

## Practical Exercise 5

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Der Code ist zusätzlich als Datei einsehbar.

### Part a:

Code:

```
1 %Task 1
2 %Part A
3
4 load ("twoClasses.mat")
5 c_0= patterns(:,1:2000);
6 c_0= transpose(c_0);
7 c_1= patterns(:,2001:4000);
8 c_1= transpose(c_1);
9
10
11 L_col=cell(1,3);           % cell array containing the labels for the columns
12 L_row=cell(2000,1);       % cell array containing the labels for the rows
13
14 L_col(1,2)="Sensor1";
15 L_col(1,3)="Sensor2";
16
17 for i=1:2000
18     L_row(i,1)=strcat("Obs", int2str(i));
19 end
20
21 % Adding labels to c_0
22 L_col(1,1)="Class0";
23 c_0_L = num2cell(c_0);
24 c_0_L = cat (2,L_row, c_0_L);
25 c_0_L = cat (1,L_col, c_0_L);
26
27 % Adding labels to c_1
28 L_col(1,1)="Class1";
29 c_1_L = num2cell(c_1);
30 c_1_L = cat (2,L_row, c_1_L);
31 c_1_L = cat (1,L_col, c_1_L);
32
33 %assembling cloud1
34 cloud1 = cat(3, c_0_L, c_1_L); %assambeling iris1
35
36 %defining help variables
37 Class0_sensor1= c_0(:,1);
38 Class0_sensor2= c_0(:,2);
39 Class1_sensor1= c_1(:,1);
40 Class1_sensor2= c_1(:,2);
41
```

### Beispielaufufe:

```
>> cloud1(1:3, 2, 1)
ans =
{
    [1,1] = Sensor1
    [2,1] = 3.6642
    [3,1] = 4.9162
}
```

```
>> cloud1(1:5,3,2)
ans =
{
    [1,1] = Sensor2
    [2,1] = 16.917
    [3,1] = 19.842
    [4,1] = 18.312
    [5,1] = 18.422
}
```

### Part b:

#### Code:

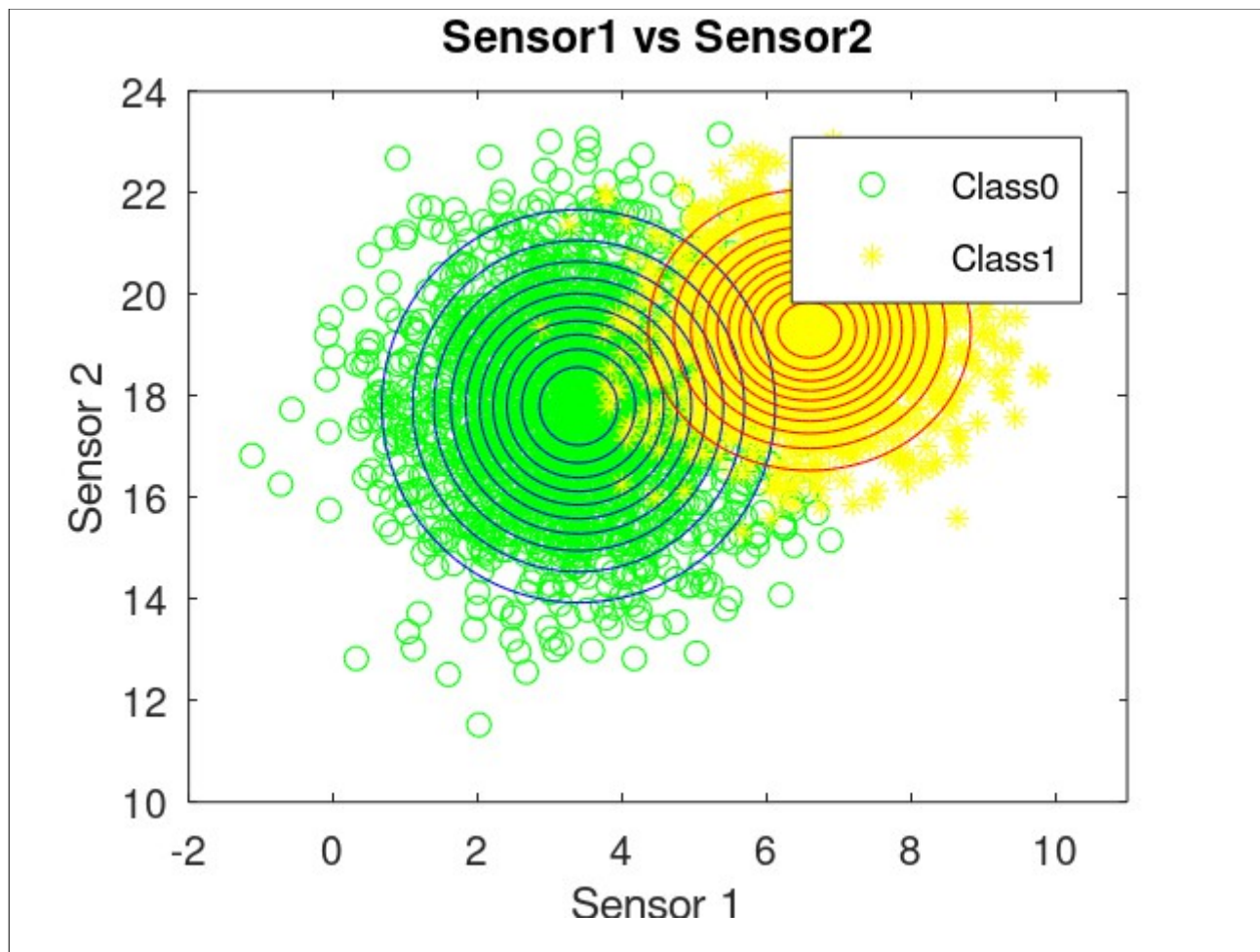
```
43 %Part B
44 % Gaussian model with independent components
45
46 %Calculating mean for class 0 and 1
47 m01 = mean(Class0_sensor1); %mean for class 0 and sensor 1
48 m02 = mean(Class0_sensor2);
49 m11 = mean(Class1_sensor1);
50 m12 = mean(Class1_sensor2);
51
52 %Calculating standard deviation for class 0 and 1
53 d01=sqrt(var(Class0_sensor1));
54 d02=sqrt(var(Class0_sensor2));
55 d11=sqrt(var(Class1_sensor1));
56 d12=sqrt(var(Class1_sensor2));
57
58 %Declaring interval of values for each sensor
59 i1= -2:0.1:11;
60 i2= 10:0.1:24;
61
62 %Calculating the maginals for class 0 and 1
63 p0_s1 = exp(-0.5*((i1-m01)./d01).^2)./(sqrt(2*pi)*d01);
64 p0_s2 = exp(-0.5*((i2-m02)./d02).^2)./(sqrt(2*pi)*d02);
65 p0_joint= p0_s2'*p0_s1;
66 p1_s1 = exp(-0.5*((i1-m11)./d11).^2)./(sqrt(2*pi)*d11);
67 p1_s2 = exp(-0.5*((i2-m12)./d12).^2)./(sqrt(2*pi)*d12);
68 p1_joint= p1_s2'*p1_s1;
69
```

```

70 %Plotting the two point distributions of the classes
71 %with contours of the two Gaussian distributions
72 figure (3);
73 plot(Class0_sensor1,Class0_sensor2,"go",Class1_sensor1,Class1_sensor2,"y*")
74     title('Sensor1 vs Sensor2');
75     xlabel ('Sensor 1');
76     ylabel ('Sensor 2');
77     legend("Class0","Class1");
78     hold on;
79     contour(i1,i2,p0_joint,"blue");
80     contour(i1,i2,p1_joint,"red");
81     hold off;
82

```

Plot:

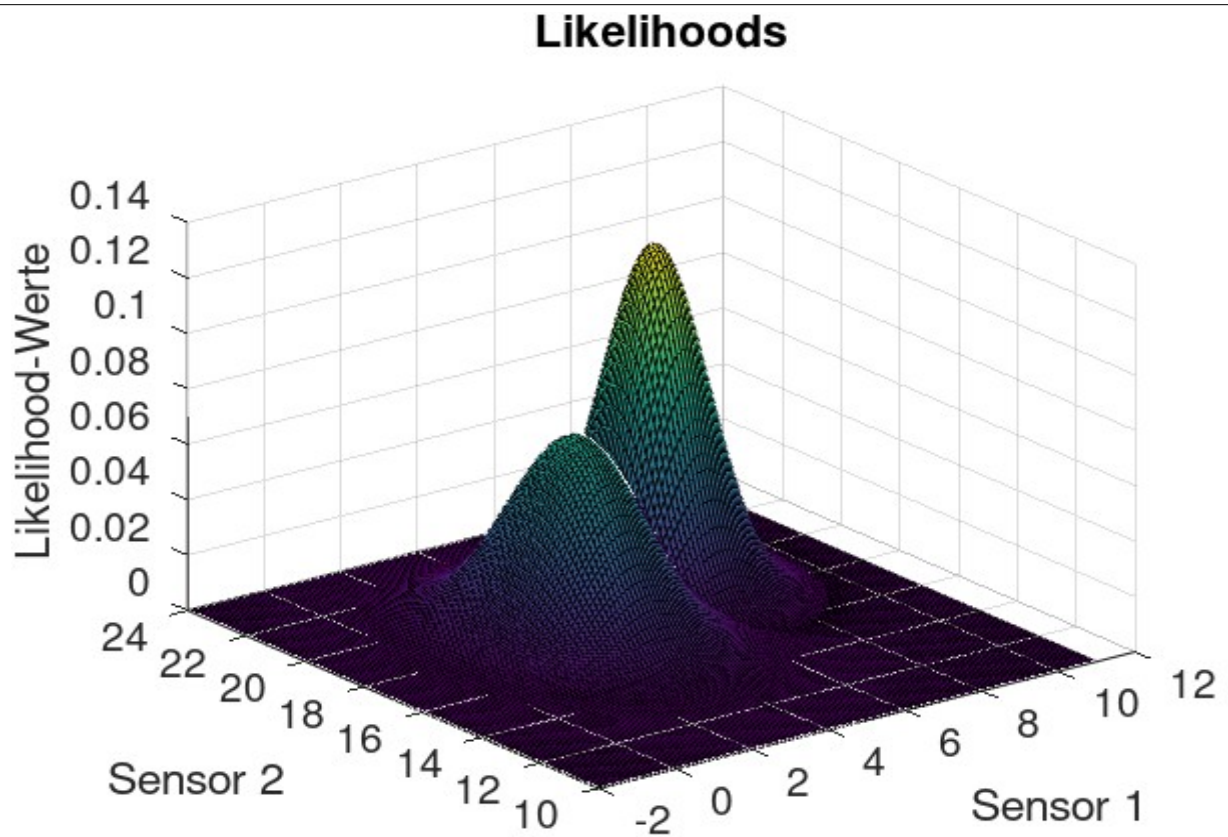


## Part c:

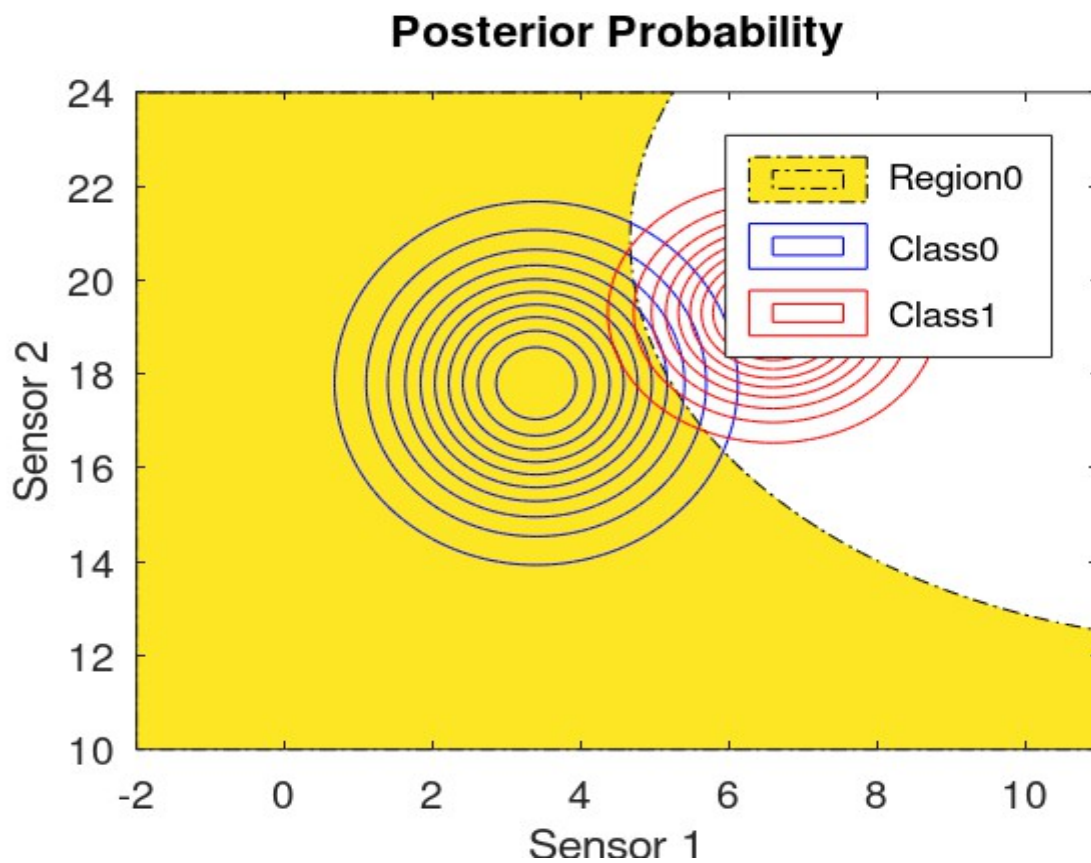
### Code:

```
84 %Part C
85 %Calculating the posterior probabilities of class 0
86 p_x_0 = p0_joint;
87 p_x_1 = p1_joint;
88
89 p_0_x = p_x_0 ./ (p_x_0 + p_x_1);
90
91
92 %Plotting the graphs
93 figure(1);
94
95 surf(i1,i2,p0_joint);
96 hold on;
97 surf(i1,i2,p1_joint);
98 title('Likelihoods');
99 xlabel ('Sensor 1');
100 ylabel ('Sensor 2');
101 zlabel ('Likelihood-Werte');
102 hold off;
103
104 figure (2);
105 contour(i1, i2, p_0_x, [0.5 0.5], 'k-.');
106 contourf(i1, i2, p_0_x, [0.5 0.5], 'k-.');
107 hold on
108 contour(i1,i2,p0_joint,"blue");
109 contour(i1,i2,p1_joint,"red");
110 title ('Posterior Probability');
111 xlabel ('Sensor 1');
112 ylabel ('Sensor 2');
113 legend("Region0","Class0","Class1");
114 hold off;
115
```

Plots:



Plot 1



Plot 2