

Task 1.

$$\text{minimize } f(\theta, \lambda) = \frac{1}{K} \sum_{k=1}^K \left[\log(1 + \exp(\theta^T x_k - \lambda)) - y_k(\theta^T x_k - \lambda) \right]$$

$$\text{Let } x = [\theta^T \lambda^T]^T, \quad g = \begin{bmatrix} x_k \\ -1 \end{bmatrix}$$

$$f(\theta, \lambda) = f(x) = \frac{1}{K} \sum_{k=1}^K \log(1 + \exp([x_k^T - 1]^T x)) - \sum_{k=1}^K \frac{y_k}{K} [x_k^T - 1]^T x$$

$$= \sum_{k=1}^K \frac{1}{K} \log(1 + \exp(g_k^T x)) + \sum_{k=1}^K \frac{y_k}{K} (-g_k)^T x$$

constant + exp function
→ is convex

log

composition of
const. + affine map
→ is convex

affine map
→ is convex

sum of convex functions
weighted by nonnegative scalars
→ is convex

composition of const. and nondecreasing
convex decreasing a convex
→ is convex

sum of convex functions weighted
by nonnegative scalars
→ is convex

sum of 2 convex functions is convex
→ is convex

$$\text{maximize}_{(\theta, \lambda) \in \mathbb{R}^d \times \mathbb{R}} h(x_1, \dots, x_K, y_1, \dots, y_K | \theta, \lambda) = \text{maximize}_{(\theta, \lambda) \in \mathbb{R}^d \times \mathbb{R}} \prod_{k=1}^K h(x_k, y_k | \theta, \lambda)$$

$$h(x_k, y_k | \theta, \lambda) = \begin{cases} \frac{\exp(\theta^T x_k - \lambda)}{1 + \exp(\theta^T x_k - \lambda)}, & y_k = 1 \\ \frac{1}{1 + \exp(\theta^T x_k - \lambda)}, & y_k = 0 \end{cases}$$

$$= \text{maximize}_{(\theta, \lambda) \in \mathbb{R}^d \times \mathbb{R}} \prod_{k=1}^K \frac{\exp(y_k(\theta^T x_k - \lambda))}{1 + \exp(\theta^T x_k - \lambda)}$$

$$\Rightarrow h(x_k, y_k | \theta, \lambda) = \frac{\exp(y_k(\theta^T x_k - \lambda))}{1 + \exp(\theta^T x_k - \lambda)}$$

$$= \text{maximize}_{(\theta, \lambda) \in \mathbb{R}^d \times \mathbb{R}} \frac{1}{K} \sum_{k=1}^K \log \frac{\exp(y_k(\theta^T x_k - \lambda))}{1 + \exp(\theta^T x_k - \lambda)}$$

$$= \text{maximize}_{(\theta, \lambda) \in \mathbb{R}^d \times \mathbb{R}} \frac{1}{K} \sum_{k=1}^K \left[y_k(\theta^T x_k - \lambda) - \log(1 + \exp(\theta^T x_k - \lambda)) \right]$$

$$= \text{minimize}_{(\theta, \lambda) \in \mathbb{R}^d \times \mathbb{R}} \frac{1}{K} \sum_{k=1}^K \left[\log(1 + \exp(\theta^T x_k - \lambda)) - y_k(\theta^T x_k - \lambda) \right]$$