

# HOMEWORK PROBLEMS 06, ANLY 561, FALL 2018

## DUE 10/20/18

### Exercises:

#### 1. Create Python functions

`logistic_objective(x,y), dlogistic_objective(x,y), d2logistic_objective(x,y)`

satisfying the following specifications:

- All functions expect two arrays/numpy arrays  $\mathbf{x}$  and  $\mathbf{y}$  where  $\mathbf{x}[i] = x_i$ ,  $\mathbf{y}[i] = y_i$ , and  $\{(x_i, y_i)\}_{i=1}^N \subset \mathbb{R} \times \{-1, 1\}$  are training data for logistic regression.
- `logistic_objective(x,y)` returns a *function*  $\mathbf{f}$  satisfying the following specifications:
  - $\mathbf{f}$  expects a single array/numpy array  $\mathbf{b}$  such that  $\mathbf{b}[0] = \beta_0$  and  $\mathbf{b}[1] = \beta_1$  for some logistic model parameters  $(\beta_0, \beta_1)$
  - $\mathbf{f}(\mathbf{b})$  computes the negative log-likelihood of the parameters  $(\beta_0, \beta_1)$ . That is,

$$\mathbf{f}(\mathbf{b}) = \ell(\beta) = \frac{1}{N} \sum_{i=1}^N \log \left( 1 + e^{-y_i(\beta_0 + \beta_1 x_i)} \right).$$

- `dlogistic_objective(x,y)` returns a *function*  $\mathbf{df}$  satisfying the following specifications:
  - $\mathbf{df}$  expects a single array/numpy array  $\mathbf{b}$  such that  $\mathbf{b}[0] = \beta_0$  and  $\mathbf{b}[1] = \beta_1$  for some logistic model parameters  $(\beta_0, \beta_1)$
  - $\mathbf{df}(\mathbf{b})$  computes the gradient of negative log-likelihood at  $(\beta_0, \beta_1)$ . That is,

$$\mathbf{df}(\mathbf{b}) = \nabla_{\beta} \ell(\beta) = \begin{pmatrix} \frac{\partial \ell}{\partial \beta_0}(\beta) \\ \frac{\partial \ell}{\partial \beta_1}(\beta) \end{pmatrix}.$$

- `d2logistic_objective(x,y)` returns a *function*  $\mathbf{d2f}$  satisfying the following specifications:
  - $\mathbf{d2f}$  expects a single array/numpy array  $\mathbf{b}$  such that  $\mathbf{b}[0] = \beta_0$  and  $\mathbf{b}[1] = \beta_1$  for some logistic model parameters  $(\beta_0, \beta_1)$
  - $\mathbf{d2f}(\mathbf{b})$  computes the Hessian of negative log-likelihood at  $(\beta_0, \beta_1)$ . That is,

$$\mathbf{d2f}(\mathbf{b}) = \nabla_{\beta}^2 \ell(\beta) = \begin{pmatrix} \frac{\partial^2 \ell}{\partial \beta_0^2}(\beta) & \frac{\partial^2 \ell}{\partial \beta_0 \partial \beta_1}(\beta) \\ \frac{\partial^2 \ell}{\partial \beta_0 \partial \beta_1}(\beta) & \frac{\partial^2 \ell}{\partial \beta_1^2}(\beta) \end{pmatrix}$$

Use these implementations and the data from Exercise 2 in Homework 05 to perform backtracking with both gradient descent and Newton increments. For the Newton increments, note that the computation of  $\left( \nabla_{\beta}^2 \ell(\beta) \right)^{-1} \nabla_{\beta} \ell(\beta)$  is carried out by the command

`dx_newt = - numpy.linalg.solve(d2f(b), df(b)).`

For each type of increment, initialize with  $\beta^{(0)} = \begin{pmatrix} 10 \\ 10 \end{pmatrix}$ , run 30 steps of backtracking and provide a  $y$ -semilog plot of the point residuals  $\{\|\beta^{(k+1)} - \beta^{(k)}\|\}_{i=1}^{30}$  and the function residuals  $\{|\ell(\beta^{(k+1)}) - \ell(\beta^{(k)})|\}_{i=1}^{30}$ . For backtracking, use  $\alpha = 0.2$  and  $\beta = 0.8$ .

#### 2. Consider the program

$$\min_{(x,y) \in \mathbb{R}^2} 2x + 3y \text{ subject to } -1 \leq x \leq 1 \text{ and } -1 \leq y \leq 1.$$

Such a program is called a **linear program** because the objective function and all the constraint functions are affine.

- (a) Exhibit the KKT conditions for this program.
- (b) Show that  $(-1, -1)$  is the only point which satisfies the KKT conditions, and that  $(1, 1)$  satisfies all the KKT conditions except dual feasibility.
- (c) Explain why the point  $(0, 0)$  is an interior point of this program, and carry out the log-barrier method to numerically produce a solution to this program using  $(0, 0)$  to initialize. Use  $M = 10$ , 10 centering steps, 5 iterations in the outer loop, and 3 iterations in each inner loop. Whenever backtracking is called, use Newton increments and the standard parameters  $\alpha = 0.2$  and  $\beta = 0.8$ . Display the answer you compute. You may find it helpful to note that

$$\nabla\phi(x, y) = \begin{pmatrix} -\frac{a}{ax+by+c} \\ -\frac{b}{ax+by+c} \end{pmatrix} \text{ and } \nabla^2\phi(x, y) = \begin{pmatrix} \frac{a^2}{(ax+by+c)^2} & \frac{ab}{(ax+bx+c)^2} \\ \frac{ab}{(ax+bx+c)^2} & \frac{b^2}{(ax+by+c)^2} \end{pmatrix}$$

for  $\phi(x, y) = -\log(-(ax + bx + c))$  on the set  $\{(x, y) \in \mathbb{R}^2 : ax + by + c < 0\}$  and where  $a, b, c \in \mathbb{R}$ .