Machine Learning (Homework 2)

Due date: 2022/11/25 23:59:59

1 Classification Problem (45%)

You are given a dataset of handwritten digits (MNIST.zip) from the MNIST dataset. The dataset contains 10 classes with 128 different images in each class. In this exercise, you need to implement

- (1) least squares for classification
- (2) logistic regression model for classification

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Note: You need to normalize the data samples before training and randomly select 32 images as test data for each class and the remaining images as training data.

- 1.1 Implement the least squares for classification. You should use a 1-of-K binary coding scheme for the target vector \mathbf{t} . Show the classification accuracy and the loss value of training and test data. (10%)
- 1.2 Implement the logistic regression model using batch GD (batch gradient descent), SGD (stochastic gradient descent) and mini-batch SGD with softmax activation. Set the initial weight vector $\mathbf{w}_k = [w_{k1}, \dots, w_{kF}]$ to be a zero vector where F is the number of features and k is the number of classes. (30%)

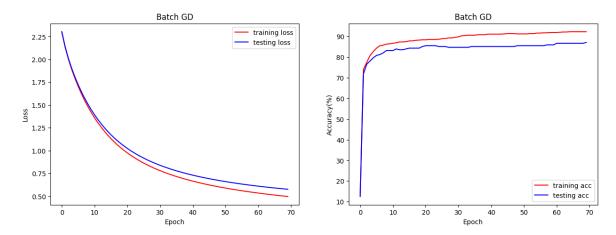
Algorithms	Batch size	Iterations in one epoch
batch GD	N	1
SGD	1	N
mini-batch SGD	B	N/B

N: number of training data, B: batch size (can be selected by yourself)

The error function is defined as

$$E(\mathbf{w}) = -\sum_{m=1}^{N} \sum_{k=1}^{K} t_{nk} \log y_{nk}.$$

(a) **Plot** the learning curves of the loss function and the classification accuracy versus the number of epochs until convergence for training data as well as test data, e.g.



- (b) Show the final classification accuracy and loss value of training and test data.
- (c) Based on your observation about the different algorithms (batch GD, SGD and minibatch SGD), please **make some discussion**.
- 1.3 Make some discussion about the difference between the results of 1.1 and 1.2. (5%)

2 Gaussian Process for Regression (55%)

In this exercise, you will implement Gaussin process (GP) for regression. The files **x.csv** and **t.csv** have input data $\mathbf{x} : \{x_1, x_2, \dots, x_{100}\}, 0 < x_i < 10$ and the corresponding target data $\mathbf{t} : \{t_1, t_2, \dots, t_{100}\}$ respectively. Please take the first 150 points as the training set and the rest as the test set. A regression function $y(\cdot)$ is used to express the target value by

$$t_n = y\left(x_n\right) + \epsilon_n$$

where the noisy signal ϵ_n is Gaussian distributed, $\epsilon_n \sim \mathcal{N}\left(0, \beta^{-1}\right)$ with $\beta^{-1} = 1$.

1. Please construct a kernel function using the basis functions in the following polynomial model and implement the Gaussian process for regression. (10%)

$$y(\mathbf{x}, \mathbf{w}) = w_0 + \sum_{i=1}^{D} w_i x_i + \sum_{i=1}^{D} \sum_{j=1}^{D} w_{ij} x_i x_j$$
 (M = 2).

Please **plot** the prediction result like Figure 6.8 of textbook for training set but one standard deviation instead of two and without the green curve. The red line shows the mean $m(\cdot)$ of the Gaussian process predictive distribution. The pink region corresponds to the band with positive and negative of one standard deviation. Training data points are shown in blue. Besides, please **show** the corresponding root-mean-square errors (shown below) for both training and test sets in **.ipynb** file.

$$E_{\text{RMS}} = \sqrt{\frac{1}{N} \left(m \left(x_n \right) - t_n \right)^2}.$$

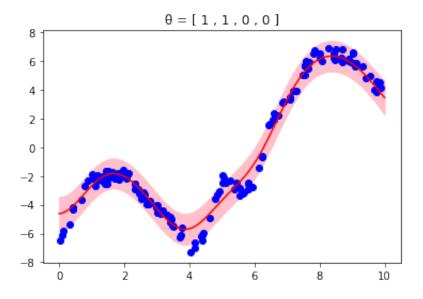
2. Repeat 1 by using the widely used exponential-quadratic kernel function given by

$$k\left(\mathbf{x}_{n}, \mathbf{x}_{m}\right) = \theta_{0} \exp \left\{-\frac{\theta_{1}}{2} \left\|\mathbf{x}_{n} - \mathbf{x}_{m}\right\|^{2}\right\} + \theta_{2} + \theta_{3} \mathbf{x}_{n}^{\mathsf{T}} \mathbf{x}_{m}$$

where the hyperparameters $\boldsymbol{\theta} = \{\theta_0, \theta_1, \theta_2, \theta_3\}$ are fixed. Please use the training set with four different combinations (25%)

- linear kernel $\boldsymbol{\theta} = \{0, 0, 0, 1\}$
- squared exponential kernel $\theta = \{1, 1, 0, 0\}$
- exponential-quadratic kernel $\theta = \{1, 1, 0, 16\}$
- exponential-quadratic kernel $\theta = \{1, 2, 16, 0\}$

Each combination needs to **plot** the prediction result where the **title of the figure** should be the value of hyperparameter used in the model and **show** the corresponding root-mean-square error. An example of figure is provided below.



- 3. Try to tune the hyperparameter in 2 to find the best combination for the dataset. Use automatic relevance determination (ARD) in Chapter 6.4.4 of textbook. (15%)
- 4. Explain your findings and **do some discussion**. (5%)

3 Rules

- Please name the assignment as hw2_StudentID.zip (e.g. hw2_0123456.zip).
- In your submission, it needs to contain two files.
 - hw2_StudentID.ipynb file which contains all the results and codes for this homework. Also, it should contain the description or explanation for this homework. (Please write all codes in one file.)
 - hw2_StudentID.py file which is downloaded from the .ipynb file
- Implementation will be graded by
 - completeness
 - algorithm Correctness
 - model description
 - discussion
- Only Python implementation is acceptable.
- Only the packages we provided is acceptable.
- DO NOT PLAGIARIZE. (We will check program similarity score.)