

# Introduction to TTE modeling: Workbook 2

Kaplan-Meier estimates and plots

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## Preliminaries for R examples

```
library(tidyverse)
library(stringr)
library(survival)
library(survminer)
library(texreg)
library(mgcv)
library(flexsurv)
library(muhaz)
library(Hmisc)

theme_set(theme_bw())
```

## Estimating the survival function

We'll use the `survfit` function to estimate the Survival function.

Using our example data from the last class, let's estimate the overall  $S(t)$ , i.e., not stratifying by any covariates.

```
load('../data/aedat.RDS')

aedat <-
  aedat %>%
  mutate(AETOXGR = factor(aedat$AETOXGR, 0:3, labels=c("None","Mild","Moderate","Severe")),
         ae_any = AETOXGR != 'None') %>%
  group_by(USUBJID) %>%
  # End of study for patients without a severe event
  mutate(TTE_SEVERE = case_when(
    STUDYID=="PROTA" ~ 2,
    STUDYID=="PROTB" ~ 6
  ),
  # Time of severe event for those that had one
  TTE_SEVERE = ifelse(AETOXGR=="Severe", TTE, TTE_SEVERE)
  )
```

```
# Both for EDA and for model-checking, it's generally helpful to have quartiles of exposure:
dat_use <-
  aedat %>% arrange(USUBJID, TTE_SEVERE) %>% slice(1) %>%
  group_by(PBO) %>%
  mutate(Quartile = ifelse(PBO == "PBO", "PBO",
                           paste0("Q", ntile(CAVGSS, n = 4)))) %>%
  ungroup() %>%
  mutate(rowid = 1:n())
```

The key elements:

- **survfit**: the function used to obtain the K-M (or Flemming-Harrington) estimate of the survival function
  - `stype=1` (the default) gives the K-M estimate
  - `stype=2` gives the Fleming-Harrington estimate
- **Surv** function for defining the outcome
  - For right-censored data, the first argument is the observed event time variable, and the second argument is the even indicator (1 or TRUE for an event; 0 or FALSE for censoring)

Obtaining the estimate and looking at some output:

```
km_est = survfit(Surv(TTE,ae_any)~1, data = dat_use)
```

```
print(km_est)
```

```
. Call: survfit(formula = Surv(TTE, ae_any) ~ 1, data = dat_use)
```

```
.
.      n events median 0.95LCL 0.95UCL
. [1,] 180      132   0.32   0.139   0.649
```

```
summary(km_est)
```

```
. Call: survfit(formula = Surv(TTE, ae_any) ~ 1, data = dat_use)
```

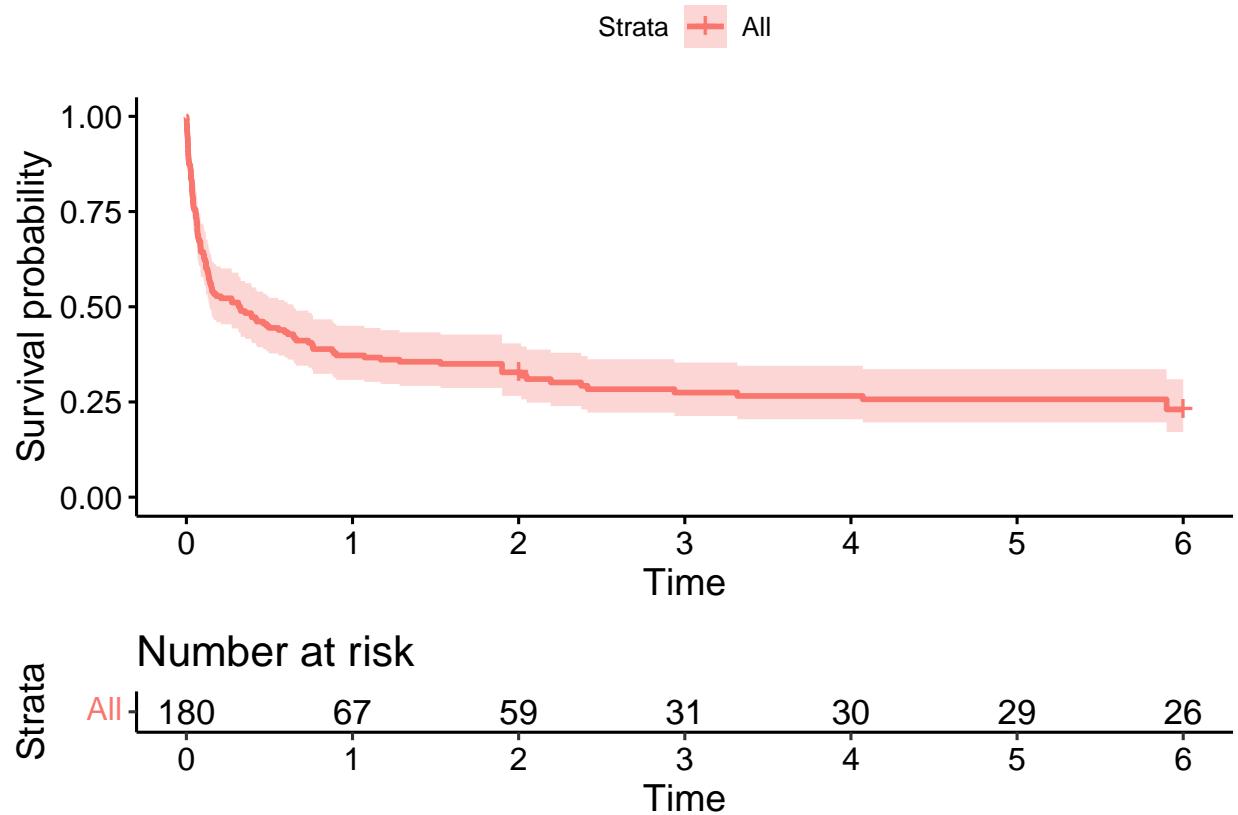
```
.
.      time n.risk n.event survival std.err lower 95% CI upper 95% CI
. 0.000805  180      1   0.994 0.00554   0.984   1.000
. 0.001935  179      1   0.989 0.00781   0.974   1.000
. 0.002245  178      1   0.983 0.00954   0.965   1.000
. 0.002302  177      1   0.978 0.01099   0.956   1.000
. 0.003077  176      1   0.972 0.01225   0.949   0.997
. 0.003199  175      1   0.967 0.01338   0.941   0.993
. 0.004871  174      1   0.961 0.01441   0.933   0.990
. 0.005289  173      1   0.956 0.01536   0.926   0.986
. 0.006043  172      1   0.950 0.01624   0.919   0.982
. 0.006828  171      1   0.944 0.01707   0.912   0.979
. 0.008177  170      1   0.939 0.01785   0.905   0.975
. 0.008267  169      1   0.933 0.01859   0.898   0.970
. 0.008707  168      1   0.928 0.01929   0.891   0.966
. 0.008826  167      1   0.922 0.01996   0.884   0.962
. 0.010071  166      1   0.917 0.02060   0.877   0.958
. 0.010159  165      1   0.911 0.02121   0.870   0.954
. 0.010784  164      1   0.906 0.02180   0.864   0.949
. 0.010916  163      1   0.900 0.02236   0.857   0.945
. 0.011968  162      1   0.894 0.02290   0.851   0.940
. 0.012724  161      1   0.889 0.02342   0.844   0.936
. 0.013258  160      1   0.883 0.02393   0.838   0.931
```

. 0.014633	159	1	0.878	0.02441	0.831	0.927
. 0.016949	158	1	0.872	0.02488	0.825	0.922
. 0.023699	157	1	0.867	0.02534	0.818	0.918
. 0.023706	156	1	0.861	0.02578	0.812	0.913
. 0.026148	155	1	0.856	0.02620	0.806	0.908
. 0.026399	154	1	0.850	0.02661	0.799	0.904
. 0.026941	153	1	0.844	0.02701	0.793	0.899
. 0.027425	152	1	0.839	0.02740	0.787	0.894
. 0.028619	151	1	0.833	0.02778	0.781	0.890
. 0.032233	150	1	0.828	0.02814	0.774	0.885
. 0.032706	149	1	0.822	0.02850	0.768	0.880
. 0.032818	148	1	0.817	0.02884	0.762	0.875
. 0.033797	147	1	0.811	0.02917	0.756	0.870
. 0.034673	146	1	0.806	0.02950	0.750	0.865
. 0.035457	145	1	0.800	0.02981	0.744	0.861
. 0.035762	144	1	0.794	0.03012	0.738	0.856
. 0.036722	143	1	0.789	0.03042	0.731	0.851
. 0.039233	142	1	0.783	0.03071	0.725	0.846
. 0.039720	141	1	0.778	0.03099	0.719	0.841
. 0.040568	140	1	0.772	0.03126	0.713	0.836
. 0.041805	139	1	0.767	0.03153	0.707	0.831
. 0.042022	138	1	0.761	0.03178	0.701	0.826
. 0.045127	137	1	0.756	0.03203	0.695	0.821
. 0.055222	136	1	0.750	0.03227	0.689	0.816
. 0.055578	135	1	0.744	0.03251	0.683	0.811
. 0.057453	134	1	0.739	0.03274	0.677	0.806
. 0.057879	133	1	0.733	0.03296	0.671	0.801
. 0.061409	132	1	0.728	0.03318	0.666	0.796
. 0.061810	131	1	0.722	0.03338	0.660	0.791
. 0.063174	130	1	0.717	0.03359	0.654	0.786
. 0.065723	129	1	0.711	0.03378	0.648	0.781
. 0.065771	128	1	0.706	0.03397	0.642	0.775
. 0.066141	127	1	0.700	0.03416	0.636	0.770
. 0.066437	126	1	0.694	0.03433	0.630	0.765
. 0.066679	125	1	0.689	0.03451	0.624	0.760
. 0.070688	124	1	0.683	0.03467	0.619	0.755
. 0.072054	123	1	0.678	0.03483	0.613	0.750
. 0.075189	122	1	0.672	0.03499	0.607	0.744
. 0.084249	121	1	0.667	0.03514	0.601	0.739
. 0.084579	120	1	0.661	0.03528	0.595	0.734
. 0.084672	119	1	0.656	0.03542	0.590	0.729
. 0.086239	118	1	0.650	0.03555	0.584	0.724
. 0.086305	117	1	0.644	0.03568	0.578	0.718
. 0.101183	116	1	0.639	0.03580	0.572	0.713
. 0.104850	115	1	0.633	0.03592	0.567	0.708
. 0.106140	114	1	0.628	0.03603	0.561	0.703
. 0.109378	113	1	0.622	0.03614	0.555	0.697
. 0.115346	112	1	0.617	0.03624	0.550	0.692
. 0.117250	111	1	0.611	0.03634	0.544	0.687
. 0.117528	110	1	0.606	0.03643	0.538	0.681
. 0.118619	109	1	0.600	0.03651	0.533	0.676
. 0.127254	108	1	0.594	0.03660	0.527	0.671
. 0.127997	107	1	0.589	0.03667	0.521	0.665
. 0.133053	106	1	0.583	0.03675	0.516	0.660

. 0.133380	105	1	0.578 0.03681	0.510	0.655
. 0.136157	104	1	0.572 0.03688	0.504	0.649
. 0.139387	103	1	0.567 0.03694	0.499	0.644
. 0.141296	102	1	0.561 0.03699	0.493	0.638
. 0.149190	101	1	0.556 0.03704	0.488	0.633
. 0.150323	100	1	0.550 0.03708	0.482	0.628
. 0.153820	99	1	0.544 0.03712	0.476	0.622
. 0.155550	98	1	0.539 0.03715	0.471	0.617
. 0.165857	97	1	0.533 0.03718	0.465	0.611
. 0.179546	96	1	0.528 0.03721	0.460	0.606
. 0.207173	95	1	0.522 0.03723	0.454	0.601
. 0.272220	94	1	0.517 0.03725	0.449	0.595
. 0.273933	93	1	0.511 0.03726	0.443	0.590
. 0.312540	92	1	0.506 0.03727	0.438	0.584
. 0.314848	91	1	0.500 0.03727	0.432	0.579
. 0.325156	90	1	0.494 0.03727	0.427	0.573
. 0.327124	89	1	0.489 0.03726	0.421	0.568
. 0.353456	88	1	0.483 0.03725	0.416	0.562
. 0.389702	87	1	0.478 0.03723	0.410	0.557
. 0.393778	86	1	0.472 0.03721	0.405	0.551
. 0.416938	85	1	0.467 0.03718	0.399	0.546
. 0.423652	84	1	0.461 0.03715	0.394	0.540
. 0.462944	83	1	0.456 0.03712	0.388	0.534
. 0.481351	82	1	0.450 0.03708	0.383	0.529
. 0.495438	81	1	0.444 0.03704	0.377	0.523
. 0.553784	80	1	0.439 0.03699	0.372	0.518
. 0.591294	79	1	0.433 0.03694	0.367	0.512
. 0.609558	78	1	0.428 0.03688	0.361	0.507
. 0.643238	77	1	0.422 0.03681	0.356	0.501
. 0.649290	76	1	0.417 0.03675	0.351	0.495
. 0.660878	75	1	0.411 0.03667	0.345	0.490
. 0.734580	74	1	0.406 0.03660	0.340	0.484
. 0.754551	73	1	0.400 0.03651	0.334	0.478
. 0.761845	72	1	0.394 0.03643	0.329	0.473
. 0.762268	71	1	0.389 0.03634	0.324	0.467
. 0.878613	70	1	0.383 0.03624	0.318	0.461
. 0.886580	69	1	0.378 0.03614	0.313	0.456
. 0.902642	68	1	0.372 0.03603	0.308	0.450
. 1.071367	67	1	0.367 0.03592	0.303	0.444
. 1.169048	66	1	0.361 0.03580	0.297	0.439
. 1.283269	65	1	0.356 0.03568	0.292	0.433
. 1.529470	64	1	0.350 0.03555	0.287	0.427
. 1.900000	63	4	0.328 0.03499	0.266	0.404
. 2.014711	37	1	0.319 0.03515	0.257	0.396
. 2.047774	36	1	0.310 0.03527	0.248	0.387
. 2.192981	35	1	0.301 0.03536	0.239	0.379
. 2.376302	34	1	0.292 0.03541	0.231	0.371
. 2.414482	33	1	0.283 0.03543	0.222	0.362
. 2.937613	32	1	0.275 0.03541	0.213	0.354
. 3.316241	31	1	0.266 0.03536	0.205	0.345
. 4.071150	30	1	0.257 0.03527	0.196	0.336
. 5.900000	29	3	0.230 0.03480	0.171	0.310

Plotting the estimate using `ggsurvplot`:

```
ggsurvplot(km_est, risk.table = TRUE)
```



Often it's helpful to add the number of subjects at risk to the bottom of the plot (a risk table).

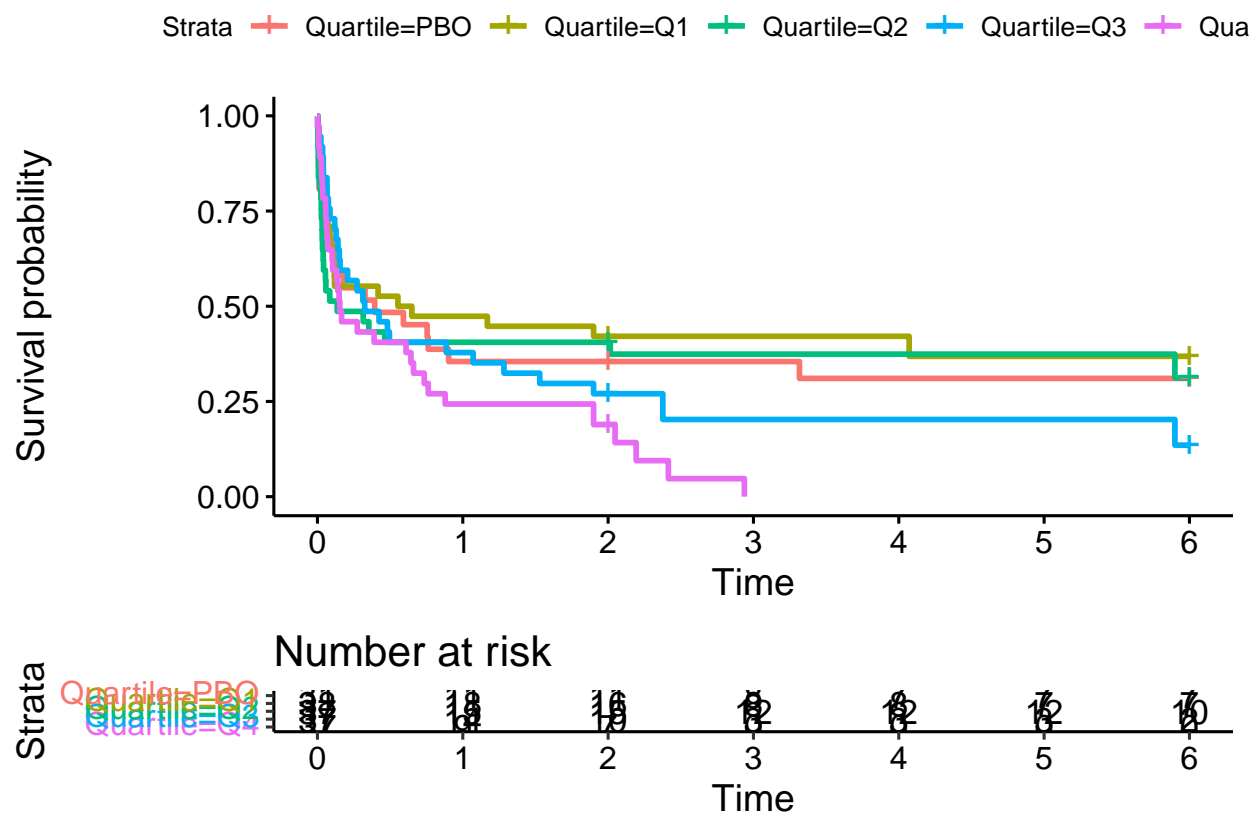
#### Exercise:

1. Obtain the Kaplan-Meier estimates for time to any AE by exposure quartile. Hint: the right hand side of the formula will be “~ Quartile”

```
km_est_css = survfit(Surv(TTE,ae_any)~Quartile, data = dat_use)
```

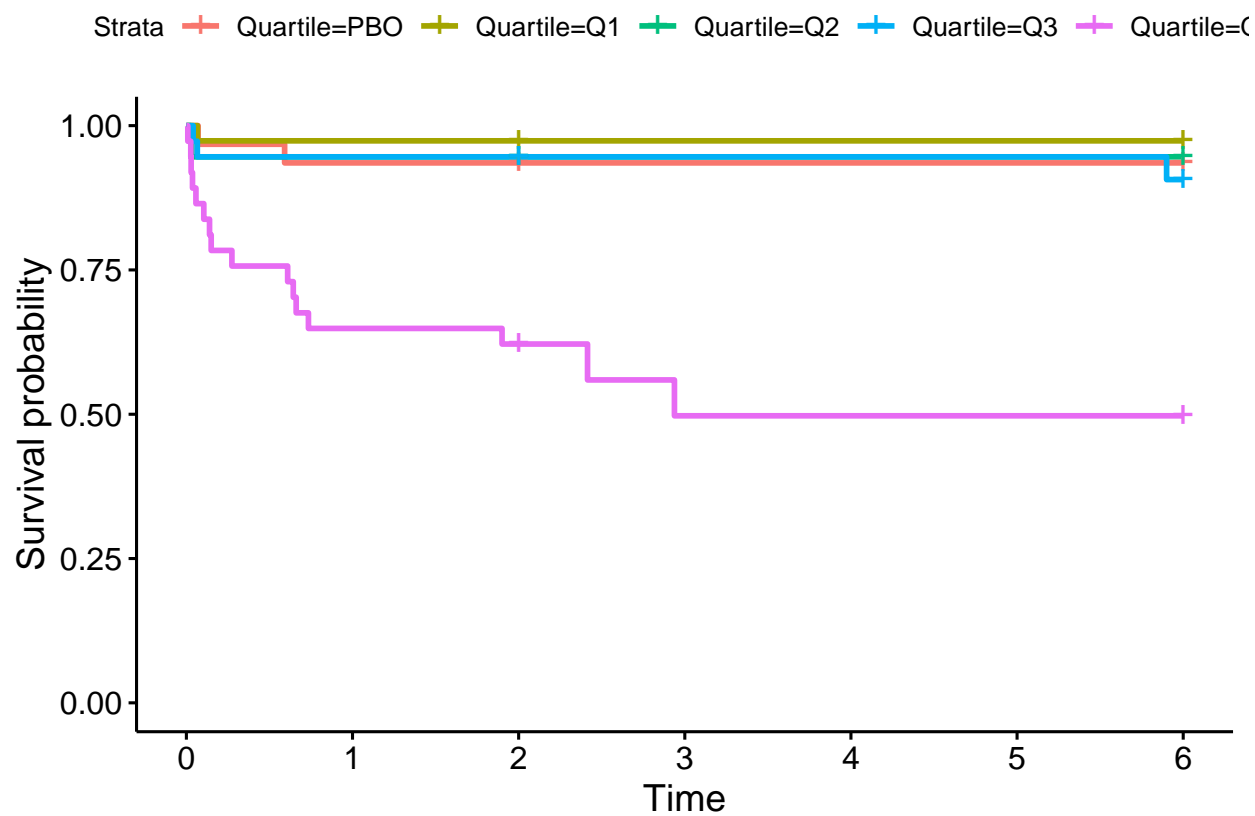
2. Plot the estimates using ggsurvplot. Do you see any evidence of an exposure-response relationship?

```
ggsurvplot(km_est_css, risk.table = TRUE)
```



- Repeat steps 1 and 2 for time to a severe AE (event time variable is TTE\_SEVERE, and the event indicator variable is AE01). Do you see any evidence of an exposure-response relationship?

```
km_est_severe_css = survfit(Surv(TTE_SEVERE,AE01)~Quartile, data = dat_use)
ggsurvplot(km_est_severe_css)
```



3. Extra: Plot the K-M estimates by patient type, faceted by exposure quartile. Hint: Fit model `Surv(TTE,AY_any)~PTTYPE` and use the `facet.by` argument to `ggsurvplot`

```
fit <- survfit(Surv(TTE_SEVERE,AE01)~PTTYPE, data=dat_use)
ggsurvplot(fit, facet.by = 'Quartile', data=dat_use)
```

