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```
%Skylar Tamke - Project 3 EM Clustering
% I have estimated 6 clusters for this set of data. This parameter
  can be
% changed easily by modifying the variable M. When using the algorithm
  the
% initial conditions are very key to how
% successful the algorithm is. If the starting conditions are bad the
% algorithm will cluster badly for a 2d array.

clc
clear
close all
warning('off','all')

%number of clusterings
M = 6;

%load data file
load("data.mat");

%initialize data set
x = data;
[N, d] = size(x);
mean = zeros(M,2);
pi_val = zeros(1,M);
pi_val(:) = 1/M;

%plot Data
scatter(data(:,1),data(:,2),3);
hold on
title("Data set [18000 points]")

%automatically finds clusters based on the kmeans method, initial
  guess
[idx1,C1] = kmeans(data,M);
% C1 = rand(6,2)*100

%getting the mean for each point. initially
for i = 1:M
    mu(i,:) = C1(i,:);
end
```

```
% scatter(C1(:,1),C1(:,2),100)

% initialize cov values for each Gaussian (M) scalable
for m = 1:M
    Em(:, :, m) = eye(2);
end

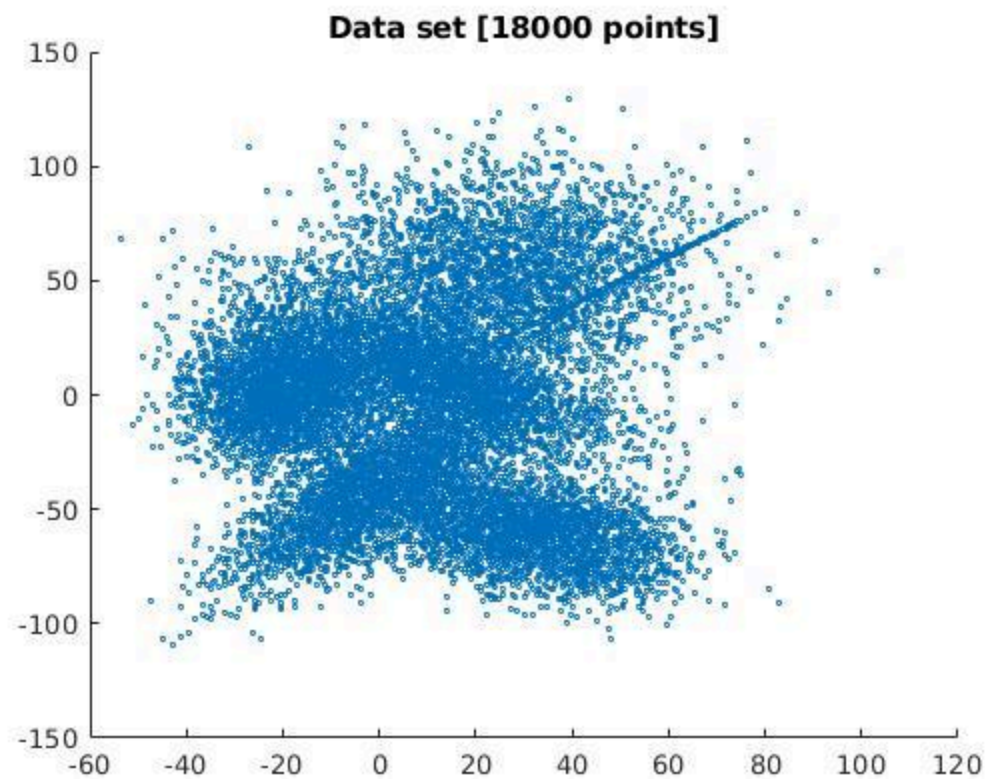
% creating booleans to use when determining if continued EM is needed
boolean = 1;
covariance_threshold = .01
loopcount = 0

covariance_threshold =

    0.0100

loopcount =

    0
```



Algorithm

The algorithm will run until the end condition is met

```
while( boolean)

    loopcount = loopcount + 1;
    % Expectation algorithm step
    cond_sum = zeros(N,1);
    for i = 1:N
        % first the probability that the data point is in a certain
        cluster
        for m=1:M
            %
            cond_prob(i,m) = mvnpdf(x(i,:),mu(m,:),Em(:, :,m));
            cond_prob(i,m) =
            (((2*pi)^(d/2))*((det(Em(:, :,m))))^(-.5))) ...
            * exp((-0.5*(x(i,:)-mu(m,:))*(inv(Em(:, :,m)))*(x(i,:)-
            (mu(m,:)))')));
            end

            for m =1:M
                cond_sum(i) = cond_prob(i,m)*pi_val(m) + cond_sum(i);
            end
            % calculate the expectation of each data point
            for m=1:M
                z(i,m) = cond_prob(i,m)*pi_val(m)/cond_sum(i);
            end
        end
        %getting rid of any NaNs, which are the result of the
        %conditional_probabilities that ended up being zero.
        for i = 1:N
            for m = 1:M
                if(isnan(z(i,m)))
                    z(i,m) = 0;
                end
            end
        end
    end

    %resetting this summation so that algorithm doesn't run away
    cov_num = zeros(2,2,6);
    % % % M step
    for m = 1:M

        z_sum = sum(z(:,m));
        for i = 1:N
            cov_num(:, :,m) = z(i,m) * (x(i,:) - mu(m,:))' * (x(i,:) -
            mu(m,:)) + cov_num(:, :,m);
        end
        %calculate new covariance matrices
        Em(:, :,m) = cov_num(:, :,m)/sum(z(:,m));

        %record old means so we can compare to check convergence
```

```

        old_mu = mu;
        for i = 1:N
            mu_num(i,:) = z(i,m) * x(i,:);
        end
        sum_num = sum(mu_num);
        %calculate new means
        mu(m,:) = sum_num/(z_sum);
        %calculate new prior probabilities
        pi_val(m) = z_sum/N;

    end

```

Threshold difference

This is where convergence of EM properties are checked to see if the algorithm is successful or not

```

    %reset convergence check
    mu_difference = 0;
    for m = 1:M
        mu_difference = abs(old_mu(m) - mu(m)) + mu_difference;
    end

    if(mu_difference < covariance_threshold)
        boolean = 0;
    end

```

Centriod plot

plot each centriod change, commented out for publish `scatter(mu(:,1),mu(:,2),50)`

```

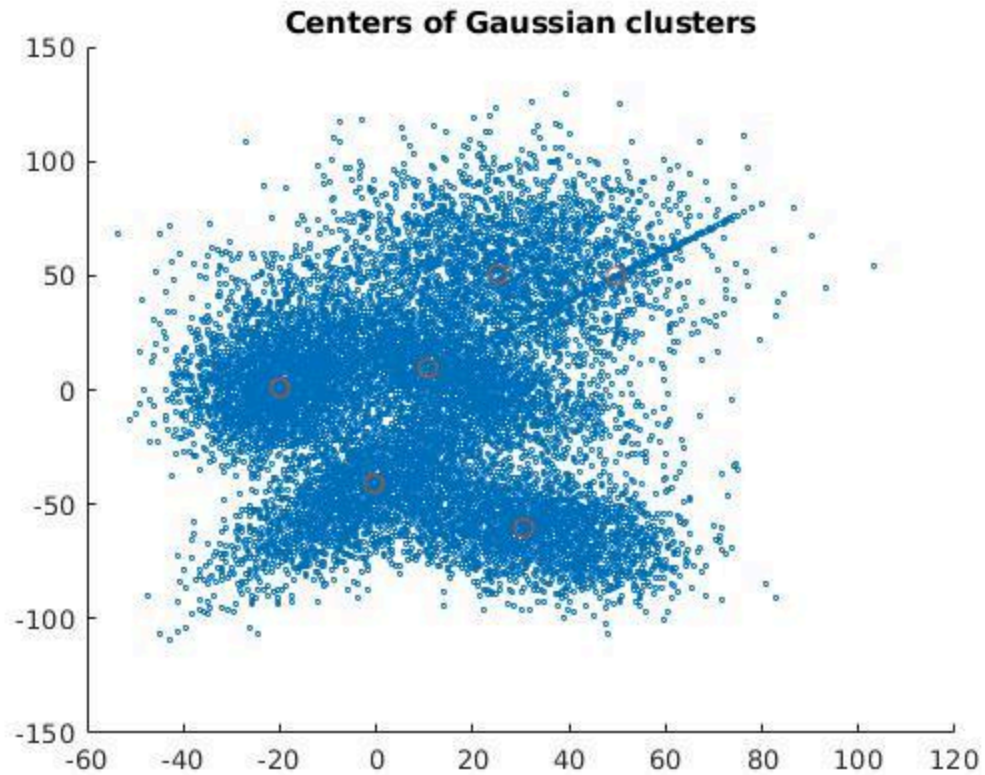
    end

    % A count of how many times the algorithm ran before finding a
    loopcount
    % A plot of the completed centers
    scatter(mu(:,1),mu(:,2),50)
    title("Centers of Gaussian clusters")

```

loopcount =

42

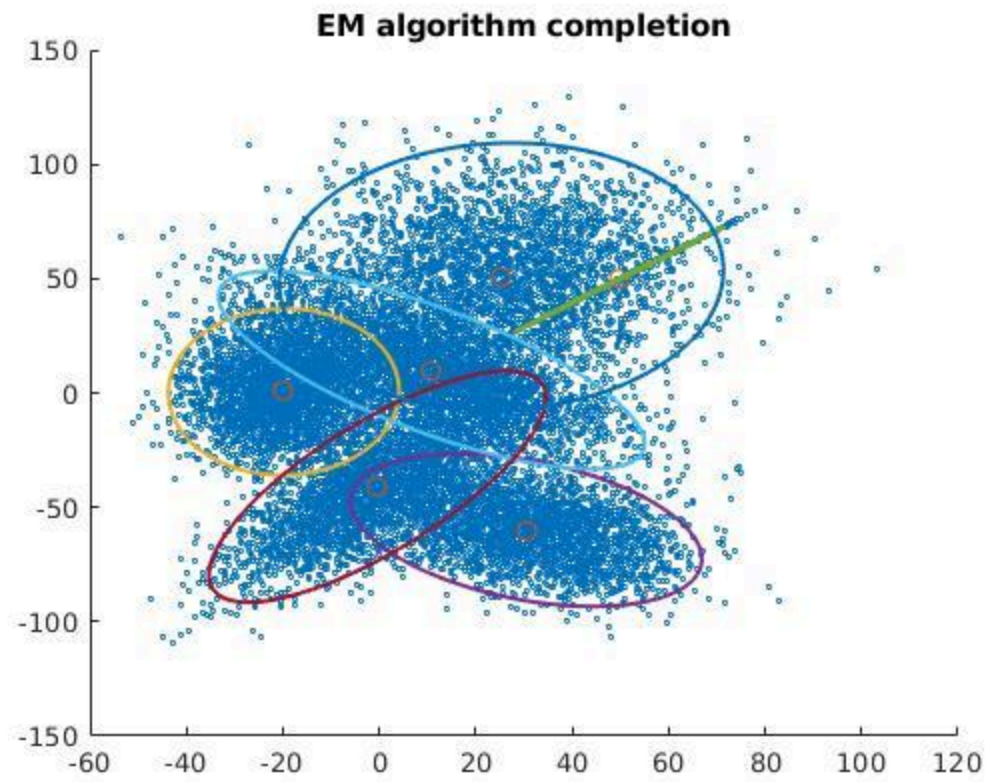


Ellipse plotting

this is a function linked for the class to use to

```
for m = 1:M
    error_ellipse(Em(:, :, m), mu(m, :))
end
title("EM algorithm completion")

%do not use
    % conditional prob. of generating xi given mth model
    %
    easyCond(i, m) =
        (((2*pi)^(d/2)) * ((det(Em(:, :, m)))^(-.5))) ...
        * exp((-0.5 * (x(i, :) - mu(m, :)) * (inv(Em(:, :, m))) * (x(i, :) -
        (mu(m, :))))));
    %using the function
```



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