Comparison of Multiclass Kernel Support Vector Machines and Neural Networks for Image Classification

Kernel-based Machine Learning and Multivariate Modeling - project

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Contents

1	Introduction					
	1.1	Task				
	1.2	Datasets				
		1.2.1 ZIP				
		1.2.2 MNIST				
		1.2.3 CIFAR-10				
2	Neu	al Networks				
	2.1	Architectures				
	2.2	Results				
	2.3	State of the art				
3	Multiclass Kernel SVMs					
	3.1	Multiclass SVM strategies				
		3.1.1 spoc-svc				
		3.1.2 kbb-svc				
		3.1.3 one-vs-all approach				
		3.1.4 tree-based approach				
	3.2	Parameter optimization				
		3.2.1 spoc-svc				
		3.2.2 kbb-svc				
		3.2.3 one-vs-all approach				
		3.2.4 tree-based approach				
	3.3	Results				

4	Conclusions			
	4.1 Difficulties of the datasets	,		
	4.2 Comparison of Neural Networks and SVMs	,		
5	Application			
	5.1 Description			
	5.2 Manual	,		

1 Introduction

1.1 Task

In this project we compare Kernel Support Vector Machines (KSVMs) and state of the art techniques like Convolutional Neural Networks (CNNs) for image classification. Since image classification is a multi-class problem, we implement our own strategies of how to use KSVMs in order to decide which class will be assigned to a given image, and we compare our strategies to the strategies which are implemented in the R-library kernlab.

Both CNNs and KSVMs have a lot of parameters, so a main part of our work was to search for parameter settings that lead to good classification results.

1.2 Datasets

We have three different datasets with varying kinds of images, image sizes and dataset sizes.

1.2.1 ZIP

The ZIP dataset was provided in the second part of the course for some homeworks. It contains 16x16-pixel grayscale images of handdrawn digits from 0 to 9. The dataset is rather small with a training set of 7291 and a test set of 2007 images. See Figure 1 for example images.

1.2.2 MNIST

The MNIST dataset is frequently used in the image classification literature and can be considered to be a standard dataset. We downloaded the dataset from [Yann LeCun's homepage]. It contains 28x28-pixel grayscale images of handdrawn digits from 0 to 9. The dataset is bigger than the ZIP dataset and comes with a training set of 60000 and a test set of 10000 images. See Figure 2 for example images.

1.2.3 CIFAR-10

The CIFAR-10 dataset contains 32x32-pixel RGB-coloured images of the following ten classes (see Figure 3 for example images):

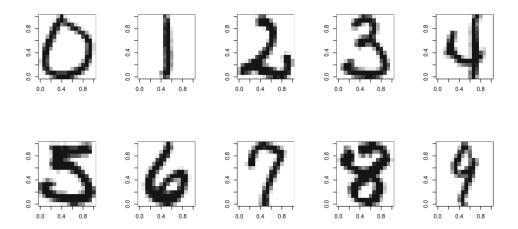


Figure 1: Example images of the 10 classes of the ZIP dataset.

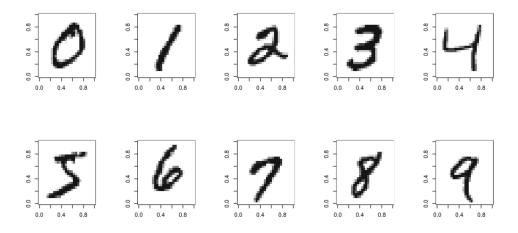


Figure 2: Example images of the 10 classes of the MNIST dataset.

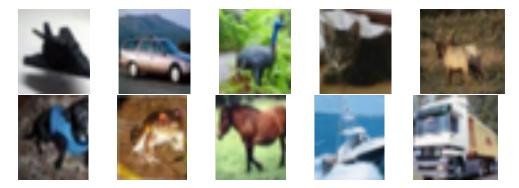


Figure 3: Example images for the 10 classes of the CIFAR-10 dataset.

- 1. airplane
- 2. automobile
- 3. bird
- 4. cat
- 5. deer
- 6. dog
- 7. frog
- 8. horse
- 9. ship
- 10. truck

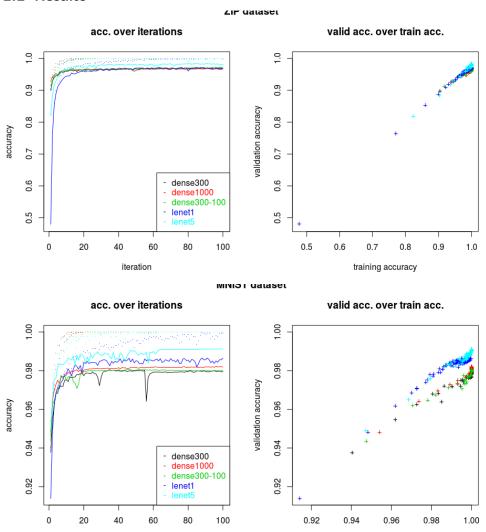
The dataset is divided into a training set of 50000 and a test set of 10000 images. All the data can be found on [Alex Krizhevsky's homepage].

2 Neural Networks

2.1 Architectures

All the network architectures are inspired by [LeCun et al., 1998].

2.2 Results



training accuracy

iteration

CIFAN IU UAIASEI

acc. over iterations valid acc. over train acc. 1.0 1.0 dense300 dense1000 dense300-100 6.0 6.0 8.0 0.8 validation accuracy 0.7 0.7 9.0 9.0 0.5 0.5 0.4 0.4 0.3 0.3 0.5 0.7 8.0 0.9 20 40 60 80 100 0.6 1.0 0.4

training accuracy

2.3 State of the art

3 Multiclass Kernel SVMs

iteration

3.1 Multiclass SVM strategies

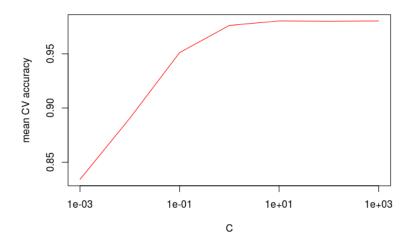
3.1.1 spoc-svc

See [Crammer and Singer, 2001] for details.

3.1.2 kbb-svc

See [Weston and Watkins] for details.

- 3.1.3 one-vs-all approach
- 3.1.4 tree-based approach
- 3.2 Parameter optimization
- 3.2.1 spoc-svc
- 3.2.2 kbb-svc
- 3.2.3 one-vs-all approach



- 3.2.4 tree-based approach
- 3.3 Results
- 4 Conclusions
- 4.1 Difficulties of the datasets
- 4.2 Comparison of Neural Networks and SVMs
- 5 Application
- 5.1 Description
- 5.2 Manual

References

[LeCun et al., 1998] Y. LeCun, L. Bottou, Y. Bengio, P. Haffner, *Gradient-Based Learning Applied to Document Recognition*, Proc. of the IEEE, November 1998.

[Alex Krizhevsky's homepage] https://www.cs.toronto.edu/~kriz/cifar.html

[Yann LeCun's homepage] http://yann.lecun.com/exdb/mnist/

[Crammer and Singer, 2001] K.Crammer, Y. Singer, On the Algorithmic Implementation of Multiclass Kernel-based Vector Machines, pp. 265-292, Journal of Machine Learning Research 2, 2001.

[Weston and Watkins] J. Weston, C.Watkins, Support Vector Machines for Multi-Class Pattern Recognition