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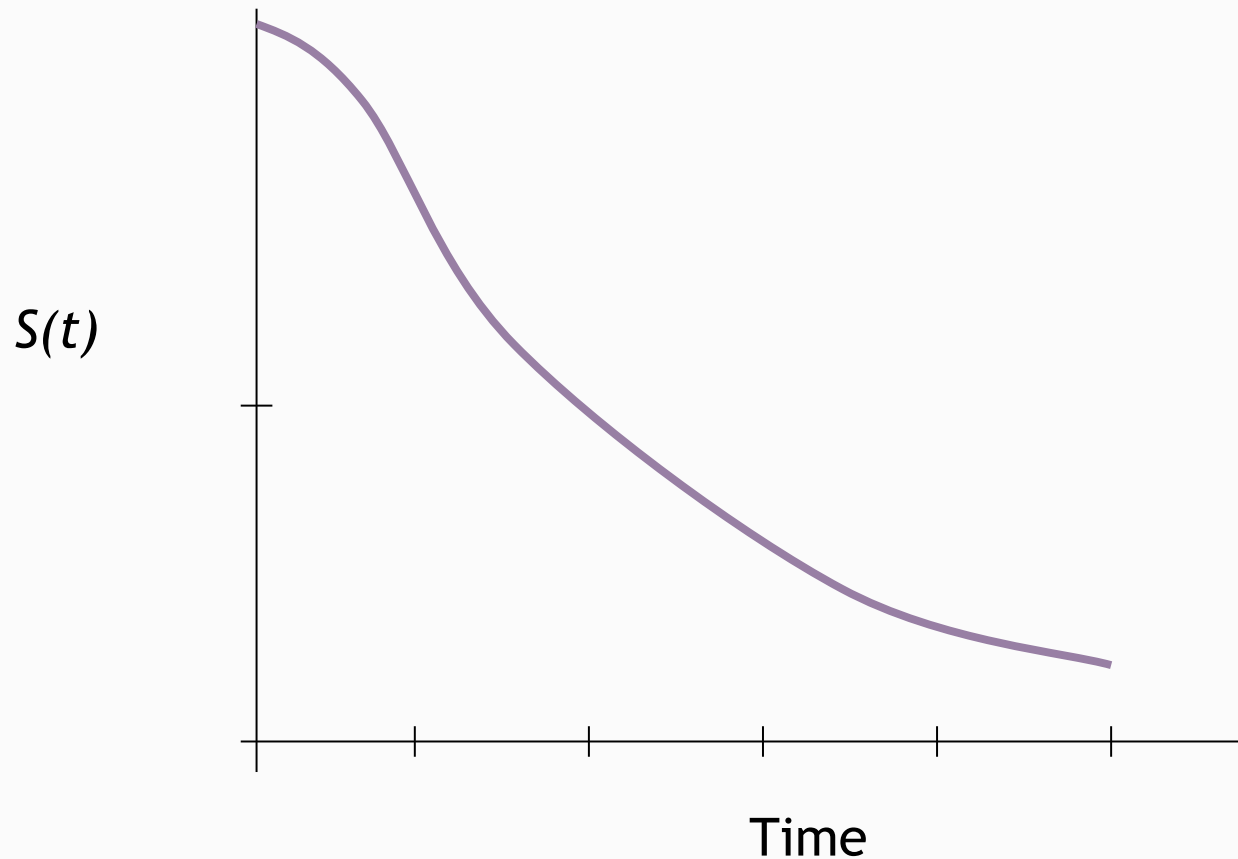
## Section B

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Estimating the Survival Curve: The Kaplan Meier Approach

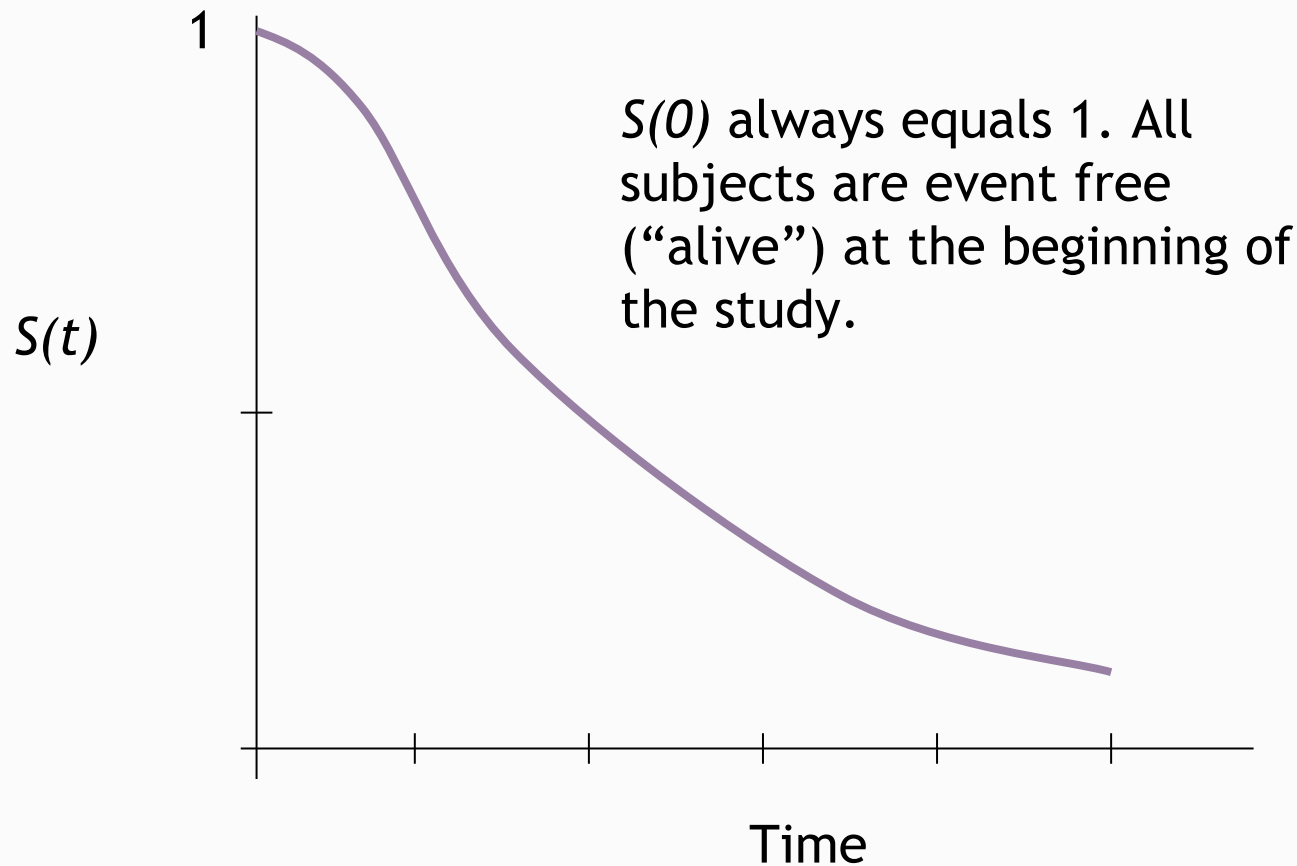
# Central Problem

- Estimation of the “survival curve”
- $S(t)$  = proportion remaining event free (surviving) at least to time  $t$  or beyond



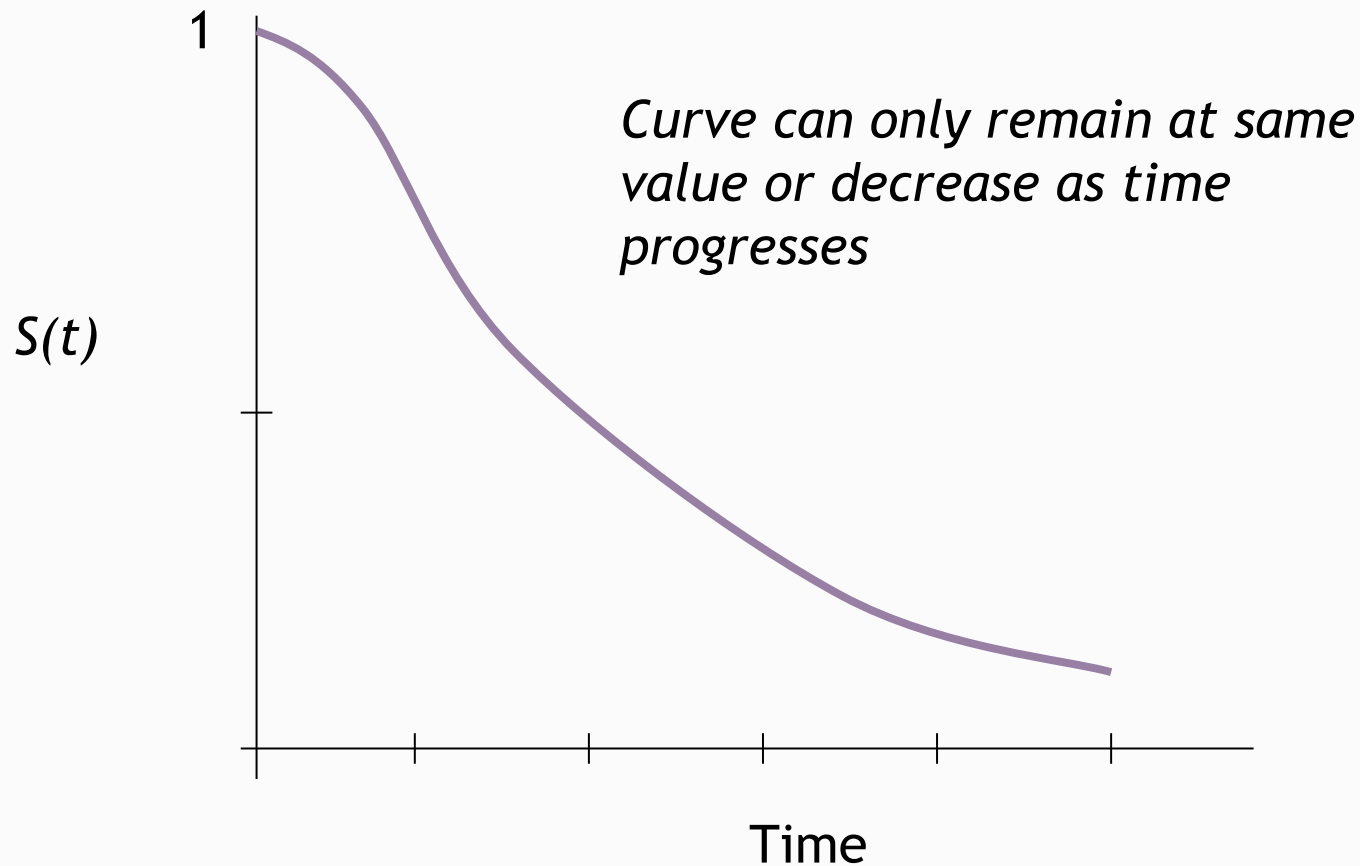
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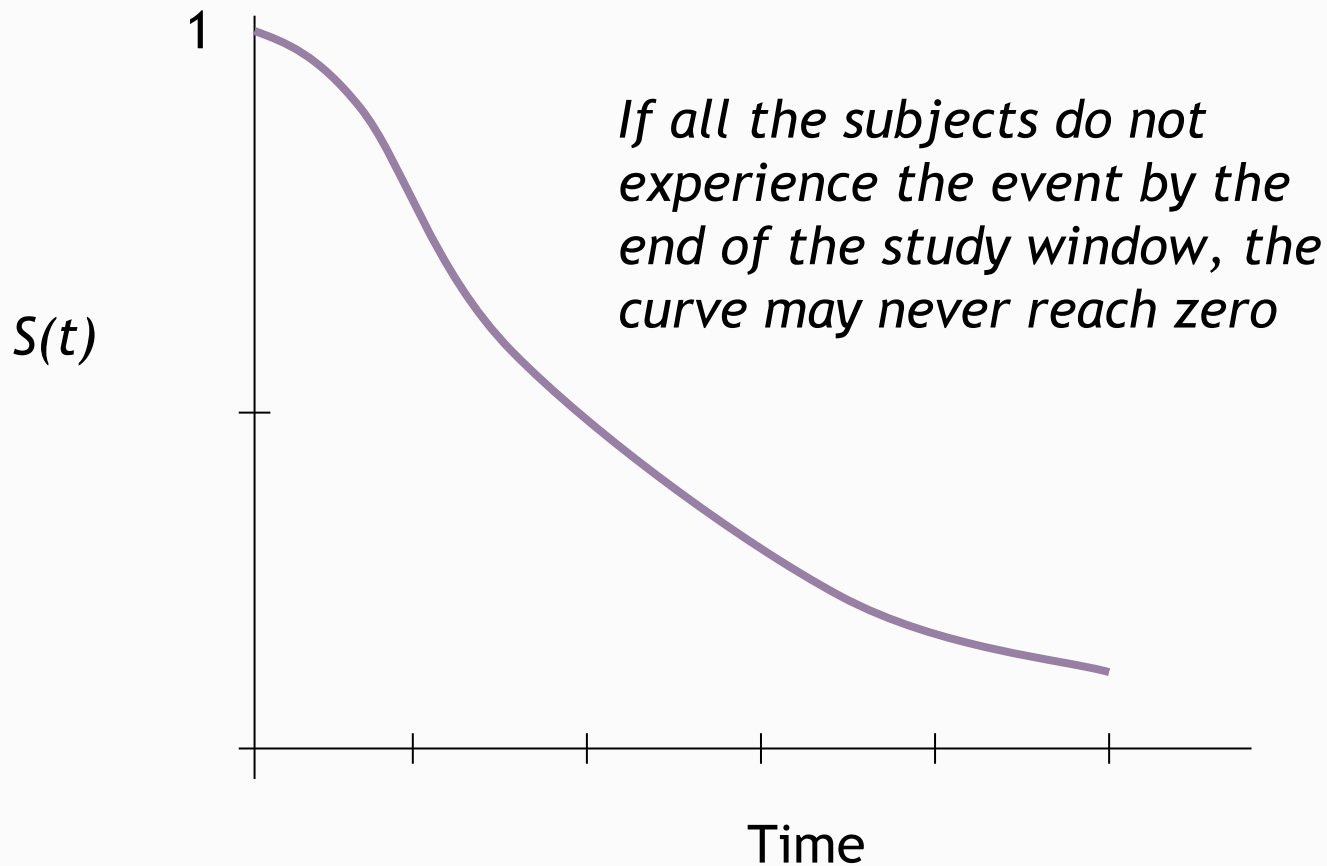
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# Central Problem

- Estimation of the “survival curve”
  - $S(t)$  = proportion remaining event free (surviving) at least to time  $t$  or beyond
  - We can estimate  $S(t)$  from a sample of data: our statistic is  $\hat{S}(t)$



# Approaches

- Life table method
  - Grouped in intervals
- Kaplan-Meier (1958)
  - Ungrouped data
  - Small samples

# Kaplan-Meier Estimate

- Example: time (months) from primary AIDS diagnosis for a sample of 12 hemophiliac patients under 40 years old at time of HIV seroconversion\*
  - Event times ( $n = 12$ ):
  - 2 3+ 5 6 7+ 10 15+ 16 16 27 30 32

Notes: \* Example based on data taken from Rosner, B. (1990). Fundamentals of biostatistics, 6<sup>th</sup> ed. (2005). Duxbury Press. (based on research by Ragni, et al. (1990). Cumulative risk for AIDS in *Journal of Acquired Immune Deficiency Syndromes*, Vol. 3.



# Kaplan-Meier Estimate

- $\hat{S}(t) = 1$ , to start
- After starting at time 0, curve can be estimated at each event time  $t$ , but not at censoring times

$$\hat{S}(t) = \left( \frac{N(t) - E(t)}{N(t)} \right) \times \hat{S}(\textit{Previous Event Time})$$

- $E(t) = \#$  events at time  $t$
- $N(t) = \#$  subjects at risk for event at time  $t$

# Kaplan-Meier Estimate

- Curve can be estimated at each event, but not at censoring times

$$\hat{S}(t) = \left( \frac{N(t) - E(t)}{N(t)} \right) \times \hat{S}(\text{Previous Event Time})$$

Proportion of original sample making it to  
time  $t$

# Kaplan-Meier Estimate

- Curve can be estimated at each event, but not at censoring times

$$\hat{S}(t) = \left( \frac{N(t) - E(t)}{N(t)} \right) \times \hat{S}(\textit{Previous Event Time})$$



Proportion surviving to time  $t$  who survive beyond time  $t$

# Kaplan-Meier Estimate

- Start estimate at first (ordered) event time
  - 2 3+ 6 6 7+ 10 15+ 15 16 27 30 32

$$\hat{S}(2) = \left( \frac{N(2) - E(2)}{N(2)} \right) = \frac{12 - 1}{12} = \frac{11}{12} = .92$$

# Kaplan-Meier Estimate

- Can estimate  $S(t)$  at each subsequent event time
  - (Censoring times inform estimate about number at risk of having the event at a time  $t$  until censoring occurs)
  - 2 3+ 6 6 7+ 10 15+ 15 16 27 30 32

$$\hat{S}(6) = \left( \frac{N(6) - E(6)}{N(6)} \right) \times \hat{S}(2) = \left( \frac{10 - 2}{10} \right) \times .92 = .80 \times .92 = .74$$

# Kaplan-Meier Estimate

- Can estimate  $S(t)$  at each subsequent event time
  - (Censoring times inform estimate about the number at risk of having the event at a time  $t$ )
  - 2 3+ 6 6 7+ 10 15+ 15 16 27 30 32

$$\hat{S}(10) = \left( \frac{N(10) - E(10)}{N(10)} \right) \times \hat{S}(6) = \left( \frac{7 - 1}{6} \right) \times .74 = .86 \times .74 = .64$$

# Kaplan-Meier Estimate

- Continue through final event time

$t$	$\hat{S}(t)$
2	.92
6	.74
10	.64
15	.52
16	.39
27	.26
30	.13
32	0

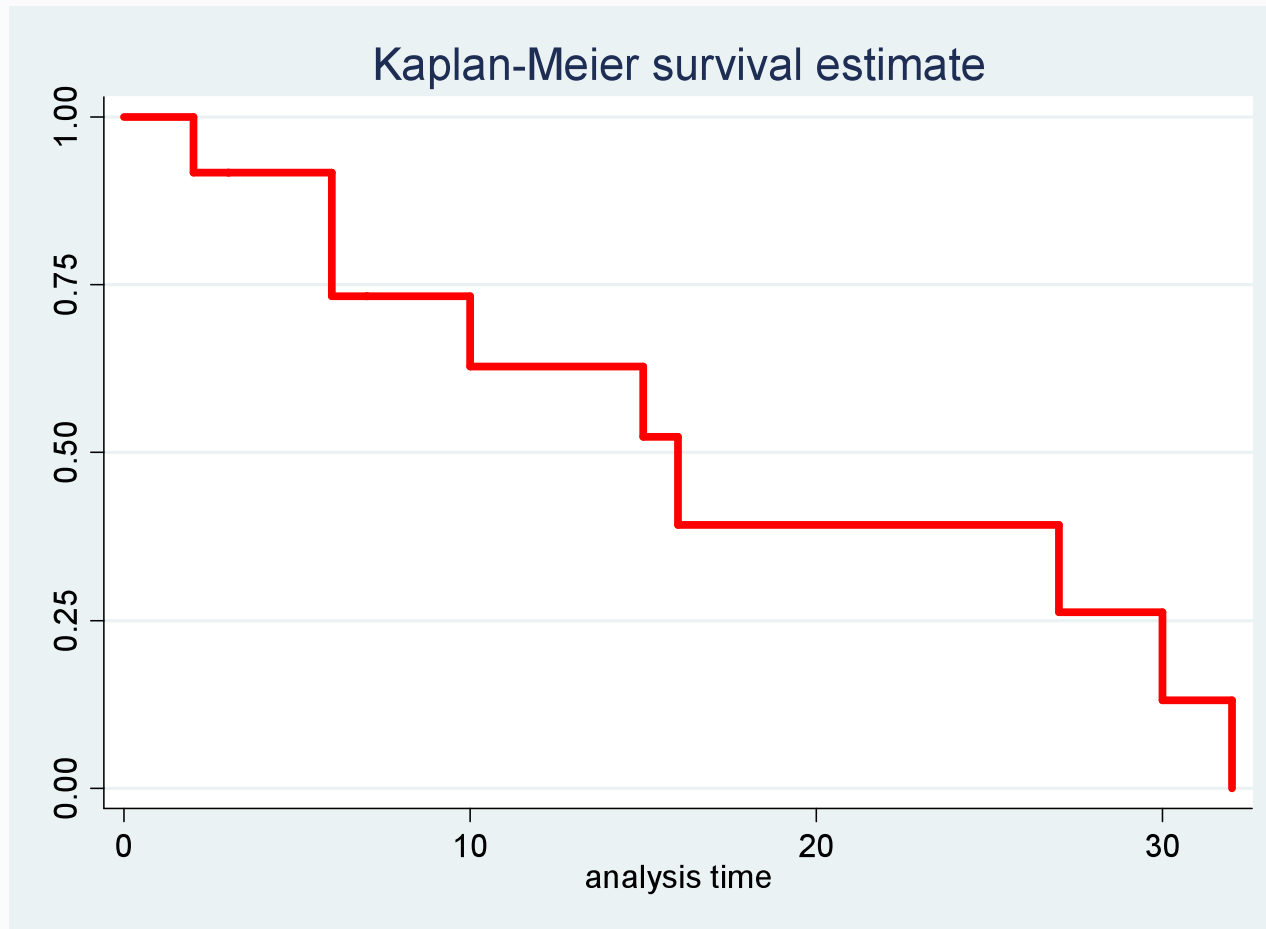
# Kaplan-Meier Estimate

- Graph is a step function
- “Jumps” at each observed event time
- Nothing is assumed about curved shape between each observed event time



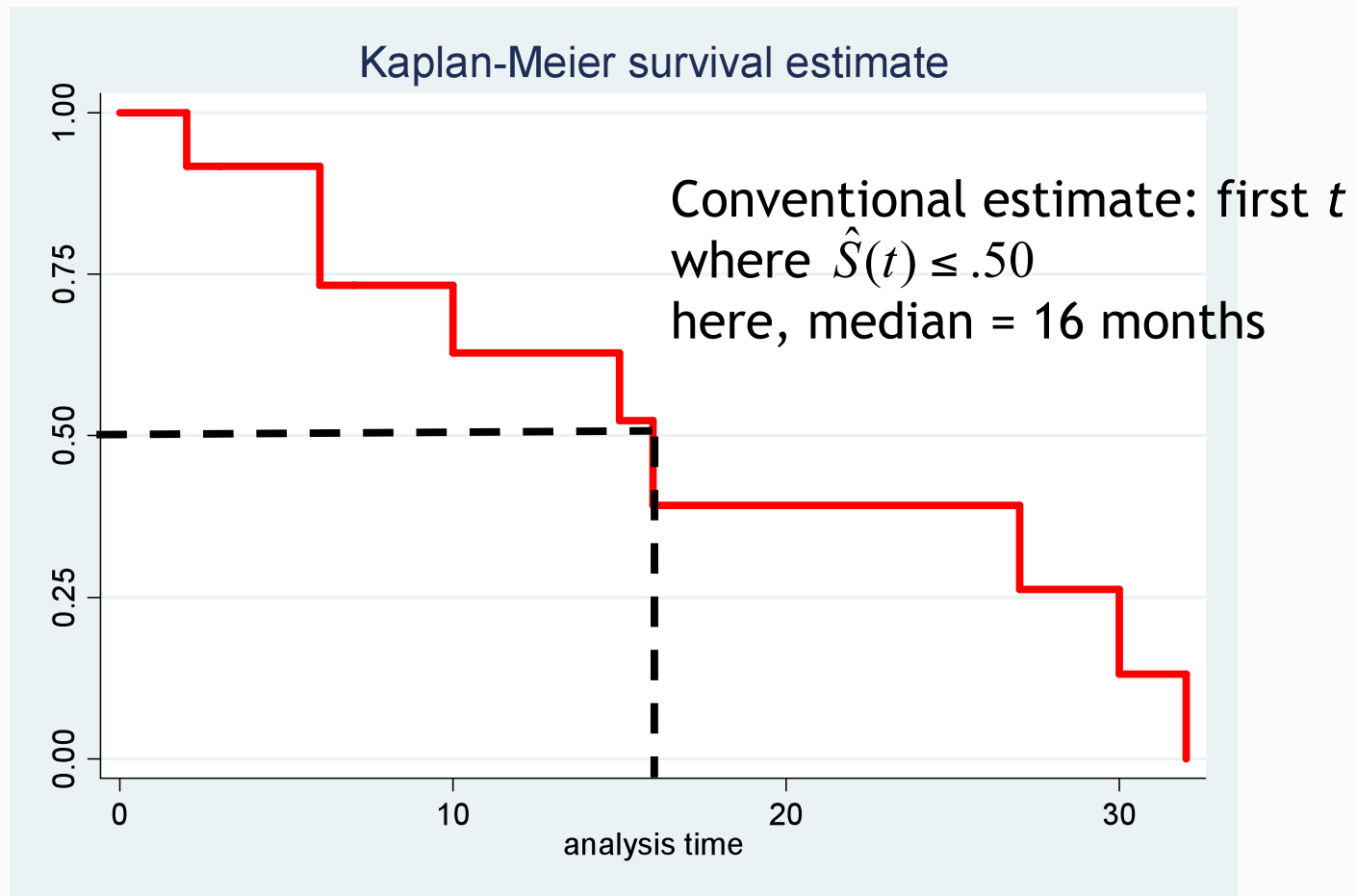
# Kaplan-Meier Estimate

- Kaplan-Meier estimate graphically presented



# Kaplan-Meier Estimate

- You can use these to estimate single number summary statistics, like the median survival time (median time remaining event free)



# Kaplan-Meier Estimate

- Example
  - Time days to resuming smoking in first month following completion of five one-hour group coaching sessions on smoking cessation (10 subjects)
  - 15   3+   30+   5   10+   30+   7   1   24+   27

# Kaplan-Meier Estimate

- Example
  - Time days to resuming smoking in first thirty day period following completion of five one-hour group coaching sessions on smoking cessation (10 subjects): ordered times
  - 1 3+ 5 7 10+ 15 24+ 27 30+ 30+

# Kaplan-Meier Estimate

- Example

- Time days to resuming smoking in first thirty day period following completion of five one-hour group coaching sessions on smoking cessation (10 subjects): ordered times

- 1 3+ 5 7 10+ 15 24+ 27 30+ 30+

- $\hat{S}(1) = \left( \frac{N(1) - E(1)}{N(1)} \right) = \frac{10 - 1}{10} = \frac{9}{10} = .90$

# Kaplan-Meier Estimate

- Example

- Time days to resuming smoking in first thirty day period following completion of five one-hour group coaching sessions on smoking cessation (10 subjects): ordered times

- 1 3+ 5 7 10+ 15 24+ 27 30+ 30+

- $\hat{S}(5) = \left( \frac{N(5) - E(5)}{N(5)} \right) \times \hat{S}(1) = \left( \frac{8 - 1}{8} \right) \times .90 = .88 \times .90 = .79$

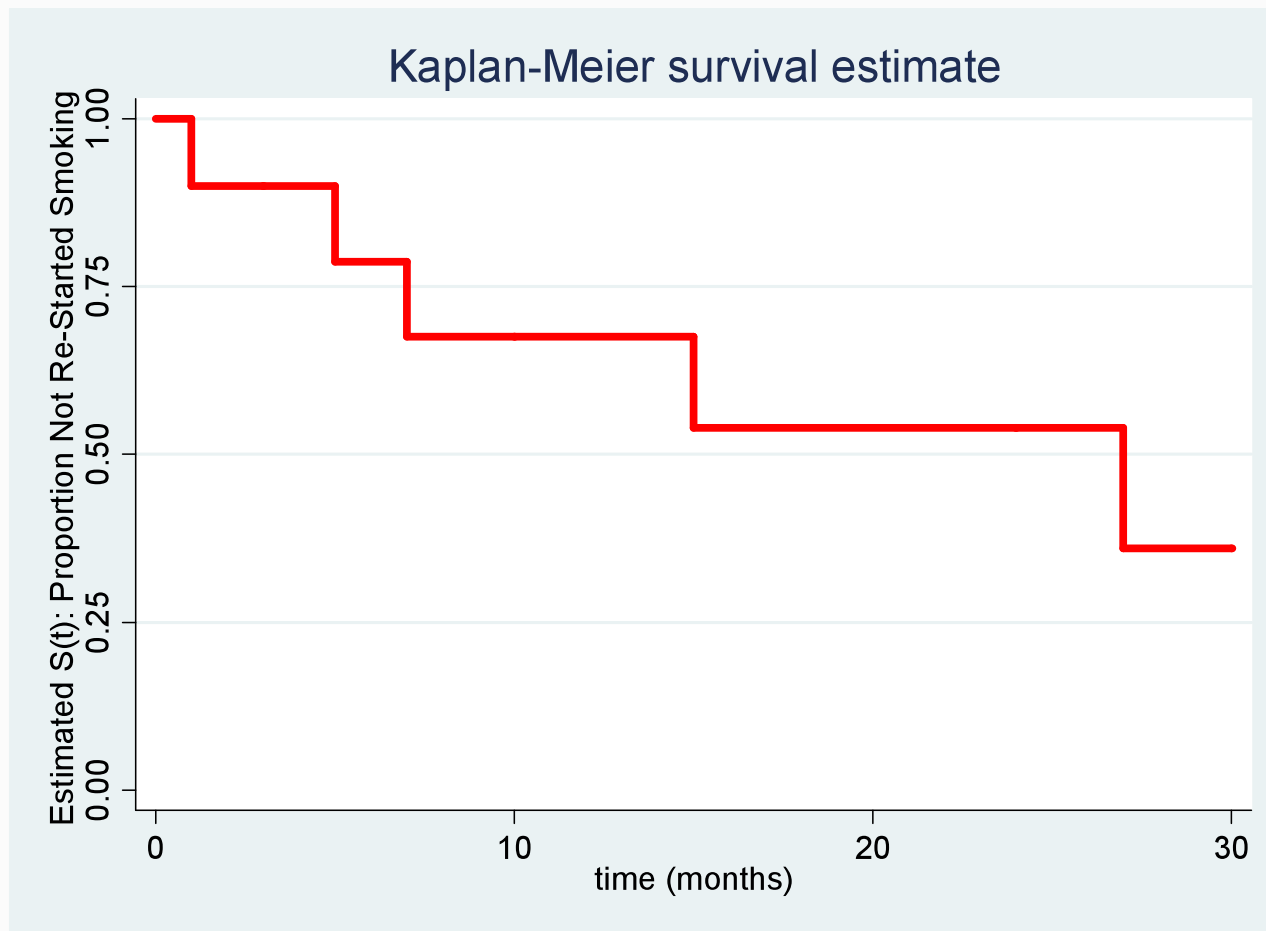
# Kaplan-Meier Estimate

- Continue through final event time: notice this estimated curve never reaches 0 because largest time values are censoring times

$t$	$\hat{S}(t)$
1	.90
5	.79
7	.68
15	.54
27	.36

# Kaplan-Meier Estimate

- Graphical presentation





# Big Assumption

- Independence of censoring and survival
- Those censored at time  $t$  have the same prognosis as those not censored at  $t$
- Examples of possible violations
  - Time to tumor—animal
  - Occupational health—loss to follow up