

Introduction to Machine Learning

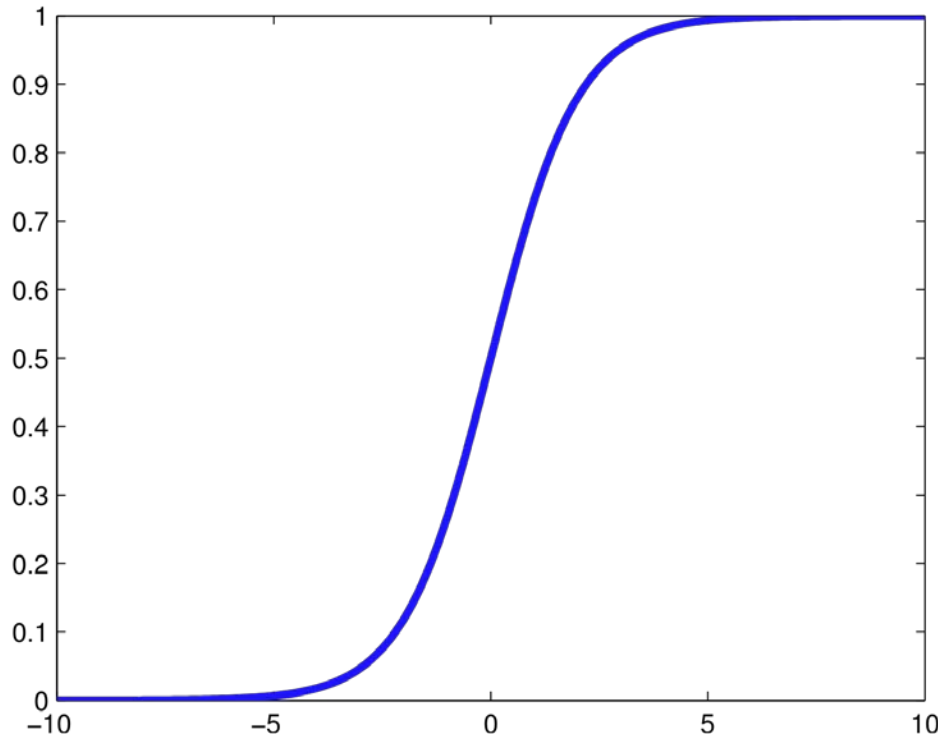
Brown University CSCI 1950-F, Spring 2011
Prof. Erik Sudderth

Lecture 11: Stochastic Gradient Methods,
Iteratively Reweighted Least Squares

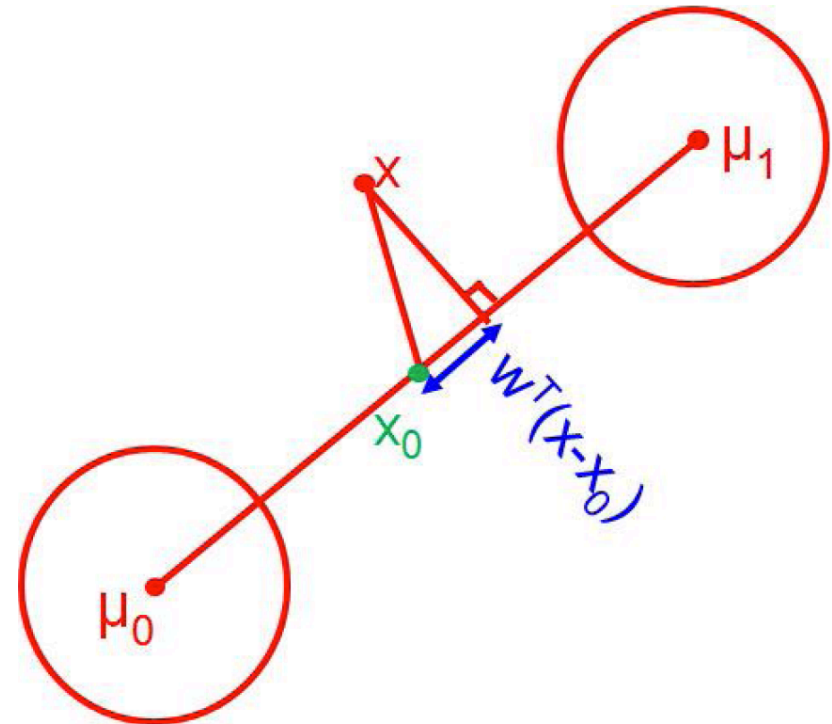
Many figures courtesy Kevin Murphy's textbook,
Machine Learning: A Probabilistic Perspective

Binary Discriminant Analysis

Logistic Function



$$\text{sigm}(\eta) := \frac{1}{1 + \exp(-\eta)} = \frac{e^\eta}{e^\eta + 1}$$

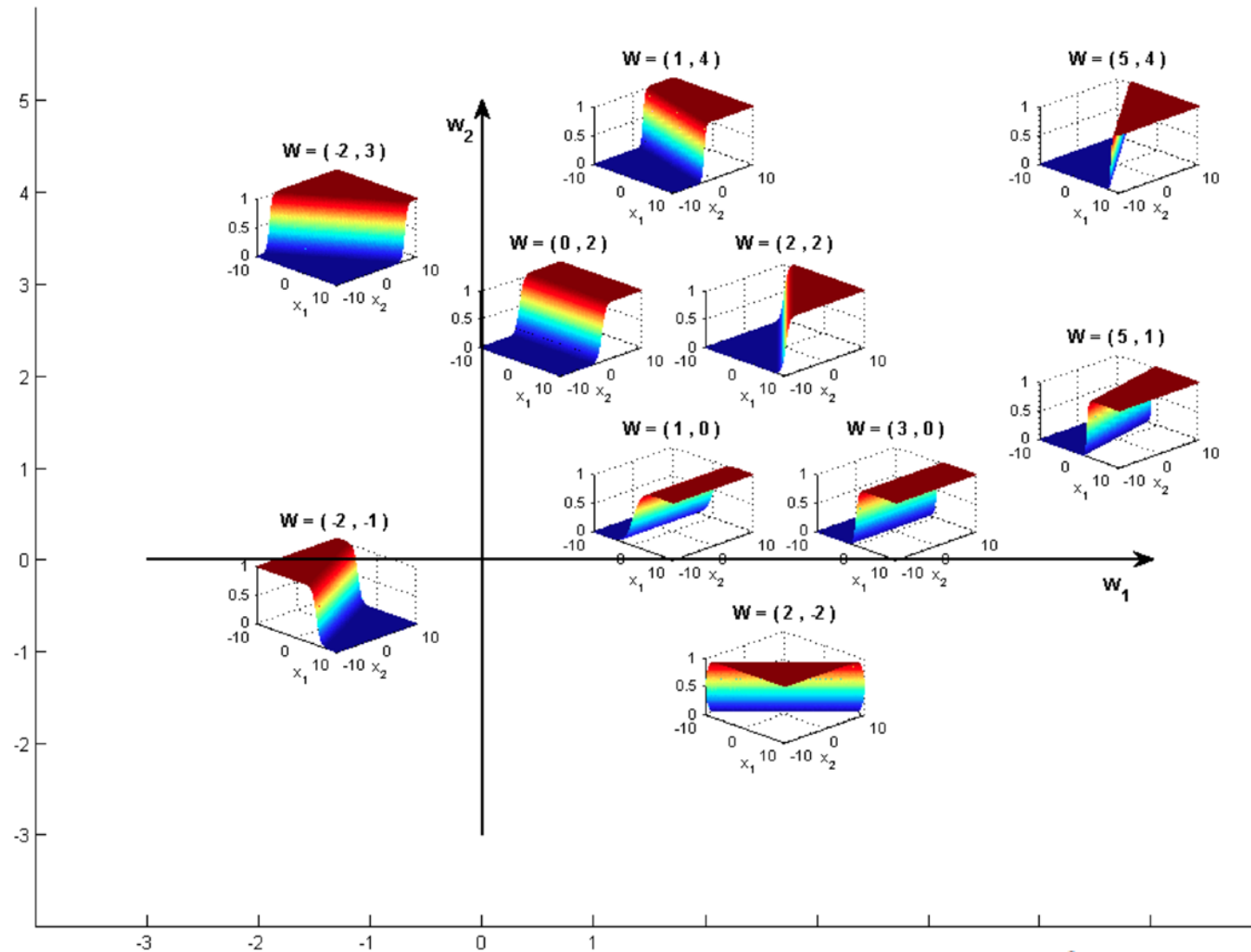


$$p(y = 1 | \mathbf{x}, \theta) = \sigma(\mathbf{w}^T (\mathbf{x} - \mathbf{x}_0))$$

$$\mathbf{w} = \beta_1 - \beta_0 = \Sigma^{-1}(\mu_1 - \mu_0)$$

$$\mathbf{x}_0 = \frac{1}{2}(\mu_1 + \mu_0) - (\mu_1 - \mu_0) \frac{\log(\pi_1/\pi_0)}{(\mu_1 - \mu_0)^T \Sigma^{-1}(\mu_1 - \mu_0)}$$

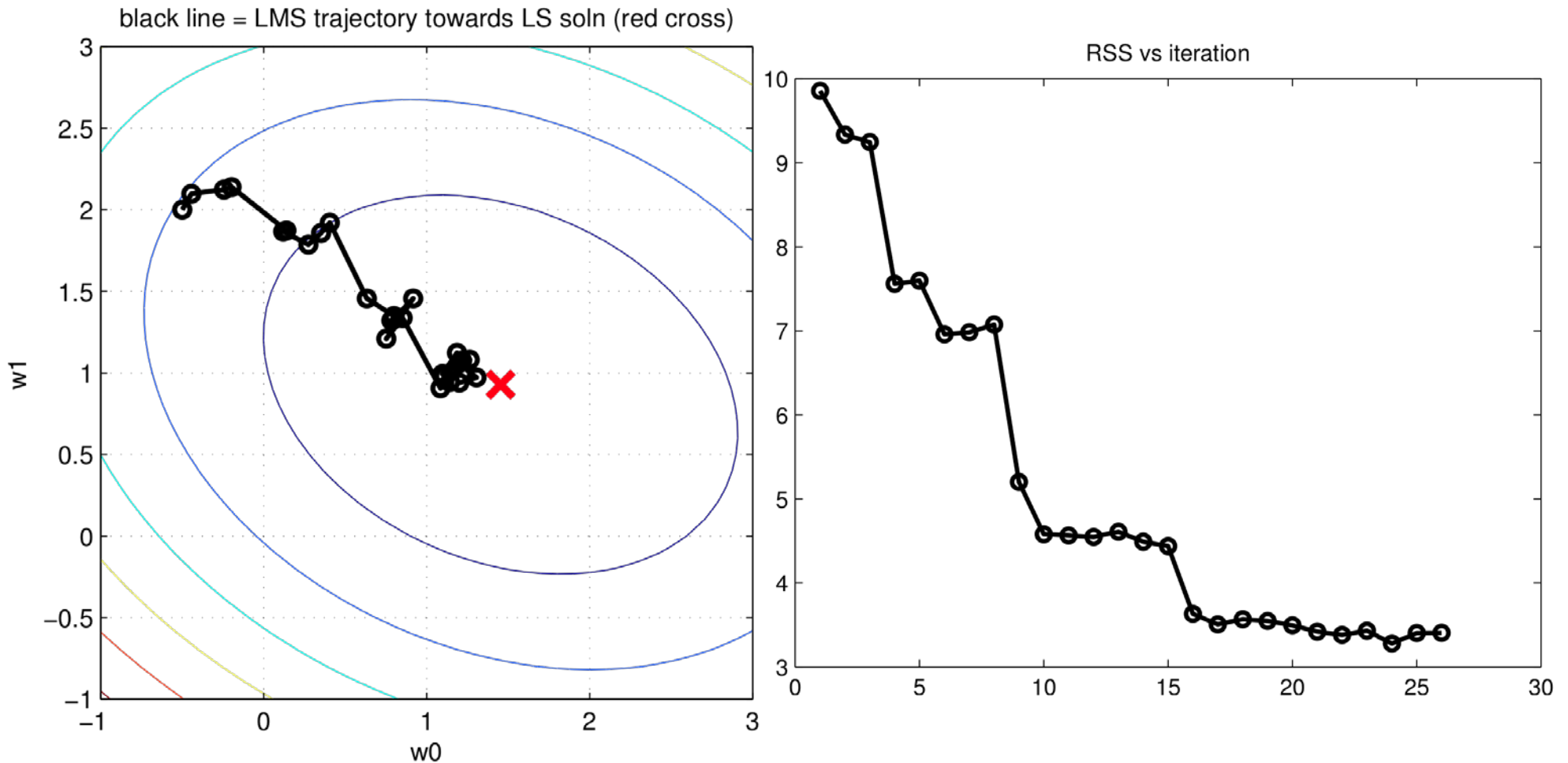
Logistic Regression



$$p(y|x, w) = \text{Ber}(y|\text{sigm}(w^T x))$$

$$\text{sigm}(\eta) := \frac{1}{1 + \exp(-\eta)} = \frac{e^\eta}{e^\eta + 1}$$

Least Mean Squares (LMS)



Stochastic gradient descent applied to linear regression model