## 1 Logistic Regression

## 1.1 Batch Gradient Descent

a.

$$NLL(w) = -\sum_{i=1}^{N} \left[ (1 - y_i) \log(1 - \sigma(\mathbf{w}^T \mathbf{x})) + y_i \log \sigma(\mathbf{w}^T \mathbf{x}) \right]$$
$$= -\sum_{i=1}^{N} [y_i \mathbf{w}^T \mathbf{x} - \log(1 + e^{\mathbf{w}^T \mathbf{x}})]$$
$$\frac{\delta NLL(w)}{\delta \mathbf{w}} = -\sum_{i=1}^{N} \mathbf{x} (y_i - \sigma(\mathbf{w}^T \mathbf{x}))$$

## 1.2 Stochastic Gradient Descent

a.

$$l(w) = \left[ (1 - y_t) \log(1 - \sigma(\mathbf{w}^T \mathbf{x_t})) + y_t \log \sigma(\mathbf{w}^T \mathbf{x_t}) \right]$$

b.

$$w_t = w_t - \eta((y_t - \sigma(\mathbf{w}^T \mathbf{x_t}))x_t)$$

c.

 $\mathbf{d}$ .

Larger values make the ROC curve closer and closer to the dotted line where as smaller values push it farther away

e.

$$w_t = w_t - \eta((y_t - \sigma(\mathbf{w}^T \mathbf{x_t})) x_t - \mu \|\mathbf{w}\|^2)$$

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2.4 c.

The two curves appear to be the same