

ELL784: Assignment 2

Maximum Marks: 6

Submission deadline: **9 October, 23:59**

1 Objective

To experiment with the use of SVMs for a multiclass classification problem, and understand the effects of varying various parameters therein.

2 Part 1 (4 marks)

2.1 Data

We provide a personalised input file that contains 3,000 labeled data points, with 25 features each. You should download your file from <http://web.iitd.ac.in/~sumeet/A2/<EntryID>.csv> (for example, <http://web.iitd.ac.in/~sumeet/A2/2010MT50608.csv>). This file contains 3,000 rows, with each row corresponding to a data point. Each row has 26 comma-separated values; the first 25 are the values of the features, and the last is the class label for that data point (there are 10 classes, denoted by the labels 0 to 9). Each data point is actually a low-dimensional representation of an image.

2.2 Methodology

Your task is to try and learn an SVM classifier for these images, using just the given features, and thereby also to assess the usefulness of the different features. Here is how you should proceed:

1. Familiarise yourself with a standard SVM library. The recommended one is LIBSVM (<http://www.csie.ntu.edu.tw/~cjlin/libsvm/>), which is available for MATLAB, Python, and many other languages. Figure out how you can set various parameters, such as the value of C for the soft-margin SVM, the choice of kernel function, and the kernel parameters (if any). You may wish to play with a simple toy data set to get a feel for using the library, before you move on to the actual data for this assignment.
2. *Binary classification:* Choose just 2 out of the 10 classes in your data, and train an SVM. Leave some data aside for validation, or ideally, use cross-validation. Study the effects of changing the different parameter values, including the type of kernel function being used. How do they affect the accuracy? Can you distinguish cases of overfitting, underfitting, and good fitting? Also try using only the first 10 features, instead of all 25, and compare the results in the two cases. Repeat this exercise for at least two more pairs of classes out of the 10 given to you. Do you consistently get the best results for the same parameter settings, or does it vary a lot depending on which pair of classes you're looking at?
3. *Multiclass classification:* Now we would like to train a classifier for all 10 classes. Figure out what method(s) your chosen library uses for multiclass classification. For instance, LIBSVM uses *one-versus-one* (see Bishop p. 183). Based on your choice, build a classifier for all the classes, and evaluate using validation or cross-validation. Again, study the effects of changing the various parameters and kernel function. Try and finetune them to obtain the best possible performance. How do these tuned values compare to what you obtained in the binary classification setting? If there are major differences, what might be the reason? Also, try training the multiclass classifier using only the first

10 features, instead of all 25. How does this affect your results? What does this tell you about the usefulness of the different features?

3 Part 2 (2 marks)

Similar to the previous leaderboard exercise in Assignment 1, we provide an extended version of the data set for the leaderboard exercise in Assignment 2. This dataset has 10,000 instances of the same 25 features (compared to the 3,000 instances in Part 1), and you are required to train a multiclass SVM model for predicting the labels on a target set of 2,000 instances (whose labels are hidden from you).

The files provided are the following:

1. http://web.iitd.ac.in/~sumeet/A2/train_set.csv, a training set of 10,000 samples of 25 features (similar to Part 1), followed by the target label for each sample. Hence, the training set has dimensionality 10000×26 .
2. http://web.iitd.ac.in/~sumeet/A2/test_set.csv, a target set of 2,000 samples, on which you will have to predict the labels using your multiclass SVM model. The target set has dimensionality 2000×25 .

Similar to the previous assignment, you are expected to experiment with hyperparameter tuning, model and feature selection and cross-validation to get an optimal accuracy score, which will be evaluated on the leaderboard (link to be shared separately via Piazza).

4 Evaluation

- You should prepare a report, compiling all your results and your interpretation of them, along with your overall conclusions. In particular, you should attempt to answer all of the questions posed in the previous sections. Any graphs or other visualisations should also be included therein. If you wish, you may also include code or other materials which are relevant, though this is not required. The submission link is <http://web.iitd.ac.in/~sumeet/submit.html>; put everything into a single zip file or tarball, and name it as per the instructions given there. The submission deadline is **October 9th, 23:59**. Any late submissions will be penalised.
- The schedule for demos/vivas will be announced by your respective TAs, in advance. If for any reason you cannot attend in your scheduled slot, you must arrange for an alternative slot with your TA well in advance. Last-minute requests for rescheduling will normally not be accepted.