

CO395 Machine Learning
CBC #2
Artificial Neural Networks

Group 1

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Implementation Details

Evaluation

The average F_1 measure per fold for both types of networks is presented below:

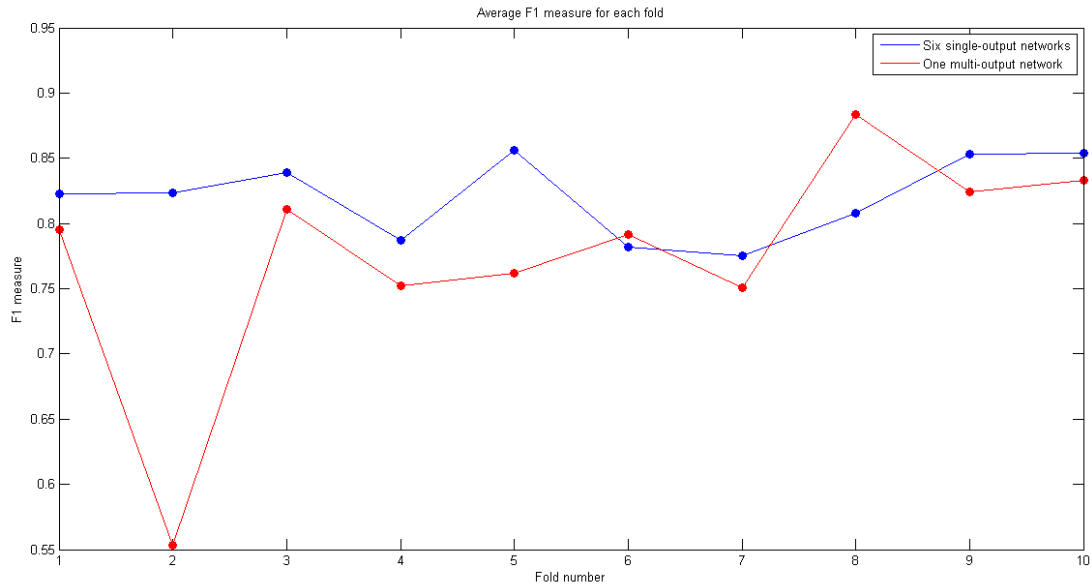


Figure 1: Average F_1 measure per fold for both types of networks

Clean dataset

Single six-output network

		Predicted class					
		1	2	3	4	5	6
Actual class	1	86	18	6	6	14	2
	2	9	167	6	5	11	0
	3	2	5	97	3	5	7
	4	0	6	1	203	4	2
	5	7	26	2	3	92	2
	6	0	3	12	2	2	188

Table 1: Confusion Matrix for single six-output ANN for the *clean* dataset

		Recall	Precision	F_1
Actual class	1	65%	83%	73%
	2	84%	74%	79%
	3	82%	78%	80%
	4	94%	91%	93%
	5	70%	72%	71%
	6	91%	94%	92%

Table 2: Recall, precision and F_1 measure for single six-output ANN for the *clean* dataset

$$C = \frac{833}{1004} = 83.0\%$$

Figure 2: Classification rate for single six-output ANN for the *clean* dataset

Six single-output networks

		Predicted class					
		1	2	3	4	5	6
Actual class	1	99	12	7	3	10	1
	2	9	168	3	6	11	1
	3	4	2	98	2	3	10
	4	2	4	2	205	1	2
	5	8	21	7	4	89	3
	6	0	3	10	5	2	187

Table 3: Confusion Matrix for six single-output ANNs for the *clean* dataset

		Recall	Precision	F_1
Actual class	1	75%	81%	78%
	2	84%	80%	82%
	3	82%	77%	80%
	4	95%	91%	93%
	5	67%	77%	72%
	6	90%	92%	91%

Table 4: Recall, precision and F_1 measure for six single-output ANNs for the *clean* dataset

$$C = \frac{846}{1004} = 84.3\%$$

Figure 3: Classification rate for six single-output ANNs for the *clean* dataset

Noisy dataset

Single six-output network

		Predicted class					
		1	2	3	4	5	6
Actual class	1	11	16	22	7	25	7
	2	4	149	18	6	6	4
	3	2	15	132	10	9	19
	4	3	7	13	175	4	7
	5	7	13	14	5	56	15
	6	1	4	15	5	9	186

Table 5: Confusion Matrix for single six-output ANN for the *noisy* dataset

		Recall	Precision	F_1
Actual class	1	13%	39%	19%
	2	80%	73%	76%
	3	71%	62%	66%
	4	84%	84%	84%
	5	51%	51%	51%
	6	85%	78%	81%

Table 6: Recall, precision and F_1 measure for single six-output ANN for the *noisy* dataset

$$C = \frac{709}{1001} = 70.8\%$$

Figure 4: Classification rate for single six-output ANN for the *noisy* dataset

Six single-output networks

		Predicted class					
		1	2	3	4	5	6
Actual class	1	15	11	25	8	23	6
	2	3	154	15	6	7	2
	3	1	11	139	8	11	17
	4	1	9	11	176	6	6
	5	4	8	19	4	67	8
	6	1	3	13	4	9	190

Table 7: Confusion Matrix for six single-output ANNs for the *noisy* dataset

		Recall	Precision	F_1
Actual class	1	17%	60%	27%
	2	82%	79%	80%
	3	74%	63%	68%
	4	84%	85%	85%
	5	61%	54%	58%
	6	86%	83%	85%

Table 8: Recall, precision and F_1 measure for six single-output ANNs for the *noisy* dataset

$$C = \frac{741}{1001} = 74.0\%$$

Figure 5: Classification rate for six single-output ANNs for the *noisy* dataset

Discussion of results

Questions

Optimal topology

The final, optimal topology which we used for the networks are:

1. Single six-output network:
 - Hidden layers: 1
 - Hidden neurons: 23
 - Transfer function: 'tansig' (sigmoid)
 - Training function: 'trainscg'
 - Learning rate: 0.01
 - Minimum gradient: $3 * 10^{-6}$
 - Goal: $2 * 10^{-3}$
2. Six single-output networks:
 - Hidden layers: 1
 - Hidden neurons: 9
 - Transfer function: 'tansig' (sigmoid)
 - Training function: 'trainscg'
 - Learning rate: 0.01
 - Minimum gradient: $5 * 10^{-6}$
 - Goal: $3 * 10^{-3}$

We arrived at these values by performing a series of simulations of the performance of networks with different values on the first fold on the *clean* dataset. In the first experiment, we tried to optimise the topology i.e. number of hidden layers, number of hidden neurons in each layer as well as the training function. We did this by repeatedly training the network with different parameters and plotted the results which are presented in **Figures 6, 7, 8, 9**. From the analysis of the results it can be concluded that the networks with just one hidden layer perform slightly better than those with two hidden layers, and so we trained our final networks with a single hidden layer. In case of the single six-output network, the peak is achieved for 32 hidden neurons, however as our final topology we selected 23 hidden neurons, since the classification rate for this topology is insignificantly smaller, yet the number of neurons is reduced by a third. For the six single-output networks, we decided to choose 9 hidden neurons.

In terms of the training function, it looks that 'trainscg' and 'trainrp' perform similarly, while 'trainlm' almost always exhibits the worst performance. Due to the speed of training and the classification rates at our chosen numbers of hidden neurons, we ultimately choose 'trainscg' as our training function for both types of networks.

We selected these values because we needed eventually to decide on the topology of the network. However, by looking at the plots, it is clear that the number of hidden layers and neurons (at least for the range of values which we investigated) does not cause the performance to vary significantly beyond some minimum number of neurons and before some maximum. As long as the number of hidden neurons is in some reasonable range (above 20 for the single six-output network and between 6 and 30 for six single-output networks), it does not matter tremendously what exact value we would choose.

Overfitting

Six-output vs. single-output networks

Ideal parameter optimisation

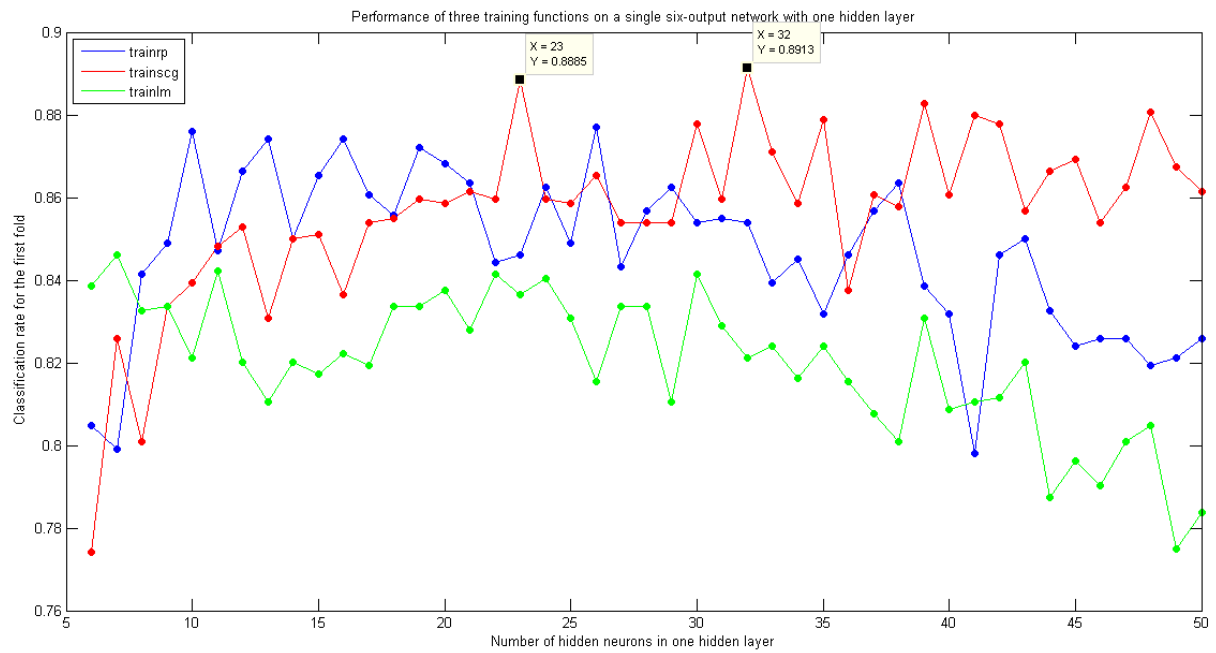


Figure 6: Classification rate for the first fold of the *clean* dataset with respect to the training function and the number of hidden neurons for a single six-output network with one hidden layer

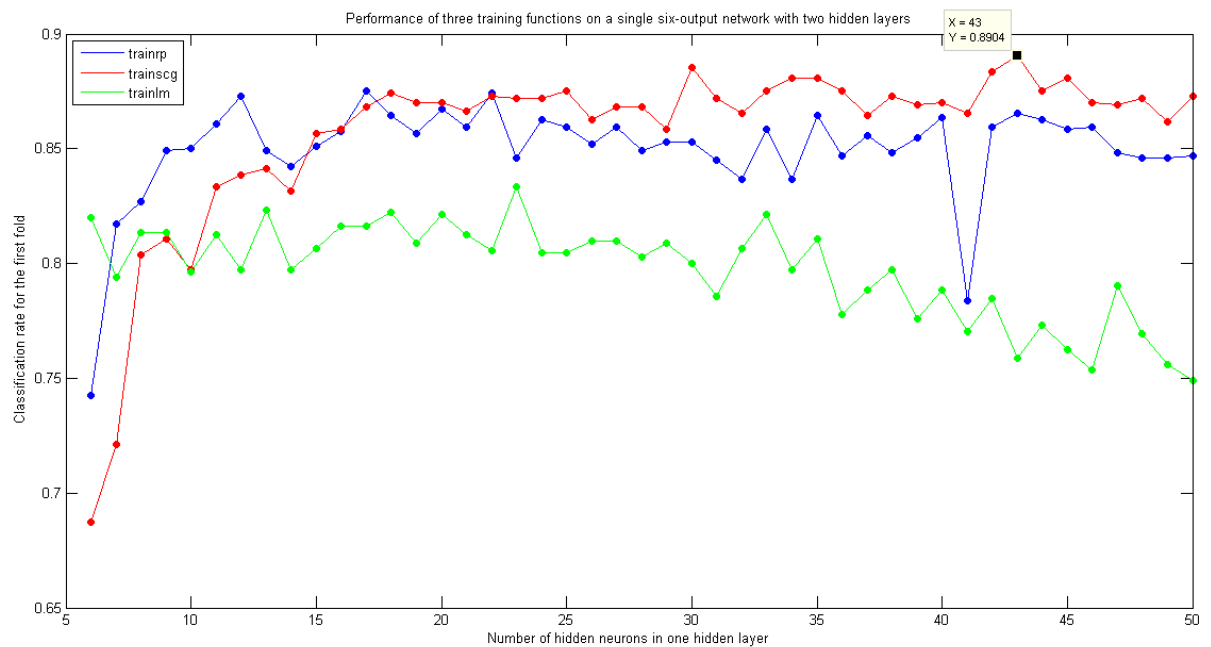


Figure 7: Classification rate for the first fold of the *clean* dataset with respect to the training function and the number of hidden neurons for a single six-output network with two hidden layers

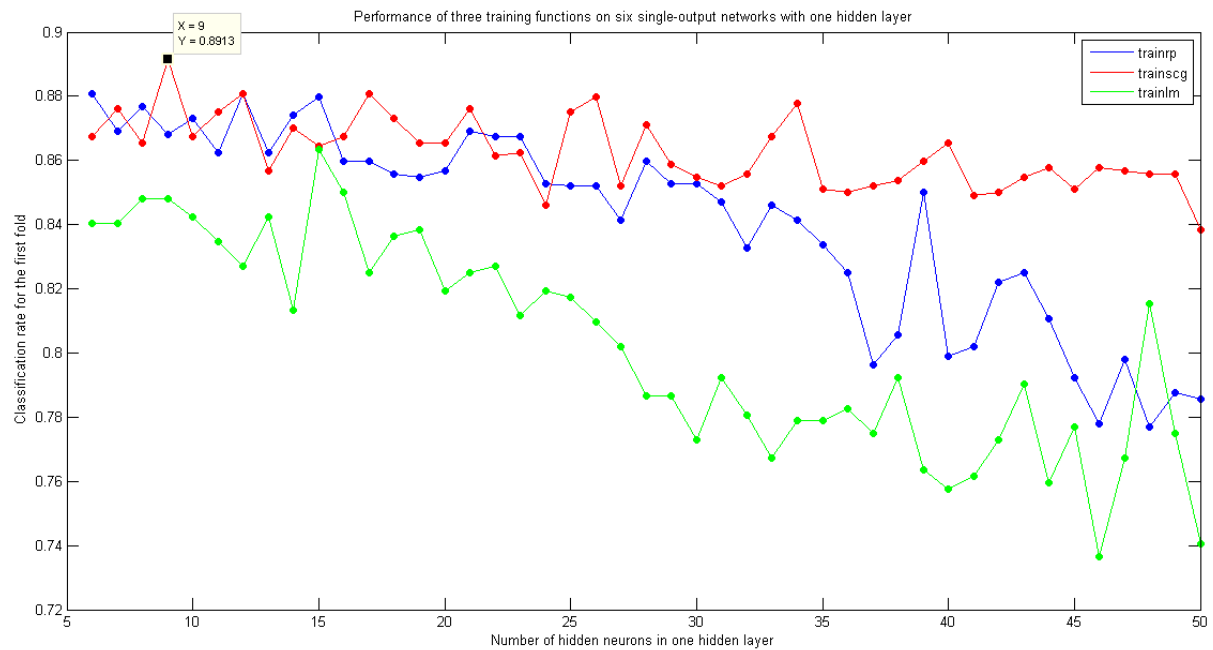


Figure 8: Classification rate for the first fold of the *clean* dataset with respect to the training function and the number of hidden neurons for six single-output networks with one hidden layer

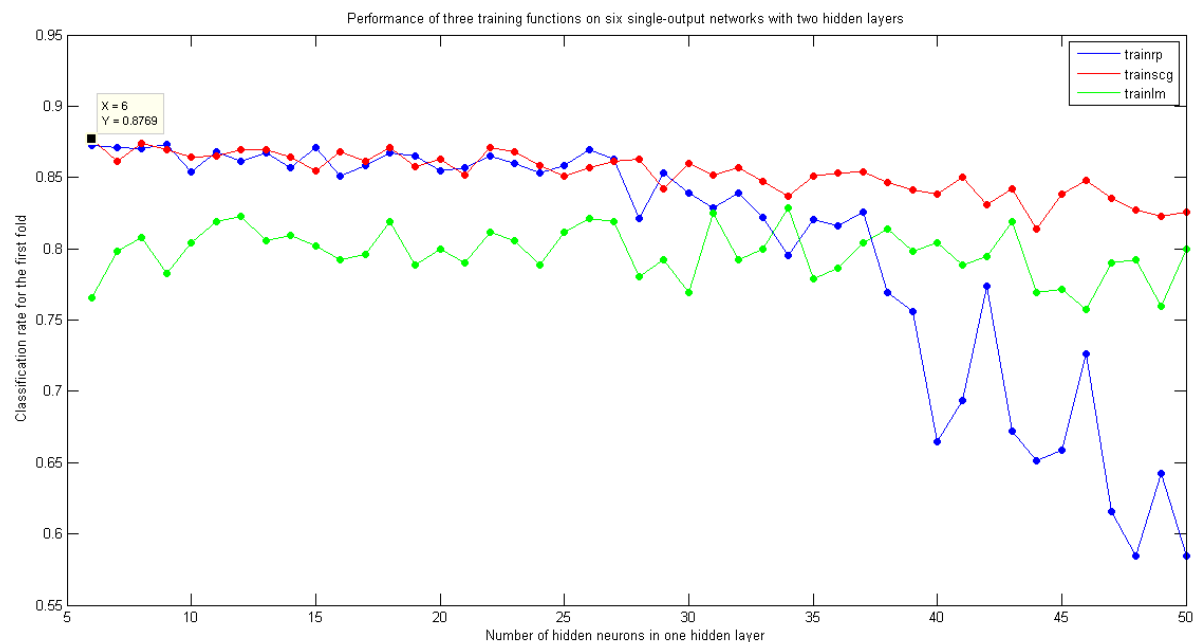


Figure 9: Classification rate for the first fold of the *clean* dataset with respect to the training function and the number of hidden neurons for six single-output networks with two hidden layers

Code Flowchart