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Nov 1

ASSIGNMENT 5

5.5 Stock market prediction

1. The linear coefficients:

$$(a_1, a_2, a_3, a_4) = (0.9452, 0.0197, -0.0136, 0.0468)$$

2. Mean squared prediction error:

$$SWE_{2000} = 13918.6329$$

$$SWE_{2001} = 3018.2678$$

I wouldn't recommend this linear model for stock market prediction since the mean squared prediction error is too big.

3. Source Code as $hw5_5.m$

5.6 Handwritten digit classification

1. Training:

Using gradient ascent: label $y_t = 0$ as number 3 and $y_t = 1$ as number 5.

Evidences are following.

(a) Converged Log-likelihood

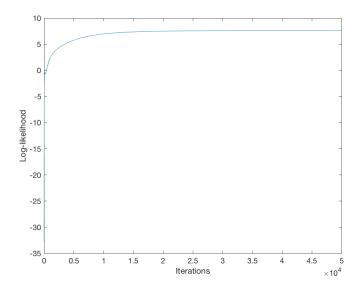


Figure 1: Converged Log-likelihood

2

(b) Converged Error Rate

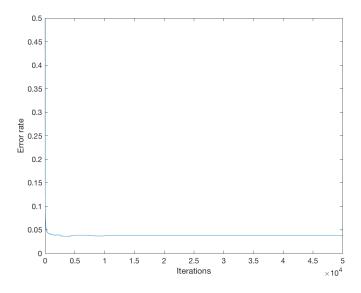


Figure 2: Converged Error Rate

(c) Weight Vector:

$$w = \begin{pmatrix} -0.6989 & -1.7908 & -1.0959 & -1.5592 & -0.6128 & -1.1959 & 0.8051 & 1.9816 \\ -0.3066 & -0.2751 & 0.3372 & -0.0348 & -0.7023 & 1.0081 & -1.5006 & -1.5139 \\ 4.5379 & 1.3988 & 1.6299 & 0.0953 & 1.0375 & -2.4793 & -2.4669 & -2.9456 \\ 0.7536 & 0.3636 & 0.7941 & -0.3658 & -0.5321 & -2.8130 & 0.5335 & -0.0648 \\ 0.6671 & 1.3347 & 0.1123 & -0.4830 & -0.6311 & -0.0300 & -0.6768 & -0.0605 \\ 1.3431 & -0.3001 & -0.4576 & -0.2282 & -0.0544 & -1.1704 & 1.0381 & -1.8978 \\ 1.7597 & -0.7811 & 1.4257 & 0.7418 & 0.5412 & -0.4761 & 0.1211 & -1.7663 \\ 0.7467 & 0.3604 & 0.7869 & 2.7175 & 0.4310 & 0.7548 & 0.9917 & -0.6337 \end{pmatrix}$$

2. Testing:

$$Error_{Three} = 7.5\%$$

$$Error_{Five} = 5.75\%$$

$$Error_{ALL} = 6.63\%$$

I think the error rate is acceptable.

3. Source Code as hw5_6.m

1 Source Code

Listing 1: hw5_5.m

```
clear all
1
   [x_0] = textread('nasdag00.txt', '%f');
3
4
5 % A 4*4
6 A = zeros(4,4);
   for t = 5: length (x_0)
        x_{t} = [x_{0}0(t-1,1); x_{0}0(t-2,1); x_{0}0(t-3,1); x_{0}0(t-4,1)];
9
        A = A + x_{t} * (x_{t})';
10 end
11
12 % b 4*1
13 b = zeros(4,1);
14 for t = 5: length(x_{-}00)
15
        for i = 1:4
16
            b(i,1) = b(i,1) + x_00(t,1) * x_00(t-i,1);
17
18
   \mathbf{end}
19
20 \ \% \ Aw = b
21 w = A^{(-1)} * b;
22
23 % SWE
24 \% 2000
25 \text{ SWE\_}00 = 0;
26 for t = 5: length(x_0)
27
        x_{t} = [x_{0}0(t-1,1); x_{0}0(t-2,1); x_{0}0(t-3,1); x_{0}0(t-4,1)];
28
        SWE_00 = SWE_00 + (x_00(t,1) - w'*x_t)^2;
29 end
30 SWE_00 = SWE_00/(length(x_00)-4);
31
32 % 2001
33 [x_01] = textread('nasdaq01.txt', '%f');
34 \text{ SWE}_{-}01 = 0;
35 for t = 5: length(x_01)
        x_{t} = [x_{0}1(t-1,1); x_{0}1(t-2,1); x_{0}1(t-3,1); x_{0}1(t-4,1)];
        SWE_01 = SWE_01 + (x_01(t,1) - w**x_t)^2;
37
38 end
   SWE_{-01} = SWE_{-01}/(length(x_{-01})-4);
```

Listing 2: $hw5_6.m$

1 clear all;

```
2 % Gradient Ascent
 3
4 \% ---- training ----
5
6 \ \% \ label \ yt = 0 \ for \ 3
 7 \% label yt = 1 for 5
8 	ext{ x3} = load('train3.txt');
9 	ext{ x5} = load('train5.txt');
10 y3 = zeros(size(x3, 1), 1);
11 y5 = ones(size(x5, 1), 1);
12 T = [x3; x5];
13 Y = [y3; y5];
14 I = ones(size(Y));
15
16 % Config of Learning
17 iter = 50000;
18 rate = 0.02/100;
19
20 % initialize weight
21 \text{ W} = \text{rand}(64, 1);
22 for i = 1: iter
23
        Grad = T'*(Y - sigmoid(T*W));
24
        W = W + rate*Grad;
25
         \operatorname{err}(i) = \operatorname{sum}(Y = (\operatorname{sigmoid}(T*W) > = 0.5))/\operatorname{length}(Y);
26
        L(i) = Y'*log(sigmoid(T*W)) - (I-Y)'*log(sigmoid(-T*W));
27 end
28 figure (1);
   plot(L(20:end)), xlabel('Iterations'), ylabel('Log-likelihood');
30 figure (2);
31 plot(err), xlabel('Iterations'), ylabel('Error_rate');
32
33 \% - testing -
34
35 	ext{ t3} = load('test3.txt');
36 	ext{ } t5 = load('test5.txt');
37 	 y3 = zeros(size(t3, 1), 1);
38 y5 = ones(size(t5, 1), 1);
39 T = [t3; t5];
40 \quad Y = [y3; y5];
   errorAll = sum(Y = (sigmoid(T*W)>=0.5))/length(Y);
42 errorOn3 = \operatorname{sum}(y3 = (\operatorname{sigmoid}(t3*W) >= 0.5))/\operatorname{length}(y3);
43 errorOn5 = \operatorname{sum}(y5 = (\operatorname{sigmoid}(t5*W)>=0.5))/\operatorname{length}(y5);
45 \% ---- sigmoid function ---
46 % function: sigmoid for a matrix at each [i][j]
47 function res = sigmoid(X)
```

```
\begin{array}{lll} 48 & & & \\ 49 & & I = {\rm ones}\,(\,\mathbf{size}\,(X)\,)\,; \\ 50 & & E = {\rm ones}\,(\,\mathbf{size}\,(X)\,)\!*\mathbf{exp}\,(\,1\,)\,; \\ 51 & & {\rm res}\,=\,I\,./\,(\,I\!+\!E.\,\hat{}\,(\,-\!X)\,)\,; \\ 52 & & \\ 53 & \mathbf{end} & & \\ \end{array}
```

Submitted by Xiaowen Mao on Nov 1.