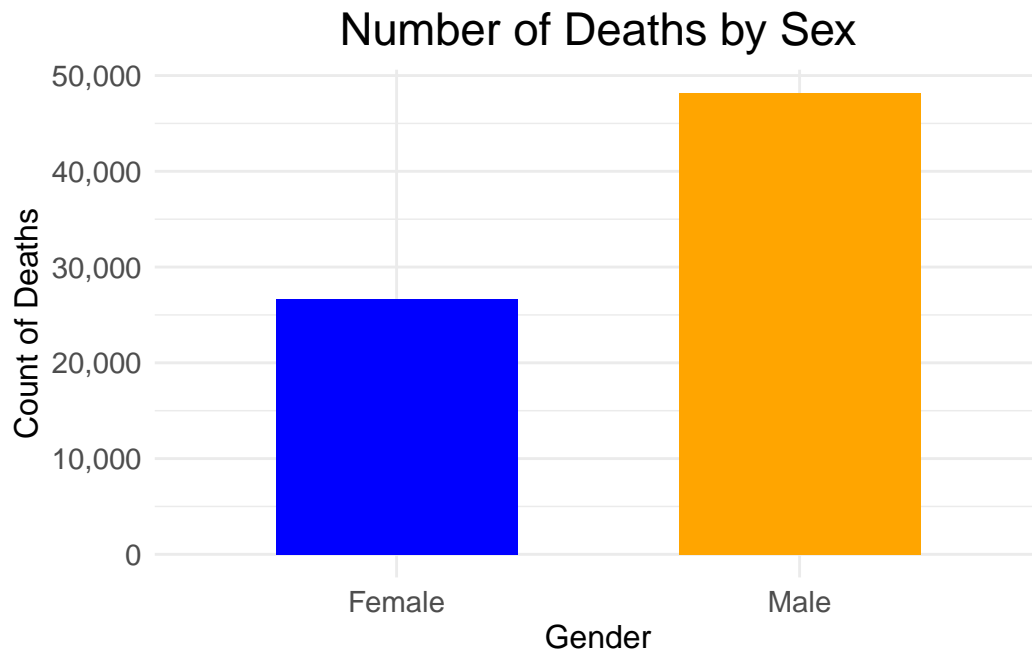
```
covid_health$SEX <- factor(covid_health$SEX, levels = c(1, 2), labels = c("Female", "Male"))
covid_health
```

```
# A tibble: 1,025,722 x 22
  USMER MEDICAL_UNIT SEX PATIENT_TYPE DATE_DIED INTUBED PNEUMONIA AGE
  <dbl> <dbl> <fct> <dbl> <chr> <dbl> <dbl> <dbl>
1 2 1 Female 1 03/05/2020 97 1 65
2 2 1 Male 1 03/06/2020 97 1 72
3 2 1 Male 2 09/06/2020 1 2 55
4 2 1 Female 1 12/06/2020 97 2 53
5 2 1 Male 1 21/06/2020 97 2 68
6 2 1 Female 2 9999-99-99 2 1 40
7 2 1 Female 1 9999-99-99 97 2 64
8 2 1 Female 1 9999-99-99 97 1 64
9 2 1 Female 2 9999-99-99 2 2 37
10 2 1 Female 2 9999-99-99 2 2 25
# i 1,025,712 more rows
# i 14 more variables: DIABETES <dbl>, COPD <dbl>, ASTHMA <dbl>, INMSUPR <dbl>,
```

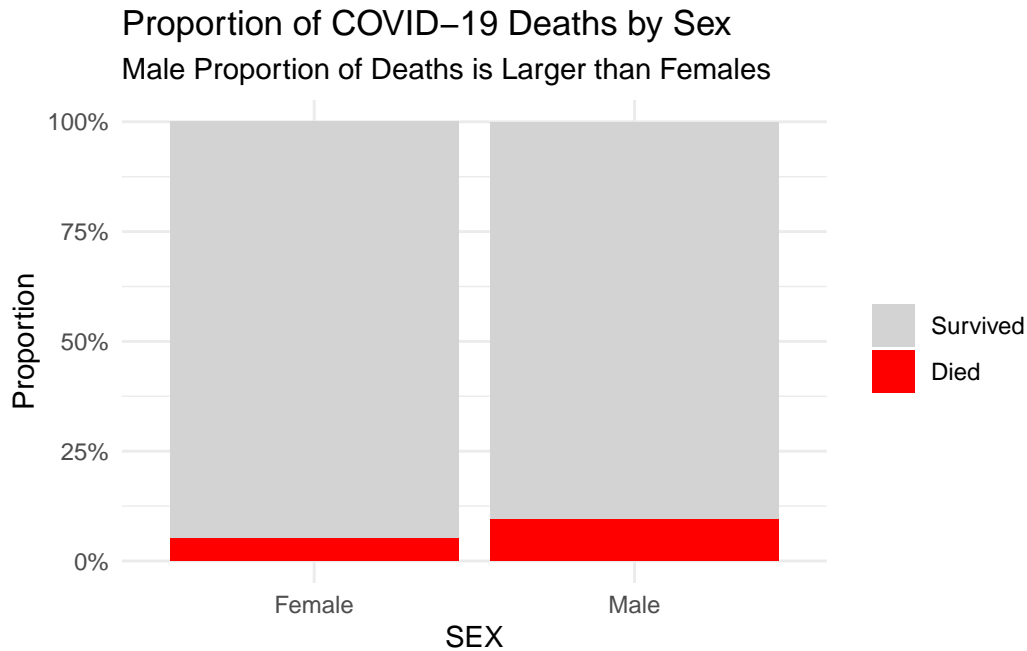
```
# HIPERTENSION <dbl>, OTHER_DISEASE <dbl>, CARDIOVASCULAR <dbl>,
# OBESITY <dbl>, RENAL_CHRONIC <dbl>, TOBACCO <dbl>,
# CLASIFFICATION_FINAL <dbl>, ICU <dbl>, died <dbl>, DIED <fct>
```

```
library(scales)
covid_health |>
filter(DIED == 1) |>
ggplot(aes(x = SEX, fill = SEX)) +
geom_bar(stat = "count", width = 0.6) + # Adjust bar width for aesthetics
scale_fill_manual(values = c("blue", "orange")) + # Change colors for clarity
labs(
title = "Number of Deaths by Sex",
x = "Gender",
y = "Count of Deaths"
) +
theme_minimal() +
theme(
text = element_text(size = 14), # Adjust text size for better readability
plot.title = element_text(hjust = 0.5), # Center the plot title
axis.title = element_text(size = 12), # Specify axis title size
legend.position = "none" # Remove legend if redundant
) +
scale_y_continuous(labels = comma)
```





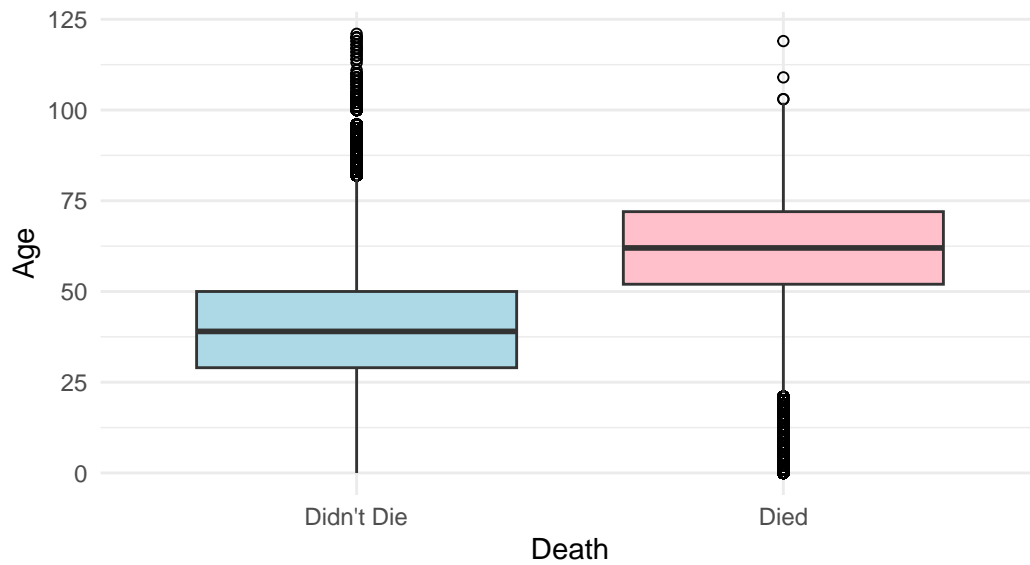
```
covid_summary <- covid_health |>
count(SEX, DIED) |>
group_by(SEX) |>
mutate(Proportion = n / sum(n))
# Plot
ggplot(covid_summary, aes(x = SEX, y = Proportion, fill = DIED)) +
  geom_col() +
  scale_y_continuous(labels = scales::percent_format()) +
  labs(
    title = "Proportion of COVID-19 Deaths by Sex",
    subtitle = "Male Proportion of Deaths is Larger than Females",
    x = "SEX",
    y = "Proportion"
  ) +
  scale_fill_manual(values = c("0" = "lightgrey", "1" = "red"),
    labels = c("Survived", "Died")) +
  theme_minimal() +
  theme(legend.title = element_blank())
```



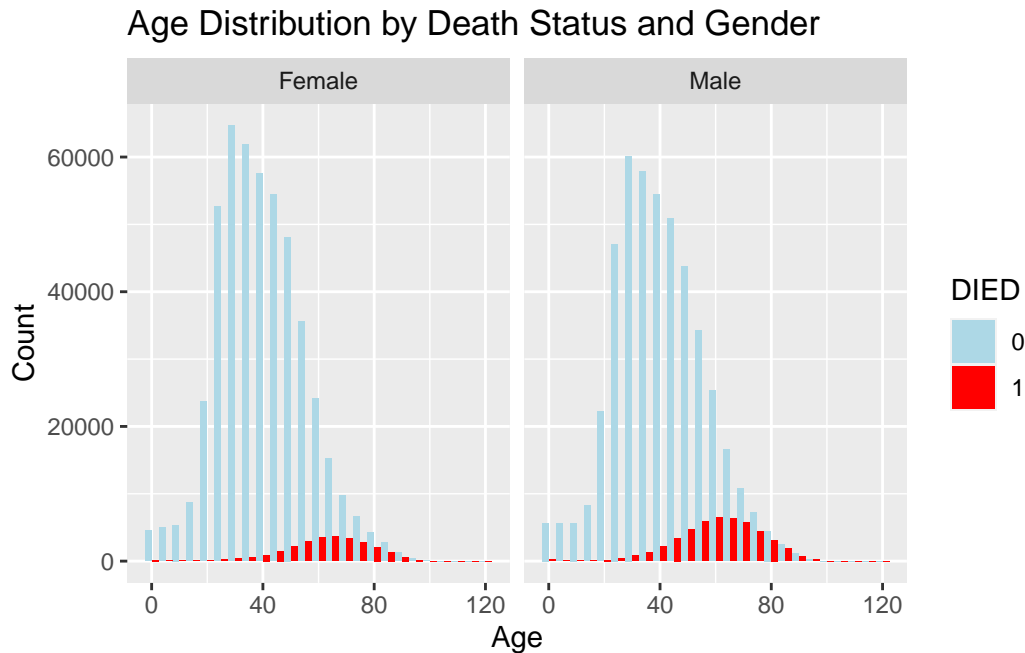
```
covid_health |>
mutate(DIED = factor(DIED, levels = c(0, 1), labels = c("Didn't Die", "Died"))) |>
ggplot(aes(x = DIED, y = AGE, fill = DIED)) +
geom_boxplot(outlier.colour = "black", outlier.shape = 1) +
labs(title = "COVID-19 Death vs. Age Correlation",
      subtitle = "Median age of the deceased patients is greater than that of the survived", x =
y = "Age") +
scale_fill_manual(values = c("Didn't Die" = "lightblue", "Died" = "pink")) +
theme_minimal() +
theme(legend.position = "none")
```

## COVID-19 Death vs. Age Correlation

Median age of the deceased patients is greater than that of the survived



```
ggplot(covid_health, aes(x = AGE, fill = DIED)) +  
geom_histogram(binwidth = 5, position = "dodge") +  
facet_wrap(~SEX) +  
labs(title = "Age Distribution by Death Status and Gender",  
x = "Age",  
y = "Count") +  
scale_fill_manual(values = c("0" = "lightblue", "1" = "red"))
```



```
logit_mod_demo <- glm(DIED ~ as.factor(SEX) + AGE, data = covid_health, family = "binomial")
tidy(logit_mod_demo)
```

# A tibble: 3 x 5

term	estimate	std.error	statistic	p.value
<chr>	<dbl>	<dbl>	<dbl>	<dbl>
1 (Intercept)	-6.75	0.0168	-401.	0
2 as.factor(SEX)Male	0.638	0.00852	74.8	0
3 AGE	0.0761	0.000263	289.	0

SEX male: Holding age constant, we predict the odds of a male patient in the COVID-19 dataset passing away to be around  $e^{0.63753001(1.8918)}$  times that of a female patient. AGE: Holding sex constant, we predict that for each additional year in age of the patient, the odds of passing away are multiplied by  $e^{0.07605202(1.0790)}$ .

```
logit_demo_aug <- augment(logit_mod_demo)
logit_demo_aug
```

# A tibble: 1,025,722 x 9

DIED	as.factor(SEX)	AGE	.fitted	.resid	.hat	.sigma	.cooksd	.std.resid
------	----------------	-----	---------	--------	------	--------	---------	------------

	<fct>	<fct>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
1	1	Female	65	-1.81	1.98	5.80e-6	0.644	1.18e-5	1.98
2	1	Male	72	-0.637	1.46	9.62e-6	0.644	6.07e-6	1.46
3	1	Male	55	-1.93	2.03	3.06e-6	0.644	7.04e-6	2.03
4	1	Female	53	-2.72	2.36	2.74e-6	0.644	1.39e-5	2.36
5	1	Male	68	-0.942	1.59	7.15e-6	0.644	6.11e-6	1.59
6	0	Female	40	-3.71	-0.220	1.62e-6	0.644	1.32e-8	-0.220
7	0	Female	64	-1.88	-0.532	5.40e-6	0.644	2.74e-7	-0.532
8	0	Female	64	-1.88	-0.532	5.40e-6	0.644	2.74e-7	-0.532
9	0	Female	37	-3.94	-0.197	1.46e-6	0.644	9.47e-9	-0.197
10	0	Female	25	-4.85	-0.125	9.52e-7	0.644	2.49e-9	-0.125

# i 1,025,712 more rows

```
logit_demo_aug <- logit_demo_aug |>
mutate(prob = exp(.fitted)/(1 + exp(.fitted)),
pred_died = ifelse(prob > 0.5, "Died", "Did Not Die")) |>
select(.fitted, prob, pred_died, DIED)
logit_demo_aug
```

# A tibble: 1,025,722 x 4

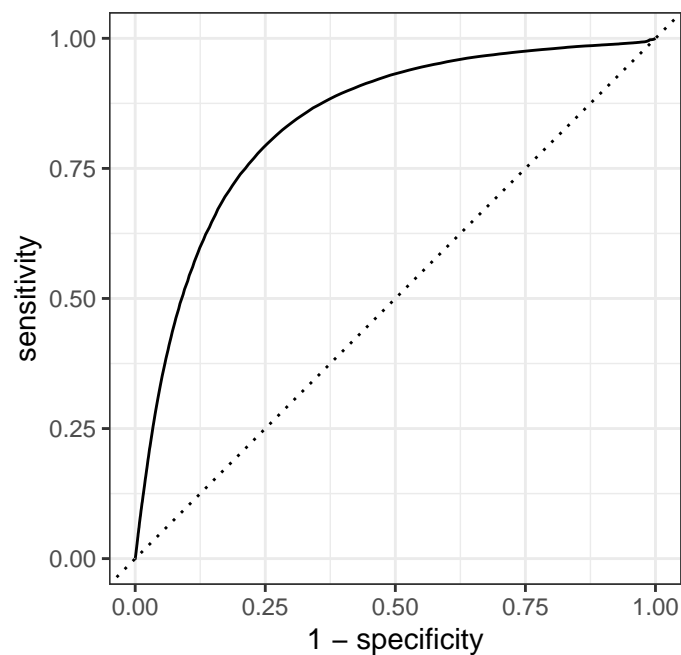
	.fitted	prob	pred_died	DIED
	<dbl>	<dbl>	<chr>	<fct>
1	-1.81	0.141	Did Not Die	1
2	-0.637	0.346	Did Not Die	1
3	-1.93	0.127	Did Not Die	1
4	-2.72	0.0618	Did Not Die	1
5	-0.942	0.281	Did Not Die	1
6	-3.71	0.0239	Did Not Die	0
7	-1.88	0.132	Did Not Die	0
8	-1.88	0.132	Did Not Die	0
9	-3.94	0.0191	Did Not Die	0
10	-4.85	0.00777	Did Not Die	0

# i 1,025,712 more rows

```
logit_demo_aug |>
roc_auc(
truth = DIED,
prob, event_level = "second"
)
```

```
# A tibble: 1 x 3
  .metric .estimator .estimate
  <chr>   <chr>       <dbl>
1 roc_auc binary      0.840
```

```
logit_demo_aug |>
roc_curve(
  truth = DIED,
  .fitted,
  event_level = "second"
) |>
autoplot()
```



The AUC I achieved is 0.8395. It means that there is approximately an 83.95% chance that the model will be able to distinguish between a patient who died and one who did not die from COVID-19.

```
library(Stat2Data)
library(pROC)
```

Type 'citation("pROC")' for a citation.

Attaching package: 'pROC'

The following objects are masked from 'package:stats':

cov, smooth, var

```
emplogitplot1(DIED ~ AGE, data = covid_health, ngroups = 20,  
main = "Linerity Satisfied Log(Odds) vs. AGE")
```



Since age is the only continuous variable in the model, we must check our linearity assumption. From the plot graphed above, we see that it displays a linear relationship between the log odds of dying and age. The graph displays points that are relatively linearly uniform and follow the trend of the line. From the graph, we do not see that it contain any distinct trend that would challenge our assumption of linearity