

# ECE/CS 559 - Fall 2017 - Homework #3

Due: 10/06/2017, the end of class.

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**Note:** All notes in the beginning of Homework #1 apply. Only a subset of the problems may be graded.

1. **Theorem:** Let  $X$  be a real  $m \times q$  matrix. There exists a unique  $q \times m$  matrix  $X^+$  with the following properties: (i)  $XX^+X = X$ , (ii)  $X^+XX^+ = X^+$ , and (iii)  $X^+X$  and  $XX^+$  are symmetric matrices. The matrix  $X^+$  is called the pseudo-inverse, or the Moore-Penrose pseudo-inverse of  $X$ .

Show that if  $X$  has linearly independent columns, then  $X^T X$  is invertible, and  $X^+ = (X^T X)^{-1} X^T$ . You may use the theorem above.

Show that if  $X$  has linearly independent rows, then  $XX^T$  is invertible, and  $X^+ = X^T (XX^T)^{-1}$ . You may use the theorem above.

2. In this computer experiment, we will implement the gradient descent method and Newton's method. Let  $f(x, y) = -\log(1 - x - y) - \log x - \log y$  with domain  $\mathcal{D} = \{(x, y) : x + y < 1, x > 0, y > 0\}$ .
  - (a) Find the gradient and the Hessian of  $f$  on paper.
  - (b) Begin with an initial point in  $w_0 \in \mathcal{D}$  with  $\eta = 1$  and estimate the global minimum of  $f$  using the Gradient descent method, which will provide you with points  $w_1, w_2, \dots$ . Report your initial point  $w_0$  and  $\eta$  of your choice. Draw a graph that shows the trajectory followed by the points at each iteration. Also, plot the energies  $f(w_0), f(w_1), \dots$ , achieved by the points at each iteration. Note: During the iterations, your point may "jump" out of  $\mathcal{D}$  where  $f$  is undefined. If that happens, change your initial starting point and/or  $\eta$ .
  - (c) Repeat part (b) using Newton's method.
  - (d) Compare the speed of convergence of gradient descent and Newton's method, i.e. how fast does each method approach the estimated global minimum?
3. Perform the following steps: **Steps (e)-(g) are optional and will not be graded, although I highly recommend you do them.**
  - (a) Let  $x_i = i, i = 1, \dots, 50$ .
  - (b) Let  $y_i = i + u_i, i = 1, \dots, 50$ , where each  $u_i$  should be chosen to be an arbitrary real number between  $-1$  and  $1$ .
  - (c) Find the linear least squares fit to  $(x_i, y_i), i = 1, \dots, 50$ . Note that the linear least squares fit is the line  $y = w_0 + w_1 x$ , where  $w_0$  and  $w_1$  should be chosen to minimize  $\sum_{i=1}^{50} (y_i - (w_0 + w_1 x_i))^2$ .
  - (d) Plot the points  $(x_i, y_i), i = 1, \dots, 50$  together with their linear least squares fit.
  - (e) Find (on paper) the gradient of  $\sum_{i=1}^{50} (y_i - (w_0 + w_1 x_i))^2$  (derivatives with respect to  $w_0$  and  $w_1$ ).
  - (f) (Re)find the linear least squares fit using the gradient descent algorithm. Compare with (c).
  - (g) Show (on paper) that a single iteration of Newton's method with  $\eta = 1$  provides the globally optimal solution (the solution in (c)) regardless of the initial point.