## Definition

 $\tilde{C}_i$ : True censoring indicator ( $\tilde{C}_i = 0$  if censored,  $\tilde{C}_i = 1$  otherwise)

 $C_i$ : Observed censoring indicator ( $C_i = 0$  if censored,  $C_i = 1$  otherwise)

## Model setting

 $\alpha_i$ : Probability that the observation i is truly censored.

$$\alpha_i = P(\tilde{C}_i = 0) \tag{4}$$

Then unconditional density(likelihood) for the observations that has  $C_i = 1$  is as follows

$$P(T_{i} = t_{i}, C_{i} = 1) = P(\tilde{C}_{i} = 0)P(T_{i} = t_{i}, C_{i} = 1|\tilde{C}_{i} = 0)$$

$$+ P(\tilde{C}_{i} = 1)P(T_{i} = t_{i}, C_{i} = 1|\tilde{C}_{i} = 1)$$

$$= \alpha_{i}S(t_{i}) + (1 - \alpha_{i})f(t_{i})$$
(5)

And unconditional density(likelihood) for the observations that has  $C_i = 0$  is as follows

$$P(T_{i} = t_{i}, C_{i} = 0) = P(\tilde{C}_{i} = 0)P(T_{i} = t_{i}, C_{i} = 0|\tilde{C}_{i} = 0)$$

$$+ P(\tilde{C}_{i} = 1)P(T_{i} = t_{i}, C_{i} = 0|\tilde{C}_{i} = 1)$$

$$= \alpha_{i}S(t_{i})$$
(6)

(in (6),  $P(T_i = t_i, C_i = 0 | \tilde{C}_i = 1) = 0$  since an observation for which the event happens before the end of observing time can not be observed as censored.)

Hence, the likelihood for our model is as follows.

$$L(t_i) = \prod_{i=1}^{N} [\alpha_i S(t_i|X,\beta) + (1-\alpha_i) f(t_i|X,\beta)]^{C_i} [\alpha_i S(t_i|X,\beta)]^{1-C_i}$$
 (7)