

Blind Source Separation

HW6-Section-3

Mohammadreza Arani : 810100511

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

Part-1:

بخش سوم)

در این تمرین می خواهیم دو روش MOD و K-SVD را برای جداسازی کور منابع اسپارس یا به عبارت دیگر یادگیری دیکشنری برای نمایش اسپارس سیگنال ها پیاده سازی کنیم.

ماتریس دیکشنری با ابعاد 10×40 ، ماتریس منابع با ابعاد 40×1500 و ماتریس مشاهدات با ابعاد 10×1500 در یک فایل با نام hw6-part3.mat قرار داده شده است.

۱- معیار mutual coherence را محاسبه و گزارش کنید.

```
Data_Q3 = load("hw6-part3.mat");  
D_Part_1 = Data_Q3.D;  
% Calculation of MU based on implemented function in previous Section:  
[MU_out, MU_Index_Out] = MU_Calc_Mine(D_Part_1);  
disp("Mutual Coherence is : "+ MU_out);
```

Mutual Coherence is : 0.86803

```
disp(" Maximum Coherence happens for column "+MU_Index_Out(1)+" and column "+MU_Index_Out(2))
```

Maximum Coherence happens for column 13 and column 38

Part-2:

ماتریس مشاهدات را داریم، روش های MOD و K-SVD را روی ماتریس مشاهدات اعمال کنید و برای دیکشنری \hat{D} و تخمین منابع \hat{S} را به دست آورید. شایان ذکر است sparsity level برابر $N_0=2$ است و د از یک الگوریتم sparse recovery استفاده کنید از OMP استفاده کنید.

MOD

```
X_part_1 = Data_Q3.X;
S_part_1 = Data_Q3.S;
N0_Part_1 = Data_Q3.N0;

S_X_p1 = size(X_part_1);
S_S_p1 = size(S_part_1);

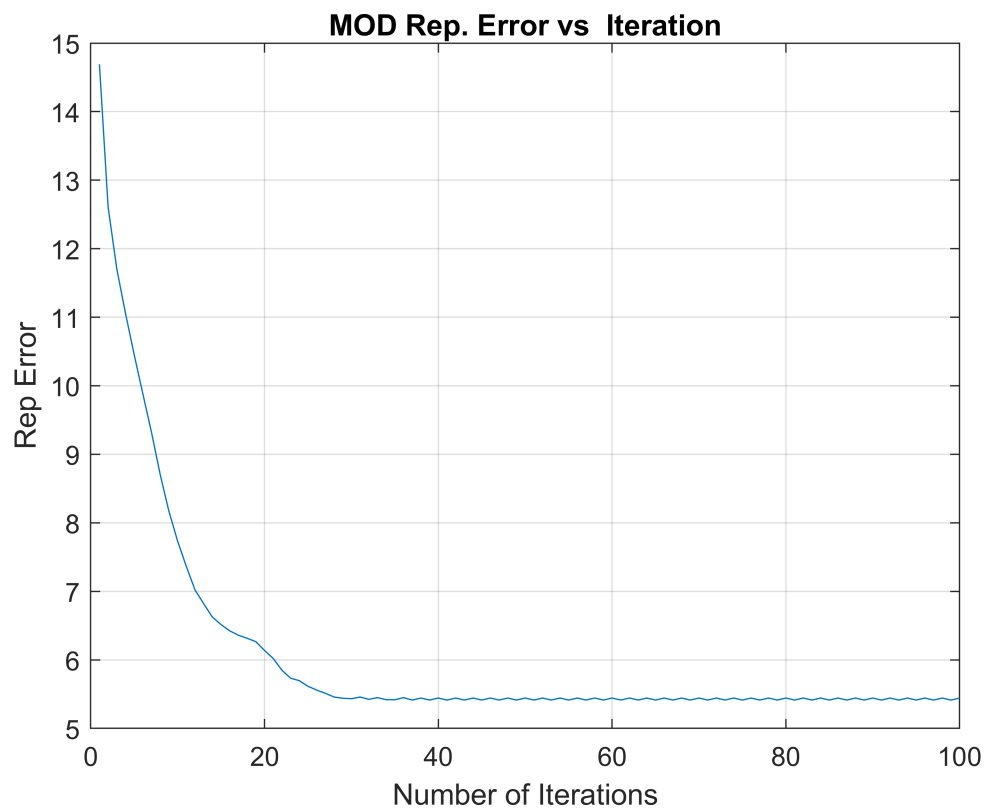
D_hat_MOD = randn(S_X_p1(1) , S_S_p1(1)) ;
D_hat_MOD = D_hat_MOD ./ sqrt(sum(D_hat_MOD.^2,1));

T = S_S_p1(2);
S_hat_MOD = randn(S_S_p1);

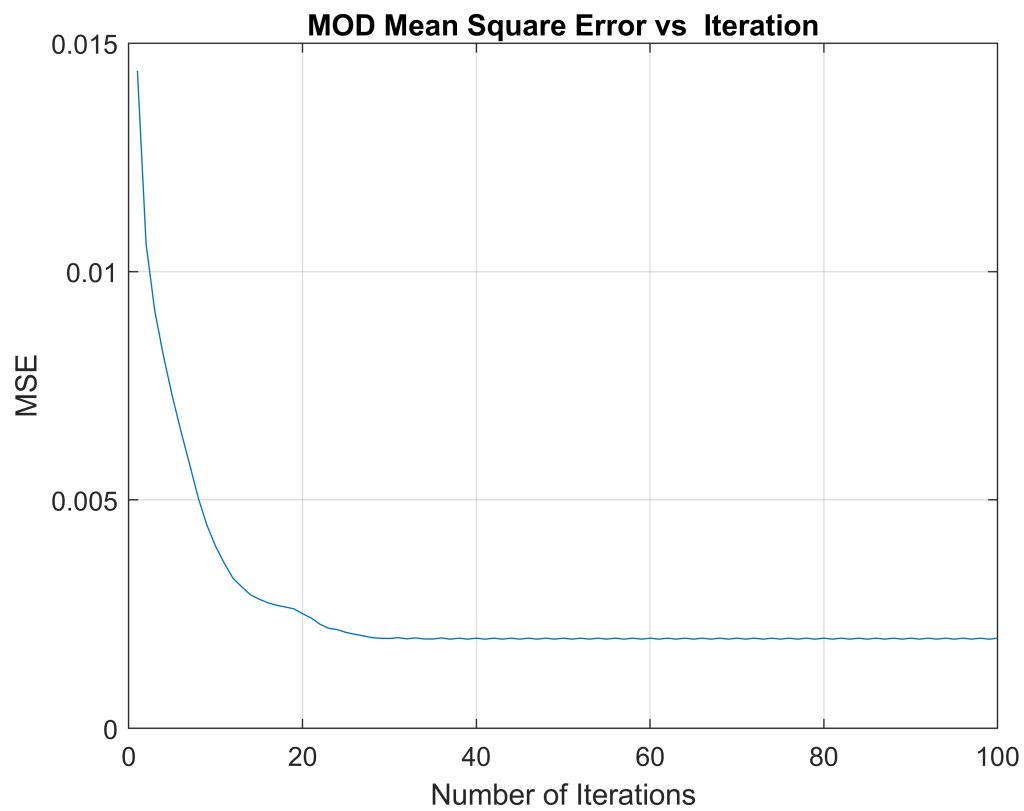
IterMax = 100;
MSE_MOD = zeros(1,IterMax);
Rep_MOD = MSE_MOD;
MOD_delta_Time = MSE_MOD;

for i=1:IterMax
    tic;
    % First Step of Alternation Minimization:
    for t=1:T
        S_hat_MOD(:,t) = OMP_Mine_CALC( X_part_1(:,t) , D_hat_MOD , N0_Part_1);
    end
    % Second Step of Alternation Minimization:
    D_hat_MOD = X_part_1*pinv(S_hat_MOD); % Second Step of Altrnation Minimization
    D_hat_MOD = D_hat_MOD ./ sqrt(sum(abs(D_hat_MOD).^2)); % Normalize Each Column of D
    % Calculate the MSE:
    MSE_MOD(i) = mse(X_part_1 , D_hat_MOD*S_hat_MOD);
    Rep_MOD(i) = norm(X_part_1 - D_hat_MOD*S_hat_MOD,"fro");
    MOD_delta_Time(i) = toc;
end

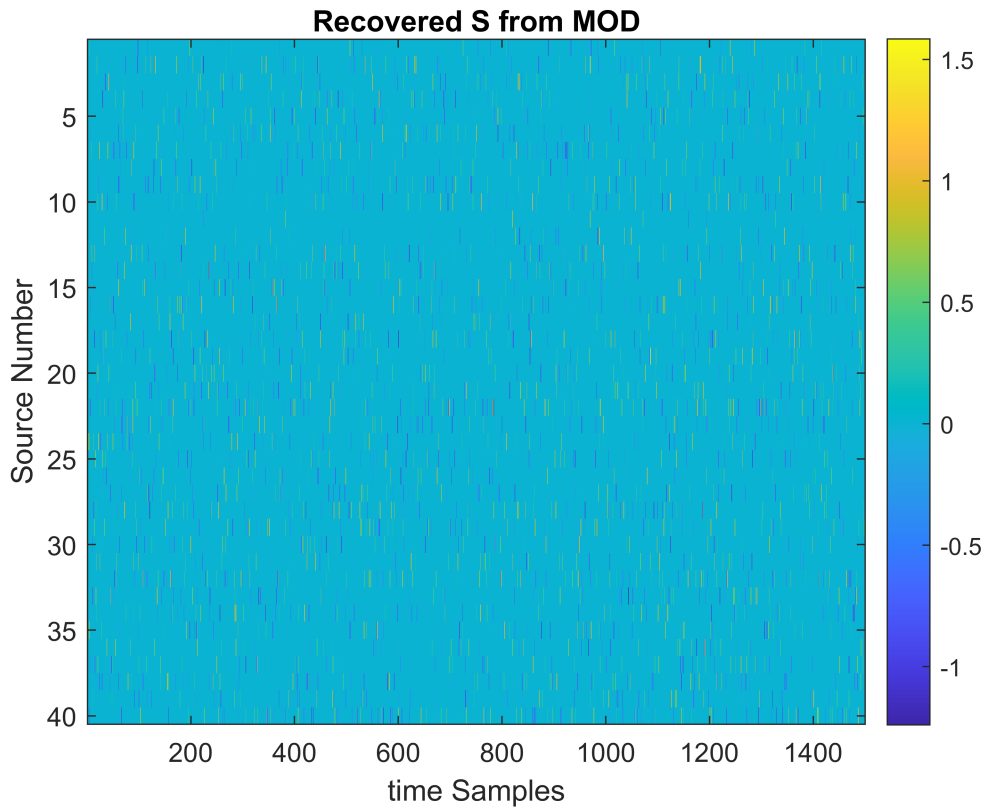
figure()
plot(Rep_MOD)
grid on
xlabel("Number of Iterations")
ylabel("Rep Error")
title("MOD Rep. Error vs Iteration")
```



```
figure()
plot(MSE_MOD)
grid on
xlabel("Number of Iterations")
ylabel("MSE")
title("MOD Mean Square Error vs Iteration")
```



```
figure()  
imagesc(S_hat_MOD)  
title("Recovered S from MOD")  
xlabel("time Samples")  
ylabel("Source Number")  
colorbar
```



K-SVD:

```

D_hat_K_SVD = randn(S_X_p1(1) , S_S_p1(1)) ;
T = S_S_p1(2);
S_hat_K_SVD = randn(S_S_p1);

IterMax = 100;
MSE_K_SVD = zeros(1,IterMax);
Rep_K_SVD = MSE_K_SVD;
K_SVD_delta_Time = MSE_K_SVD;

for i=1:IterMax
    tic;
    % First Step of Alternation Minimization:
    for t=1:T
        S_hat_K_SVD(:,t) = OMP_Mine_CAlc( X_part_1(:,t) , D_hat_K_SVD , N0_Part_1);
    end
    % Second Step of Alternation Minimization:

```

```

for n=1:S_S_p1(1)

    % Choose corresponding Index:
    D_ind_SVD = 1:S_S_p1(1);
    D_ind_SVD(n) = [];

    % Calculate the Residual X:
    X_res = X_part_1 - D_hat_K_SVD(:, D_ind_SVD)*S_hat_K_SVD(D_ind_SVD,:) ;
    [U,S,V] = svd(X_res);

    % Sort:
    [S_sorted, idx] = sort(diag(S), 'descend');

    % Reorder the columns of U and rows of V based on the sorted indices
    U_sorted = U(:, idx);
    V_sorted = V(:, idx);

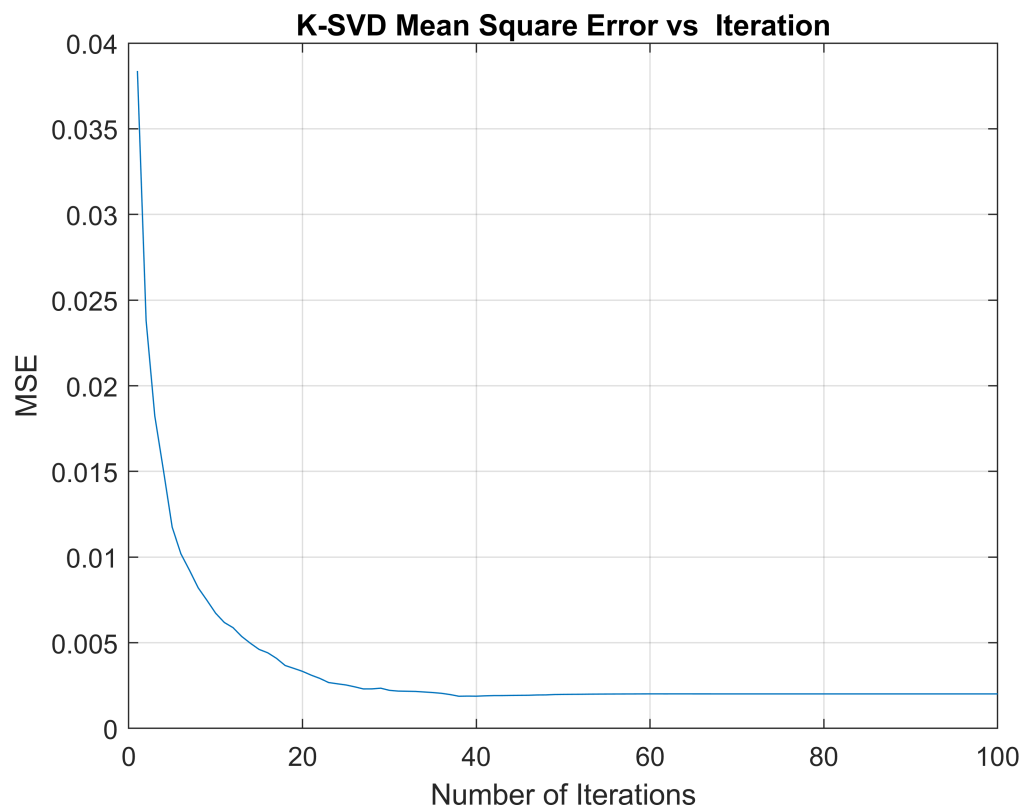
    % Assign U and V to D and S:
    D_hat_K_SVD(:,n) = U_sorted(:,1);
    ZeroIndex = find(abs(S_hat_K_SVD(n,:))<1e-6);
    S_hat_K_SVD(n,:) = S_sorted(1)*V_sorted(:,1);
    S_hat_K_SVD(n,ZeroIndex) = 0;
end
% Calculate the MSE
MSE_K_SVD(i) = mse(X_part_1 , D_hat_K_SVD*S_hat_K_SVD);
Rep_K_SVD(i) = norm(X_part_1 - D_hat_K_SVD*S_hat_K_SVD,"fro" );
K_SVD_delta_Time(i) = toc;
end

```

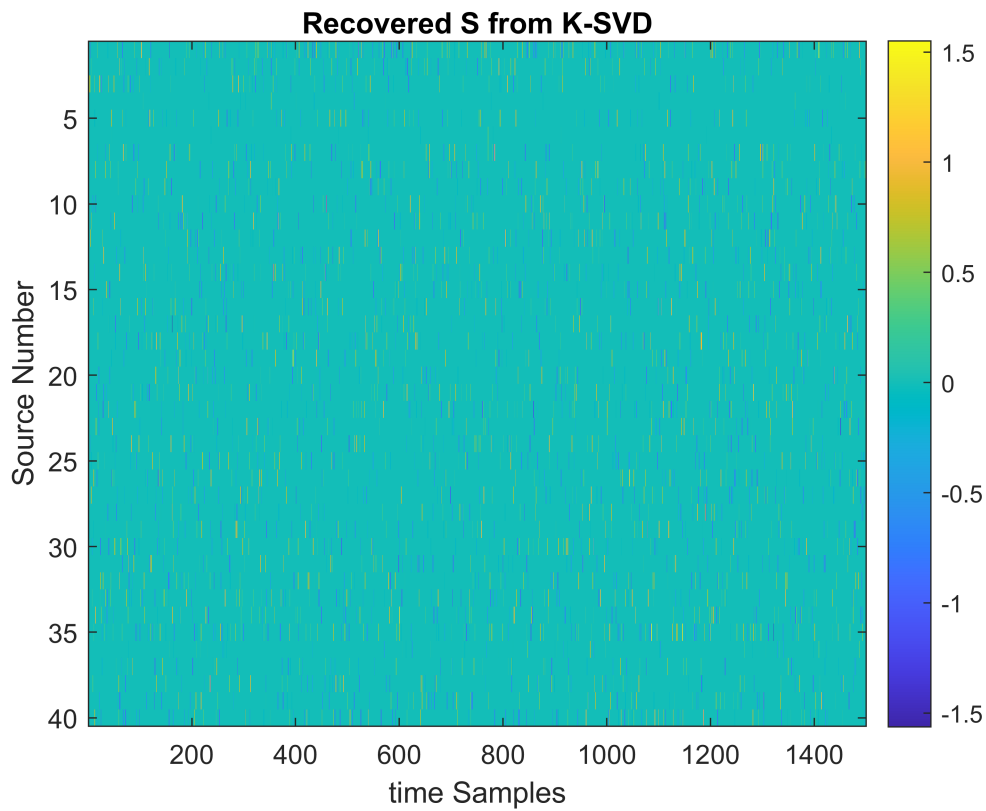
```

figure()
plot(MSE_K_SVD)
grid on
xlabel("Number of Iterations")
ylabel("MSE")
title("K-SVD Mean Square Error vs Iteration")

```

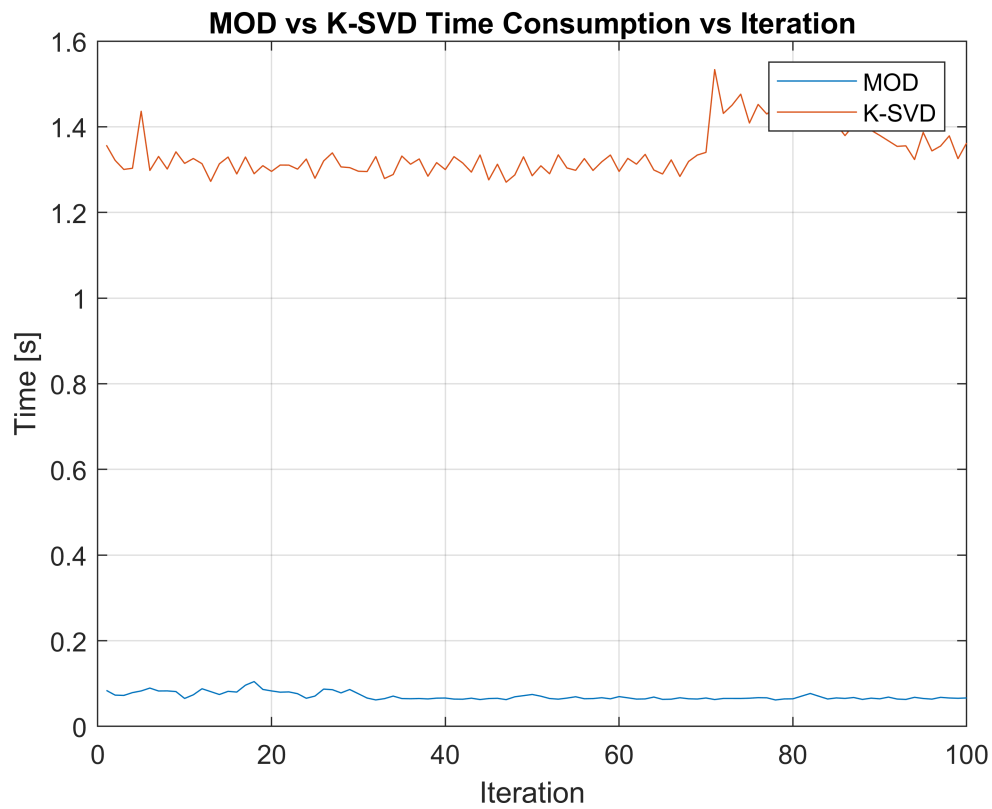


```
figure()  
imagesc(S_hat_K_SVD)  
title("Recovered S from K-SVD")  
xlabel("time Samples")  
ylabel("Source Number")  
colorbar
```

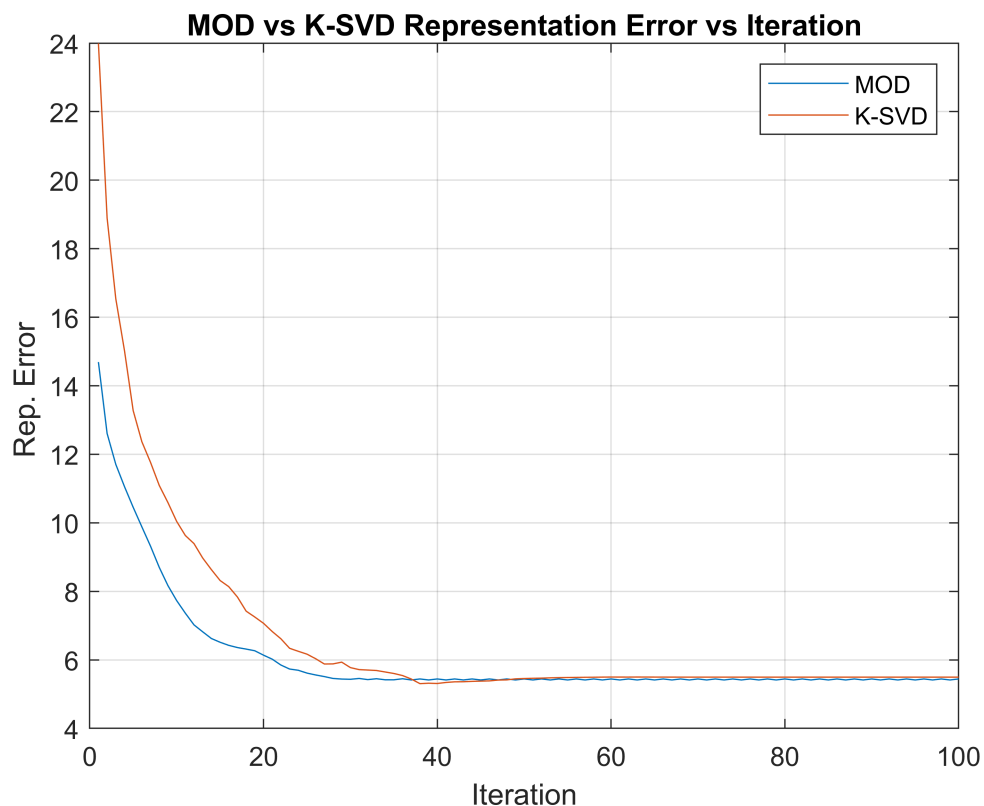


Comparison of MOD and K-SVD

```
figure()
plot(MOD_delta_Time)
hold on
plot(K_SVD_delta_Time)
title("MOD vs K-SVD Time Consumption vs Iteration")
grid on
xlabel("Iteration")
ylabel("Time [s]")
legend("MOD", "K-SVD")
hold off
```

```
figure()
plot(Rep_MOD)
hold on
plot(Rep_K_SVD)
hold off
title("MOD vs K-SVD Representation Error vs Iteration")
grid on
xlabel("Iteration")
ylabel("Rep. Error ")
legend("MOD", "K-SVD")
```



Part-5:

۵- برای مقایسه ی کیفیت یادگیری دیکشنری، برای هر روش به صورت جداگانه، به ترتیب از ستون (اتم) اول تا ستون (اتم) آخر \hat{D} ، قدر مطلق correlation هر اتم \hat{D} را با همه ی اتم های D محاسبه کنید. در صورتی که ماکزیمم این مقادیر از 0.98 بیشتر بود، فرض کنید اتمی از D که ماکزیمم correlation را داده است به درستی recover شده است. توجه داشته باشید که برای تکرار این روش و محاسبات بعدی، اتمی از D که recover شده است را حذف کنید. برای هر روش، چند درصد از ستون های ماتریس D را به درستی Recover کردید؟ به این درصد successful recovery rate می گوییم.

```
% [index_remove_Recovered_SVD,counter_SVD] = Recovery_Calc_D(D_hat_K_SVD , D_Part_1 )
[D_hat_K_SVD,Succ_Rec_Num_K_SVD] = Recovery_Calc_D(D_hat_K_SVD , D_Part_1 )
```

D_hat_K_SVD = 10×40

0.1782	0.0616	0.0067	0.1625	0.3639	0.4327	0.0122	0.1649 ...
-0.6215	-0.1760	-0.3174	0.0084	-0.2324	-0.1279	-0.4431	0.0825
0.2242	-0.0384	-0.3175	0.2171	-0.3800	0.1295	0.3095	0.3399
0.2124	-0.1355	-0.5826	0.5912	0.0028	0.3417	0.2711	-0.0910
-0.2676	-0.1822	-0.1205	-0.3646	0.2652	0.3546	-0.1804	-0.4486
0.3716	-0.0659	-0.3122	-0.1084	0.6184	0.0031	0.0841	0.0360
-0.3596	-0.4665	-0.1029	-0.1599	0.0002	0.1226	-0.2405	0.2894
0.2836	-0.1813	-0.0590	-0.0752	0.1848	0.1050	-0.2951	0.2645
-0.1171	0.7003	-0.5078	0.1144	-0.3329	-0.2997	-0.6311	-0.4214
0.2314	-0.4085	-0.2739	0.6228	-0.2672	0.6490	0.2267	0.5553

Succ_Rec_Num_K_SVD = 35

```
disp(Succ_Rec_Num_K_SVD/size(D_Part_1,2)*100+" % of the Columns of D where recovered successfully")
```

87.5 % of the Columns of D where recovered successfully!

```
[D_hat_MOD,Succ_Rec_Num_MOD] = Recovery_Calc_D(D_hat_MOD , D_Part_1 )
```

```
D_hat_MOD = 10x40
-0.1758 -0.0923 0.0390 0.2260 -0.3966 -0.4335 0.3994 -0.1268 ...
0.6312 0.2277 -0.3149 0.1461 0.2639 0.1159 0.3185 -0.0980
-0.2250 -0.0150 -0.3193 0.1594 0.3539 -0.1148 -0.1668 -0.3396
-0.1637 0.1880 -0.5239 0.0376 -0.0423 -0.3057 -0.0217 0.1679
0.2384 0.2389 -0.1397 0.0554 -0.2338 -0.4027 0.3025 0.4656
-0.4267 0.0726 -0.3082 0.1902 -0.6326 -0.0306 0.2271 -0.0517
0.3760 0.4176 -0.1242 -0.5923 0.0434 -0.1187 -0.4292 -0.3287
-0.2806 0.1437 -0.0507 0.0940 -0.2106 -0.1346 -0.0215 -0.2818
0.1273 -0.7189 -0.5549 0.2522 0.3271 0.3057 -0.1286 0.4117
-0.1343 0.3604 -0.2871 0.6620 0.1946 -0.6350 0.6054 -0.5044
Succ_Rec_Num_MOD = 34
```

```
% [index_remove_Recovered_MOD,counter_MOD] = Recovery_Calc_D(D_hat_MOD, D_Part_1 )
```

```
disp(Succ_Rec_Num_MOD/size(D_Part_1,2)*100+" % of the Columns of D where recovered successfully")
```

85 % of the Columns of D where recovered successfully!

Part-6:

۶-ابهام ترتیب و scale منابع را برطرف کرده و سپس عبارت زیر را برای هر دو روش گزارش کنید.

$$E = \frac{\|\hat{S} - S\|_F^2}{\|S\|_F^2}$$

```
S_hat_K_SVD_OMP = S_hat_K_SVD;
S_hat_MOD_OMP = S_hat_MOD;
for t=1:T
    S_hat_K_SVD_OMP(:,t) = OMP_Mine_Calc( X_part_1(:,t) , D_hat_K_SVD , N0_Part_1);
    S_hat_MOD_OMP(:,t) = OMP_Mine_Calc( X_part_1(:,t) , D_hat_MOD , N0_Part_1);
end
[Erroes_S_Hat_K_SVD,S_Hat_Chosen_K_SVD] = Perm_AMP_Disamb(S_hat_K_SVD_OMP,S_part_1);

disp("Error of S_K_SVD equals to: "+ min(Erroes_S_Hat_K_SVD));
```

Error of S_K_SVD equals to: 1.4798

```
[Erroes_S_Hat_MOD,S_Hat_Chosen_MOD] = Perm_AMP_Disamb(S_hat_MOD_OMP,S_part_1);
disp("Error of S_MOD equals to: "+ min(Erroes_S_Hat_MOD));
```

Error of S_MOD equals to: 1.8142

```
function [Erroes_B,B_Hat_Chosen] = Perm_AMP_Disamb(B_Hat,S_Amp) %% Perm_AMP_Disamb
Perm_1 = B_Hat;
Perm_2 = -B_Hat;
Perm_3 = B_Hat([end, 1:end-1],:);
Perm_4 = -B_Hat([end, 1:end-1],:);
Perms = { Perm_1,Perm_2,Perm_3,Perm_4 };
% Estimate S:
S_hat_1 = Perm_1;
S_hat_1_Normalised = S_hat_1/norm(S_hat_1,'fro');
S_hat_2 = Perm_2;
S_hat_2_Normalised = S_hat_2/norm(S_hat_2,'fro');
S_hat_3 = Perm_3;
S_hat_3_Normalised = S_hat_3/norm(S_hat_3,'fro');
S_hat_4 = Perm_4;
S_hat_4_Normalised = S_hat_4/norm(S_hat_4,'fro');
% Calc Error:
S_Amp_Normalized = S_Amp/norm(S_Amp,'fro');

Error_1 = (norm((S_hat_1_Normalised)-(S_Amp_Normalized),'fro'))^2/norm((S_Amp_Normalized),'fro');
Error_2 = (norm((S_hat_2_Normalised)-(S_Amp_Normalized),'fro'))^2/norm((S_Amp_Normalized),'fro');
Error_3 = (norm((S_hat_3_Normalised)-(S_Amp_Normalized),'fro'))^2/norm((S_Amp_Normalized),'fro');
Error_4 = (norm((S_hat_4_Normalised)-(S_Amp_Normalized),'fro'))^2/norm((S_Amp_Normalized),'fro');

Erroes_B = [Error_1,Error_2,Error_3,Error_4];
[~, index] = min(Eroes_B);
B_Hat_Chosen = Perms{index};
end

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%

function [D_Hat_sorted,Counter] = Recovery_Calc_D(D_hat , D )
%index_remove_Recovered

% D_hat = D_KSVD;
D_Hat_sorted = D_hat;
Counter = 0;
for i=1:size(D,2)
    for j = 1 : size(D_hat,2)
        if(abs(D(:,i)'*D_hat(:,j)) > 0.98)
            Counter = Counter + 1;
            D_Hat_sorted(:,i) = D_hat(:,j);
            D_hat(:,j)=[];
            break;
        end
    end
end
end
```

```
Perm_1 = B_Hat;
```

```
Perm_3 = B_Hat([end, 1:end-1],:);
```

```
Perms = { Perm_1,Perm_2,Perm_3,Perm_4 };
```

% Estimate S:

```
S_hat_1_Normalised = S_hat_1/norm(S_hat_1,'fro');
```

```
S_hat_2 = Perm_2;
```

```
S_hat_3 = Perm_3;
```

```
S_hat_4 = Perm_4;
```

% Calc Error:

```
Error_1 = (norm((S_hat_1_Normalised)-(S_Amp_Normalized),'fro'))^2/norm((S_Amp_Normalized),'fro')
```

```
Error_2 = (norm((S_hat_2_Normalised)-(S_Amp_Normalized), 'fro'))^2/norm((S_Amp_Normalized), "fro"
```

```
Error_3 = (norm((S_hat_3_Normalised)-(S_Amp_Normalized),'fro'))^2/norm((S_Amp_Normalized),'fro')
```

```
Error_4 = (norm((S_hat_4_Normalised)-(S_Amp_Normalized),'fro'))^2/norm((S_Amp_Normalized),'fro')
```

```
Erroes_B = [Error_1,Error_2,Error_3,Error_4];
```

```
[~, index] = min(Erroes_B);
```

```
B_Hat_Chosen = Perms{index};
```

end

[illegible]

```
function [D_Hat_sorted,Counter] = Recovery_Calc_D(D_hat , D )
```

```
%index_remove_Recovered
```

```
% D_hat = D_KSVD;
```

```
D_Hat_sorted = D_hat;
```

```
Counter = 0;
```

```
for i=1:size(D,2)
```

```
for j = 1 : size(D_hat,2)
```

```
if(abs(D(:,i)'*D_hat(:,j)) > 0.98)
```

```
Counter = Counter + 1;
```

```
D_Hat_sorted(:,i) = D_hat(:,j);
```

```
D_hat(:,j)=[ ];
```

```
break;
```

end

end

end

```

%     [~, Col_D] = size(D_hat);
%     Temp_D_check = D;
%
%     index_remove_Recovered = [];
%     counter = 0;
%     for i=1:Col_D
%         Cor_Col_D = D_hat(:,i)'*Temp_D_check ;
%         %max(abs(Cor_Col_D))
%         if(max(abs(Cor_Col_D)) > 0.98) % Check to see whether we have recovered this column
%             [~, index_remove_Recovered(counter+1)] = max(Cor_Col_D);
%             Temp_D_check(index_remove_Recovered) = [];
%             counter = counter+1;
%         end
%     end
%
%     end

end

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%

function [MU_out, MU_Index_Out] = MU_Calc_Mine(D)

Temp      = (ones(size(D'*D)) - eye(size(D'*D))).* abs( D'*D );
[MU_Vals , MU_Index] = max(Temp) ;
[MU_out , MU_Index2] = max(MU_Vals);

MU_Index_Out = [MU_Index2 , MU_Index(MU_Index2)];
end

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%

function s_OMP = OMP_Mine_CALC( x , D , N0)

xr = x;
Dr = D;
idx_OMP = zeros(1, N0);
for i = 1 : N0
    [~, idx_OMP(i)] = max(abs(xr'*Dr),[],'omitnan');
    xr = xr - x'*Dr(:, idx_OMP(i))*Dr(:,idx_OMP(i));
    if i > 1
        xr = xr - D(:,idx_OMP(1:i))*(pinv(D(:,idx_OMP(1:i)))*xr);
    end
    Dr(:,idx_OMP(i)) = nan;
end

s_OMP = zeros(size(D, 2), 1);
s_OMP(idx_OMP) = pinv(D(:, idx_OMP))*x;

```

```

%
% xr = X;
% [~, Col_D] = size(D);
% Chosen_IDX = inf+zeros(1,Col_D);
% S_hat = zeros(Col_D,1);
% for i=1:N0
%
%     B = D;
%
%     if(i>1)
%         B(:,Chosen_IDX(1:i-1)) = []; % Removing previously chosen Indices!
%     end
%
%     Corr_matrix = abs(repmat(xr,1,Col_D-i+1).*B);
%     Corr_sum = sum(Corr_matrix,1); % Summation for each Column --> a Row Vector
%     [Value,Idx] = max(Corr_sum); % Choosing Best Fitted
%
%     if(sum(Idx>Chosen_IDX)>0)
%         Idx = Idx + sum(Idx>Chosen_IDX) ; % Update the Index with respect to the Original
%     end
%
%     Chosen_IDX(i) = Idx;
%     xr = xr - X'*D(:,Idx)*D(:,Idx); %  $xr = x - \langle x, di \rangle di$ 
%
%     % Update Coefficients:
%     if(i>1)
%         D_sub_omp = D(:,Chosen_IDX(1:i)) ;
%         xr = xr - D(:,Chosen_IDX(1:i))*(pinv(D_sub_omp)*xr);
%     end
%
%     % Stop Criteria:
%     %% if(norm(x - D(:,Chosen_IDX_3(1:i)),2)<thresh )
%     %%     break;
%     %% end
%     S_hat(Chosen_IDX(i)) = pinv(D(:,i))*X;
% end

% delta_t_OMP = toc;
% disp("Elapsed Time (OMP): "+ delta_t_OMP+"(s)");
end

```