## Lagrange Optimization

Minimize with constraints

$$\min_{w} f(w)$$
, s.t.  $g_i(w) \le 0$ ,  $i = 1, ..., k$ , and  $h_i(w) = 0$ ,  $i = 1, ..., l$ 

Lagrange function:  $L_p(w, \alpha, \beta) = f(w) + \sum_{i=1}^k \alpha_i g_i(w) + \sum_{i=1}^l \beta_i h_i(w)$ 

## SVM Optimization Function - Linearly Separable Case

Cost function:  $\Phi(\mathbf{w}) = \frac{1}{2}\mathbf{w}^T\mathbf{w}$ 

Constraint:  $y_n(\mathbf{w}^T\mathbf{x}_i + \tilde{b}) \ge 1$  for i = 1, 2, ..., NOptimization Problem:  $\min \frac{1}{2} ||\mathbf{w}||_2^2$ , s.t.  $y_n(\mathbf{w}^T\mathbf{x} + w_0) \ge 1$ , n = 1, ..., N

**Step 1:** Formulate Lagrangian function (primal problem)

$$L_p = \frac{1}{2} \|\mathbf{w}\|_2^2 - \sum_{n=1}^{N} \alpha_n \left[ y_n(\mathbf{w}^T \mathbf{x}_n + w_0) - 1 \right]$$

(The "-" sign is because the sign of inequality in the constrain is now flipped)

**Step 2:** Set gradient of Lagrangian to 0 wrt primal variables  $\mathbf{w}, w_0$ 

$$\frac{\partial L_p}{\partial \mathbf{w}} = 0 \implies \mathbf{w} = \sum_{n=1}^N \alpha_n y_n \mathbf{x_n}$$

$$\frac{\partial L_p}{\partial w_0} = 0 \quad \Rightarrow \quad \sum_{n=1}^{N} \alpha_n y_n = 0$$

Step 3: Substitute the primal variables  $\mathbf{w}, w_0$  into the Lagrangian and express in terms of dual variables  $\alpha_n$ 

$$L_d = \frac{1}{2} \|\mathbf{w}\|_2^2 - \mathbf{w}^T \sum_{n=1}^N \alpha_n y_n \mathbf{x_n} - w_0 \sum_{n=1}^N \alpha_n y_n + \sum_{n=1}^N \alpha_n$$
$$= -\frac{1}{2} \mathbf{w}^T \mathbf{w} + \sum_{n=1}^N \alpha_n$$
$$= -\frac{1}{2} \sum_{n=1}^N \sum_{m=1}^M \alpha_n \alpha_m y_n y_m \mathbf{x_n}^T \mathbf{x_m} + \sum_{n=1}^N \alpha_n$$

Step 4: Set gradient of Lagrangian to 0 wrt to dual variables (dual problem)

$$\max_{\alpha_n} L_d = \max_{\alpha_n} \left\{ -\frac{1}{2} \sum_{n=1}^{N} \sum_{m=1}^{N} \alpha_n \alpha_m y_n y_m \mathbf{x}_{\mathbf{n}}^T \mathbf{x}_{\mathbf{m}} + \sum_{n=1}^{N} \alpha_n \right\}$$
s.t. 
$$\sum_{n=1}^{N} \alpha_n y_n = 0 \text{ and } \alpha_n \ge 0, \text{ for } n = 1, \dots, N$$

**Dual problem** whose cost depends on data size N, rather than data dimensionality D. Most of the  $\alpha_n$  will vanish. (See outline in slides for solving the problem) Step 5: Substitute the multipliers to the weight

## SVM Optimization Function - Non-Linearly Separable Case

Cost function:  $\min_{\mathbf{w}, \boldsymbol{\xi}} \left[ \frac{1}{2} \|\mathbf{w}\|_2^2 + C \sum_{n=1}^N \xi_n \right]$ 

Constraint:  $y_n(\mathbf{w}^T\mathbf{x} + w_0) \ge 1 - \xi_n$  and  $\xi_n \ge 0$ ,  $\forall n$ 

Step 1: Formulate Lagrangian function (primal problem)

$$L_p = \frac{1}{2} \|\mathbf{w}\|_2^2 + C \sum_{n=1}^N \xi_n - \sum_{n=1}^N \alpha_n \left[ y_n(\mathbf{w}^T \mathbf{x}_n + w_0) - 1 + \xi_n \right] - \sum_{n=1}^N \mu_n \xi_n$$

**Step 2:** Set gradient of Lagrangian to 0 wrt primal variables  $\mathbf{w}, w_0$ 

$$\frac{\partial L_p}{\partial \mathbf{w}} = 0 \implies \mathbf{w} = \sum_{n=1}^N \alpha_n y_n \mathbf{x_n}$$

$$\frac{\partial L_p}{\partial w_0} = 0 \quad \Rightarrow \quad \sum_{n=1}^N \alpha_n y_n = 0$$

$$\frac{\partial L_p}{\partial \xi_n} = 0 \quad \Rightarrow \quad C - \alpha_n - \mu_n = 0$$

**Step 3:** Substitute the primal variables  $\mathbf{w}, w_0$  into the Lagrangian and express in terms of dual variables  $\alpha_n$ 

$$L_{d} = \frac{1}{2} \|\mathbf{w}\|_{2}^{2} + C \sum_{n=1}^{N} \xi_{n} - \mathbf{w}^{T} \sum_{n=1}^{N} \alpha_{n} y_{n} \mathbf{x}_{n}$$

$$- w_{0} \sum_{n=1}^{N} \alpha_{n} y_{n} + \sum_{n=1}^{N} \alpha_{n} - \sum_{n=1}^{N} \alpha_{n} \xi_{n} - \sum_{n=1}^{N} \mu_{n} \xi_{n}$$

$$= \frac{1}{2} \sum_{n=1}^{N} \sum_{m=1}^{N} \alpha_{n} \alpha_{m} y_{n} y_{m} \mathbf{x}_{n}^{T} \mathbf{x}_{m} + \sum_{n=1}^{N} \xi_{n} (C - \alpha_{n} - \mu_{n}) + \sum_{n=1}^{N} \alpha_{n}$$

$$= \sum_{n=1}^{N} \alpha_{n} - \frac{1}{2} \sum_{n=1}^{N} \sum_{m=1}^{N} \alpha_{n} \alpha_{m} y_{n} y_{m} \mathbf{x}_{n}^{T} \mathbf{x}_{m}$$

Step 4: Set gradient of Lagrangian to 0 wrt to dual variables (dual problem)

$$\max_{\alpha_n} L_d = \max_{\alpha_n} \left\{ -\frac{1}{2} \sum_{n=1}^N \sum_{m=1}^N \alpha_n \alpha_m y_n y_m \mathbf{x}_n^T \mathbf{x}_m + \sum_{n=1}^N \alpha_n \right\}$$
  
s.t. 
$$\sum_{n=1}^N \alpha_n y_n = 0 \text{ and } \alpha_n \ge 0, \text{ for } n = 1, \dots, N$$

**Dual problem** solved with approximate methods.

**Step 5:** Substitute the multipliers to the weight