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Common application: in important cases, one may be able to write the dual directly.

SVM

$$\begin{array}{l} \min \quad \frac{w'w}{2} \\ \text{st } y_i (w'x_i + b) \geq 1 \end{array} \quad \left. \vphantom{\begin{array}{l} \min \\ \text{st} \end{array}} \right\} \begin{array}{l} \text{Primal form,} \\ \text{Separable} \end{array}$$

$$\mathcal{L}_P(w, \lambda) = \frac{w'w}{2} - \sum_i \lambda_i \{ [y_i (w'x_i + b)] - 1 \}$$

$$\nabla_w \mathcal{L} = 0 = w - \sum_i \lambda_i \{ [y_i x_i] \}$$

$$\nabla_b \mathcal{L} = 0 = - \sum_i \lambda_i y_i$$

⑥

Subst

$$L_0 = \sum_i \lambda_i - \frac{1}{2} \sum_{ij} \lambda_i \lambda_j [y_i y_j (x_i^T x_j)]$$

Notice constraints

$$\sum_i \lambda_i y_i = 0$$

$$\lambda_i \geq 0$$

and we must max this in  $\lambda$

If there is an fp for primal, the  
max is soln to primal

i.e.  $\text{Value(Dual)} = \text{Value(Primal)}$

What if data is not separable? (7)

$$\begin{array}{l} \min \frac{w'w}{2} + C \sum_i \xi_i \\ \text{st} \quad y_i (w'x_i + b) \geq 1 - \xi_i \\ \quad \quad \xi_i \geq 0 \end{array} \quad \left. \vphantom{\begin{array}{l} \min \\ \text{st} \end{array}} \right\} \text{Primal prob}$$

$\xi_i$  are slack variables

$$\mathcal{L}_p = \frac{w'w}{2} + C \sum_i \xi_i - \sum_i \lambda_i [y_i (w'x_i + b) - 1 + \xi_i] - \sum_i \mu_i \xi_i$$

$$\nabla_w \mathcal{L}_p = w - \sum_i \lambda_i y_i x_i = 0$$

$$\nabla_b \mathcal{L}_p = 0 = -\sum_i \lambda_i y_i$$

$$\nabla_{\xi_i} \mathcal{L}_p = C - \lambda_i - \mu_i = 0 \quad \left. \vphantom{\nabla_{\xi_i} \mathcal{L}_p} \right\} \rightarrow \text{this gets rid of } \xi_i$$

So we have

$$L_D = \sum_i \lambda_i - \frac{1}{2} \sum_{ij} y_i y_j \lambda_i \lambda_j x_i' x_j$$

subject to

$$\sum_i \lambda_i y_i = 0$$

$$0 \leq \lambda_i \leq C$$

Notice that  $\xi_i$  can be interpreted  
as a loss

$$\text{hinge loss} \left( \frac{y_i y_j}{2} \right) = \max(0, 1 - y_i y_j)$$