

3) Argue that $p = \min_i |\langle \theta^*, \tilde{x}_i \rangle| > 0$ = distance from hyperplane θ^* to closest x_i in training set

$$p = \min_i |(\theta^*)^T \tilde{x}_i| = \min_i |(\omega^*)^T x_i + b^*|$$

$$p = \min_i \frac{|\omega^T x_i + b|}{\|\omega\|_2} = \min_i \frac{|\omega^T x_i + b|}{\sqrt{\omega^T \omega}}$$

$|\omega^T x_i + b| > 0$ & $\sqrt{\omega^T \omega} > 0$ by definition

So $p = \min_i |\langle \theta^*, \tilde{x}_i \rangle| > 0$

a)

b) $\langle \theta^j, \theta^* \rangle \geq \langle \theta^{j-1}, \theta^* \rangle + p$

$$\langle \theta^{j-1} + \frac{y_i - \text{sign}((\theta^{j-1})^T \tilde{x}_i)}{2} \tilde{x}_i, \theta^* \rangle \geq \langle \theta^{j-1}, \theta^* \rangle + \min_i \frac{|\omega^T x_i + b|}{\|\omega\|_2}$$

$$(\theta^{j-1})^T \theta^* + \left[\frac{y_i - \text{sign}((\theta^{j-1})^T \tilde{x}_i)}{2} \tilde{x}_i \right]^T \theta^* \geq (\theta^{j-1})^T \theta^* + \min_i \frac{|\omega^T x_i + b|}{\|\omega\|_2}$$

$$\frac{1}{2} [y_i - \text{sign}((\theta^{j-1})^T \tilde{x}_i)] \tilde{x}_i^T \theta^* \geq \min_i \frac{|\omega^T x_i + b|}{\|\omega\|_2}$$

if $y_i \neq \text{sign}((\theta^{j-1})^T \tilde{x}_i) \rightarrow$
(mislabel requiring update)

$$\frac{1}{2} (y_i - \text{sign}((\theta^{j-1})^T \tilde{x}_i)) \tilde{x}_i^T \theta^* \geq \min_i \frac{|\omega^T x_i + b|}{\|\omega\|_2}$$

$$y_i ((\omega^*)^T x_i + b^*) \geq \min_i \frac{|\omega^T x_i + b|}{\|\omega\|_2}$$

Therefore

$$\langle \theta^j, \theta^* \rangle \geq \langle \theta^{j-1}, \theta^* \rangle + p$$

$$\langle \theta^{j-1}, \theta^* \rangle + p \geq j p$$

$$\langle \theta^{j-1}, \theta^* \rangle p + p^2 \geq (R^2 + 1) p^2$$

$$p \langle \theta^j, \theta^* \rangle \geq p (\langle \theta^{j-1}, \theta^* \rangle + p) \geq (R^2 + 1) p^2$$

$\frac{(R^2 + 1) \|\theta^*\|^2}{p^2}$
b/c mislabel θ^{j-1}

$$\textcircled{3} \text{ c) } \|\theta^j\|_2^2 \leq \|\theta^{j-1}\|_2^2 + \|\tilde{x}_{ij}\|_2^2$$

$$\|\theta^{j-1} + y_i \tilde{x}_{ij}\|_2^2 \leq \|\theta^{j-1}\|_2^2 + \|\tilde{x}_{ij}\|_2^2$$

$$\text{if } y_i = 1 \rightarrow \|\theta^{j-1} + \tilde{x}_{ij}\|_2^2 \leq \|\theta^{j-1}\|_2^2 + \|\tilde{x}_{ij}\|_2^2 \quad \checkmark$$

$$\text{if } y_i = -1 \rightarrow \|\theta^{j-1} - \tilde{x}_{ij}\|_2^2 \leq \|\theta^{j-1}\|_2^2 + \|\tilde{x}_{ij}\|_2^2 \quad \checkmark$$

$$\text{d) } \|\theta^j\|_2^2 \leq \|\theta^{j-1}\|_2^2 + \|\tilde{x}_{ij}\|_2^2 \leq j(1+R^2)$$

$$\|\theta^{j-1} + (1+x_i)\|_2^2 \leq \frac{(R^2+1)\|\theta^*\|_2^2}{\rho^2} (1+R^2)$$

$$\rho^2(\theta^{j-1} + (1+x_i)) \leq \theta^* (1 + \max_i x_i)^2$$

$$\min_i |\langle \theta^*, \tilde{x}_i \rangle|^2 (\theta^{j-1} + 1 + x_i) \leq \theta^* (1 + \max_i x_i)^2$$

$$\max_i x_i > x_i$$

$$1 + \max_i x_i > 1 + x_i$$

$$(1 + \max_i x_i)^2 = 1 + 2(\max_i x_i) + (\max_i x_i)^2 > 1 + x_i + \theta^{j-1}$$

$$\theta^* (1 + 2 \max_i x_i + (\max_i x_i)^2) > (1 + x_i + \theta^{j-1}) \min_i |\langle \theta^*, \tilde{x}_i \rangle|^2$$

$$\text{e) Cauchy-Schwarz: } |\langle u, v \rangle|^2 \leq \langle u, u \rangle \cdot \langle v, v \rangle$$

$$\langle \theta^j, \theta^* \rangle \geq j\rho \quad \text{and } \|\theta^j\|_2^2 \leq j(1+R^2)$$

$$j \leq \frac{\langle \theta^j, \theta^* \rangle}{\rho} \quad \text{need } \|\langle \theta^j, \theta^* \rangle\|_2^2 = \frac{(1+R^2)\|\theta^*\|_2^2}{\rho^2}$$

$$j \leq \frac{j(1+R^2)\|\theta^*\|_2^2}{\rho}$$

$$j^{\frac{1}{2}} \leq \frac{[(1+R^2)\|\theta^*\|_2^2]^{\frac{1}{2}}}{\rho^{\frac{1}{2}}}$$

$$j \leq \frac{(1+R^2)\|\theta^*\|_2^2}{\rho^2}$$

$$|\langle \theta^j, \theta^* \rangle|^2 \leq \langle \theta^j, \theta^j \rangle \cdot \langle \theta^*, \theta^* \rangle$$

$$= (\theta^j)^T \theta^j \cdot (\theta^*)^T \theta^*$$

$$\|\theta^j\|_2^2 \cdot \|\theta^*\|_2^2$$

$$< j(1+R^2)\|\theta^*\|_2^2$$

$$j \cdot \langle \theta^j, \theta^* \rangle \leq (j(1+R^2)\|\theta^*\|_2^2)^{\frac{1}{2}}$$