

### Batch Gradient Descent Algorithm for Logistic Regression

- Assume 4-D dataset

$$\mathbf{w}^T = [w_0 \quad w_1 \quad w_2 \quad w_3 \quad w_4]$$
$$\mathbf{x}^T = [x_0 \quad x_1 \quad x_2 \quad x_3 \quad x_4] \text{ where } x_0 = 1$$

$$\sigma(\mathbf{w}^T \mathbf{x}) = 1 / (1 + \exp(-(w_0 x_0 + w_1 x_1 + w_2 x_2 + w_3 x_3 + w_4 x_4)))$$

### Training

repeat until *convergence* {

// Update equation

$$w_j = w_j - \eta \frac{\partial J(\mathbf{w})}{\partial w_j}$$

$j = 0, \dots, n$  where  $n$  is the number of features

$$\frac{\partial J(\mathbf{w})}{\partial w_j} = \frac{1}{m} \sum_{i=1}^m \left( (\sigma(\mathbf{w}^T \mathbf{x}^{(i)}) - y_i) \cdot x_j^{(i)} \right)$$

$$temp0 = w_0 - \eta \frac{\partial J(\mathbf{w})}{\partial w_0} = w_0 - \eta \cdot \frac{1}{m} \sum_{i=1}^m \left( (\sigma(\mathbf{w}^T \mathbf{x}^{(i)}) - y_i) \cdot x_0^{(i)} \right) \text{ where } x_0^{(i)} = 1$$

$$temp1 = w_1 - \eta \frac{\partial J(\mathbf{w})}{\partial w_1} = w_1 - \eta \cdot \frac{1}{m} \sum_{i=1}^m \left( (\sigma(\mathbf{w}^T \mathbf{x}^{(i)}) - y_i) \cdot x_1^{(i)} \right)$$

...

$$temp4 = w_4 - \eta \frac{\partial J(\mathbf{w})}{\partial w_4} = w_4 - \eta \cdot \frac{1}{m} \sum_{i=1}^m \left( (\sigma(\mathbf{w}^T \mathbf{x}^{(i)}) - y_i) \cdot x_4^{(i)} \right)$$

$$w_0 = temp0, w_1 = temp1, \dots, w_4 = temp4$$

$m$  is the number of training examples

$\eta$  is the learning rate

// Compute the cost function

$$J = - \sum_{i=1}^m \left( \left( y_i \log(\sigma(\mathbf{w}^T \mathbf{x}^{(i)})) \right) + \left( (1 - y_i) \log(1 - \sigma(\mathbf{w}^T \mathbf{x}^{(i)})) \right) \right)$$

}

### Prediction

$$\text{if } \sigma(\mathbf{w}^T \mathbf{x}) \geq 0.5 \rightarrow \hat{y} = 1 \text{ else } \hat{y} = 0$$

where  $\mathbf{w}$  is the optimal weight vector obtained after training and  $\mathbf{x}$  is the test instance.