

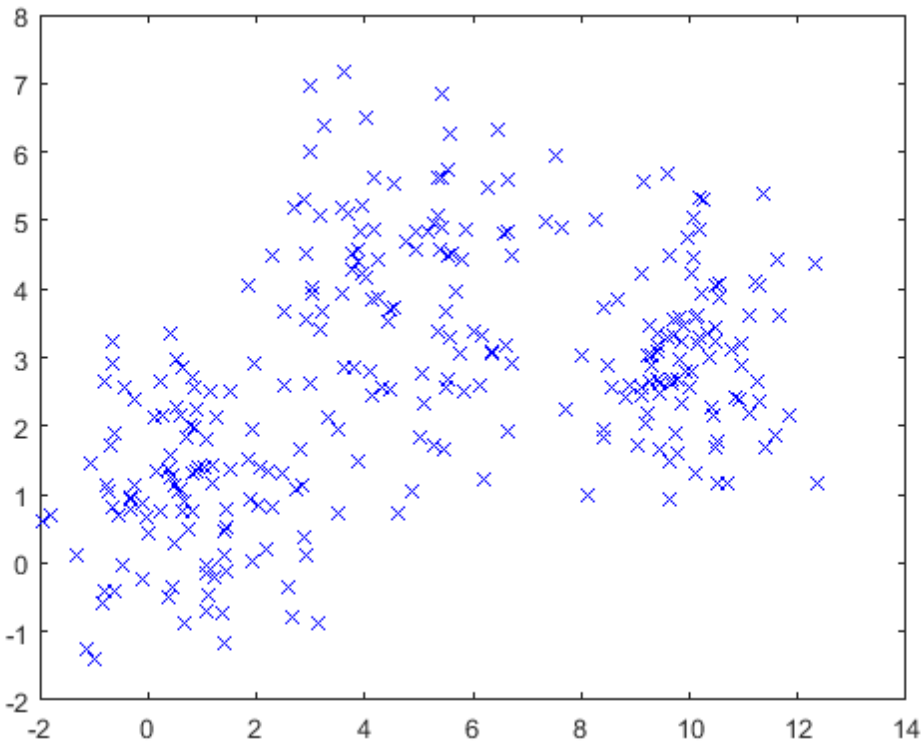
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```
close all;  
clear all;  
clc;
```

产生样本点

```
N=300;  
noise=0;  
[data]=data_gen(N,noise);  
% 对数据展示  
plot(data(:,1),data(:,2),'bx')
```



进行参数估计并比对结果

```
disp('-----原始参数值-----');  
mu1 = [10, 3]'; %数学期望  
mu2 = [1, 1]';  
mu3 = [5, 4]';  
sigma1=[1, 0; 0, 1]; %协方差矩阵  
sigma2=[1.5, 0; 0, 1.5];
```

```

sigma3=[2,0;0,2];
mu_ori=[mu1,mu2,mu3]
sigma1=[1,0;0,1];%协方差矩阵
sigma2=[3,0;0,3];
sigma3=[2,0;0,2];
sigma_ori=[sigma1,sigma2,sigma3]
w=[1,1,1]./3

disp('-----手动初始化-----');
epson =1e-10;
% 手动初始化
mu1 = [3,5]';%数学期望
mu2 = [2,1]';
mu3 = [0,3]';
mu=[mu1,mu2,mu3]
sigma1=[1,0;0,1];%协方差矩阵
sigma2=[3,0;0,3];
sigma3=[2,0;0,2];
sigma=[sigma1,sigma2,sigma3]
phi=[0.5,0.4,0.1]
disp('-----估计值-----')
[L,mu_1,sigma_1,weight_1]=EM_GMM(data,mu,sigma,phi,epson);

mu_1
sigma_1
weight_1
figure
plot(L,'R*')
title('EM算法估计GMM参数-似然函数曲线');

disp('-----完全无差别的初始值设置-----')
mu1 = [0,0]';%数学期望
mu2 = [0,0]';
mu3 = [0,0]';
mu=[mu1,mu2,mu3]
sigma1=[1,0;0,1];%协方差矩阵
sigma2=[1,0;0,1];
sigma3=[1,0;0,1];
sigma=[sigma1,sigma2,sigma3]
phi=[1/3,1/3,1/3]

disp('-----无差别初始值时的估计值-----')
[L,mu_1,sigma_1,weight_1]=EM_GMM(data,mu,sigma,phi,epson);

mu_1
sigma_1
weight_1
figure
plot(L,'R')
title('EM算法估计GMM参数-无差别初始值-似然函数曲线');

```

-----原始参数值-----

```

mu_ori =

    10     1     5
     3     1     4

```

sigma\_ori =

1	0	3	0	2	0
0	1	0	3	0	2

w =

0.3333	0.3333	0.3333
--------	--------	--------

-----手动初始化-----

mu =

3	2	0
5	1	3

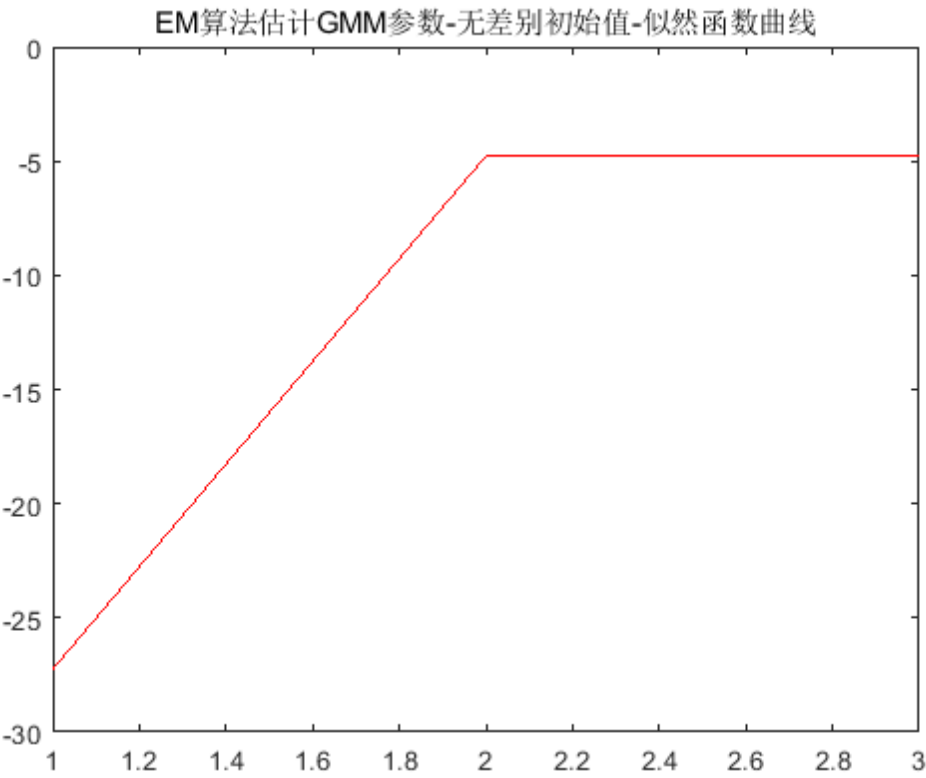
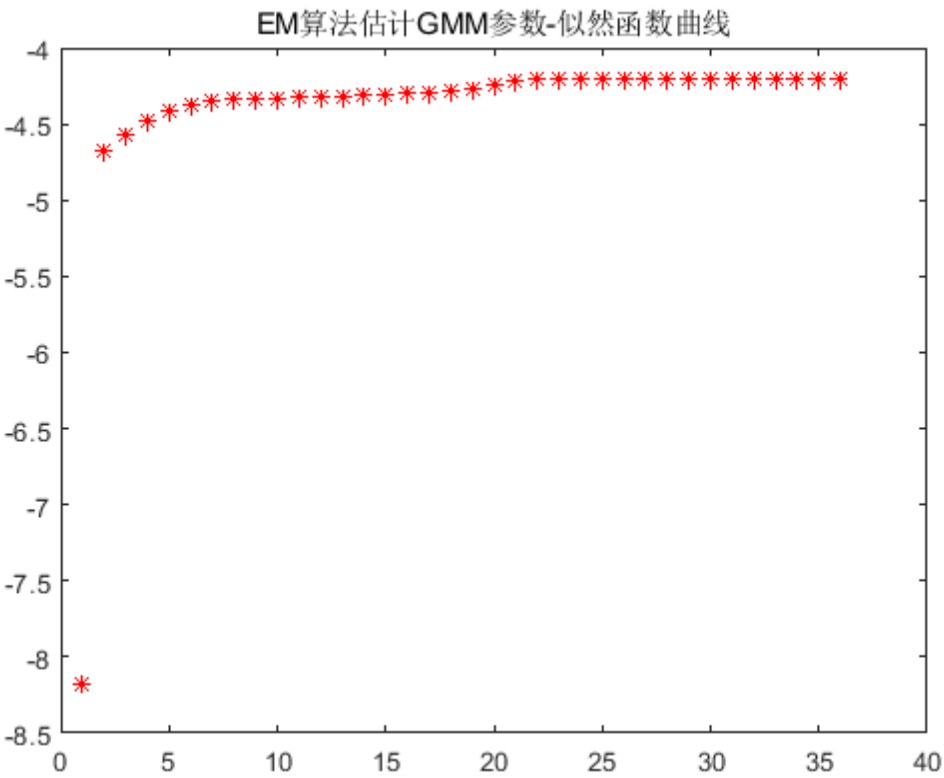
sigma =

1	0	3	0	2	0
0	1	0	3	0	2

phi =

0.5000	0.4000	0.1000
--------	--------	--------

-----估计值-----



随机初始化参数并比对结果

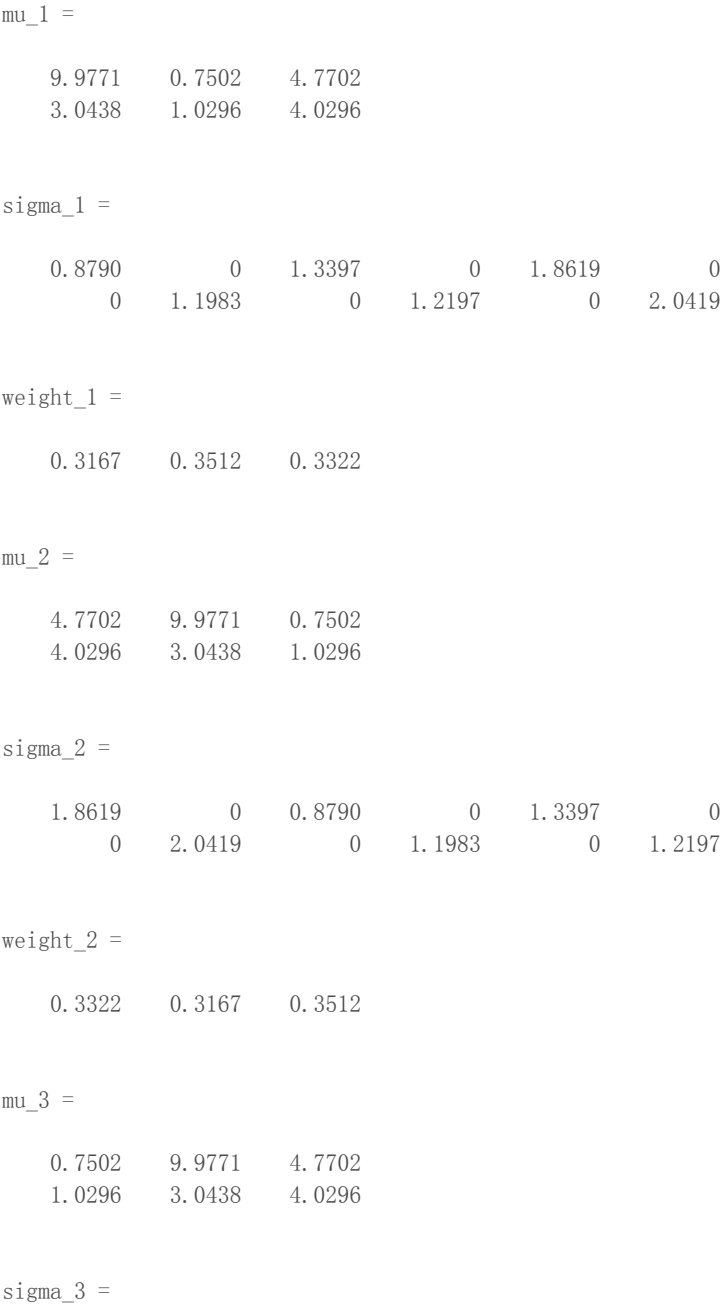
```
% 第一组
[mu, sigma, weight]=param_gen();
[Ll, mu_1, sigma_1, weight_1]=EM_GMM(data, mu, sigma, phi, epson);
mu_1
sigma_1
weight_1
% 第二组
```

```
[mu, sigma, weight]=param_gen()
[L2, mu_2, sigma_2, weight_2]=EM_GMM(data, mu, sigma, phi, epon);
mu_2
sigma_2
weight_2
% 第三组
[mu, sigma, weight]=param_gen()
[L3, mu_3, sigma_3, weight_3]=EM_GMM(data, mu, sigma, phi, epon);
mu_3
sigma_3
weight_3

% 对比不同结果
figure
plot(L1', '*');
hold on
plot(L2', 's');
plot(L3', 'x');

legend(' 第一组对数似然值L1', ' 第二组对数似然值L2', ' 第三组对数似然值L3');

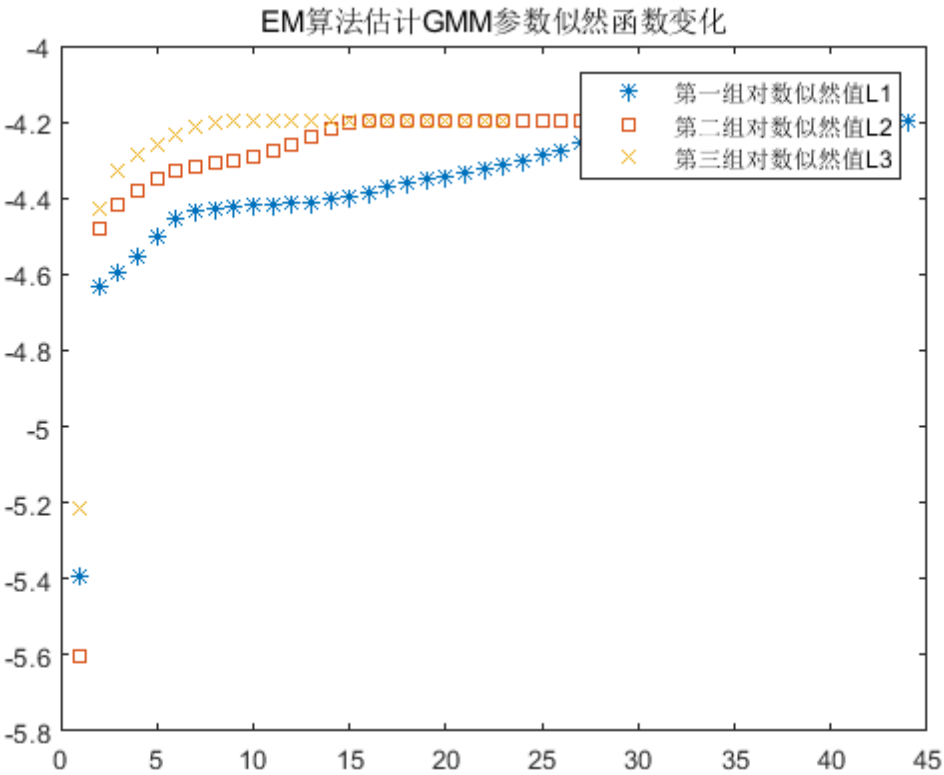
title(' EM算法估计GMM参数似然函数变化');
```



1.3397            0        0.8790            0        1.8619            0  
0        1.2197            0        1.1983            0        2.0419

weight\_3 =

0.3512        0.3167        0.3322



对数据增加噪声并分析结果

```
disp('-----噪声对估计的影响-----')
% 设置初始参数值
mu1 = [3,5]';%数学期望
mu2 = [2,1]';
mu3 = [0,3]';
mu=[mu1,mu2,mu3];
sigma=[1,0;0,1];%协方差矩阵
sigma2=[3,0;0,3];
sigma3=[2,0;0,2];
sigma=[sigma,sigma2,sigma3];
phi=[0.5,0.4,0.1];
% 原始数据组
data;
[L1,~,~,~]=EM_GMM(data,mu,sigma,phi,epson);

% 加噪声第一组
[m,n]=size(data);
noise=0.01;
ndata=noise*rand(m,n);
data=data+ndata;
[L2,~,~,~]=EM_GMM(data,mu,sigma,phi,epson);
% 加噪声第二组
noise=0.1;
ndata=noise*rand(m,n);
```

```

data=data+ndata;
[L3,~,~,~]=EM_GMM(data,mu,sigma,phi,epson);

% 加噪声第二组
noise=2;
ndata=noise*rand(m,n);
data=data+ndata;
[L4,~,~,~]=EM_GMM(data,mu,sigma,phi,epson);

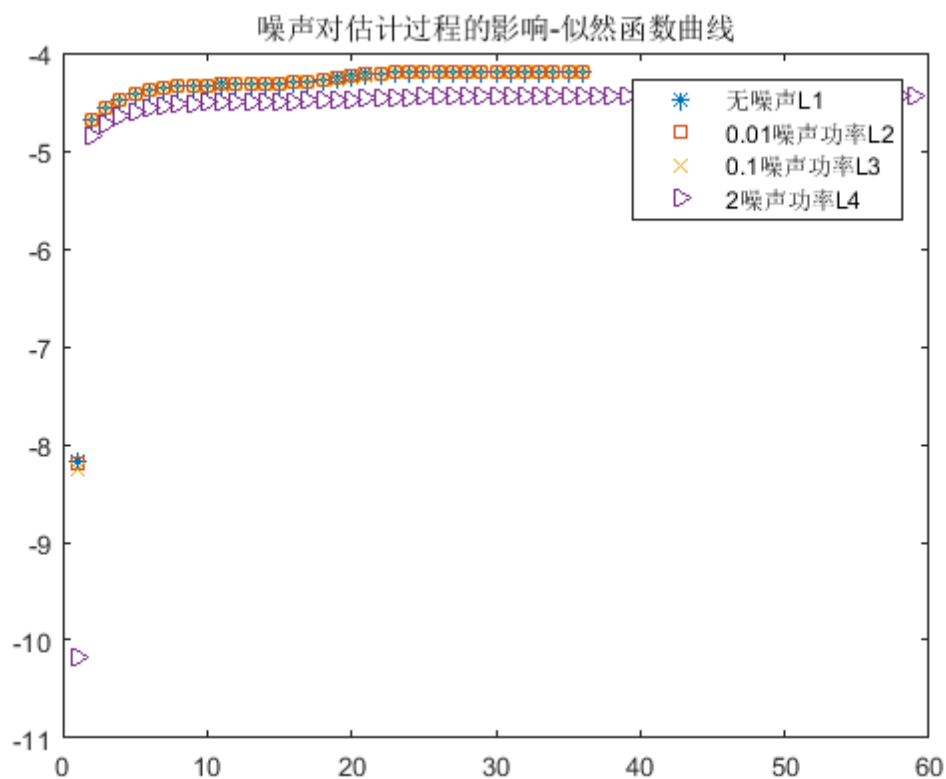
figure
plot(L1','*');
hold on
plot(L2','s');
plot(L3','x');
plot(L4','>');

legend('无噪声L1','0.01噪声功率L2','0.1噪声功率L3','2噪声功率L4');

title('噪声对估计过程的影响-似然函数曲线');

```

-----噪声对估计的影响-----



## 生成数据函数

```

function [data]=data_gen(N,noise)
mu1 = [10,3]';%数学期望
mu2 = [1,1]';
mu3 = [5,4]';
sigma1=[1,0;0,1];%协方差矩阵
sigma2=[1.5,0;0,1.5];
sigma3=[2,0;0,2];
w=[1,1,1]./3;

```

```

r1=mvnrnd(mu1, sigma1, N);
r2=mvnrnd(mu2, sigma2, N);
r3=mvnrnd(mu3, sigma3, N);
rawdata=[r1, r2, r3];
% 产生随机数
rate = floor(rand(N, 1)*3)+1;
data=zeros(N, 2);
% 取数据
    for i=1:3
        idx=find(rate==i);
        data(idx,:)=rawdata(idx, i*2-1:i*2);
    end
% 增加噪声
[m, n]=size(data);
ndata=noise*rand(m, n);
data=data+ndata;

end
```

随机初始化参数

```

function [mu, sigma, weight]=param_gen()
% 产生均值
mu=floor(rand(2, 3)*10);
%协方差矩阵
s1=ceil(rand(1)*10)*eye(2);
s2=ceil(rand(1)*10)*eye(2);
s3=ceil(rand(1)*10)*eye(2);
sigma=[s1, s2, s3];
% 权值
w_i=rand(1, 3);
weight=w_i/sum(w_i);
end
```

mu =

2	6	3
8	6	0

sigma =

10	0	7	0	3	0
0	10	0	7	0	3

weight =

0.0747	0.1950	0.7304
--------	--------	--------

mu =

2	9	4
4	1	6

sigma =

6	0	7	0	3	0
---	---	---	---	---	---



0      6      0      7      0      3

weight =

0.1966      0.3708      0.4325

## 使用EM算法对GMM参数进行估计

```
function [L, mu_s, sigma_s, weight_s]=EM_GMM(data, mu, sigma, phi, epson)
N=length(data);
T=5000;
w = zeros(N, 3);
muarr=[];
sigmarr=[];
phiarr=[];
error=100;
L(1)=1;
pos=2;

% while(true)
for y=1:1000
    % Expectation
    for k = 1 : 3
        w(:, k) = phi(k)*mvnpdf(data, mu(:, k)', sigma(:, k*2-1:k*2)); % 对于每一个样本，对参数w进行估计
    end
    % 中间穿插计算似然函数
    L(pos)=sum(log(sum(w, 2)))/N;
    err=L(pos)-L(pos-1);
    if(abs(err)<epson)
        break;
    end
    pos=pos+1;

    w = w ./ repmat(sum(w, 2), [1 3]);

    % Maximization
    for k = 1 : 3
        mu(:, k) = (w(:, k)'*data / sum(w(:, k)))';
        sigma(:, k*2-1:k*2) = w(:, k)'*((data-mu(:, k))'*(data-mu(:, k))) / sum(w(:, k));
        temp= kron(w(:, k), ((data-mu(:, k))'*(data-mu(:, k)))) / sum(w(:, k)); % 对方差求解
        % 有一个矩阵求和过程
        sigma(:, k*2-1:k*2)= sqrt(reshape(sum(reshape(temp', 4, N), 2), 2, 2));

    % 直接构造矩阵
        cov_mat=diag(w(:, k)'*((data-mu(:, k))'*(data-mu(:, k))))/sum(w(:, k));
        sigma(:, k*2-1:k*2)=cov_mat;
        phi(k) = sum(w(:, k)) / N;
    end
    muarr = [muarr; reshape(mu, 1, 6)];
    sigmarr= [sigmarr; reshape(sigma, 1, 12)];
    phiarr= [phiarr; phi];
    y=y+1;
end

mu_s=mu;
sigma_s=sigma;
weight_s=phi;
L=L(2:end);
end
```

mu\_1 =

4.7702	9.9771	0.7502
4.0296	3.0438	1.0296

sigma\_1 =

1.8619	0	0.8790	0	1.3397	0
0	2.0419	0	1.1983	0	1.2197

weight\_1 =

0.3322	0.3167	0.3512
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-----完全无差别的初始值设置-----

mu =

0	0	0
0	0	0

sigma =

1	0	1	0	1	0
0	1	0	1	0	1

phi =

0.3333	0.3333	0.3333
--------	--------	--------

-----无差别初始值时的估计值-----

mu\_1 =

5.0073	5.0073	5.0073
2.6639	2.6639	2.6639

sigma\_1 =

15.5717	0	15.5717	0	15.5717	0
0	3.0892	0	3.0892	0	3.0892

weight\_1 =

0.3333	0.3333	0.3333
--------	--------	--------