

CS 189: Introduction to Machine Learning

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1 Perceptron Algorithm (cont'd)

- linear decision function $f(x) = w \cdot x$
- decision boundary $\{x : f(x) = 0\}$
- sample points $X_1, X_2, \dots, X_n \in \mathbb{R}^d$; classifications $y_1, \dots, y_n = \pm 1$
- goal: find weights w such that $y_i X_i \cdot w \geq 0$
- goal, rewritten: find w that minimizes $R(w) = \sum_{i \in V} -y_i X_i \cdot w$ where V is the set of indices i for which $y_i X_i \cdot w < 0$.

Objects in x -space transform to objects in w -space:

| x -space | w -space |
|-------------------------------------|-------------------------------------|
| hyperplane: $\{z : w \cdot z = 0\}$ | point: w |
| point: x | hyperplane: $\{z : x \cdot z = 0\}$ |

If we want to enforce inequality $x \cdot w \geq 0$, that means:

- in x -space, x should be on the same side of $\{z : z \cdot w = 0\}$ as w
- in w -space, w should be on the same side of $\{z : x \cdot z = 0\}$ as x