

SPCOM homework 3

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In this report, we will show our results in channel and symbol estimation.

I. PILOT-BASED CHANNEL ESTIMATION

We first decompose \mathbf{X} using this equation:

$$\mathbf{X} = \mathbf{H}\mathbf{S}_L + \mathbf{N} \quad (1)$$

After that, based on the information of \mathbf{S}_L , which is constructed by data sequence \mathbf{s} and signal \mathbf{X} from antennas, we can estimate the channel by pseudo inverse operation. The results are shown in Fig 2. As we can see, both the real and

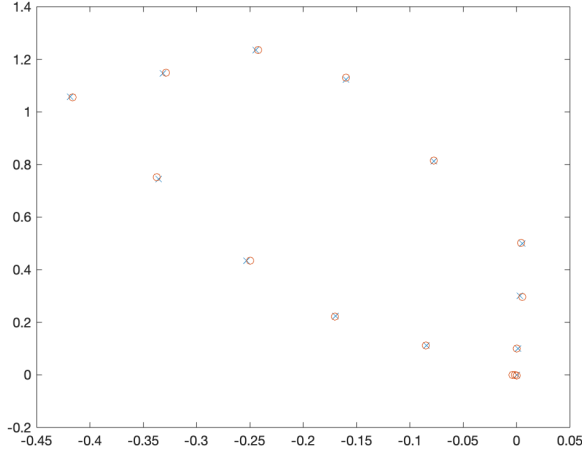


Fig. 1: True channel (x) versus Estimated channel (o)

imaginary part of the channel estimation are close to the true one.

II. BLIND CHANNEL ESTIMATION

In order to estimate the channel \mathbf{h} from the column null-space of \mathbf{X} , we can first establish this noiseless data model:

$$\mathbf{X} = \mathbf{H}\mathbf{S}_L \quad (2)$$

And the singular value decomposition of \mathbf{X} is:

$$\mathbf{X} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^H \quad (3)$$

where \mathbf{U} can be written as $[\mathbf{U}_s \mathbf{U}_n]$ and \mathbf{U}_n is the columns of \mathbf{U} corresponding to the 0 singular values. Therefore, we have:

$$\mathbf{U}_n^H \mathbf{X} = \mathbf{0} \quad (4)$$

Since \mathbf{H} and \mathbf{X} have the same column space, then we can rewrite Equation 4 as:

$$\mathbf{U}_n^H \mathbf{H} = \mathbf{0} \quad (5)$$

After this, we will look into the structure of \mathbf{H} matrix. $\mathbf{H} = [\mathbf{h}_0 \mathbf{h}_1 \dots \mathbf{h}_{L-1}]$. We can make all of the \mathbf{h} vectors into one column and correspondingly re-arrange the Equation 5, which is shown in Equation 6.

$$\begin{bmatrix} \mathbf{U}_n^H & 0 & 0 \\ 0 & \mathbf{U}_n^H & 0 \\ 0 & 0 & \mathbf{U}_n^H \end{bmatrix} \begin{bmatrix} \mathbf{h}_0 \\ \mathbf{h}_1 \\ \mathbf{h}_2 \end{bmatrix} = \mathbf{0} \quad (6)$$

We denote the first matrix is $\mathbf{U}_{n,T}$. Therefore, the re-arranged \mathbf{h} vector is in the null space of the $\mathbf{U}_{n,T}$. Then we can do singular value decomposition of this matrix and find the null-space.

Here, the basis of the null space has 9 vectors, thus there are infinity solutions of \mathbf{h} , we cannot estimate channel using blind estimation.

III. BLIND SYMBOL ESTIMATION

Here we can use our data model again shown in Equation 2. But now we focus on matrix \mathbf{S}_L .

We can first do singular value decomposition of \mathbf{X} , after that, we divide \mathbf{V} into two parts instead of \mathbf{U} . The \mathbf{X} has rank = 3, and the division is $[\mathbf{V}_s \mathbf{V}_n]$ where \mathbf{V}_s corresponds to the non-zero singular values. So we will have:

$$\mathbf{X}\mathbf{V}_n = \mathbf{0} \quad (7)$$

Since \mathbf{X} and \mathbf{S}_L have the same row space, thus we have:

$$\mathbf{S}_L \mathbf{V}_n = \mathbf{0} \quad (8)$$

By investigating the structure of \mathbf{S}_L , in our case we can re-arrange Equation 8 into the following one: We do singular

$$[\zeta_{-2} \ \zeta_{-1} \ \zeta_0 \ \dots \ \zeta_{N-1}] \cdot \begin{bmatrix} \boxed{V_n} & \boxed{\begin{matrix} \sigma & \sigma \\ 0 & \sigma \end{matrix}} & \boxed{\begin{matrix} \sigma \\ \sigma \end{matrix}} \end{bmatrix} = \mathbf{0}$$

value decomposition of the second matrix. We found that there is a singular value is equal to 0 (actually it's not 0 but a value that smaller than any other singular values). Then we just pick up this column as the estimation of symbols.

However, there is still a scalar between the estimated symbol vector and the true symbol vector. We address this problem by moving the cluster in the first quadrant to $(1+0j)$, and use the ratio to map all the other values. The result is shown in Fig 2. They do match.

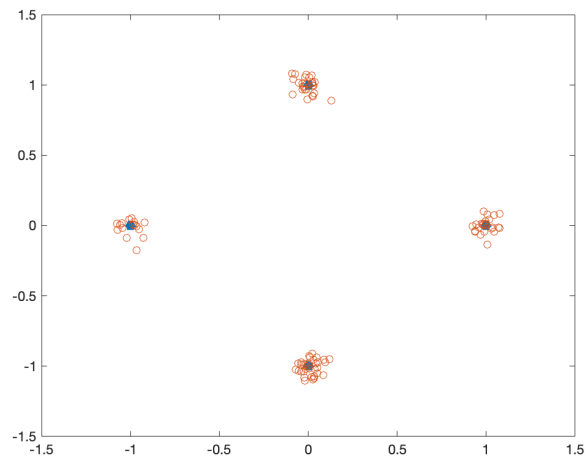


Fig. 2: The estimated symbols (red) and the true symbols (blue)

APPENDIX A

PILOT

```

1 function h = pilot(X,s)
2 % just compute the pinv?
3 [P,N] = size(X);
4 L = 3;
5 S_L = zeros(L,N);
6 for i = 1:L
7     for j = 1:N
8         S_L(i,j+i-1) = s(j);
9     end
10 end
11 S_L = S_L([1:L]), ([1:N]);
12 H = X*pinv(S_L);
13 [a1,a2] = size(H);
14 h = zeros(L*P,1);
15 ll = 1;
16 for i = 1:a2
17     for j = 1:a1
18         h(ll) = H(j,i);
19         ll = ll + 1;
20     end
21 end
22 end

```

APPENDIX B

TEST 1

```

1 %% test 1
2 %% question 1
3 tau = [0.1,0.6]';
4 % generate beta vector
5 ph1 = rand()*2*pi;
6 ph2 = rand()*2*pi;
7 beta = [exp(1i*ph1),0.7*exp(1i*ph2)]';
8 %% generate x and X
9 P = 5;
10 N = 100;
11 L = 3;
12 SNR = 30;
13 s = source(N);
14 h = channel(tau,beta,L,P);
15 X = gen_data1(h,s,P,N,SNR);
16 h_ = pilot(X,s);
17 figure(1)
18 title('True versus Estimated')
19 plot(h,'x');
20 hold on
21 plot(h_,'o');

```

APPENDIX C

BLIND CHANNEL

```

1 function h = blind_channel(X)
2 %
3 [S,V,D] = svd(X);
4 U_n = S(:,([4:5]));
5 % Construct U_T
6 [a1,a2] = size(U_n');

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7 U_T = zeros(6,3*a2);
8 U_T([1:2]),([1:5])=U_n';
9 U_T([3:4]),([6:10])=U_n';
10 U_T([5:6]),([11:15])=U_n';
11 [S_,V_,D_] = svd(U_T);
12 end

```

APPENDIX D BLIND SYMBOL

```

1 function s_ = blind_symbol(X)
2 %
3 [S,V,D] = svd(X);
4 V_n = D(:, [4:100]);
5 V_t = zeros(102,97*3);
6 V_t([1:100],[1:97]) = V_n;
7 V_t([2:101],[98:2*97]) = V_n;
8 V_t([3:102],[2*97+1:3*97]) = V_n;
9 [s,v,d]=svd(V_t);
10 s_ = s([3:102],102);
11 end

```

APPENDIX E TEST 2

```

1 %% test 2
2 %% question 2
3 tau = [0.1,0.6]';
4 % generate beta vector
5 ph1 = rand()*2*pi;
6 ph2 = rand()*2*pi;
7 beta = [exp(1i*ph1),0.7*exp(1i*ph2)]';
8 %% generate x and X
9 P = 5;
10 N = 100;
11 L = 3;
12 SNR = 30;
13 s = source(N);
14 h = channel(tau,beta,L,P);
15 X = gen_data1(h,s,P,N,SNR);
16 % h_ = blind_channel(X);
17 s_ = blind_symbol(X);
18 cluster = zeros();
19 k = 1;
20 for i = 1:N
21     if real(s_(i))>0 && imag(s_(i))>0
22         cluster(k) = s_(i);
23         k = k + 1;
24     end
25 end
26 s_est = mean(cluster);
27 Ratio = 1/s_est;
28 figure(1)
29 title('True versus Estimated')
30 plot(s,'^','markersize',5,'linewidth',10);
31 hold on
32 plot(s_*Ratio,'o');

```