# Q1 Variations of the Two-Spiral Task

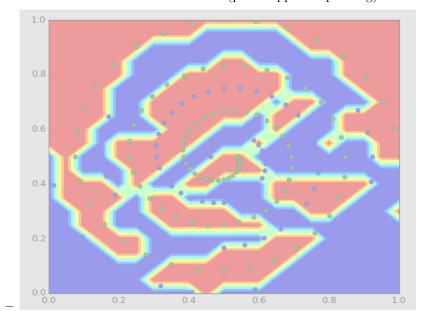
### a) Original Dataset

### Steps

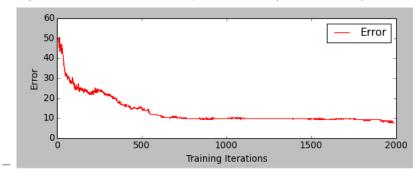
- 1. data obtained from http://wiki.cs.brynmawr.edu/?page=TwoSpiralsProblem
  - 1. version on blackboard did not contain class identifiers
  - 2. according to the code in the original paper, this seems to be the correct format
- 2. converted spaces to tabs
- 3. processed with Pybrain (pybrain-classify.py)
  - 1. followed tutorial
  - used two binary output neurons (dataset.\_convertToOneOfMany(bounds=[0.,1.]))
  - 3. used ideas from Beherey et al.
    - 1. network layout: 2 hidden layers with 77 neurons each
    - 2. activation: tanh for hidden layers, linear for output
    - 3. RPROP as training algorithm, because it converges faster than back propagation

#### Result

- reproduce by running pybrain-classify.py
- visualization of final result not available (plot stopped responding)



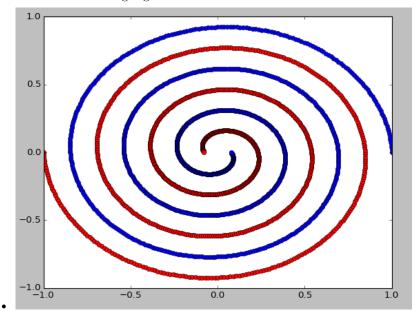
- training error achieved after 5000 epochs: 0.52% (1 misclassified)



## b) Self-generated dataset

### ${\bf Steps}$

1. generated data set using algorithm in blackboard

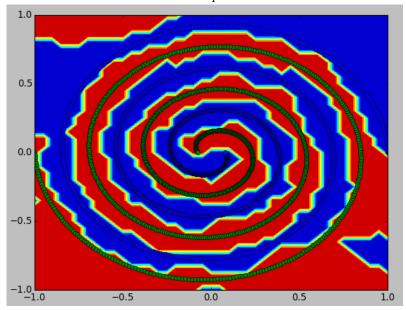


- 1. far denser spirals
- 2. 1920 (10x as many) data points
- 2. trained feed-forward net with same characteristics as in a) on new data

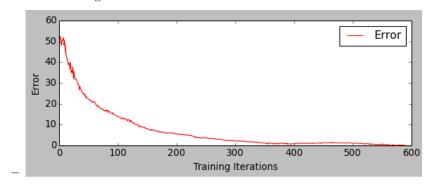
### Result

 $\bullet~10$  times as many data points leads to longer training times per epoch

- faster conversion
  - zero classification errors after 598 epochs



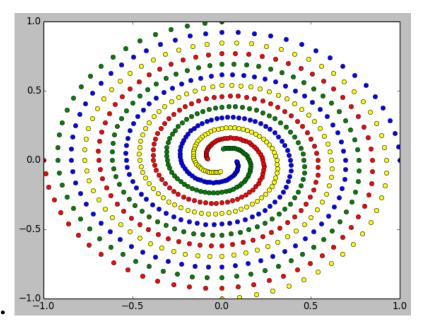
• smoother learning curve



# c) Four Spirals

### Steps

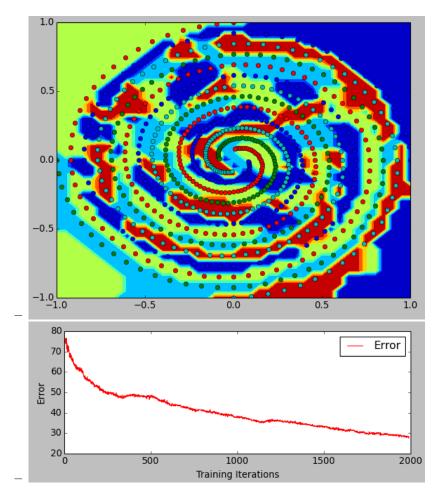
1. adapted spiral generation script to generate two additional spirals (rotated 90 degrees against original ones)



2. trained feed-forward net with same characteristics as in a) (but 4 classes instead of only two) on new data

### $\mathbf{Result}$

- $\bullet\,$  due to time constraints canceled training after 2000 epochs
  - classification error at this point: 28.42%



• up to this point promising: with enough time, the ANN should hopefully generalize

## d) ANNs vs SVMs

#### General Discussion

- $\bullet$  as discussed in class, SVMs can be seen as a generalisation of neural networks
  - $-\,$  with a good kernel, the spiral data can be transformed into a linearly separable form

### Results

- as suggested in the background reading paper, we used radial basis function kernels
- $\bullet\,$  far lower training times than ANNs for the spiral task