Table of Contents

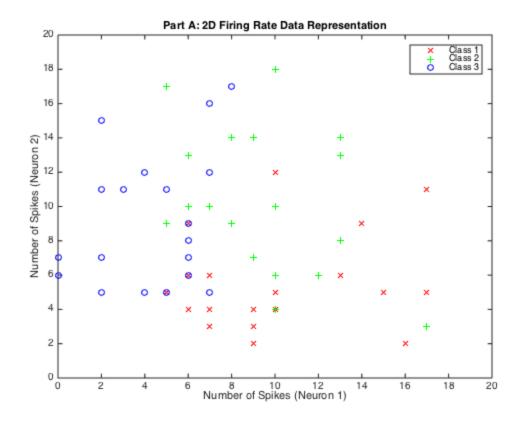
	1
Problem 3: Simulated Neural Data	1
Part A: 2D Plot	1
Part B: ML Parameters	2
Part C: Firing Rate Means	Δ
Part D: Means, Covariance Ellipses	5
Part E: Means, Covariance Ellipses and Decision Boundaries	7
% EE239AS Homework 4	
clc	
clear	
close all	

Problem 3: Simulated Neural Data

```
ps3_data = importdata('ps4_simdata.mat');
% 20x3 struct
% rows = data point
% columns = class
```

Part A: 2D Plot

```
D_trial = 2;
n_class = size(ps3_data,2);
n_trial = size(ps3_data,1);
data = cell(1, n_class);
for i = 1:n_trial
    for j = 1:n_class
        data{1,j} = [data{1,j}, ps3_data(i,j).x];
        % organize data into cell for easier access
    end
end
figure(1)
plotData(data)
% plot neurons in each class in different colors
title('Part A: 2D Firing Rate Data Representation')
xlabel('Number of Spikes (Neuron 1)')
ylabel('Number of Spikes (Neuron 2)')
legend('Class 1','Class 2','Class 3')
```



Part B: ML Parameters

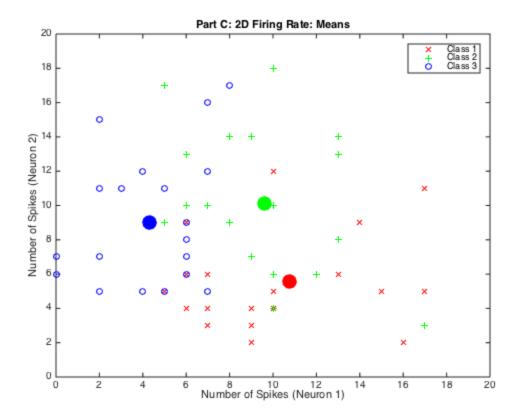
```
% Model (i) Gaussian, Shared Covariance
N_k = n_trial;
N = N_k*n_class;
P_Ck = N_k/(n_class*N_k);
 % calculate the prior probabilities of each class (equal for all classes)
mu_i = zeros(D_trial, n_class);
S k i = cell(1, n class);
sigma_i = zeros(D_trial, D_trial);
 for i = 1:n_class
                mu_i(:,i) = 1/(N_k)*sum(data{1,i},2);
                cov_trial_i = zeros(D_trial, D_trial);
                 for j = 1:n_trial
                                 cov_trial_i = cov_trial_i + (data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,i))*(data\{1,i\}(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)*(data\{1,i\}(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)-mu_i(:,j)
                                  % sum the (x-mu)*(x-mu)' matrices for each trial
                 end
                 S_k_i\{i\} = 1/N_k * cov_trial_i;
                 % calculate the S_k for each class and store into cell
                 sigma_i = sigma_i + N_k/N * S_k_i\{i\};
                 % calculate sigma (weighted sum of S_k)
 end
```

```
fprintf('Model (i) Gaussian, Shared Covariance\n----\n\n')
disp('Probability of Each Class:')
disp(P_Ck)
disp('Means:')
disp(mu_i)
disp('Covariance Matrix:')
disp(sigma_i)
% Model (ii) Gaussian, Class Specific Covariance
% Class probabilities and mean are the same as Model (i).
% The covariance matrices are specific to each class, as opposed to the
% weighted sum in Model (i).
fprintf('Model (ii) Gaussian, Class Specific Covariance\n-----
disp('Probability of Each Class:')
disp(P_Ck)
disp('Means:')
disp(mu_i)
disp('Covariance Matrix (Class 1):')
disp(S_k_i\{1\})
disp('Covariance Matrix (Class 2):')
disp(S_k_i\{2\})
disp('Covariance Matrix (Class 3):')
disp(S_k_i{3})
% Model (iii) Poisson
% Class probabilities and mean firing rate are the same as Model (i).
fprintf('Model (iii) Poisson, Class Specific Covariance\n-----
disp('Probability of Each Class:')
disp(P Ck)
disp('Mean Firing Rates:')
disp(mu_i)
Model (i) Gaussian, Shared Covariance
Probability of Each Class:
    0.3333
Means:
   10.7500
             9.6000
                       4.3000
   5.5500
            10.1000
                       9.0000
Covariance Matrix:
   11.9792
            -0.0242
   -0.0242
            12.5125
Model (ii) Gaussian, Class Specific Covariance
Probability of Each Class:
    0.3333
```

```
Means:
  10.7500 9.6000 4.3000
   5.5500 10.1000 9.0000
Covariance Matrix (Class 1):
  20.9875 2.1375
           7.2475
   2.1375
Covariance Matrix (Class 2):
   9.5400 -4.7100
  -4.7100 15.7900
Covariance Matrix (Class 3):
   5.4100 2.5000
   2.5000 14.5000
Model (iii) Poisson, Class Specific Covariance
Probability of Each Class:
   0.3333
Mean Firing Rates:
           9.6000 4.3000
  10.7500
   5.5500 10.1000 9.0000
```

Part C: Firing Rate Means

```
figure(2)
plotData(data)
title('Part C: 2D Firing Rate: Means')
xlabel('Number of Spikes (Neuron 1)')
ylabel('Number of Spikes (Neuron 2)')
legend('Class 1','Class 2','Class 3')
hold on
% plot means of each class (same for each model)
plotMeans(mu_i)
hold off
```



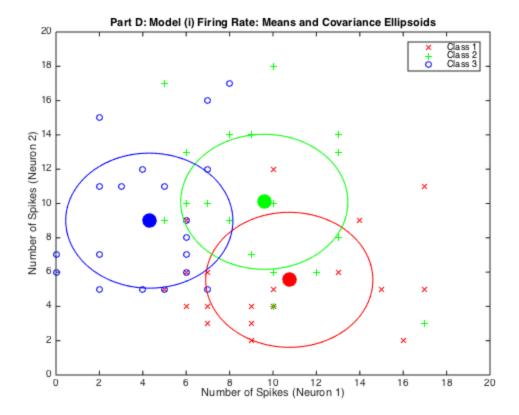
Part D: Means, Covariance Ellipses

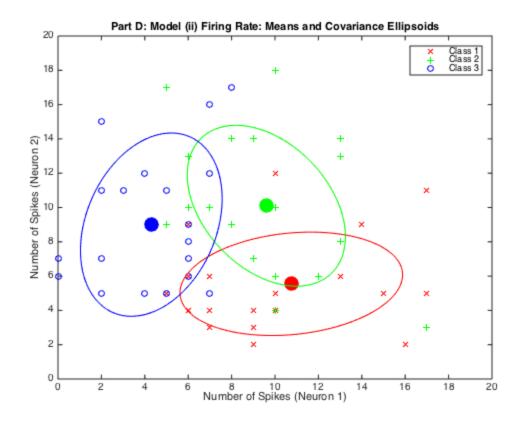
```
% Model (i) Gaussian, Shared Covariance
figure(3)
plotData(data)
title('Part D: Model (i) Firing Rate: Means and Covariance Ellipsoids')
xlabel('Number of Spikes (Neuron 1)')
ylabel('Number of Spikes (Neuron 2)')
legend('Class 1','Class 2','Class 3')
hold on
% plot means of each class (same for each model)
plotMeans(mu_i)
% plot covariance ellipses for each class (shared covariance)
plotContour(mu_i(:,1)',sigma_i,'r');
plotContour(mu_i(:,2)',sigma_i,'g');
plotContour(mu_i(:,3)',sigma_i,'b');
hold off
% Model (ii) Gaussian, Class Specific Covariance
figure(4)
```

```
plotData(data)
title('Part D: Model (ii) Firing Rate: Means and Covariance Ellipsoids')
xlabel('Number of Spikes (Neuron 1)')
ylabel('Number of Spikes (Neuron 2)')
legend('Class 1','Class 2','Class 3')

hold on
% plot means of each class (same for each model)
plotMeans(mu_i)
plotContour(mu_i(:,1)',S_k_i{1},'r');
plotContour(mu_i(:,2)',S_k_i{2},'g');
plotContour(mu_i(:,3)',S_k_i{3},'b');
```

hold off



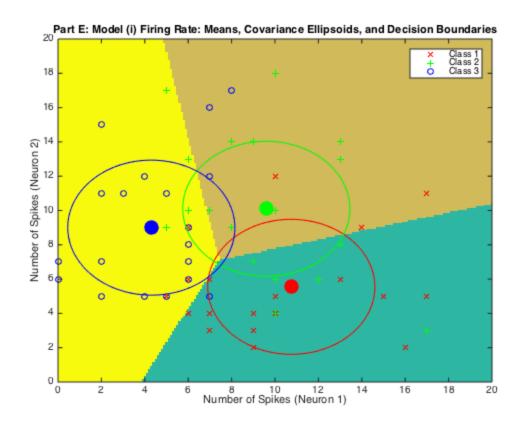


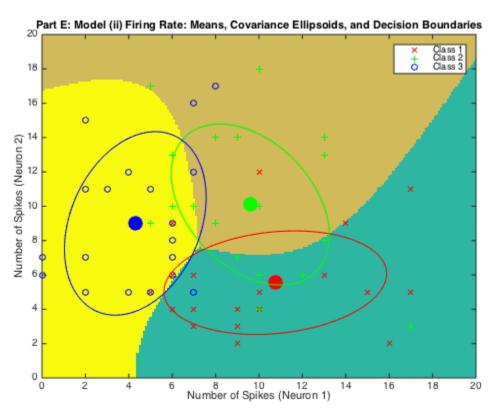
Part E: Means, Covariance Ellipses and Decision Boundaries

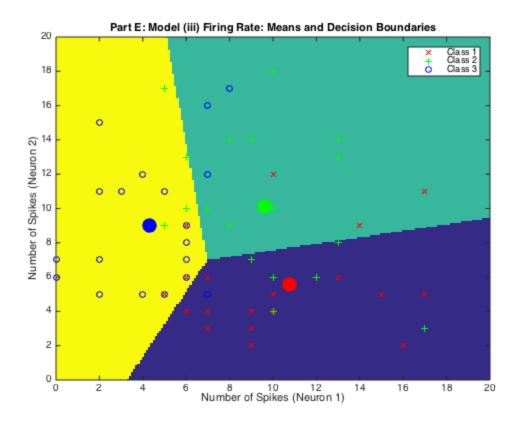
```
% Model (i) Gaussian, Shared Covariance
x = 0:0.1:20;
y = 0:0.1:20;
[X Y] = meshgrid(x,y);
xy = [X(:) Y(:)];
1 = length(xy);
img_size = size(X);
for j = 1:1
for i = 1:n_class
    k(j,i) = log(P_Ck) + mu_i(:,i)'*inv(sigma_i)*xy(j,:)'-0.5*mu_i(:,i)'...
        *inv(sigma_i)*mu_i(:,i);
end
end
[m,idx] = max(k, [], 2);
% reshape the idx (which contains the class label) into an image.
decisionmap = reshape(idx, img_size);
```

```
figure;
                                                   % show the image
imagesc(x,y,decisionmap);
hold on;
set(gca,'ydir','normal');
% colormap for the classes:
% class 1 = light red, 2 = light green, 3 = light blue
% cmap = [1 0.8 0.8; 0.95 1 0.95; 0.9 0.9 1]
% colormap(cmap);
% Plot the class training data
hold on
plotData(data)
plotMeans(mu_i)
plotContour(mu_i(:,1)',sigma_i,'r');
plotContour(mu_i(:,2)',sigma_i,'g');
plotContour(mu_i(:,3)',sigma_i,'b');
title('Part E: Model (i) Firing Rate: Means, Covariance Ellipsoids, and Decision B
xlabel('Number of Spikes (Neuron 1)')
ylabel('Number of Spikes (Neuron 2)')
legend('Class 1','Class 2','Class 3')
% Model (ii) Gaussian, Class Specific Covariance
x = 0:0.1:20;
y = 0:0.1:20;
[X Y] = meshgrid(x,y);
xy = [X(:) Y(:)];
l = length(xy);
img_size = size(X);
for j = 1:1
for i = 1:n_class
          k(j,i) = log(P_Ck) + mu_i(:,i)'*inv(S_k_i\{i\})*xy(j,:)'-0.5*mu_i(:,i)'*...
                    inv(S_k_i\{i\})*mu_i(:,i) - 0.5*xy(j,:)*inv(S_k_i\{i\})*xy(j,:)'-0.5*log(det(S_k_i\{i\})*xy(j,:)'-0.5*log(det(S_k_i\{i\})*xy(j,:)'-0.5*log(det(S_k_i\{i\})*xy(j,:)'-0.5*log(det(S_k_i\{i\})*xy(j,:)'-0.5*log(det(S_k_i\{i\})*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'-0.5*log(det(S_k_i[i])*xy(j,:)'
end
end
[m,idx] = max(k, [], 2);
% reshape the idx (which contains the class label) into an image.
decisionmap = reshape(idx, img_size);
figure;
%show the image
imagesc(x,y,decisionmap);
hold on;
set(gca,'ydir','normal');
% colormap for the classes:
% class 1 = light red, 2 = light green, 3 = light blue
% cmap = [1 0.8 0.8; 0.95 1 0.95; 0.9 0.9 1]
% colormap(cmap);
% plot the class training data.
```

```
plotData(data)
plotMeans(mu i)
plotContour(mu_i(:,1)',S_k_i{1},'r');
plotContour(mu_i(:,2)',S_k_i{2},'g');
plotContour(mu_i(:,3)',S_k_i{3},'b');
title('Part E: Model (ii) Firing Rate: Means, Covariance Ellipsoids, and Decision
xlabel('Number of Spikes (Neuron 1)')
ylabel('Number of Spikes (Neuron 2)')
legend('Class 1','Class 2','Class 3')
% Model (iii): Poisson
x = 0:0.1:20;
y = 0:0.1:20;
[X Y] = meshgrid(x,y);
xy = [X(:) Y(:)];
l = length(xy);
imq size = size(X);
k_iii = [];
for j = 1:1
for i = 1:n_class
  k_{iii}(j,i) = xy(j,:)*log(mu_{i}(:,i)) - sum(mu_{i}(:,i)');
end
end
[m,idx] = max(k iii, [], 2);
% reshape the idx (which contains the class label) into an image.
decisionmap = reshape(idx, img_size);
figure;
%show the image
imagesc(x,y,decisionmap);
hold on;
set(gca,'ydir','normal');
%colormap for the classes:
%class 1 = light red, 2 = light green, 3 = light blue
% cmap = [1 0.8 0.8; 0.95 1 0.95; 0.9 0.9 1]
% colormap(cmap);
% plot the class training data.
plotData(data)
plotMeans(mu_i)
title('Part E: Model (iii) Firing Rate: Means and Decision Boundaries')
xlabel('Number of Spikes (Neuron 1)')
ylabel('Number of Spikes (Neuron 2)')
legend('Class 1','Class 2','Class 3')
```







Published with MATLAB® R2014b