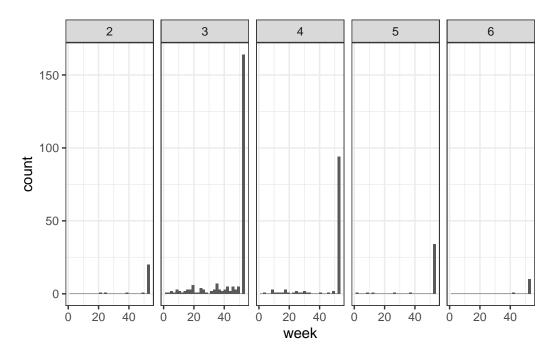
Nowacek HW 4

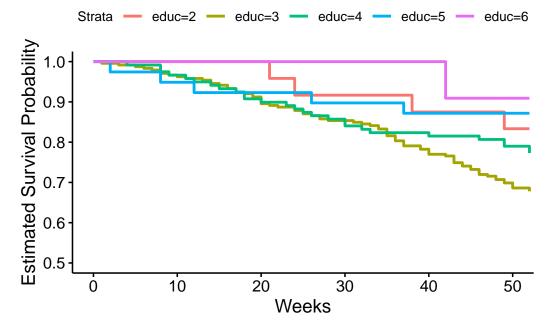
```
d <- read_csv("https://www2.stat.duke.edu/courses/Fall24/sta490.01/data/hw4.csv")

Rows: 432 Columns: 3
-- Column specification ------
Delimiter: ","
dbl (3): week, arrest, educ

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.

ggplot(aes(x = week), data = d) +
    geom_histogram() +
    facet_grid(~ educ) +
    theme_bw()</pre>
```





```
model <- survdiff(Surv(week, arrest) ~ educ, data = d)
model</pre>
```

Call:
survdiff(formula = Surv(week, arrest) ~ educ, data = d)

	N	Observed	Expected	$(0-E)^2/E$	$(0-E)^2/V$
educ=2	24	4	6.81	1.157	1.24
educ=3	239	77	61.54	3.884	8.49
educ=4	119	27	31.59	0.668	0.93
educ=5	39	5	10.78	3.102	3.45
educ=6	11	1	3.28	1.581	1.64

Chisq= 10.5 on 4 degrees of freedom, p= 0.03

This model shows us that there are significant differences in the survival of groups based on education. To get a better idea of this effect, I will condense the education variable into two levels: no high school (groups 2 and 3) and at least some high school (groups 4-6).

```
d <- d |>
  mutate(educ2 = ifelse(educ %in% c(2, 3), 0, 1))

model2 <- survdiff(Surv(week, arrest) ~ educ2, data = d)
model2</pre>
```

Call:

```
survdiff(formula = Surv(week, arrest) ~ educ2, data = d)
```

	N	Observed	Expected	$(0-E)^2/E$	$(0-E)^2/V$
educ2=0	263	81	68.3	2.34	5.88
educ2=1	169	33	45.7	3.51	5.88

Chisq= 5.9 on 1 degrees of freedom, p=0.02

This p-value tells us that there is a significant difference between these two groups, indicating that some high-school education vs non-high school education makes a significant impact on time to recidivism.