
Mustererkennung WiSe 12/13

Übung 8

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1 Aufgabe 1: Online vs. Batch

1.1 Code

1.1.1 a1.m

Durchführung der Algorithmen und abspeichern der Resultate.

```
1 penTrainingData = load("-ascii", "pendigits-training.txt");
2 ionTrainingData = load("-ascii", "ionosphere.data");
3
4
5 penFeaturesTraining = penTrainingData(:,1:end - 1);
6 penLabelsTraining = penTrainingData(:, end);
7
8 ionFeaturesTraining = ionTrainingData(:,1:end - 1);
9 ionLabelsTraining = ionTrainingData(:, end);
10
11 [penTrainingPCA, penMove, penCov, penU] = normalize(penFeaturesTraining);
12 [ionTrainingPCA, ionMove, ionCov, ionU] = normalize(ionFeaturesTraining);
13
14 % die labelmatrix zum Trainieren so aufbauen, dass an der entsprechenden ...
    Stelle eine 1 steht
15 penLabels = [];
16 for label = penLabelsTraining '
17     row = zeros(1,10);
18     row(label + 1) = 1;
19     penLabels = [penLabels; row];
20 end
21 penLabelsTraining = penLabels;
22
23
24 % die unwichtigsten Hauptkomponenten wegwerfen
25 penTrainingPCAn = penTrainingPCA(:, 1:14);
26
27 % die unwichtigsten Hauptkomponenten wegwerfen
28 ionTrainingPCAn = ionTrainingPCA(:, 1:14);
29
30 % arguments for online()/batch(): input, outputs, #hidden, maxError, ...
    gamma, subsetSize
31 [ionDigitsWeightsOnline, ionDigitsErrorsOnline] = ...
    online(ionTrainingPCAn, ionLabelsTraining, 35, 0.5, 0.05, 100)
32 save ionOnline.mat ionDigitsWeightsOnline ionDigitsErrorsOnline
33
34 [ionDigitsWeightsBatch, ionDigitsErrorsBatch] = batch(ionTrainingPCAn, ...
    ionLabelsTraining, 35, 0.5, 0.05, 100)
35 save ionBatch.mat ionDigitsWeightsBatch ionDigitsErrorsBatch
36
37 [penDigitsWeightsOnline, penDigitsErrorsOnline] = ...
    online(penTrainingPCAn, penLabelsTraining, 35, 0.8, 0.05, 100)
38 save penOnline.mat penDigitsWeightsOnline penDigitsErrorsOnline
39
```

```
40 [penDigitsWeightsBatch, penDigitsErrorsBatch] = batch(penTrainingPCAn, ...  
    penLabelsTraining, 35, 0.8, 0.05, 100)  
41 save penBatch.mat penDigitsWeightsBatch penDigitsErrorsBatch
```

1.1.2 aldisp.m

Plotten der Resultate.

```
1  
2 load penOnline.mat  
3 load penBatch.mat  
4  
5 load ionOnline.mat  
6 load ionBatch.mat  
7  
8 hold on  
9 figure  
10  
11 subplot(2, 2, 1)  
12 semilogy(penDigitsErrorsOnline);  
13 grid on  
14 title("pendigits errors online");  
15  
16  
17 subplot(2, 2, 2)  
18 semilogy(penDigitsErrorsBatch);  
19 grid on  
20 title("pendigits errors batch");  
21  
22  
23 subplot(2, 2, 3)  
24 semilogy(ionDigitsErrorsOnline);  
25 grid on  
26 title("ionosphere errors online");  
27  
28  
29 subplot(2, 2, 4)  
30 semilogy(ionDigitsErrorsBatch);  
31 grid on  
32 title("ionosphere errors batch");  
33  
34 hold off  
35  
36 print("al.png");  
37  
38 pause
```

1.2 Plots

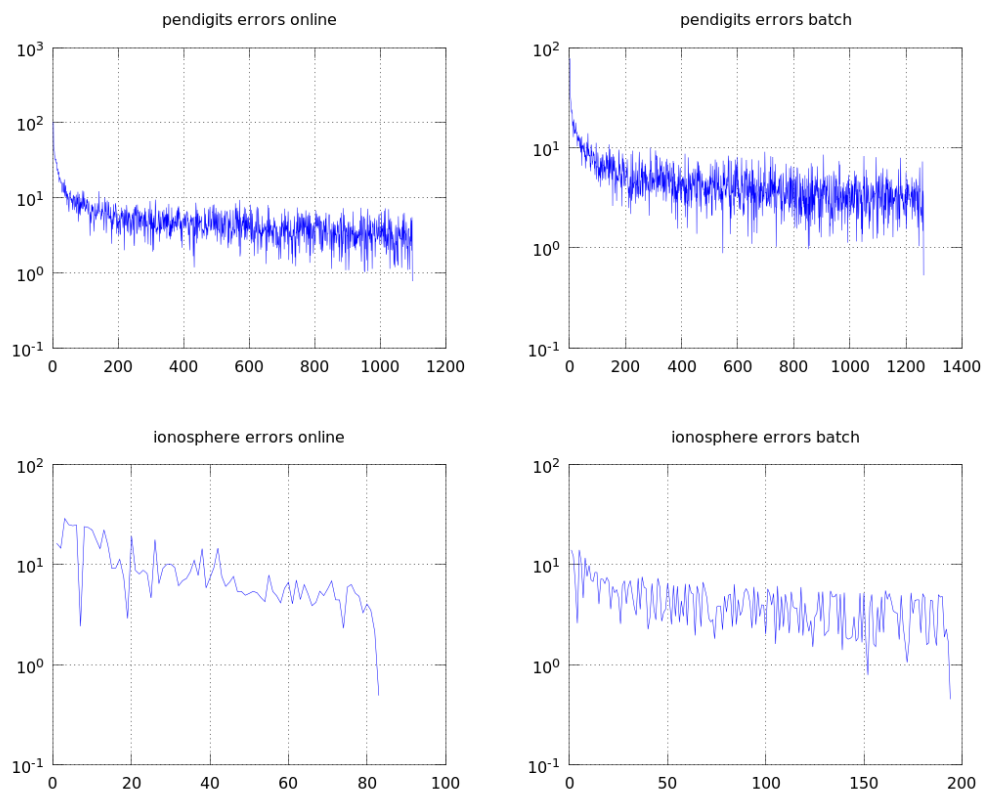


Figure 1: Iterationen vs. Fehler

1.3 Weiterer Code

1.3.1 online.m

```
1 function [weights, errors] = online(inputData, outputs, hiddenCnt, ...
   maxError, gamma, subsetSize)
2   inputWidth = size(inputData, 2)
3   outputWidth = size(outputs, 2)
4   W1 = rand(inputWidth + 1, hiddenCnt) ./ 0.5;
5   W2 = rand(hiddenCnt + 1, outputWidth) ./ 0.5;
6
7   errors = [];
8
9   error = inf
10  while (error > maxError)
11      error = 0;
12      correction1 = zeros(size(W1));
13      correction2 = zeros(size(W2));
14
15      % generate a subset of inputs to fasten up the computation
16      r = ceil((size(inputData, 1) - subsetSize) * rand());
17      subsetInputs = inputData(r:(r + subsetSize),:);
18      subsetOutputs = outputs(r:(r + subsetSize),:);
19
20      subset = [subsetInputs, subsetOutputs];
21
22      for input = subset'
23          % seperate the input from expected output
24          output = input(end - (outputWidth - 1): end);
25          input = augmentWithOnes(input(1:end - outputWidth)')';
26
27          % forward propagation
28          o1 = sigmoid(input' * W1);
29          o2 = sigmoid(augmentWithOnes(o1) * W2)';
30
31          err = (o2 ./ output);
32          error += sum(0.5 .* err .^ 2);
33
34          D1 = diag((o1 .* (1 ./ o1)));
35          D2 = diag((o2 .* (1 ./ o2)));
36
37          % the small W2 matrix
38          W2s = W2(2:end, :);
39
40          delta2 = D2 * err;
41          delta1 = D1 * W2s * delta2;
42
43          correction1 -= (gamma * delta1 * input')';
44          correction2 -= (gamma * delta2 * augmentWithOnes(o1))';
45      end
46
47      W1 += correction1;
48      W2 += correction2;
49
50      error'
51      errors = [errors, error];
52  end
53
54  weights = {W1, W2};
55  end
```

1.3.2 batch.m

```
1 function [weights, errors] = batch(inputData, outputs, hiddenCnt, ...
   maxError, gamma, subsetSize)
2   inputWidth = size(inputData, 2)
3   outputWidth = size(outputs, 2)
4   W1 = rand(inputWidth + 1, hiddenCnt) ./ 0.5;
5   W2 = rand(hiddenCnt + 1, outputWidth) ./ 0.5;
6
7   errors = [];
8
9   error = inf
10  while (error > maxError)
11      error = 0;
12
13      % generate a subset of inputs to fasten up the computation
14      r = ceil((size(inputData, 1) - subsetSize) * rand());
15      subsetInputs = inputData(r:(r + subsetSize),:);
16      subsetOutputs = outputs(r:(r + subsetSize),:);
17
18      subset = [subsetInputs, subsetOutputs];
19
20      for input = subset '
21          % seperate the input from expected output
22          output = input(end - (outputWidth - 1): end);
23          input = augmentWithOnes(input(1:end - outputWidth)')';
24
25          % forward propagation
26          o1 = sigmoid(input' * W1);
27          o2 = sigmoid(augmentWithOnes(o1) * W2)';
28
29          err = (o2 ./ output);
30          error += sum(0.5 .* err .^ 2);
31
32          D1 = diag((o1 .* (1 ./ (1 - o1))));
33          D2 = diag((o2 .* (1 ./ (1 - o2))));
34
35          % the small W2 matrix
36          W2s = W2(2:end, :);
37
38          delta2 = D2 * err;
39          delta1 = D1 * W2s * delta2;
40
41          W1 -= (gamma * delta1 * input')';
42          W2 -= (gamma * delta2 * augmentWithOnes(o1))';
43
44      end
45
46      error '
47      errors = [errors, error];
48
49  end
50
51  weights = {W1, W2};
52 end
```

1.3.3 normalize.m

```
1 function [ret, mean, cov, u] = normalize(inputData)
2     % Mittelwert der Daten berechnen
3     mean = [mean(inputData(:,1:end-1)) 0];
4     inputData = inputData - repmat(mean, size(inputData, 1), 1);
5
6     % compute mu and covariance matrix
7     [mu, cov] = gauss(inputData);
8
9     [u,s,v] = svd(cov);
10
11     ret = inputData * u;
12 end
```

2 Aufgabe 2: RPROP

2.1 Code

2.1.1 a2.m

Durchführung der Algorithmen und abspeichern der Resultate.

```
1 penTrainingData = load("-ascii", "pendigits-training.txt");
2 ionTrainingData = load("-ascii", "ionosphere.data");
3
4
5 penFeaturesTraining = penTrainingData(:,1:end - 1);
6 penLabelsTraining = penTrainingData(:, end);
7
8 ionFeaturesTraining = ionTrainingData(:,1:end - 1);
9 ionLabelsTraining = ionTrainingData(:, end);
10
11 [penTrainingPCA, penMove, penCov, penU] = normalize(penFeaturesTraining);
12 [ionTrainingPCA, ionMove, ionCov, ionU] = normalize(ionFeaturesTraining);
13
14 % die labelmatrix zum Trainieren so aufbauen, dass an der entsprechenden ...
    Stelle eine 1 steht
15 penLabels = [];
16 for label = penLabelsTraining '
17     row = zeros(1,10);
18     row(label + 1) = 1;
19     penLabels = [penLabels; row];
20 end
21 penLabelsTraining = penLabels;
22
23
24 % die unwichtigsten Hauptkomponenten wegwerfen
25 penTrainingPCAn = penTrainingPCA(:, 1:14);
26
27 % die unwichtigsten Hauptkomponenten wegwerfen
28 ionTrainingPCAn = ionTrainingPCA %(:, 1:14);
29
30
31 [ionWeightsRPROP, ionErrorsRPROP] = rprop(ionTrainingPCAn, ...
    ionLabelsTraining, 35, 0.05, 0.0001, 0.5, 1.2, 0.5, 100)
32 [penWeightsRPROP, penErrorsRPROP] = rprop(penTrainingPCAn, ...
    penLabelsTraining, 35, 0.505, 0.0001, 0.5, 1.2, 0.5, 100)
33 save rprop.mat ionErrorsRPROP penErrorsRPROP
```

2.1.2 a2disp.m

Plotten der Resultate.

```
1
2 load rprop.mat
3
4 hold on
5 figure
6
7 subplot(2,1,1)
8 semilogy(ionErrorsRPROP);
9 grid on
10 title("ionospere errors rprop");
11
12 subplot(2,1,2)
13 semilogy(penErrorsRPROP);
14 grid on
15 title("pendigits errors rprop");
16
17
18 hold off
19
20 print("a2.png");
21
22 pause
```


2.2 Plots

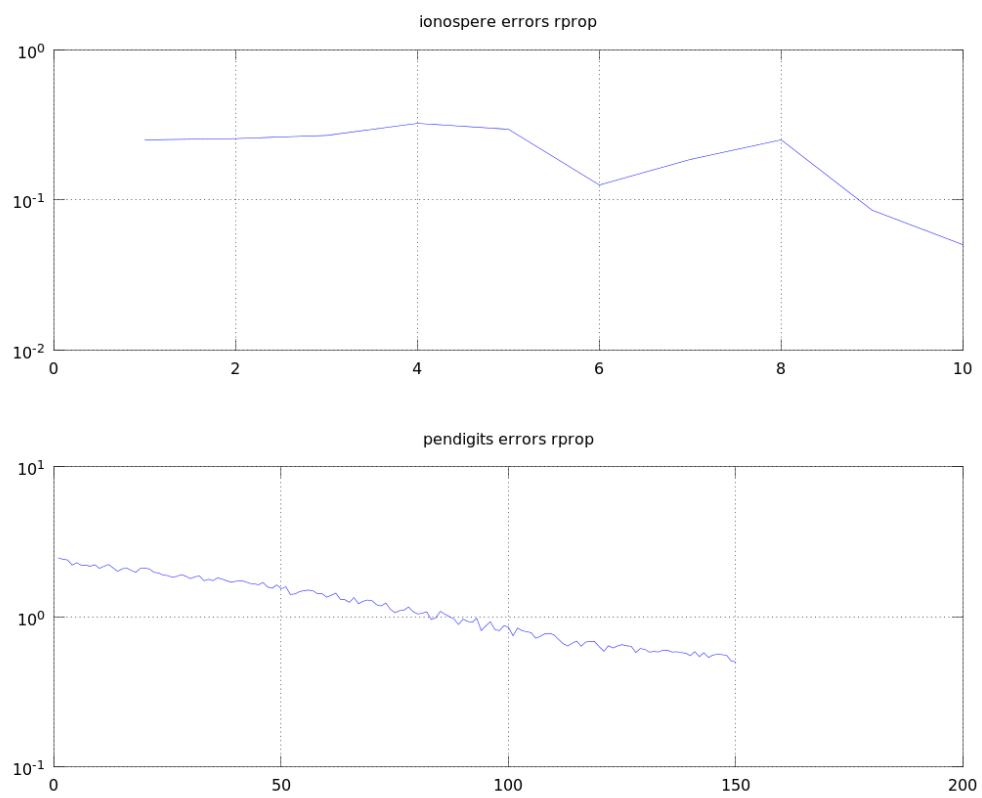


Figure 2: Iterationen vs. Fehler

2.3 Weiterer Code

2.3.1 rprop.m

```
1 function [weights, errors] = rprop(inputData, outputs, hiddenCnt, ...
2   maxError, gammaMin, gammaMax, u, d, subsetSize)
3   inputWidth = size(inputData, 2)
4   outputWidth = size(outputs, 2)
5   W1 = (rand(inputWidth + 1, hiddenCnt) .- 0.5) .* 10;
6   W2 = (rand(hiddenCnt + 1, outputWidth) .- 0.5) .* 10;
7
8   gammaMax1 = repmat(gammaMax, size(W1));
9   gammaMax2 = repmat(gammaMax, size(W2));
10
11   gammaMin1 = repmat(gammaMin, size(W1));
12   gammaMin2 = repmat(gammaMin, size(W2));
13
14   gamma1 = gammaMin1;
15   gamma2 = gammaMin2;
16
17   errors = [];
18
19   error = 999
20   while (error > maxError)
21       error = 0;
22
23       % generate a subset of inputs to fasten up the computation
24       r = ceil((size(inputData, 1) - subsetSize) * rand());
25       subsetInputs = inputData(r:(r + subsetSize),:);
26       subsetOutputs = outputs(r:(r + subsetSize),:);
27
28       subset = [subsetInputs, subsetOutputs];
29
30       last_correction11 = zeros(size(W1));
31       last_correction21 = zeros(size(W2));
32
33       last_correction12 = zeros(size(W1));
34       last_correction22 = zeros(size(W2));
35
36       for input = subset'
37           % seperate the input from expected output
38           output = input(end - (outputWidth - 1): end);
39           input = augmentWithOnes(input(1:end - outputWidth))';
40
41           % forward propagation
42           o1 = sigmoid(input' * W1);
43           o2 = sigmoid(augmentWithOnes(o1) * W2)';
44
45           err = (o2 .- output);
46           error += (sum(0.5 .* err .^ 2) ./ subsetSize);
47
48           D1 = diag((o1 .* (1 .- o1)));
49           D2 = diag((o2 .* (1 .- o2)));
50
51           W2s = W2(2:end, :);
52
53           delta2 = D2 * err;
54           delta1 = D1 * W2s * delta2;
55
56           correction1 = (delta1 * input')';
57           correction2 = (delta2 * augmentWithOnes(o1))';
58
59           gamma1 = (min(gamma1 * u, gammaMax1) .* ...
60             ((last_correction11 .* last_correction12) < 0)) .+ ...
```

```

59         (max(gamma1 * d, gammaMin1) .* ...
        ((last_correction11 .* last_correction12) > ...
        0)) .+ ...
60         (gamma1 .* ((last_correction11 .* ...
        last_correction12) == 0));
61
62     gamma2 = (min(gamma2 * u, gammaMax2) .* ...
63             ((last_correction21 .* last_correction22) < 0)) .+ ...
64             (max(gamma2 * d, gammaMin2) .* ...
65             ((last_correction21 .* last_correction22) > ...
66             0)) .+ ...
67             (gamma2 .* ((last_correction21 .* ...
68             last_correction22) == 0));
69
70
71     last_correction12 = last_correction11;
72     last_correction22 = last_correction21;
73
74     last_correction11 = correction1;
75     last_correction21 = correction2;
76
77     W1 -= gamma1 .* sign(correction1);
78     W2 -= gamma2 .* sign(correction2);
79 end
80
81 error '
82 errors = [errors , error];
83
84 end
85
86 weights = {W1, W2};
87 end

```