

## A.

From two conditions  $P(y_i=1|s_i=0)=P(y_i=1)$  and  $P(\mathbf{x}|s_i=1)=P(\mathbf{x}|y_i=1)$ ,

$$\begin{aligned} & P(f_i < f_j \mid s_i = 1, s_j = 0) \\ &= P(f_i < f_j \mid y_i = 1, s_j = 0) \\ &= P(y_j = 0 \mid y_i = 1, s_j = 0)P(f_i < f_j \mid y_i = 1, y_j = 0, s_j = 0) \\ &\quad + P(y_j = 1 \mid y_i = 1, s_j = 0)P(f_i < f_j \mid y_i = 1, y_j = 1, s_j = 0) \\ &= P(y_j = 0 \mid y_i = 1)P(f_i < f_j \mid y_i = 1, y_j = 0) \\ &\quad + P(y_j = 1 \mid y_i = 1)P(f_i < f_j \mid y_i = 1, y_j = 1). \end{aligned} \quad (20)$$

Similarly,

$$\begin{aligned} & \frac{1}{2}P(f_i = f_j \mid s_i = 1, s_j = 0) \\ &= \frac{1}{2}P(y_j = 0 \mid y_i = 1)P(f_i = f_j \mid y_i = 1, y_j = 0) \\ &\quad + \frac{1}{2}P(y_j = 1 \mid y_i = 1)P(f_i = f_j \mid y_i = 1, y_j = 1). \end{aligned} \quad (21)$$

Combining them, we obtain

$$\begin{aligned} & R_X(i, j) \\ &= P(f_i < f_j \mid s_i = 1, s_j = 0) + \frac{1}{2}P(f_i = f_j \mid s_i = 1, s_j = 0) \\ &= (1 - \pi_{ij})R(i, j) + \pi_{ij}R_{-X}(i, j). \end{aligned} \quad (22)$$

## B.

From (13),

$$\begin{aligned} & P(y_i = 1, y_j = 0)R(i, j) \\ &= \frac{P(y_i = 1, y_j = 0)}{P(y_j = 0 \mid y_i = 1)} \left\{ R_X(i, j) - P(y_j = 1 \mid y_i = 1)R_{-X}(i, j) \right\} \\ &= P(y_i = 1)R_X(i, j) - P(y_i = 1, y_j = 1)R_{-X}(i, j), \end{aligned} \quad (23)$$

Plugging the equation above and (14) into (6), we obtain

$$\begin{aligned} & \mathcal{L}_{\text{rank}} \\ &= \sum_{1 \leq i < j \leq m} P(y_i = 1, y_j = 0)R(i, j) + P(y_i = 0, y_j = 1)R(j, i) \\ &= \sum_{1 \leq i < j \leq m} P(y_i = 1)R_X(i, j) + P(y_j = 1)R_X(j, i) \\ &\quad - P(y_i = 1, y_j = 1) \left\{ R_{-X}(i, j) + R_{-X}(j, i) \right\} \\ &= \sum_{1 \leq i < j \leq m} P(y_i = 1)R_X(i, j) + P(y_j = 1)R_X(j, i) \\ &\quad - P(y_i = 1, y_j = 1), \end{aligned} \quad (24)$$

## C.

Because

$$\begin{aligned} & P(y_i = 1) - P(s_i = 1, s_j = 0) \\ &= P(y_i = 1, s_i = 0) + P(y_i = 1, s_i = 1) - P(y_i = 1, s_i = 1, s_j = 0) \\ &= P(y_i = 1, s_i = 0) + P(s_i = 1, s_j = 1), \end{aligned} \quad (25)$$

from (15) and (9), we obtain

$$\begin{aligned} & \mathcal{L}_{\text{rank}} - \hat{\mathcal{L}}_{\text{rank}} \\ &= \sum_{1 \leq i < j \leq m} P(y_i = 1)R_X(i, j) + P(y_j = 1)R_X(j, i) - \text{const} \\ &\quad - \sum_{1 \leq i < j \leq m} P(s_i = 1, s_j = 0)R_X(i, j) + P(s_i = 0, s_j = 1)R_X(j, i) \\ &= \sum_{1 \leq i < j \leq m} P(s_i = 1, s_j = 1) \left\{ R_X(i, j) + R_X(j, i) \right\} \\ &\quad + P(y_i = 1, s_i = 0)R_X(i, j) + P(y_j = 1, s_j = 0)R_X(j, i) \\ &\quad + \text{const}. \end{aligned} \quad (26)$$

The first term is proportional to the ratio of samples having both  $i$ -th and  $j$ -th label. Second and third terms are proportional to ratio of samples, which is not labeled even if they are positive.

## D.

When labels are given completely, the loss function which should be minimized is (6),

$$\begin{aligned} & \mathcal{L}_{\text{rank}} \\ &= \sum_{1 \leq i < j \leq m} P(y_i = 1, y_j = 0)R(i, j) + P(y_i = 0, y_j = 1)R(j, i). \end{aligned} \quad (27)$$

However, the loss function derived based on “case-controlled” assumption is (15)

$$\begin{aligned} & \mathcal{L}_{\text{rank}} \\ &= \sum P(y_i = 1)R_X(i, j) + P(y_j = 1)R_X(j, i) - P(y_i = 1, y_j = 1) \end{aligned} \quad (28)$$

Considering the case in which label deficit does not exist, because  $R(i, j) = R_X(i, j)$ ,

$$\begin{aligned} & \mathcal{L}_{\text{rank-false}} \\ &= \sum P(y_i = 1)R(i, j) + P(y_j = 1)R(j, i) - P(y_i = 1, y_j = 1). \end{aligned} \quad (29)$$

Their mutual difference is

$$\begin{aligned} & \mathcal{L}_{\text{rank-false}} - \mathcal{L}_{\text{rank}} \\ &= \sum P(y_i = 1, y_j = 1) \left\{ R(i, j) + R(j, i) - 1 \right\}. \end{aligned} \quad (30)$$

This error is proportional to the joint probability that both  $i$ -th class and  $j$ -th class are positive. However, this error is cancelled if we use symmetric surrogate loss.