- For C > 2, using maximal value method
- Augmented space

$$- g_k(\underline{x}) = \underline{w}_k^T \underline{x}, \quad k = 1, 2, \dots, C$$

Decision rule:

$$g_k(\underline{x}) > g_j(\underline{x}) \ \forall j \neq k \implies \underline{x} \in S_k$$

Algorithm:

- 1. Shuffle the order of training data points
- 2. For each data point $\underline{x}^{(k)}$ [Given $\underline{x}^{(k)} \in S_k$]:

If
$$g_k(\underline{x}^{(k)}) > g_j(\underline{x}^{(k)}) \quad \forall j \neq k$$

then: $\underline{w}^{(m)}(i+1) = \underline{w}^{(m)}(i) \quad \forall m$
else:

$$\underline{w}^{(k)}(i+1) = \underline{w}^{(k)}(i) + \eta(i)\underline{x}^{(k)}$$
Let $l = \operatorname{arg\,max}_{j \neq k} \left\{ g_j(\underline{x}^{(k)}) \right\}$

(If more than one possible value of l, pick any one)

$$\underline{w}^{(l)}(i+1) = \underline{w}^{(l)}(i) - \eta(i)\underline{x}^{(k)}$$

$$\underline{w}^{(m)}(i+1) = \underline{w}^{(m)}(i) \quad \forall m \neq l, k$$

until: all training data points are correctly classified.

Convergence: Convergence is proven in DHS 5.12.2 for $\eta(i) = \text{constant} > 0$, for linearly separable data.