## OPTIMIZATION. HOMEWORK 12

## OSCAR DALMAU

(1) Consider the optimization problem

$$F(\theta) = \frac{1}{N} \sum_{i=1}^{N} (h_{\theta}(\boldsymbol{x}_i) - y_i)^2$$
 (1)

where  $(\boldsymbol{x}_i, y_i), \, \boldsymbol{x}_i \in \mathbb{R}^n, \, y_i \in \{0, 1\}$ ,  $i = 1, 2, \cdots, N$  are given and

$$h_{\theta}(\boldsymbol{x}) = f_{\boldsymbol{a},b}(g_{\boldsymbol{c},\boldsymbol{d}}(\boldsymbol{x}))$$
  
 $g_{\boldsymbol{c},\boldsymbol{d}} : \mathbb{R}^n \to \mathbb{R}^m$   
 $f_{\boldsymbol{a},b} : \mathbb{R}^m \to \mathbb{R}$ 

 $m \in \mathbb{N}$  is known,

$$g_{c,d}(\boldsymbol{x}) = [\sigma(\boldsymbol{c}_j^T \boldsymbol{x} + d_j)]_{j=1}^m$$

$$f_{\boldsymbol{a},b}(\boldsymbol{z}) = \sigma(\boldsymbol{a}^T \boldsymbol{z} + b)$$

$$\sigma(t) = \frac{1}{1 + e^{-t}}, \ t \in \mathbb{R}$$

and  $\theta$  corresponds to the set of parameters  $\boldsymbol{a}, b, \boldsymbol{c}, \boldsymbol{d}$ .

Implement the following algorithms:

- Gradient descent with fixed step size and gradient approximation.
- BFGS with gradient approximation.

Compare the previous algorithms with respect to number of iteration and computational time using the mnist dataset and selecting only two digits.

**Note:** You can use any library to load mnist dataset. However, you can use dataset mnist.pkl.gz

• train\_set[0] is a matrix of size (50000, 784) where n = 50000 is the number of observations and each row represents an observation  $\boldsymbol{x}_i \in \mathbb{R}^{784}$ 

• train\_set[1] is a vector of size (50000) where each entry  $y_i \in \{0, 1, \dots, 9\}$ 

Select from train\_set[0] and train\_set[1] the set of observations  $S = \{(\boldsymbol{x}_i, y_i)\}$  with  $\boldsymbol{x}_i \in \mathbb{R}^{784}$  and  $y_i \in \{0, 1\}$  and estimate the parameters  $\hat{\theta}$  that minimizes the function  $F(\theta)$  in equation (1).

Select from test\_set[0] and test\_set[1] the set  $\mathcal{T} = \{(\boldsymbol{x}_i, y_i)\}$  such that  $\boldsymbol{x}_i \in \mathbb{R}^{784}$  and  $y_i \in \{0, 1\}$  and compute the error

error = 
$$\frac{1}{|\mathcal{T}|} \sum_{(\boldsymbol{x}_i, y_i) \in \mathcal{T}} |\mathbf{1}_{h_{\hat{\theta}}(x_i) > 0.5}(\boldsymbol{x}_i) - y_i|$$

where  $|\mathcal{T}|$  represents the number of elements of the set  $\mathcal{T}$ .