## A. PROOF OF THEOREM 2.2

Consider a loss function  $L(\boldsymbol{y}, \boldsymbol{t}) = \sum_{c=1}^K t_c \Psi_c(\boldsymbol{y})$  and  $\Psi_c(\boldsymbol{y}) = f(\sum_{k=1}^K \phi(y_c - y_k))$ . We firstly prove that if  $t_i < t_j$ , then  $y_i < y_j$  in the minimization optimized loss function  $L^*(\boldsymbol{t})$ .

We can take i=1 and j=2. Consider an optimal prediction  $\boldsymbol{y}=[y_1,y_2,y_3,...,y_K]$  and a new prediction  $\boldsymbol{y}'=[y_2,y_1,y_3,...,y_K]$ . Let  $\phi(\xi)$  be a non-increasing function and assume  $y_1>y_2$ . We have

$$L(\mathbf{y}', \mathbf{t}) - L(\mathbf{y}, \mathbf{t}) = \sum_{c=1}^{K} t_c f(\sum_{k=1}^{K} \phi(y_c' - y_k')) - \sum_{c=1}^{K} t_c f(\sum_{k=1}^{K} \phi(y_c - y_k))$$

$$= \sum_{c=1}^{K} t_c (f(\sum_{k=1}^{K} \phi(y_c' - y_k')) - f(\sum_{k=1}^{K} \phi(y_c - y_k)))$$

$$= y_1 (f(\sum_{k=1}^{K} \phi(y_2 - y_k)) - f(\sum_{k=1}^{K} \phi(y_1 - y_k))) + y_2 (f(\sum_{k=1}^{K} \phi(y_1 - y_k)) - f(\sum_{k=1}^{K} \phi(y_2 - y_k)))$$

$$= (y_2 - y_1) (f(\sum_{k=1}^{K} \phi(y_1 - y_k)) - f(\sum_{k=1}^{K} \phi(y_2 - y_k))) < 0.$$
(1)

According to Equation (1), L(y',t) < L(y,t). However, this is a contradiction to the optimality  $L(y,t) = L^*(t)$ . Therefore,  $y_1 \le y_2$ . Now we assume that  $\phi(\xi)$  is differentiable. We know the first order condition  $\frac{\partial}{\partial y_c} L(y,t) = 0$  at the optimal solution. Then, We have

$$\frac{\partial}{\partial y_{c}} L(\mathbf{y}, \mathbf{t}) = \frac{\partial}{\partial y_{c}} \sum_{c'=1}^{K} t_{c'} f(\sum_{k=1}^{K} \phi(y_{c'} - y_{k}))$$

$$= t_{c} \frac{\partial}{\partial y_{c}} f(\sum_{k=1}^{K} \phi(y_{c} - y_{k})) + \sum_{c'=1}^{K} t_{c'} \frac{\partial}{\partial y_{c}} f(\sum_{k=1}^{K} \phi(y_{c'} - y_{c}))$$

$$= t_{c} (\sum_{k=1}^{K} \phi'(y_{c} - y_{k})) f'(\sum_{k=1}^{K} \phi(y_{c} - y_{k})) - \sum_{c'=1}^{K} t_{c'} \phi'(y_{c'} - y_{c}) f'(\sum_{k=1}^{K} \phi(y_{c'} - y_{k})) = 0.$$
(2)

Therefore,

$$t_c(\sum_{k=1}^K \phi'(y_c - y_k))f'(\sum_{k=1}^K \phi(y_c - y_k)) = \sum_{c'=1}^K t_{c'}\phi'(y_{c'} - y_c)f'(\sum_{k=1}^K \phi(y_{c'} - y_k)).$$
(3)

Let us assume  $y_1 = y_2 = y$ , then the above equality implies that

$$t_1(\sum_{k=1}^K \phi'(y-y_k))f'(\sum_{k=1}^K \phi(y-y_k)) = t_2(\sum_{k=1}^K \phi'(y-y_k))f'(\sum_{k=1}^K \phi(y-y_k)).$$
(4)

This is not possible since  $\sum_{k=1}^{K} \phi'(y-y_k) \leq 2\phi'(0) < 0$ . Therefore,  $y_1 < y_2$ .

Assume  $t_1 < t_2$ . We show that

$$\inf\{L(y,t): y_1 \ge y_2\} > L^*(t).$$
 (5)

This implies ISC. If the claim is not true, then we can find a new prediction  $\boldsymbol{y}^{(m)}$  such that  $y_1^{(m)} \geq y_2^{(m)}$  and  $\lim_m L(\boldsymbol{y}^{(m)}, \boldsymbol{t}) = L^*(\boldsymbol{t})$ . However, we already proved that if  $t_1 < t_2$ , then  $y_1 < y_2$ . Now, we can say that the function  $\Psi_c(\boldsymbol{y})$  is ISC when  $f(\xi)$  is differentiable,  $f'(\xi) > 0$ ,  $\phi(\xi)$  is differentiable,  $\phi(\xi) \geq 0$ ,  $\phi'(\xi) \leq 0$ , and  $\phi'(0) < 0$ .