

K-M Life Table

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```
dat <- read_csv("./data/heart_failure.csv") %>%
  arrange(TIME) %>% janitor::clean_names()

## Rows: 299 Columns: 13
## -- Column specification -----
## Delimiter: ","
## dbl (13): TIME, Event, Gender, Smoking, Diabetes, BP, Anaemia, Age, Ejection...
##
## 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.
```

dat

```
## # A tibble: 299 x 13
##   time event gender smoking diabetes bp anaemia age ejection_fraction
##   <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1     4     1     1     0     0     1     0    75         20
## 2     6     1     1     0     0     0     0    55         38
## 3     7     1     1     1     0     0     0    65         20
## 4     7     1     1     0     0     0     1    50         20
## 5     8     1     0     0     1     0     1    65         20
## 6     8     1     1     1     0     1     1    90         40
## 7    10     1     1     0     0     0     1    75         15
## 8    10     1     1     1     1     0     1    60         60
## 9    10     1     0     0     0     0     0    65         65
## 10   10     1     1     1     0     1     1    80         35
## # i 289 more rows
## # i 4 more variables: sodium <dbl>, creatinine <dbl>, pletelets <dbl>,
## #   cpk <dbl>
```

Fleming-Harrington Survival Estimate

$$\hat{S}_F(t) = \begin{cases} \prod_{t_i \leq t} \exp\left(-\frac{d_i}{n_i}\right) & \text{if } t > t_1 \\ 1 & \text{otherwise} \end{cases}$$

```
surv_object <- Surv(time = dat$time, event = dat$event)

# Calculate Fleming-Harrington table
```

```
fh_table <- survfit(surv_object ~ 1)

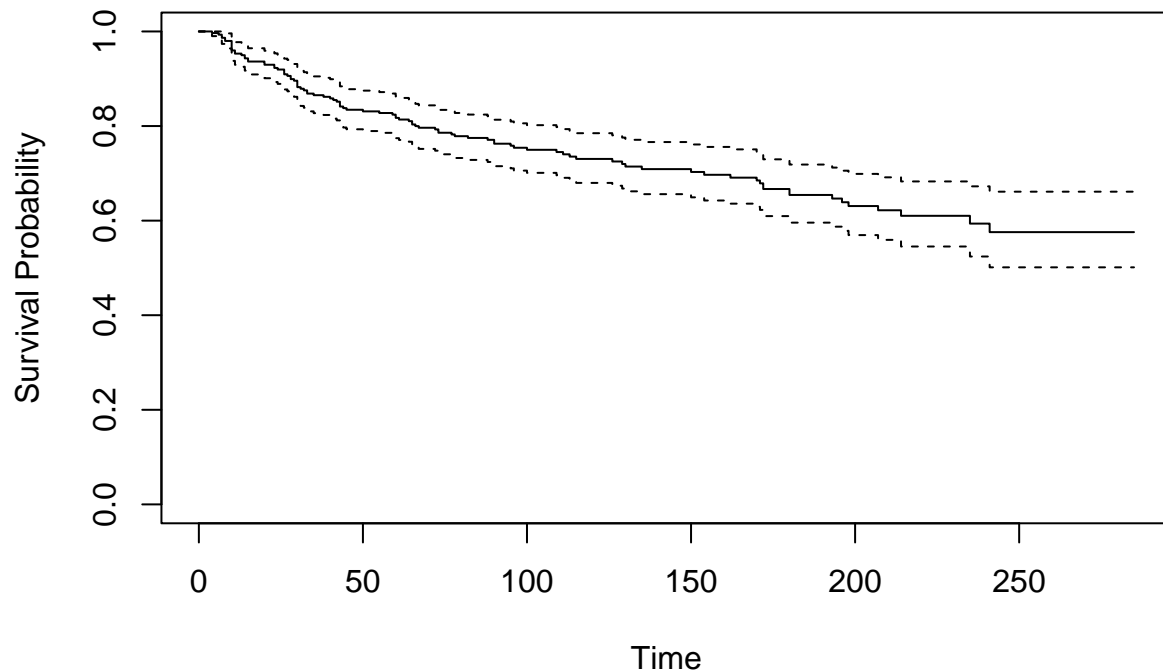
# Extract survival times and number of events
time_points <- fh_table$time
n_events <- fh_table$n.event
n_at_risk <- fh_table$n.risk

# Calculate Fleming-Harrington weights
fh_weights <- cumprod(exp(n_events / n_at_risk))

# Create a data frame for Fleming-Harrington statistics
fh_stats <- data.frame(Time = time_points,
                       N_events = n_events, N_at_risk = n_at_risk,
                       FH_Weights = fh_weights)

# Plotting the Fleming-Harrington survival curve
plot(fh_table, xlab = "Time", ylab = "Survival Probability",
     main = "Fleming-Harrington Survival Curve")
```

Fleming-Harrington Survival Curve



Kaplan-Meier Survival Estimate

$$\hat{S}_K(t) = \begin{cases} \prod_{t_i \leq t} \left(1 - \frac{d_i}{n_i}\right) & \text{if } t > t_1 \\ 1 & \text{otherwise} \end{cases}$$

```
surv_object <- Surv(time = dat$time, event = dat$event)
km_fit <- survfit(surv_object ~ 1)
summary(km_fit)
```

```
## Call: survfit(formula = surv_object ~ 1)
##
##   time n.risk n.event survival std.err lower 95% CI upper 95% CI
##   4      299      1    0.997 0.00334      0.990      1.000
##   6      298      1    0.993 0.00471      0.984      1.000
##   7      297      2    0.987 0.00664      0.974      1.000
##   8      295      2    0.980 0.00811      0.964      0.996
##  10      293      6    0.960 0.01135      0.938      0.982
##  11      287      2    0.953 0.01222      0.930      0.977
##  13      284      1    0.950 0.01263      0.925      0.975
##  14      283      2    0.943 0.01340      0.917      0.970
##  15      281      2    0.936 0.01412      0.909      0.964
##  20      278      2    0.930 0.01480      0.901      0.959
##  23      275      2    0.923 0.01545      0.893      0.954
##  24      273      1    0.920 0.01575      0.889      0.951
##  26      272      3    0.909 0.01663      0.877      0.943
##  27      269      1    0.906 0.01691      0.873      0.940
##  28      268      2    0.899 0.01745      0.866      0.934
##  29      266      1    0.896 0.01771      0.862      0.931
##  30      264      4    0.882 0.01869      0.846      0.920
##  31      259      1    0.879 0.01893      0.843      0.917
##  32      258      1    0.875 0.01916      0.839      0.914
##  33      257      2    0.869 0.01961      0.831      0.908
##  35      254      1    0.865 0.01983      0.827      0.905
##  38      253      1    0.862 0.02004      0.823      0.902
##  40      252      1    0.858 0.02025      0.820      0.899
##  41      251      1    0.855 0.02046      0.816      0.896
##  42      250      1    0.852 0.02066      0.812      0.893
##  43      249      3    0.841 0.02124      0.801      0.884
##  44      246      1    0.838 0.02143      0.797      0.881
##  45      245      1    0.834 0.02161      0.793      0.878
##  50      244      1    0.831 0.02179      0.789      0.875
##  55      241      1    0.828 0.02197      0.786      0.872
##  59      240      1    0.824 0.02215      0.782      0.869
##  60      239      2    0.817 0.02250      0.774      0.863
##  61      236      1    0.814 0.02267      0.771      0.859
##  64      234      1    0.810 0.02283      0.767      0.856
##  65      233      2    0.803 0.02316      0.759      0.850
##  66      231      1    0.800 0.02332      0.755      0.847
##  67      230      1    0.796 0.02348      0.752      0.844
##  72      227      1    0.793 0.02364      0.748      0.841
##  73      225      2    0.786 0.02394      0.740      0.834
##  77      217      1    0.782 0.02411      0.736      0.831
##  78      216      1    0.779 0.02427      0.732      0.828
##  82      207      1    0.775 0.02444      0.728      0.824
##  88      194      1    0.771 0.02464      0.724      0.821
##  90      189      2    0.763 0.02504      0.715      0.813
##  95      180      1    0.758 0.02526      0.711      0.810
##  96      175      1    0.754 0.02548      0.706      0.806
## 100      173      1    0.750 0.02571      0.701      0.802
## 109      159      1    0.745 0.02597      0.696      0.798
## 111      155      1    0.740 0.02625      0.691      0.794
```

##	113	152	1	0.735	0.02652	0.685	0.789
##	115	150	1	0.730	0.02679	0.680	0.785
##	126	136	1	0.725	0.02713	0.674	0.780
##	129	135	1	0.720	0.02746	0.668	0.776
##	130	134	1	0.714	0.02777	0.662	0.771
##	135	132	1	0.709	0.02808	0.656	0.766
##	150	118	1	0.703	0.02848	0.649	0.761
##	154	117	1	0.697	0.02886	0.643	0.756
##	162	116	1	0.691	0.02923	0.636	0.751
##	170	115	1	0.685	0.02959	0.629	0.745
##	171	114	1	0.679	0.02993	0.623	0.740
##	172	113	2	0.667	0.03059	0.610	0.730
##	180	106	2	0.654	0.03128	0.596	0.719
##	193	86	1	0.647	0.03183	0.587	0.712
##	196	83	1	0.639	0.03238	0.578	0.706
##	198	79	1	0.631	0.03297	0.569	0.699
##	207	71	1	0.622	0.03368	0.559	0.692
##	214	53	1	0.610	0.03503	0.545	0.683
##	235	37	1	0.594	0.03776	0.524	0.673
##	241	33	1	0.576	0.04068	0.501	0.661

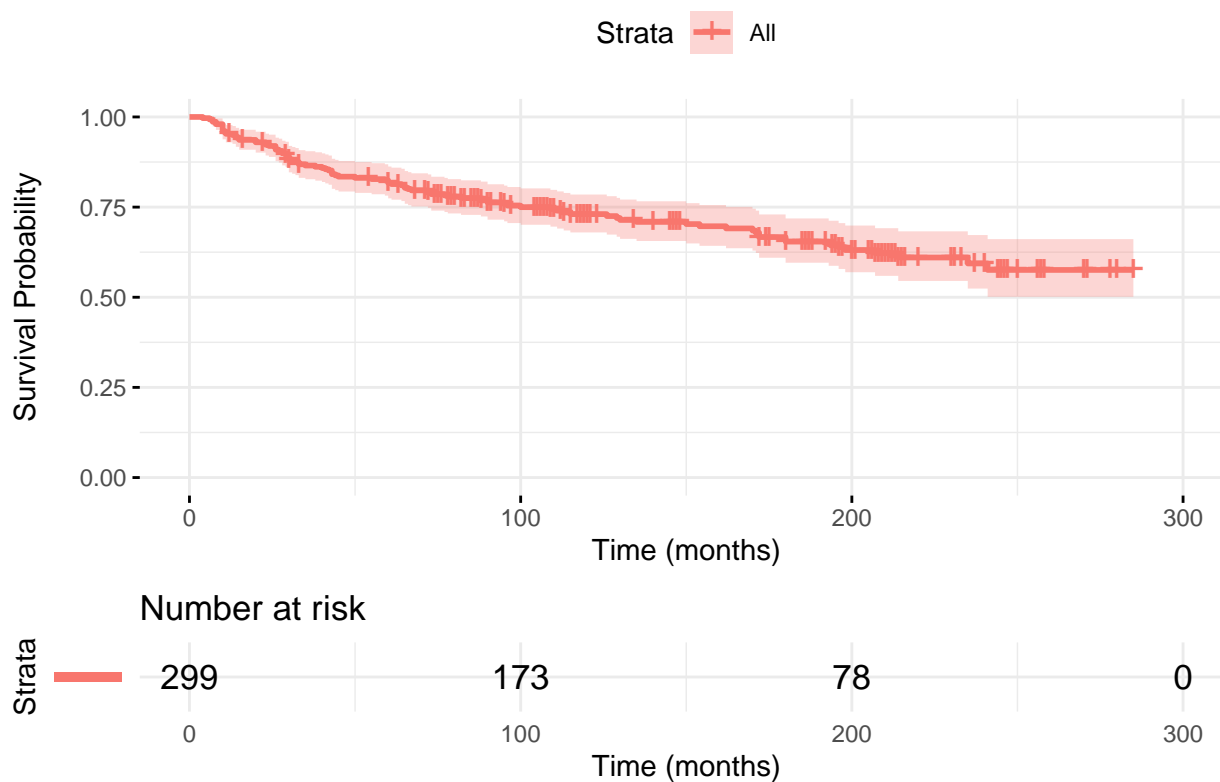
Life Table

```
surv_object <- Surv(time = dat$time, event = dat$event)
life_table <- survfit(surv_object ~ 1)
summary(life_table, times = c(1:10)*30) # Monthly intervals
```

```
## Call: survfit(formula = surv_object ~ 1)
##
##   time n.risk n.event survival std.err lower 95% CI upper 95% CI
##   30    264     35    0.882  0.0187    0.846    0.920
##   60    239     19    0.817  0.0225    0.774    0.863
##   90    189     15    0.763  0.0250    0.715    0.813
##  120    145      7    0.730  0.0268    0.680    0.785
##  150    118      5    0.703  0.0285    0.649    0.761
##  180    106      8    0.654  0.0313    0.596    0.719
##  210     62      4    0.622  0.0337    0.559    0.692
##  240     34      2    0.594  0.0378    0.524    0.673
##  270      6      1    0.576  0.0407    0.501    0.661
```

```
ggsurvplot(km_fit, data = dat, title = "Kaplan-Meier Survival Curve",
  xlab = "Time (months)", ylab = "Survival Probability",
  risk.table = TRUE, risk.table.y.text = FALSE,
  ggtheme = theme_minimal())
```

Kaplan–Meier Survival Curve

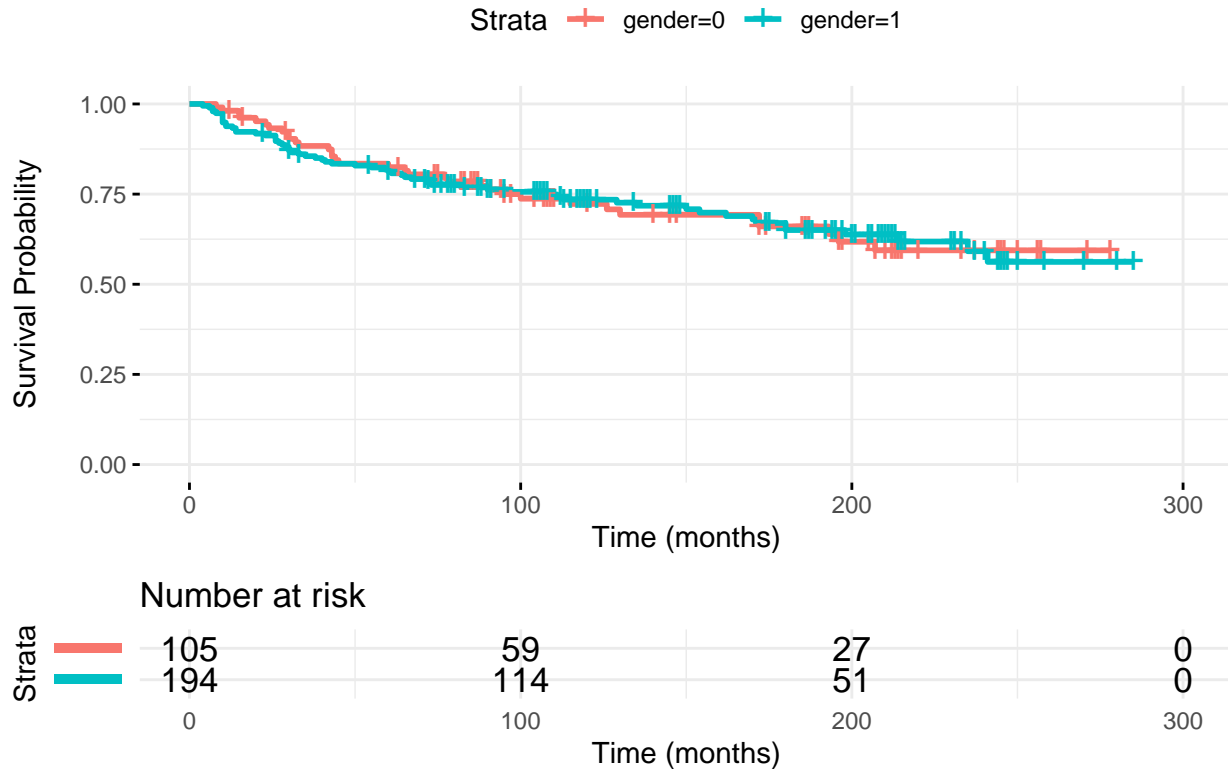


Life Table Stratified by Gender

Plot

```
surv_object <- Surv(time = dat$time, event = dat$event)
km_fit_gender <- survfit(surv_object ~ gender, data = dat)
ggsurvplot(km_fit_gender, data = dat,
  title = "Kaplan-Meier Survival Curve by Gender",
  xlab = "Time (months)", ylab = "Survival Probability",
  risk.table = TRUE, risk.table.y.text = FALSE,
  ggtheme = theme_minimal())
```

Kaplan–Meier Survival Curve by Gender



Log Rank Test

We would like to test if there is difference in survival between genders.

```
logrank_test <- survdiff(surv_object ~ gender, data = dat)
print(logrank_test)
```

```
## Call:
## survdiff(formula = surv_object ~ gender, data = dat)
##
##           N Observed Expected (O-E)^2/E (O-E)^2/V
## gender=0 105      34    34.3   0.00254   0.00397
## gender=1 194      62    61.7   0.00141   0.00397
##
##  Chisq= 0  on 1 degrees of freedom, p= 0.9
```

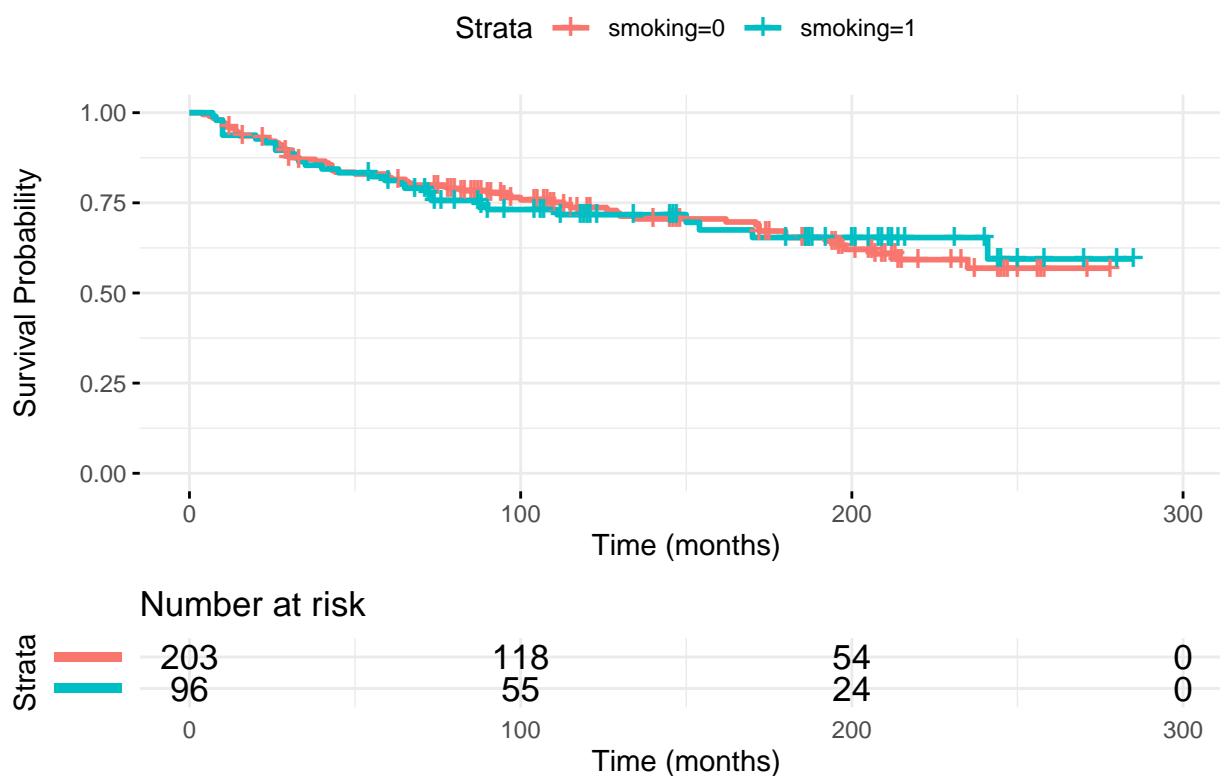
With a p-value of 0.9, there is no sufficient evidence to reject the null hypothesis. In this case, there is no difference in survival between the two genders based on this dataset.

Life Table Stratified by Smoking Status

Plot

```
surv_object <- Surv(time = dat$time, event = dat$event)
km_fit_smoking <- survfit(surv_object ~ smoking, data = dat)
ggsurvplot(km_fit_smoking, data = dat,
  title = "Kaplan-Meier Survival Curve by Smoking Status",
  xlab = "Time (months)", ylab = "Survival Probability",
  risk.table = TRUE, risk.table.y.text = FALSE,
  ggtheme = theme_minimal())
```

Kaplan-Meier Survival Curve by Smoking Status



Log Rank Test

We would like to test if there is difference in survival between smoking status.

```
logrank_test <- survdiff(surv_object ~ smoking, data = dat)
print(logrank_test)
```

```
## Call:
## survdiff(formula = surv_object ~ smoking, data = dat)
##
##           N Observed Expected (O-E)^2/E (O-E)^2/V
```

```
## smoking=0 203      66      65.8  0.00064  0.00204
## smoking=1  96      30      30.2  0.00139  0.00204
##
##  Chisq= 0  on 1 degrees of freedom, p= 1
```

With a p-value of 1, there is no sufficient evidence to reject the null hypothesis. In this case, there is no difference in survival between smokers and non-smokers based on this dataset.