

Survival

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```
echocardiogram <- read.csv("~/Desktop/GCU/DSC_520/Data/echocardiogram/echocardiogram.data", header=FALSE)
colnames(echocardiogram) <- c('survival','still-alive','age-at-heart-attack',
                              "pericardial-effusion","fractional-shortening",
                              "epss","lvdd","wall-motion-score",
                              "wall-motion-index","mult","name",
                              "group","alive-at-1")
echocardiogram[,c('name', 'group')] <- list(NULL)

suppressWarnings(echocardiogram$survival <- as.numeric(echocardiogram$survival))

echocardiogram$`still-alive` <- factor(echocardiogram$`still-alive`, levels = c(0,1))

suppressWarnings(echocardiogram$`age-at-heart-attack` <- as.numeric(echocardiogram$`age-at-heart-attack`))

echocardiogram$`pericardial-effusion` <- factor(echocardiogram$`pericardial-effusion`, levels = c(0,1))

suppressWarnings(echocardiogram$`fractional-shortening` <- as.numeric(echocardiogram$`fractional-shortening`))

suppressWarnings(echocardiogram$epss <- as.numeric(echocardiogram$epss))

suppressWarnings(echocardiogram$lvdd <- as.numeric(echocardiogram$lvdd))

suppressWarnings(echocardiogram$`wall-motion-score` <- as.numeric(echocardiogram$`wall-motion-score`))

suppressWarnings(echocardiogram$`wall-motion-index` <- as.numeric(echocardiogram$`wall-motion-index`))

suppressWarnings(echocardiogram$mult <- as.numeric(echocardiogram$mult))

echocardiogram <- na.omit(echocardiogram)

echocardiogram$`alive-at-1` <- ifelse((echocardiogram$survival >= 12 | echocardiogram$`still-alive` == 1),1,0)

echocardiogram$`alive-at-1` <- factor(echocardiogram$`alive-at-1`, levels = c(0,1))
echocardiogram <- na.omit(echocardiogram)
head(echocardiogram)
```

```
## survival still-alive age-at-heart-attack pericardial-effusion
## 1      11          0          71          0
## 2      19          0          72          0
## 3      16          0          55          0
## 4      57          0          60          0
## 5      19          1          57          0
## 6      26          0          68          0
## fractional-shortening epss lvdd wall-motion-score wall-motion-index mult
## 1              0.260 9.000 4.600              14              1.00 1.000
## 2              0.380 6.000 4.100              14              1.70 0.588
## 3              0.260 4.000 3.420              14              1.00 1.000
## 4              0.253 12.062 4.603              16              1.45 0.788
## 5              0.160 22.000 5.750              18              2.25 0.571
## 6              0.260 5.000 4.310              12              1.00 0.857
## alive-at-1
## 1      0
## 2      1
## 3      1
## 4      1
## 5      1
## 6      1
```

First I must clean the dataset, so that it can be used. I assigned every factor its desired data type, and fixed the alive at 1 year factor as it can be easily manually filled. I then removed the data with missing values since they tended to have multiple missing data points if they were missing one.

```
# Load in libraries
library(knitr)
library(devtools)
```

```
## Loading required package: usethis
```

```
library(survival)
library(tidyverse)
```

```
## — Attaching core tidyverse packages — tidyverse 2.0.0 —
## ✓ dplyr      1.1.4      ✓ readr      2.1.5
## ✓ forcats    1.0.0      ✓ stringr    1.5.1
## ✓ ggplot2     3.5.1      ✓ tibble     3.2.1
## ✓ lubridate  1.9.3      ✓ tidyr      1.3.1
## ✓ purrr       1.0.2
```

```
## — Conflicts — tidyverse_conflicts() —
## * dplyr::filter() masks stats::filter()
## * dplyr::lag()     masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(survminer)
```

```
## Loading required package: ggpubr
##
## Attaching package: 'survminer'
##
## The following object is masked from 'package:survival':
##
##      myeloma
```

```
library(ggplot2)
library(ggfortify)
```

```
kable(echocardiogram)
```

	survival	still- alive	age-at- heart- attack	pericardial- effusion	fractional- shortening	epss	lvdd	wall- motion- score	wall- motion- index	alive- mult	alive- at-1
1	11.00	0	71.000	0	0.260	9.000	4.600	14.00	1.000	1.000	0
2	19.00	0	72.000	0	0.380	6.000	4.100	14.00	1.700	0.588	1
3	16.00	0	55.000	0	0.260	4.000	3.420	14.00	1.000	1.000	1
4	57.00	0	60.000	0	0.253	12.062	4.603	16.00	1.450	0.788	1
5	19.00	1	57.000	0	0.160	22.000	5.750	18.00	2.250	0.571	1
6	26.00	0	68.000	0	0.260	5.000	4.310	12.00	1.000	0.857	1
7	13.00	0	62.000	0	0.230	31.000	5.430	22.50	1.875	0.857	1
8	50.00	0	60.000	0	0.330	8.000	5.250	14.00	1.000	1.000	1
9	19.00	0	46.000	0	0.340	0.000	5.090	16.00	1.140	1.003	1
10	25.00	0	54.000	0	0.140	13.000	4.490	15.50	1.190	0.930	1
11	10.00	1	77.000	0	0.130	16.000	4.230	18.00	1.800	0.714	1
12	52.00	0	62.000	1	0.450	9.000	3.600	16.00	1.140	1.003	1
13	52.00	0	73.000	0	0.330	6.000	4.000	14.00	1.000	1.000	1
14	44.00	0	60.000	0	0.150	10.000	3.730	14.00	1.000	1.000	1
15	0.50	1	62.000	0	0.120	23.000	5.800	11.67	2.330	0.358	1
16	24.00	0	55.000	1	0.250	12.063	4.290	14.00	1.000	1.000	1
17	0.50	1	69.000	1	0.260	11.000	4.650	18.00	1.640	0.784	1
18	0.50	1	62.529	1	0.070	20.000	5.200	24.00	2.000	0.857	1

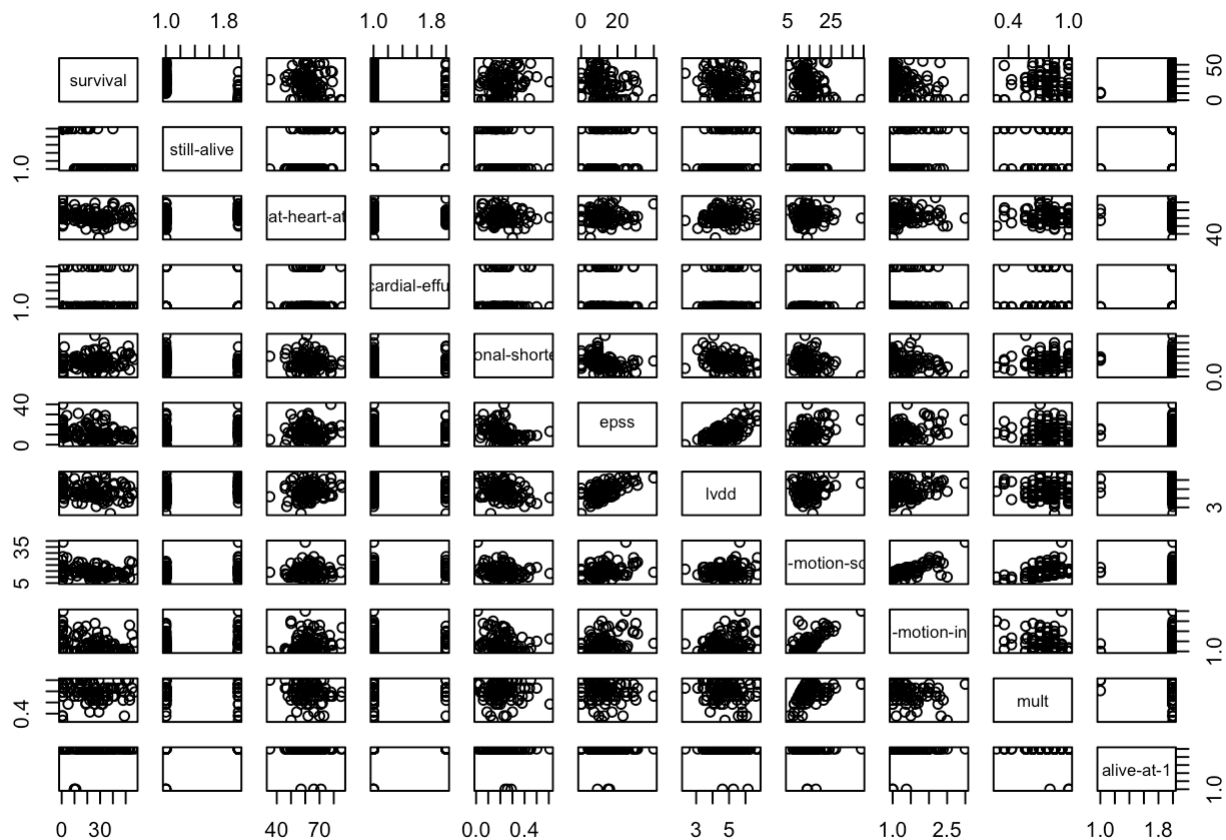
	survival	still- alive	age-at- heart- attack	pericardial- effusion	fractional- shortening	epss	lvdd	wall- motion- score	wall- motion- index	mult	alive- at-1
19	22.00	1	66.000	0	0.090	17.000	5.819	8.00	1.333	0.429	1
20	1.00	1	66.000	1	0.220	15.000	5.400	27.00	2.250	0.857	1
21	0.75	1	69.000	0	0.150	12.000	5.390	19.50	1.625	0.857	1
22	0.75	1	85.000	1	0.180	19.000	5.460	13.83	1.380	0.710	1
23	0.50	1	73.000	0	0.230	12.733	6.060	7.50	1.500	0.360	1
24	5.00	1	71.000	0	0.170	0.000	4.650	8.00	1.000	0.570	1
25	48.00	0	64.000	0	0.190	5.900	3.480	10.00	1.110	0.640	1
26	29.00	0	54.000	0	0.300	7.000	3.850	10.00	1.667	0.430	1
27	29.00	0	35.000	0	0.300	5.000	4.170	14.00	1.000	1.000	1
30	36.00	0	55.000	1	0.210	4.200	4.160	14.00	1.560	0.640	1
32	1.00	1	52.000	1	0.170	17.200	5.320	14.00	1.170	0.857	1
34	27.00	0	47.000	0	0.400	5.120	3.100	12.00	1.000	0.857	1
36	26.00	0	61.000	0	0.610	13.100	4.070	13.00	1.625	0.571	1
41	32.00	0	54.000	0	0.350	9.300	3.630	11.00	1.222	0.640	1
42	16.00	0	70.000	1	0.270	4.700	4.490	22.00	2.000	0.786	1
43	40.00	0	79.000	0	0.150	17.500	4.270	13.00	1.300	0.714	1
45	2.00	1	67.000	1	0.440	9.000	3.960	17.50	1.450	0.857	1
48	20.00	1	59.000	0	0.030	21.300	6.290	17.00	1.310	0.928	1
54	10.00	0	57.000	0	0.240	14.800	5.260	18.00	1.380	0.812	0
55	12.00	0	58.000	0	0.300	9.400	3.490	14.00	1.000	1.000	1
56	1.00	1	60.000	0	0.010	24.600	5.650	39.00	3.000	0.928	1
57	10.00	0	66.000	0	0.290	15.600	6.150	14.00	1.000	1.000	0
58	45.00	0	63.000	0	0.150	13.000	4.570	13.00	1.080	0.857	1
59	22.00	0	57.000	0	0.130	18.600	4.370	12.33	1.370	0.642	1
60	53.00	0	70.000	0	0.100	9.800	5.300	23.00	2.300	0.714	1
62	26.00	0	79.000	0	0.170	11.900	5.150	10.50	1.050	0.714	1
64	26.00	0	72.000	0	0.187	12.000	5.020	13.00	1.180	0.785	1
65	0.50	1	59.000	0	0.130	16.400	4.960	17.83	1.370	0.928	1
66	12.00	0	67.000	1	0.110	10.300	4.680	11.00	1.000	0.785	1

	survival	still- alive	age-at- heart- attack	pericardial- effusion	fractional- shortening	epss	lvdd	wall- motion- score	wall- motion- index	mult	alive- at-1
67	49.00	0	51.000	0	0.160	13.200	5.260	11.00	1.000	0.786	1
68	0.75	1	50.000	0	0.140	11.400	4.750	10.00	2.500	0.280	1
69	49.00	0	70.000	1	0.250	9.700	5.570	5.50	1.100	0.357	1
70	47.00	0	65.000	0	0.360	8.800	5.780	12.00	1.000	0.857	1
71	41.00	0	78.000	0	0.060	16.100	5.620	13.67	1.367	0.714	1
72	0.25	1	86.000	0	0.225	12.200	5.200	24.00	2.180	0.786	1
73	33.00	0	56.000	0	0.250	11.000	4.720	11.00	1.000	0.785	1
74	29.00	0	60.000	0	0.120	10.200	4.310	15.00	1.670	0.640	1
75	41.00	0	59.000	0	0.290	7.500	4.750	13.00	1.080	0.857	1
76	26.00	0	50.000	0	0.060	30.100	5.950	21.50	2.390	0.643	1
77	15.00	0	54.000	0	0.217	17.900	4.540	16.50	1.180	1.000	1
78	0.25	1	68.000	0	0.220	21.700	4.850	15.00	1.150	0.928	1
80	12.00	0	64.000	0	0.200	7.100	4.580	14.00	1.000	1.000	1
81	32.00	0	63.000	0	0.200	5.000	5.200	8.00	1.000	0.570	1
82	32.00	0	65.000	0	0.060	23.600	6.740	12.00	1.090	0.785	1
83	27.00	0	54.000	1	0.070	16.800	4.160	18.00	1.500	0.857	1
84	23.00	0	62.000	0	0.250	6.000	4.480	11.00	1.000	0.786	1
85	0.75	1	78.000	0	0.050	10.000	4.440	15.00	1.360	0.786	1
87	34.00	0	52.000	0	0.140	25.000	6.210	11.50	1.150	0.714	1
88	1.00	1	73.000	0	0.050	14.800	4.140	15.50	1.410	0.786	1
89	21.00	1	70.000	1	0.160	19.200	5.250	11.00	1.000	0.786	1
90	55.00	0	55.000	0	0.280	5.500	4.480	22.00	1.830	0.857	1
91	15.00	1	60.000	0	0.180	8.700	4.560	13.50	1.040	0.928	1
92	0.50	1	67.000	0	0.155	11.300	5.160	13.00	1.000	0.928	1
93	35.00	0	64.000	0	0.300	6.600	4.360	14.00	1.270	0.786	1
94	53.00	0	59.000	0	0.344	9.100	4.040	9.00	1.000	0.643	1
95	33.00	0	46.000	0	0.272	16.500	5.360	12.67	1.060	0.857	1
97	33.00	0	63.000	0	0.250	5.600	3.870	18.00	1.500	0.857	1
98	40.00	1	74.000	0	0.200	4.800	4.560	12.50	1.040	0.857	1

	survival	still- alive	age-at- heart- attack	pericardial- effusion	fractional- shortening	epss	lvdd	wall- motion- score	wall- motion- index	mult	alive- at-1
99	33.00	0	59.000	0	0.500	9.100	3.420	18.00	1.500	0.857	1
100	5.00	1	65.000	1	0.160	8.500	5.470	16.00	1.450	0.786	1
101	4.00	1	58.000	0	0.170	28.900	6.730	26.08	2.010	0.928	1
104	22.00	0	70.000	0	0.380	0.000	4.550	10.00	1.000	0.714	1
105	25.00	0	62.000	0	0.258	11.800	4.870	11.00	1.000	0.786	1
106	1.25	1	63.000	0	0.300	6.900	3.520	18.16	1.510	0.857	1
107	24.00	0	59.000	0	0.170	14.300	5.490	13.50	1.500	0.643	1
108	25.00	0	57.000	0	0.228	9.700	4.290	11.00	1.000	0.786	1
109	24.00	0	57.000	0	0.036	7.000	4.120	13.50	1.230	0.786	1
110	0.75	1	78.000	0	0.230	40.000	6.230	14.00	1.400	0.714	1
111	3.00	1	62.000	0	0.260	7.600	4.420	14.00	1.000	1.000	1
112	27.00	0	62.000	0	0.220	12.100	3.920	11.00	1.000	0.785	1
113	13.00	0	66.000	0	0.240	13.600	4.380	22.00	2.200	0.714	1
114	36.00	0	61.000	0	0.270	9.000	4.060	12.00	1.000	0.857	1
115	25.00	0	59.000	1	0.400	9.200	5.360	12.00	1.000	0.857	1
116	27.00	0	57.000	0	0.290	9.400	4.770	9.00	1.000	0.640	1
117	34.00	0	62.000	1	0.190	28.900	6.630	19.50	1.950	0.714	1
119	34.00	0	54.000	0	0.430	9.300	4.790	10.00	1.000	0.714	1
120	28.00	1	62.000	1	0.240	28.600	5.860	21.50	1.950	0.786	1
122	17.00	0	64.000	0	0.150	6.600	4.170	14.00	1.270	0.786	1
123	38.00	0	57.000	1	0.120	0.000	2.320	16.50	1.375	0.857	1
124	31.00	0	61.000	0	0.180	0.000	4.480	11.00	1.375	0.570	1
125	12.00	0	61.000	1	0.190	13.200	5.040	19.00	1.730	0.786	1
126	36.00	0	48.000	0	0.150	12.000	3.660	10.00	1.000	0.714	1
128	21.00	0	61.000	0	0.140	25.500	5.160	14.00	1.270	0.786	1
129	7.50	1	64.000	0	0.240	12.900	4.720	12.00	1.000	0.857	1
130	41.00	0	64.000	0	0.280	5.400	5.470	11.00	1.100	0.714	1
131	36.00	0	69.000	0	0.200	7.000	5.050	14.50	1.210	0.857	1
132	22.00	0	57.000	0	0.140	16.100	4.360	15.00	1.360	0.786	1

		still- survival	alive	age-at- heart- attack	pericardial- effusion	fractional- shortening	epss	lvdd	wall- motion- score	wall- motion- index	mult	alive- at-1
133	20.00	0	62.000	0		0.150	0.000	4.510	15.50	1.409	0.786	1

```
plot(echocardiogram)
```



We can see that there are a handful of possibly correlated factors, and there are no visual extreme outliers.

In this we will aim to answer if there is a difference in survival rates between individuals with pericardial effusion and without pericardial effusion.

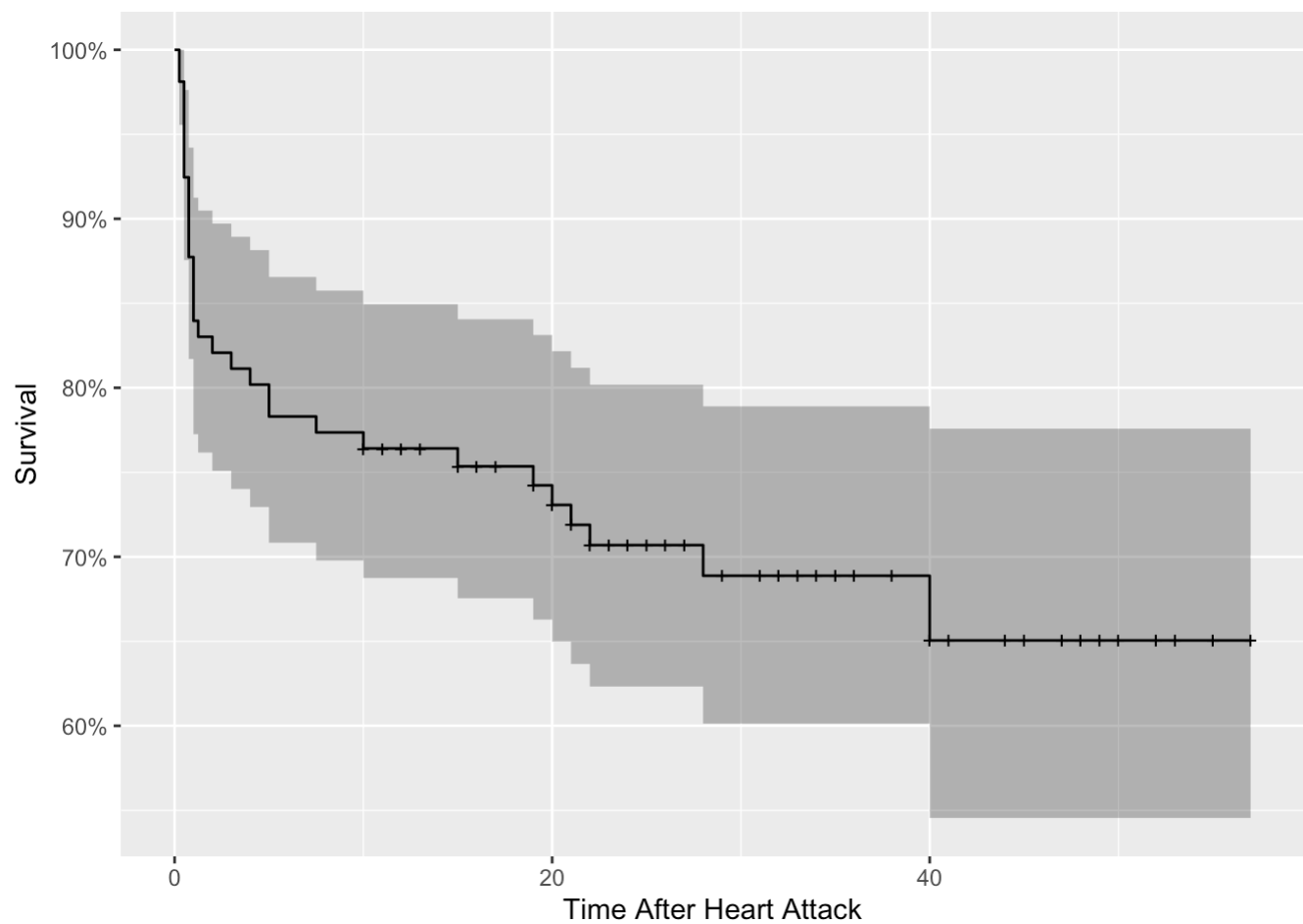
```
sur <- Surv(echocardiogram$survival, as.numeric(as.character(echocardiogram$`still-alive`)))
print(sur)
```

```
## [1] 11.00+ 19.00+ 16.00+ 57.00+ 19.00 26.00+ 13.00+ 50.00+ 19.00+ 25.00+
## [11] 10.00 52.00+ 52.00+ 44.00+ 0.50 24.00+ 0.50 0.50 22.00 1.00
## [21] 0.75 0.75 0.50 5.00 48.00+ 29.00+ 29.00+ 36.00+ 1.00 27.00+
## [31] 26.00+ 32.00+ 16.00+ 40.00+ 2.00 20.00 10.00+ 12.00+ 1.00 10.00+
## [41] 45.00+ 22.00+ 53.00+ 26.00+ 26.00+ 0.50 12.00+ 49.00+ 0.75 49.00+
## [51] 47.00+ 41.00+ 0.25 33.00+ 29.00+ 41.00+ 26.00+ 15.00+ 0.25 12.00+
## [61] 32.00+ 32.00+ 27.00+ 23.00+ 0.75 34.00+ 1.00 21.00 55.00+ 15.00
## [71] 0.50 35.00+ 53.00+ 33.00+ 33.00+ 40.00 33.00+ 5.00 4.00 22.00+
## [81] 25.00+ 1.25 24.00+ 25.00+ 24.00+ 0.75 3.00 27.00+ 13.00+ 36.00+
## [91] 25.00+ 27.00+ 34.00+ 34.00+ 28.00 17.00+ 38.00+ 31.00+ 12.00+ 36.00+
## [101] 21.00+ 7.50 41.00+ 36.00+ 22.00+ 20.00+
```

```
fit <- survfit(sur ~ 1 , data = echocardiogram )
summary(fit)
```

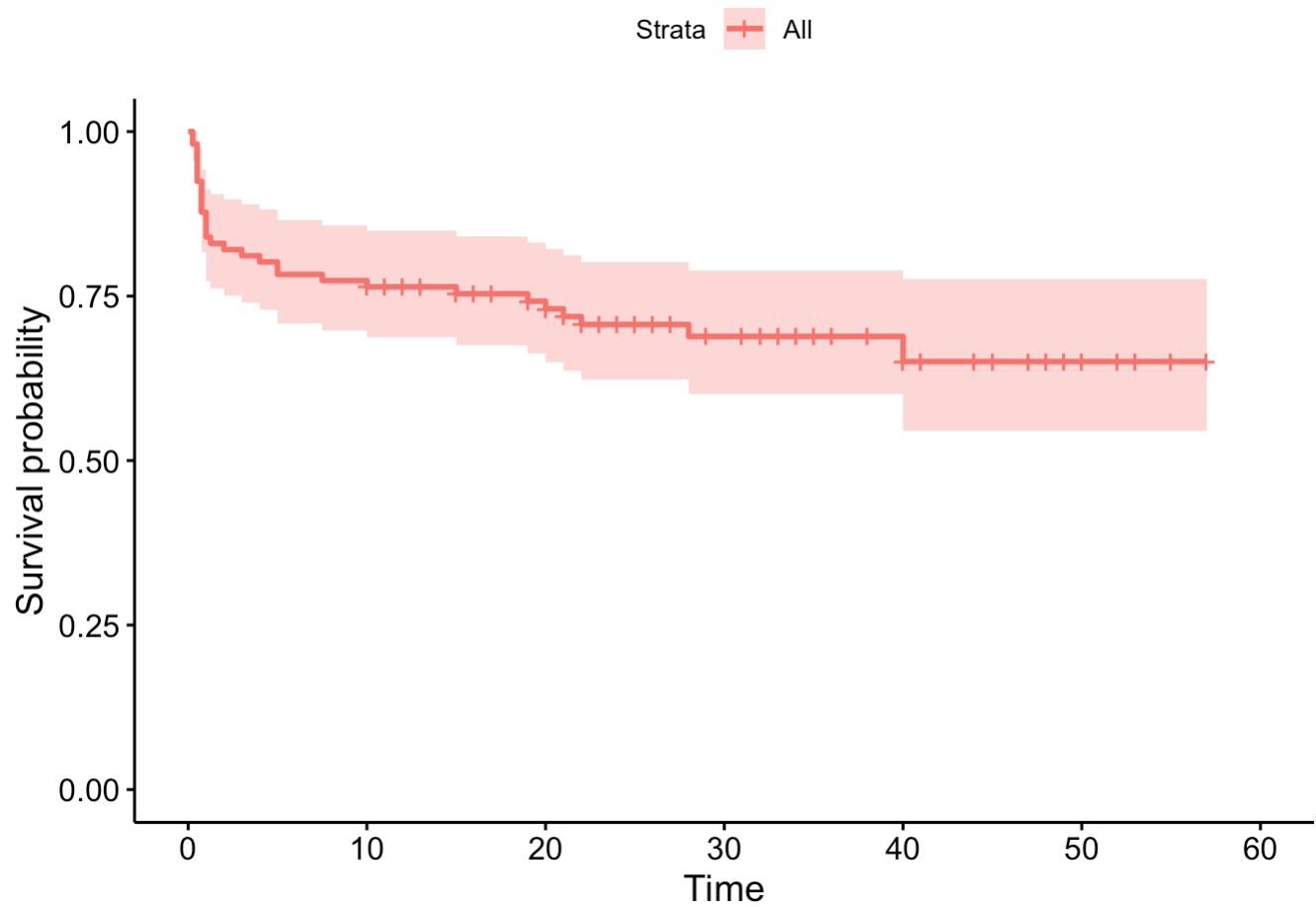
```
## Call: survfit(formula = sur ~ 1, data = echocardiogram)
##
##      time n.risk n.event survival std.err lower 95% CI upper 95% CI
##  0.25    106      2    0.981  0.0132      0.956      1.000
##  0.50    104      6    0.925  0.0257      0.876      0.976
##  0.75     98      5    0.877  0.0319      0.817      0.942
##  1.00     93      4    0.840  0.0356      0.773      0.912
##  1.25     89      1    0.830  0.0365      0.762      0.905
##  2.00     88      1    0.821  0.0373      0.751      0.897
##  3.00     87      1    0.811  0.0380      0.740      0.889
##  4.00     86      1    0.802  0.0387      0.729      0.881
##  5.00     85      2    0.783  0.0400      0.708      0.866
##  7.50     83      1    0.774  0.0406      0.698      0.858
## 10.00     82      1    0.764  0.0412      0.687      0.849
## 15.00     72      1    0.754  0.0420      0.676      0.841
## 19.00     67      1    0.742  0.0429      0.663      0.831
## 20.00     64      1    0.731  0.0437      0.650      0.822
## 21.00     62      1    0.719  0.0446      0.637      0.812
## 22.00     60      1    0.707  0.0454      0.623      0.802
## 28.00     39      1    0.689  0.0477      0.601      0.789
## 40.00     18      1    0.651  0.0584      0.546      0.776
```

```
autoplot(fit, ylab = 'Survival', xlab = 'Time After Heart Attack', legLabs = c('dead','alive'))
```

When analyzing survival over time we can see that the more time after the heart attack the less likely the survival is.

```
ggsurvplot(fit = fit, data = echocardiogram)
```

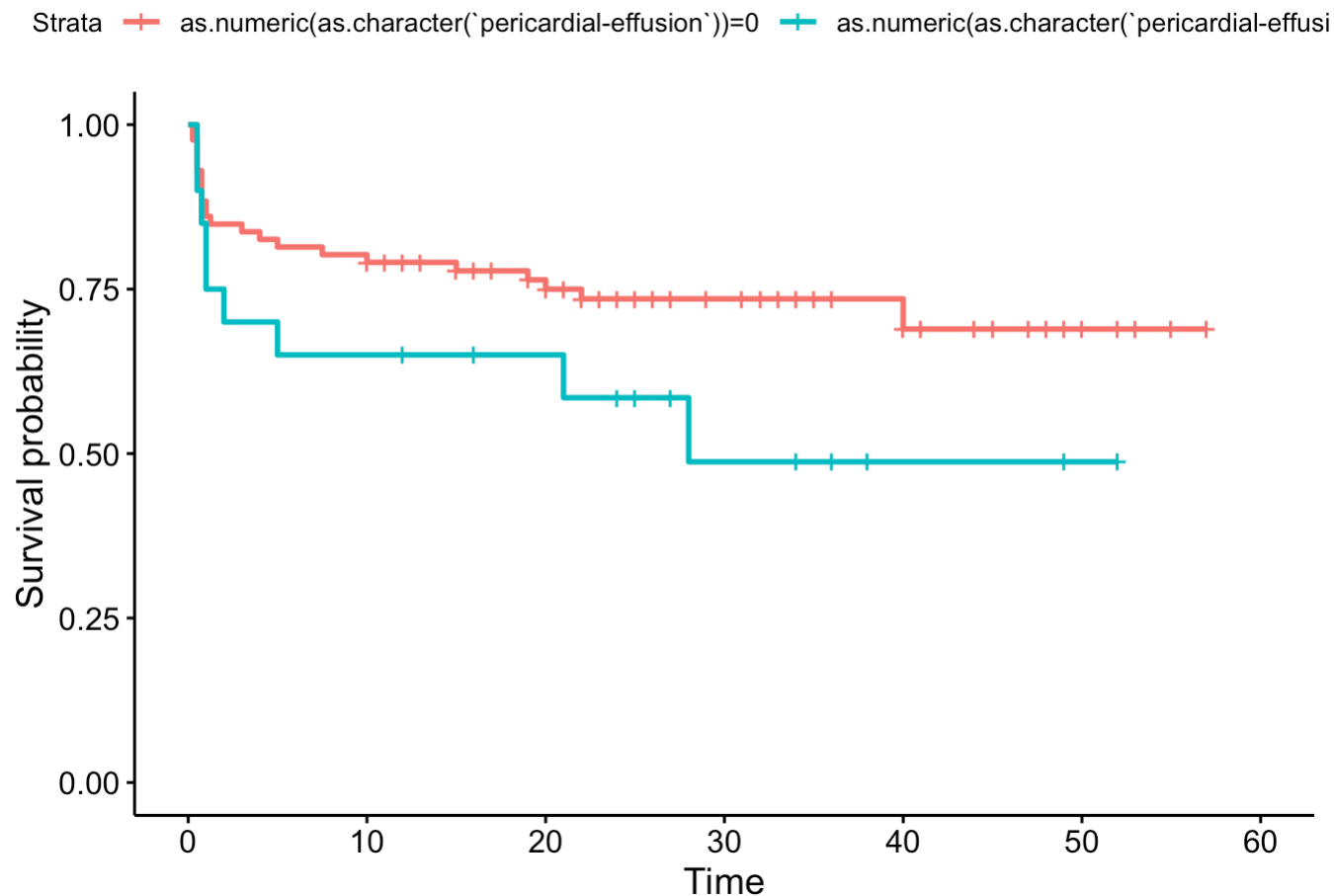


This shows a similar story with the error bar included.

```
fit2 <- survfit(sur ~ as.numeric(as.character(`pericardial-effusion`)) , data = echocard
iogram )
summary(fit2)
```

```
## Call: survfit(formula = sur ~ as.numeric(as.character(`pericardial-effusion`)),
##      data = echocardiogram)
##
##               as.numeric(as.character(`pericardial-effusion`))=0
##   time n.risk n.event survival std.err lower 95% CI upper 95% CI
##   0.25    86      2   0.977  0.0163    0.945    1.000
##   0.50    84      4   0.930  0.0275    0.878    0.986
##   0.75    80      4   0.884  0.0346    0.819    0.954
##   1.00    76      2   0.860  0.0374    0.790    0.937
##   1.25    74      1   0.849  0.0386    0.776    0.928
##   3.00    73      1   0.837  0.0398    0.763    0.919
##   4.00    72      1   0.826  0.0409    0.749    0.910
##   5.00    71      1   0.814  0.0420    0.736    0.900
##   7.50    70      1   0.802  0.0429    0.722    0.891
##  10.00    69      1   0.791  0.0439    0.709    0.882
##  15.00    61      1   0.778  0.0450    0.694    0.871
##  19.00    57      1   0.764  0.0463    0.679    0.860
##  20.00    54      1   0.750  0.0475    0.662    0.849
##  22.00    51      1   0.735  0.0488    0.646    0.837
##  40.00    16      1   0.689  0.0638    0.575    0.826
##
##               as.numeric(as.character(`pericardial-effusion`))=1
##   time n.risk n.event survival std.err lower 95% CI upper 95% CI
##   0.50    20      2   0.900  0.0671    0.778    1.000
##   0.75    18      1   0.850  0.0798    0.707    1.000
##   1.00    17      2   0.750  0.0968    0.582    0.966
##   2.00    15      1   0.700  0.1025    0.525    0.933
##   5.00    14      1   0.650  0.1067    0.471    0.897
##  21.00    10      1   0.585  0.1141    0.399    0.857
##  28.00     6      1   0.488  0.1302    0.289    0.823
```

```
ggsurvplot(fit = fit2, data = echocardiogram)
```

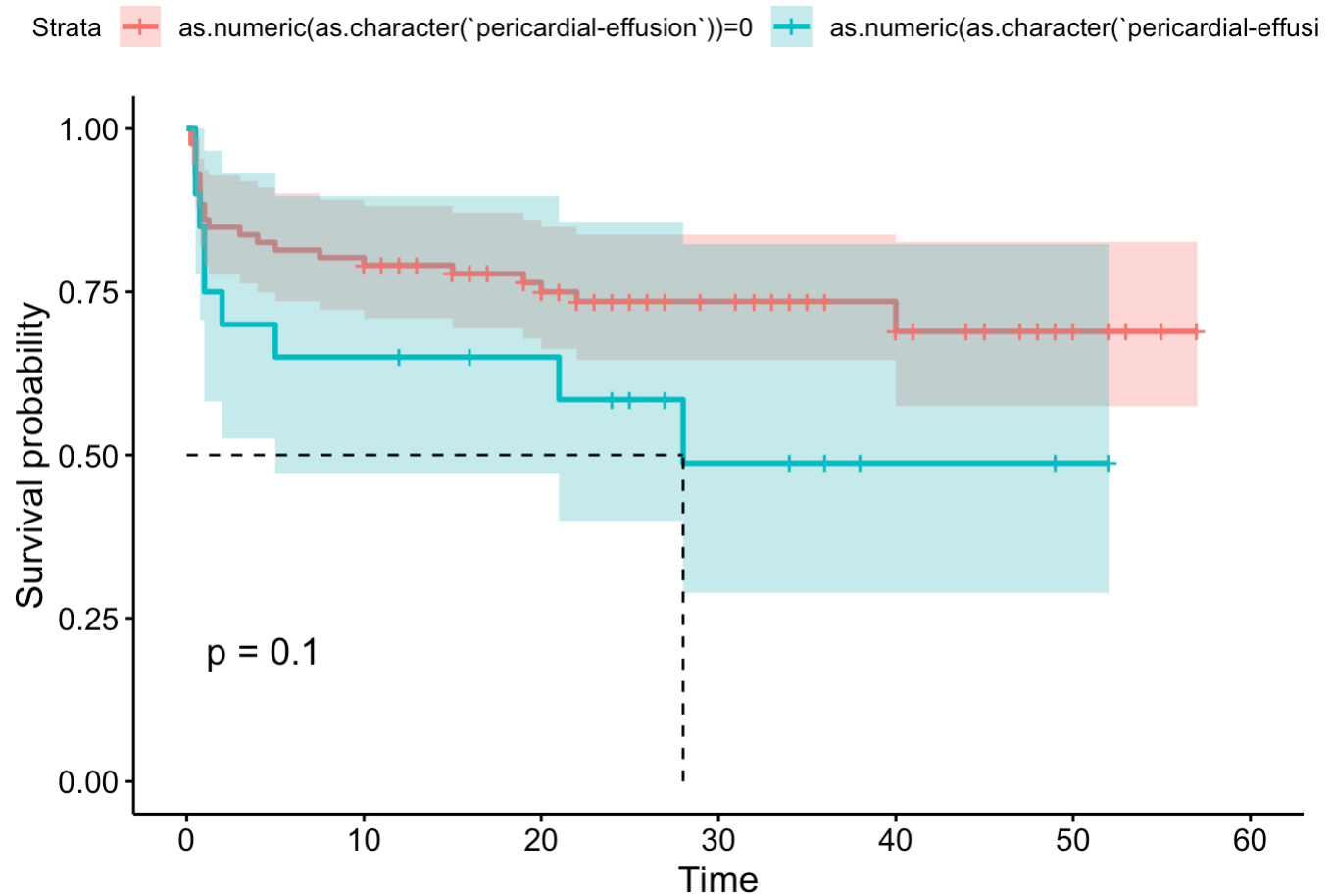


When splitting it between paricardial effusion we can see that having pericardial effusion tends to leave to a lower survival chance over time.

```
dif <- survdiff(sur ~ as.numeric(as.character(`pericardial-effusion`)) , data = echocard
iogram)
dif
```

```
## Call:
## survdiff(formula = sur ~ as.numeric(as.character(`pericardial-effusion`)),
##   data = echocardiogram)
##
##
##               N Observed Expected
## as.numeric(as.character(`pericardial-effusion`))=0 86      23    26.48
## as.numeric(as.character(`pericardial-effusion`))=1 20       9     5.52
##
##               (0-E)^2/E (0-E)^2/V
## as.numeric(as.character(`pericardial-effusion`))=0  0.457    2.71
## as.numeric(as.character(`pericardial-effusion`))=1  2.190    2.71
##
## Chisq= 2.7  on 1 degrees of freedom, p= 0.1
```

```
ggsurvplot(fit = fit2, data = echocardiogram, surv.median.line = 'hv', pval = TRUE, conf.
int = TRUE)
```



When including the error bars we can see that there is a lot of overlap so, there may not be a significant difference in the two groups. By then looking at the log rank test we can see that there is p-value of .1, so there is not significant evidence to prove that the two survival curves are different, at a critical value of .05.

```
write.csv(echocardiogram, 'echo.csv')
```

Python

```
import lifelines as lf
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
```

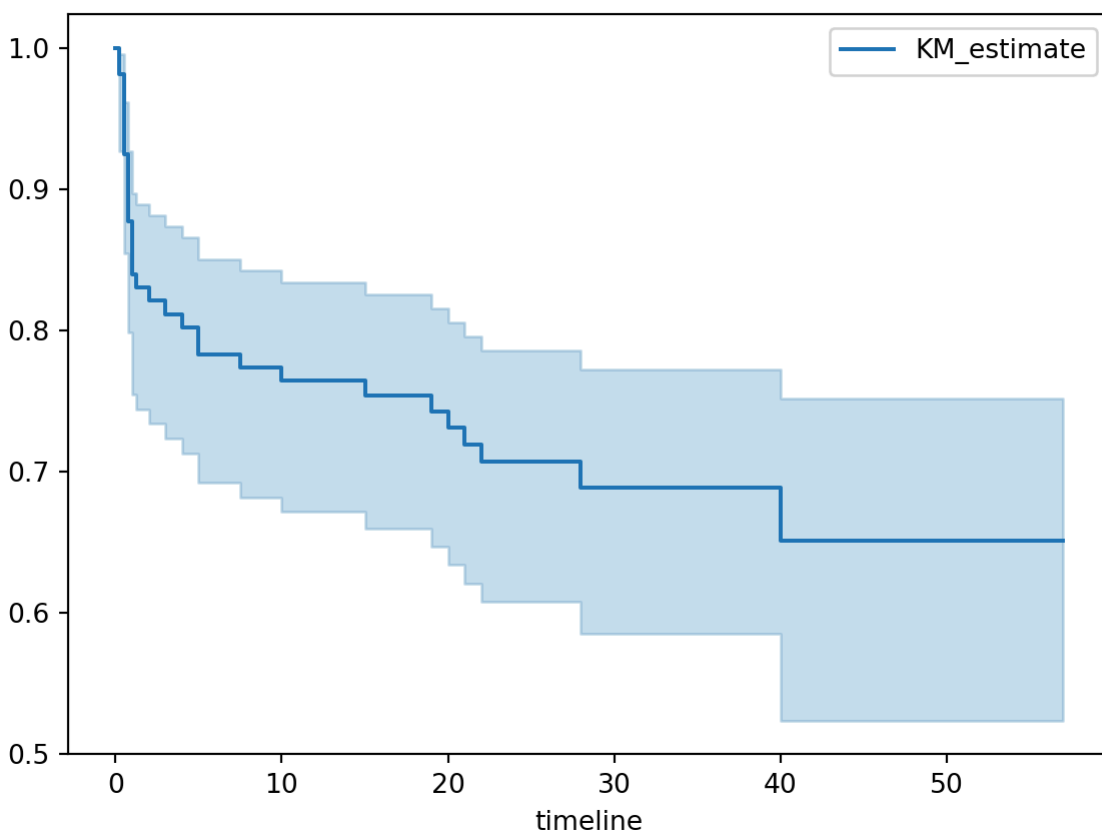
```
echo = pd.read_csv('echo.csv', index_col = 0)
echo.head()
```

```
##      survival  still-alive  ...    mult  alive-at-1
## 1      11.0           0  ...   1.000           0
## 2      19.0           0  ...   0.588           1
## 3      16.0           0  ...   1.000           1
## 4      57.0           0  ...   0.788           1
## 5      19.0           1  ...   0.571           1
##
## [5 rows x 11 columns]
```

```
kmf = lf.KaplanMeierFitter()
kmf.fit(echo['survival'],echo['still-alive'])
```

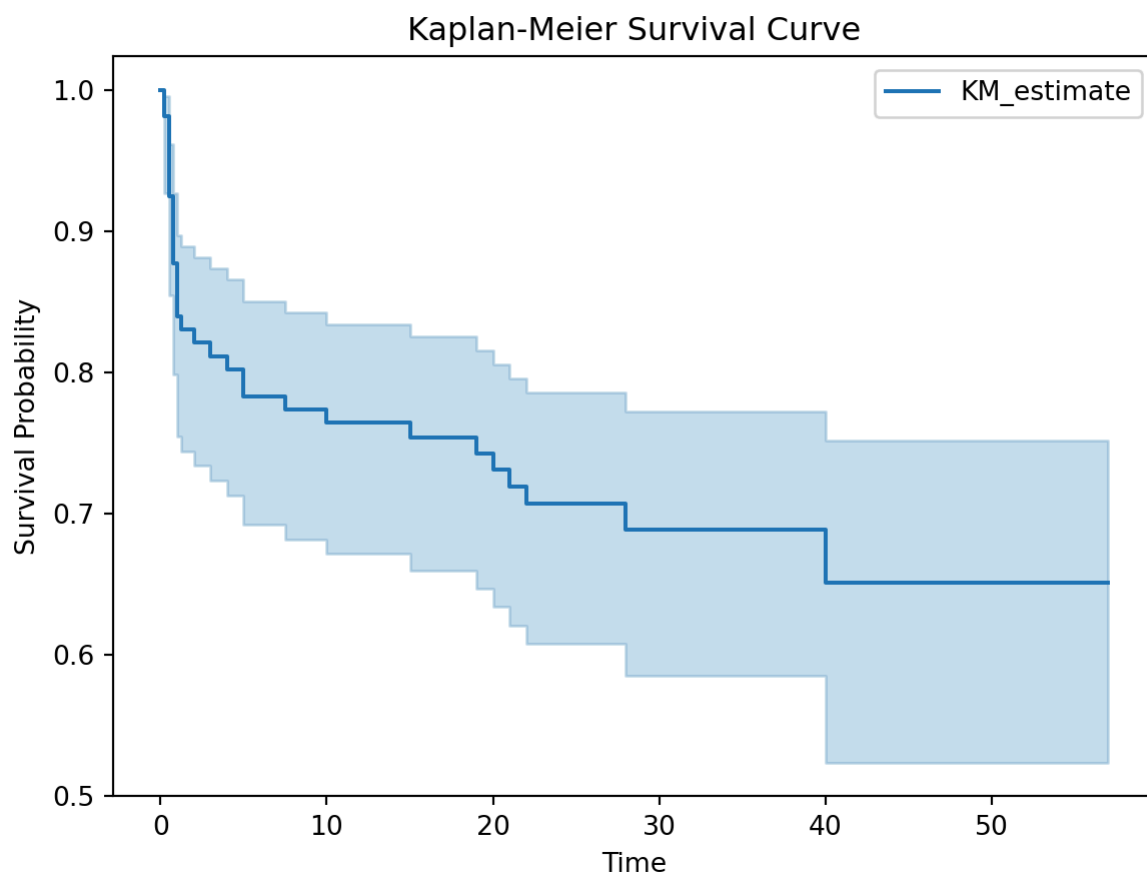
```
## <lifelines.KaplanMeierFitter:"KM_estimate", fitted with 106 total observations, 74 right-censored observations>
```

```
kmf.plot_survival_function()
plt.show()
```



```
plt.close()
```

```
kmf.plot_survival_function(title= 'Kaplan-Meier Survival Curve', xlabel='Time', ylabel
='Survival Probability')
plt.show()
```



```
plt.close()
```

```
ax = plt.subplot(111)
```

```
kmf2 = lf.KaplanMeierFitter()
kmf2.fit(echo.loc[echo['pericardial-effusion'] == 1]['survival'],echo.loc[echo['pericardial-effusion'] == 1]['still-alive'], label = 'Pericardial Effusion')
```

```
## <lifelines.KaplanMeierFitter:"Pericardial Effusion", fitted with 20 total observations, 11 right-censored observations>
```

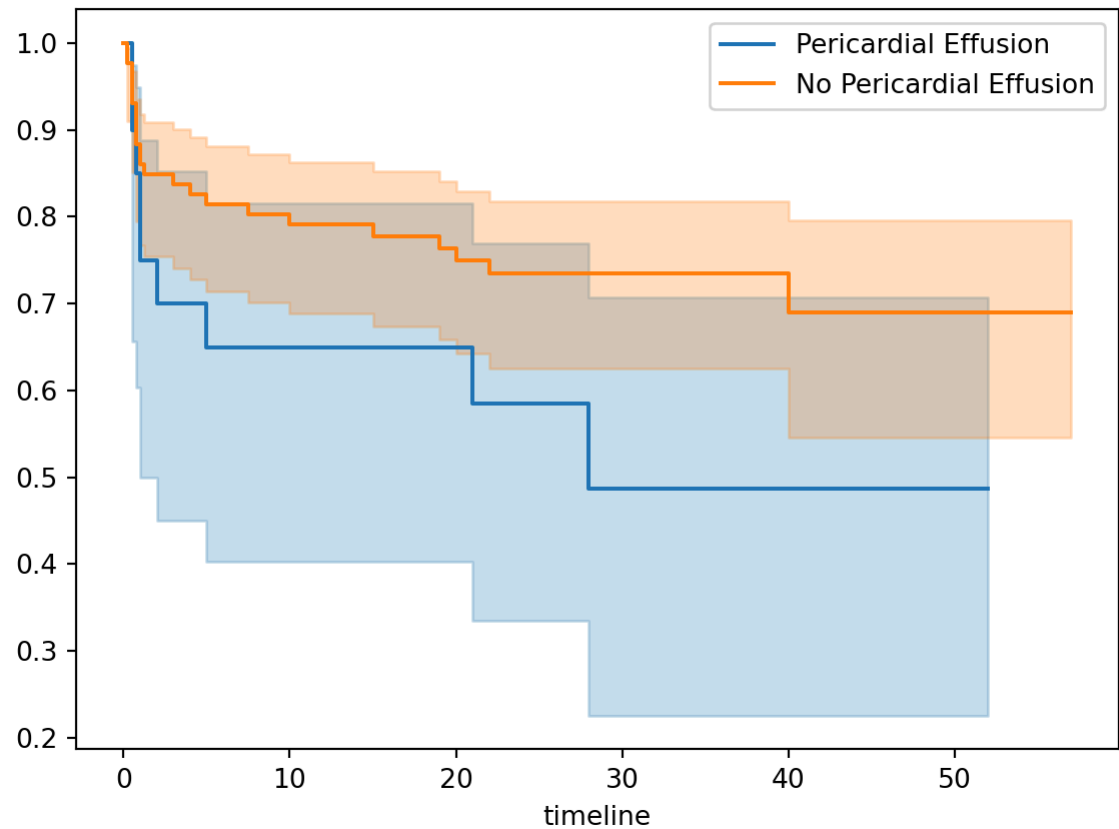
```
kmf2.plot_survival_function(ax=ax)
```

```
kmf2.fit(echo.loc[echo['pericardial-effusion'] == 0]['survival'],echo.loc[echo['pericardial-effusion'] == 0]['still-alive'], label = 'No Pericardial Effusion')
```

```
## <lifelines.KaplanMeierFitter:"No Pericardial Effusion", fitted with 86 total observations, 63 right-censored observations>
```

```
kmf2.plot_survival_function(ax=ax)

plt.show()
```



```
plt.close()
```

```
results = lf.statistics.logrank_test(echo.loc[echo['pericardial-effusion'] == 1]['survival'], echo.loc[echo['pericardial-effusion'] == 0]['survival'], event_observed_A=echo.loc[echo['pericardial-effusion'] == 1]['still-alive'], event_observed_B=echo.loc[echo['pericardial-effusion'] == 0]['still-alive'])
```

results

t_0	-1
null_distribution	chi squared
degrees_of_freedom	1
test_name	logrank_test
test_statistic	p-log2(p)
02.71	0.103.33


```
ax = plt.subplot(111)
```

```
kmf2 = lf.KaplanMeierFitter()  
kmf2.fit(echo.loc[echo['pericardial-effusion'] == 1]['survival'],echo.loc[echo['pericardial-effusion'] == 1]['still-alive'], label = 'Pericardial Effusion')
```

```
## <lifelines.KaplanMeierFitter:"Pericardial Effusion", fitted with 20 total observations, 11 right-censored observations>
```

```
kmf2.plot_survival_function(ax=ax)
```

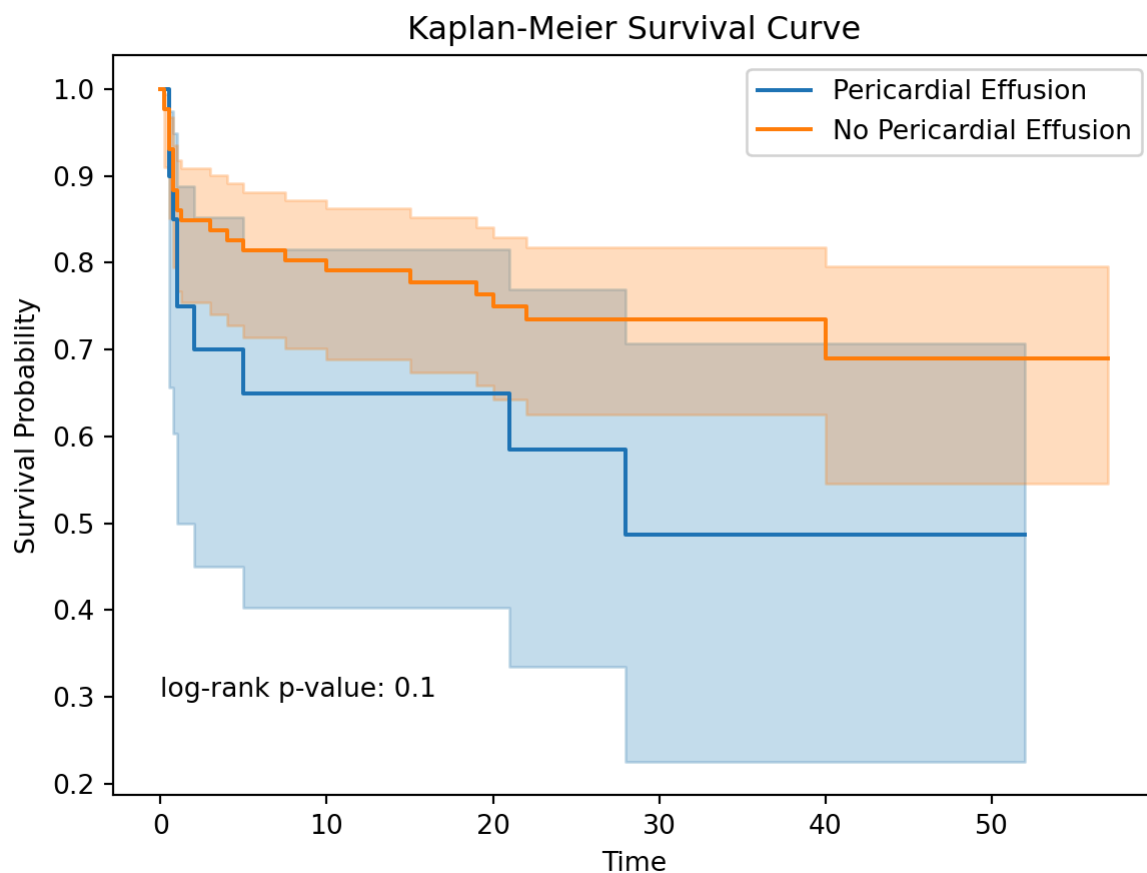
```
kmf2.fit(echo.loc[echo['pericardial-effusion'] == 0]['survival'],echo.loc[echo['pericardial-effusion'] == 0]['still-alive'], label = 'No Pericardial Effusion')
```

```
## <lifelines.KaplanMeierFitter:"No Pericardial Effusion", fitted with 86 total observations, 63 right-censored observations>
```

```
kmf2.plot_survival_function(ax=ax,title= 'Kaplan-Meier Survival Curve', xlabel='Time', ylabel='Survival Probability')
```

```
plt.text(s = "log-rank p-value: {:.2}".format(results.p_value), x = 0, y= .3 )
```

```
plt.show()
```



```
plt.close()
```

We came to the same conclusion as in R where there is not a significant difference in the two groups. There is a p-value of .1 which is not significant at the .05 level.

In the end we found that there is no significant difference in survival rates in the groups with pericardial effusion and without pericardial effusion. The dataset is fairly small, so we might be able to make a more definitive conclusion if we were able to increase the sample size.

References

Echocardiogram. (1989). UCI Machine Learning Repository. <https://doi.org/10.24432/C5QW24> (<https://doi.org/10.24432/C5QW24>).