

## CHEAT SHEET

# Perceptron

<b>Algorithm Name</b>	Perceptron
<b>Description</b>	The Perceptron attempts to converge to a weight vector $\mathbf{w}$ that separates the data from two classes. In essence, it iterates through the whole dataset and when it encounters a misclassified point, it updates using $\mathbf{w} \leftarrow \mathbf{w} + y\mathbf{x}$ .
<b>Applicability</b>	Binary classification problems.
<b>Assumptions</b>	Data are linearly separable and the labels have to be $\pm 1$ .
<b>Underlying Mathematical Principles</b>	A hyperplane is defined as $\mathbf{w}^T \mathbf{x} = 0$ , where $\mathbf{w}$ determines the orientation of the hyperplane. Predictions are the sign of $h(\mathbf{x}) = \text{sign}(\mathbf{w}^T \mathbf{x})$ .
<b>Additional Details</b>	<ul style="list-style-type: none"> <li>• Perceptron is a mistake-driven algorithm in the sense that it will update the parameter <math>\mathbf{w}</math> when it incorrectly classifies any data.</li> <li>• You can append bias <math>b</math> to <math>\mathbf{w}</math> by appending a 1 to all data points.</li> <li>• Depending on how our data points are ordered, the parameter <math>\mathbf{w}</math> will be different. For example, if two points in different classes are misclassified at time <math>t</math>, then updating <math>\mathbf{w}</math> with the first might fix classification for the second as well, in which case perceptron algorithm will stop. In another run, if updating <math>\mathbf{w}</math> with the second point fixes classification for the first and perceptron algorithm stops, we end up with a <math>\mathbf{w}</math> different from the first one.</li> </ul>
<b>Example</b>	You want to classify cats and dogs based on number of meows (x axis) and number of woofs (y axis). The data should be linearly separable.

