

## CS281A/Stat241A Homework Assignment 3 (due 5pm October 14, 2009)

### 1. (Logistic regression)

Suppose that we wish to model an unordered discrete response variable  $Y$  (such as the outcome of a potential customer's visit to a website), conditioned on a vector  $X$  of real variables (such as characteristics of the advertisements presented to the potential customer). We could model this kind of relationship using

$$\Pr(Y = y|X = x) = \frac{\exp(-\beta'_y x)}{1 + \sum_{i=1}^{k-1} \exp(-\beta'_i x)}, \quad (1)$$

where  $y \in \{1, \dots, k-1\}$ ,  $x \in \mathbb{R}^d$ ,  $\beta_i \in \mathbb{R}^d$  and  $k$  is the number of distinct responses.

Suppose that we have data  $(x_1, y_1), \dots, (x_n, y_n)$  generated i.i.d. from the model (1).

- Write down the log likelihood and its first and second derivatives.
  - Describe (in pseudocode) a Newton-Raphson algorithm for maximizing the log likelihood.
  - Suppose that the dimension  $d$  of the data is so large that it is impractical to store more than a constant number of vectors in  $\mathbb{R}^d$ , let alone manipulate second derivative matrices. Suggest an online steepest ascent algorithm.
- ### 2. (ML Estimation)

On the course website, there is a data set (hw3-2.data), consisting of 100 pairs,  $(v_1, y_1), \dots, (v_{100}, y_{100})$ . Each  $v_i$  is a vector in  $\mathbb{R}^2$ , and each  $y_i$  is a number in  $\{1, \dots, 4\}$ . Line  $i$  of the file contains the two components of  $v_i$ , followed by  $y_i$ . Using the algorithm that you proposed in question 1b, calculate the maximum likelihood estimate for the parameters of the model (1) for this data, with  $x_i = (1, v_{i1}, v_{i2})'$ . Plot the data (with four different symbols for the  $y$  values) and the contours

$$C_y = \left\{ v \in \mathbb{R}^2 : \Pr \left( Y = y \mid X = \begin{pmatrix} 1 \\ v \end{pmatrix} \right) = 1/2 \right\}$$

for  $y = \{1, \dots, 4\}$ .

- ### 3. (IPF) Consider the undirected graphical model

$$p(x) = \frac{1}{Z} \prod_{(i,j) \in E} \psi_{i,j}(x_i, x_j),$$

with binary variables  $x_1, \dots, x_k$ , where the  $\psi_{i,j}$  are non-negative functions. The data in the file **hw3-3.data** on the course website consists of  $n$  binary vectors of length  $k = 5$ . Implement the IPF algorithm, and use your implementation on this data to estimate the model parameters for the following graphs:

- $E = \{(1, 2), (2, 3), (3, 4), (4, 5), (5, 1)\}$ ,
- $E = \{(1, 2), (2, 3), (3, 4), (4, 1), (2, 5)\}$ ,
- $E = \{(1, 2), (2, 3), (3, 4), (2, 5)\}$ .

Which model fits the data best?