# CT509:

# Direction of Arrival Estimation Algorithms

November 1 2020

# 1 Spectral Resolution and Leakage

# 1.1 Case1:

K=3 and angle of arrivals are [-1° 0° 1°]. In all the following figures,

- PSD indicates FFT approach
- MF indicates Matched Filter approach
- Window indicates FIR filter approach
- EM stands for Expectation Maximization
- ESPRIT is an algebraic algorithm.

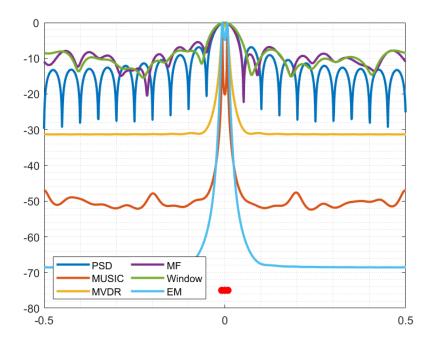


Figure 1: PSD in dB

Algorithm	Estimated AoA
MVDR	[-1 1]
MUSIC	[-1 1]
EM	[-1 1]
ESPRIT	[-1 0 1]

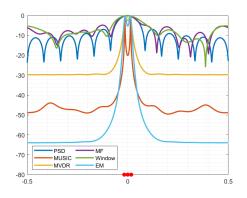
Table 1: Estimated Angle of arrivals for different algorithms

#### **Observations**

- FFT,MF,FIR filter methods doesn't perform well
- We can see peaks in MUSIC, MVDR, EM algorithms, but as frequencies are very close [-0.0087 0 0.0087] we were able to see only two peaks.
- MUSIC outperforms other algorithms in terms of resolution.
- Leakage is least in EM, followed by MUSIC and MVDR.
- ESPRIT calculates angle of arrivals accurately. We can generalize that ESPRIT can calculate AoA precisely for other higher angles (and K=3)

# 1.2 Case2:

K=3 and angles are [-d 0 d] where  $d \in \{2,4,5,6,8,10\}$ 

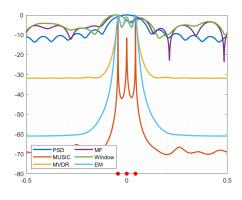


-10
-20
-30
-40
-50
-60
-70
-MUSIC Window
MVDR EM
-0.5

Figure 2: [-2 0 2] (d=2)

Figure 3: [-4 0 4] (d=4)

- In the above two cases signals are inseparable.
- These are typical examples of smearing



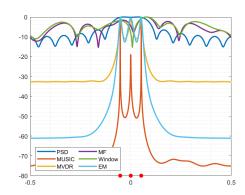
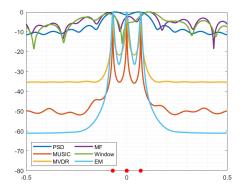


Figure 4: [-5 0 5] (d=5)

Figure 5: [-6 0 6] (d=6)

- As d is increasing, MUSIC is clearly able to distinguish different users.
- Expectation maximization is starting to produce better results



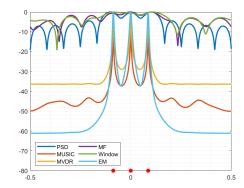


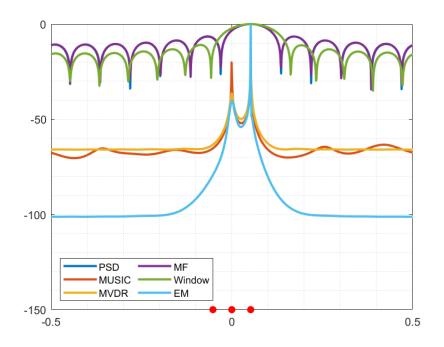
Figure 6: [-8 0 8] (d=8)

Figure 7: [-10 0 10] (d=10)

- MUSIC, MVDR and EM are totally able to separate the signals.
- FFT based approach is also shaping out but rest are not upto the mark.

# 1.3 Case3:

In this case, d is chosen as 6 and power levels are set to  $\{-20dB\ 0dB\ 20dB\}$ 

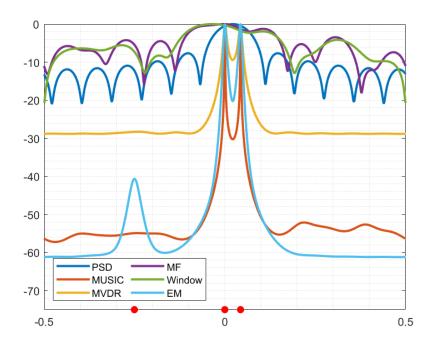


# **Observations:**

- In MUSIC,MVDR and EM, we observe two peaks.No algorithm is able to separate three signals.
- ESPRIT is able to calculate AoA's precisely
- EM,MVDR are sensitive to received power signals

# 1.4 Case4:

K=3, angles [-30 0 5] (in degrees), power levels are set to {-20dB 0dB 0dB}



# **Observations:**

- Again, Similar comments can be made on MUSIC, MVDR, EM.
- However, EM is showing a faint presence of third user, other two were able to distinguish only two users.

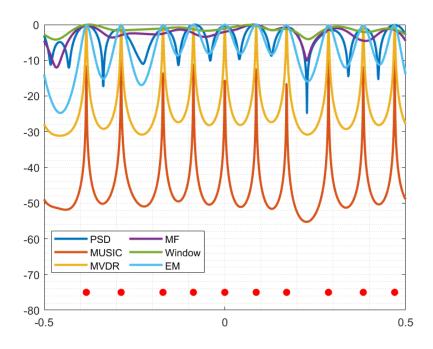
# 2 A Performance Study:

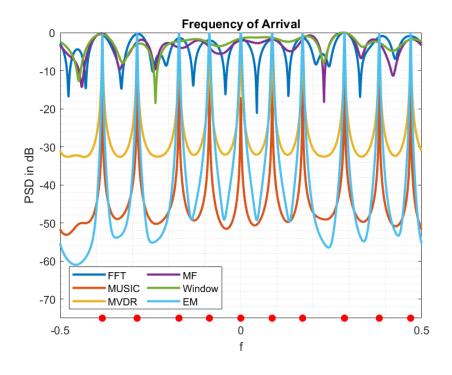
# 2.1 Performance of all algorithms with increase in number of elements

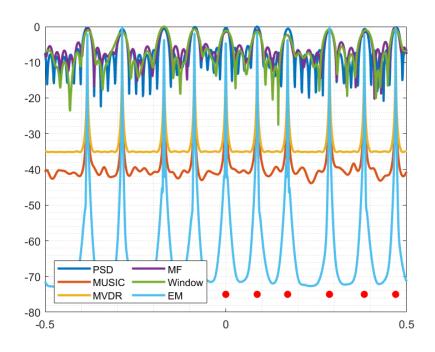
Number of users are set to 10, DoAs are the following angles and N is varied as [15 20 50 80] respectively

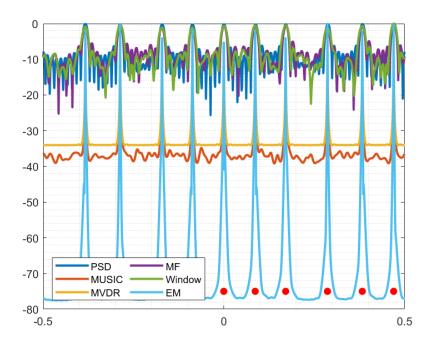
$$\theta^{\circ} = \{-50^{\circ}, -35^{\circ}, -20^{\circ}, -10^{\circ}, 0^{\circ}, 10^{\circ}, 20^{\circ}, 35^{\circ}, 50^{\circ}, 70^{\circ}\}$$

And signals are received at power level of 1W









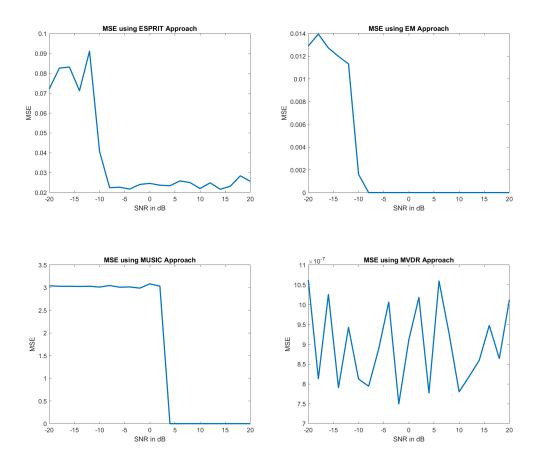
#### **Observations**

- With increase in number of Elements(N) in ULA, We observe,
  - Peaks become more prominent, i.e. better resolution
  - Spectral leakage is reduced. i.e The power leaked in the sidelobes is reduced.
- Expectation maximization algorithm gives accurate results with increase in number of elements
- However, with increase in number of elements, receivers become expensive.
- MUSIC algorithm gives expected results with lesser number of elements.

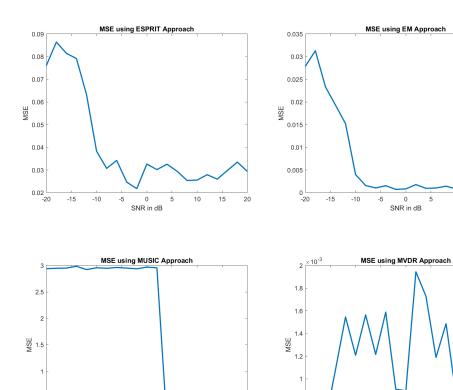
# 2.2 Monte-Carlo Simulations

- Number of sources be 5, K=5
- SNR is varied from -20 to 20 (in the steps of 2)
- Number of MonteCarlo Simulations is considered to be 50.

# case1 let the separation between any two closest sources be 20°.



# case2 let the separation between any two closest sources be 15°.



0 SNR in dB

0.5

0.6

-15 -10

0 SNR in dB 15

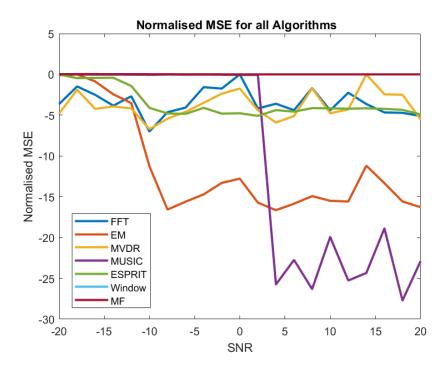


Figure 8: MSE vs SNR for all algorithms when  $\Delta=20\ensuremath{\,^\circ}$ 

# **Calculations**

• MSE is calculated as follows

$$MSE = \sum_{k=1}^{K} (\theta_{estimated}^{k} - \theta_{actual}^{k})^{2}$$
 (1)

Normalised MSE

$$NormalisedMSE = 10log_{10}(MSE/max(MSE))$$
 (2)

# 3 Inferences

- MVDR algorithm has relatively better resolution, compared to MF,FFT and Window algorithms.
- MVDR and EM algorithms depend on the power of the received signals, while MUSIC and ESPRIT are less dependent.
- All algorithms are dependent on the number of signals processed
- Both MUSIC and ESPRIT produce better results in detecting signals that are close. i.e. when the DoAs are close to each other.
- These two algorithms have superior resolution and accuracy.
- In case of Low SNR, performance of MUSIC deteriorates.

# 4 Appendix

#### **4.1 CODE**

```
clearvars;close all;clc;
N=20; n=0:N-1; K=5;
a_k = ones(K,1); mse = zeros(1,21);
_{4} M=1024; t=0;
5 \text{ snr} = -20:2:20;
6 mse_psd=zeros(1,length(snr));
7 mse_PEM=zeros(1,length(snr))
s ;mse_mvdr=zeros(1,length(snr));
9 mse_music=zeros(1,length(snr));
mse_esprit=zeros(1,length(snr));
for ii =1:length(snr)
_{12} for nsim = 1:50
SNR_dB=snr(ii); gamma=10^(SNR_dB/10); sigma=1/sqrt(gamma);
theta1 = 15*rand()-60;
theta2 = 15*rand()-30;
16 theta3 = 15*rand();
theta4 = -15*rand()+30;
18 theta5 = -15*rand()+60;
thetass =[theta1 theta2 theta3 theta4 theta5];
theta=thetass.*(pi/180);
f=sin(theta)/2;
phi_k=2*pi*rand(K,1)-pi;f=f(:);
xn_k = a_k.*exp(1j*2*pi*theta').*exp(1j*2*pi*f*(0:N-1));
24 %% PSD
  for kk=1:N
     xnk=xn_k(:,kk);
     R=xnk*xnk';
     t=t+R;
     f_k=f.';
     W=\exp(-1j*2*pi*n'*f_k);
     u = sigma*randn(N,1)+1j*randn(N,1);
      xn = W*xnk+u;
33 end
```

```
R_s=t/N;
  Rss = diag(a_k.^2); Rxx=W*R_ss*W'+sigma^2*eye(N);
  fo = linspace(-0.5, 0.5, M);
  r = fftshift(fft(Rxx(1,:),M));
  r1 = abs(r); r1 = 10*log10(r1/max(r1));
  a1 = [-70; -50]; a2 = [-40; -20]; a3 = [-10; 10]; a4 = [20; 40]; a5 =
      [50;70];
  f_{one} = sin([a1 a2 a3 a4 a5]*pi/180)/2;
  aa1 = find(abs(fo-f_one(1,1))<0.0005);ab1 =
      find(abs(fo-f_one(2,1))<0.0005);
  aa2 = find(abs(fo-f_one(1,2))<0.0005);ab2 =
      find(abs(fo-f_one(2,2))<0.0005);
  aa3 = find(abs(fo-f_one(1,3))<0.0005);ab3 =
      find(abs(fo-f_one(2,3))<0.0005);
  aa4 = find(abs(fo-f_one(1,4))<0.0005);ab4 =
      find(abs(fo-f_one(2,4))<0.0005);
  aa5 = find(abs(fo-f_one(1,5))<0.0005);ab5 =
      find(abs(fo-f_one(2,5))<0.0005);
  for oo = 1:length(fo)
   if oo>=aa1 && oo<ab1</pre>
      r11(oo) = r1(oo);
  else
      r11(oo) = -80;
52 end
53 end
[mm,nn] = max(r11); f1_est = fo(nn);
for oo = 1:length(fo)
  if oo>=aa2 && oo<ab2
      r11(oo) = r1(oo);
  else
      r11(oo) = -80;
  end
60
  end
  [mm,nn]=max(r11);f2\_est = fo(nn);
  for oo = 1:length(fo)
  if oo>=aa3 && oo<ab3
      r11(00) = r1(00);
```

```
66 else
      r11(oo) = -80;
   end
69 end
70 [mm,nn]=max(r11);f3_est = fo(nn);
   for oo = 1:length(fo)
12 if 00>=aa4 && 00<ab4
      r11(oo) = r1(oo);
   else
      r11(00) = -80;
   end
76
   end
   [mm,nn]=max(r11);f4\_est = fo(nn);
  for oo = 1:length(fo)
   if oo>=aa5 && oo<ab5
      r11(oo) = r1(oo);
   else
      r11(oo) = -80;
   end
85
86 end
[mm,nn]=max(r11);f5_est = fo(nn);
88 f_estimated = [f1_est f2_est f3_est f4_est f5_est];
aoa = asin(2*f_estimated)*180/pi;
90 theta11 = sort(thetass)';
91 aoal1 = sort(aoa);
mmse_psd(nsim,ii) = (1/K)*sum((theta11'-aoa11).^2);
93 %% MVDR
W1 = \exp(-1j*2*pi*n'*fo);
95 for i = 1:M
96 P_MVDR(i) = 1/(W1(:,i)'*inv(Rxx)*W1(:,i));
97 end
   Pmvdr=10*log10(abs(P_MVDR)/max(abs(P_MVDR))); %Spatial spectrum
      function
   for oo = 1:length(fo)
   if oo>=aa1 && oo<ab1</pre>
      r11(oo) = Pmvdr(oo);
101
102 else
```

```
r11(00) = -80;
103
   end
104
   end
   [mm,nn]=max(r11);f1_est = fo(nn);
   for oo = 1:length(fo)
   if oo>=aa2 && oo<ab2</pre>
       r11(oo) = Pmvdr(oo);
109
   else
       r11(00) = -80;
   end
   end
113
   [mm,nn]=max(r11);f2_est = fo(nn);
114
   for oo = 1:length(fo)
   if oo>=aa3 && oo<ab3
116
       r11(oo) = Pmvdr(oo);
117
   else
       r11(oo) = -80;
119
   end
120
   end
   [mm,nn]=max(r11);f3_est = fo(nn);
   for oo = 1:length(fo)
   if oo>=aa4 && oo<ab4
       r11(oo) = Pmvdr(oo);
125
   else
126
       r11(oo) = -80;
127
   end
   [mm,nn]=max(r11);f4_est = fo(nn);
131
   for oo = 1:length(fo)
   if oo>=aa5 && oo<ab5</pre>
       r11(oo) = Pmvdr(oo);
134
   else
       r11(00) = -80;
136
   end
137
   end
   [mm,nn]=max(r11);f5_est = fo(nn);
   f_estimated = [f1_est f2_est f3_est f4_est f5_est];
```

```
aoa = asin(2*f_estimated);
   theta11 = sort(theta)';
   aoal1 = sort(aoa);
   mmse_mvdr(nsim,ii) = (1/K)*sum((thetall'-aoall).^2);
   %% MUSIC
146
147 [Q,D]=eig(Rxx);
   k=sum(sum(round(D)\sim=0));
_{149} K1 = N-k;
V = Q(:,[k+1:end]);
   W1 = \exp(-1j*2*pi*n'*fo);
   for i = 1:length(fo)
   PsePSD(i) = 1/((W1(:,i)'*(V*V')*W1(:,i)));
   end
154
   Pmusic=10*log10(abs(PsePSD))/max(abs(PsePSD))); %Spatial spectrum
       function
   for oo = 1:length(fo)
   if oo>=aa1 && oo<ab1
       r11(oo) = Pmusic(oo);
158
   else
159
       r11(00) = -80;
   end
161
162 end
   [mm,nn]=max(r11);f1\_est = fo(nn);
   for oo = 1:length(fo)
   if oo>=aa2 && oo<ab2</pre>
       r11(oo) = Pmusic(oo);
166
   else
       r11(00) = -80;
168
   end
169
   end
   [mm,nn]=max(r11);f2\_est = fo(nn);
171
   for oo = 1:length(fo)
   if oo>=aa3 && oo<ab3
       r11(oo) = Pmusic(oo);
174
   else
       r11(00) = -80;
176
   end
177
```

```
end
   [mm,nn]=max(r11);f3_est = fo(nn);
   for oo = 1:length(fo)
   if oo>=aa4 && oo<ab4
      r11(oo) = Pmusic(oo);
182
   else
183
      r11(oo) = -80;
   end
   end
186
   [mm,nn]=max(r11);f4\_est = fo(nn);
188
   for oo = 1:length(fo)
189
   if oo>=aa5 && oo<ab5
      r11(oo) = Pmusic(oo);
191
   else
192
      r11(00) = -80;
   end
194
   end
   [mm,nn]=max(r11);f5_est = fo(nn);
   f_estimated = [f1_est f2_est f3_est f4_est f5_est];
  aoa = asin(2*f_estimated);
   theta11 = sort(theta)';
   aoal1 = sort(aoa);
   mmse_music(nsim,ii) = (1/K)*sum((thetall'-aoall).^2);
201
   %% EM
W1 = \exp(-1j*2*pi*n'*fo); Reference filter bank
   %Random initialization of Rzz
   rnf = rand(M,1);
   Rzz = diag(rnf.^2);
   Xn = W*xn_k+sigma^2*eye(N);
   % Rzz = eye(M);
   for i = 1:20
209
      Rxx1 = W1*Rzz*W1';
210
      Rxx1 = Rxx1/N;
211
      Rxx_inv = inv(Rxx1);
      for mm = 1:M
         W_m = W1(:,mm);
214
         w_mvdr = Rxx_inv*w_m/(w_m'*Rxx_inv*w_m);
```

```
y_em = w_mvdr'*Xn;
216
                  for aa = 1:12
217
               y_em(:,aa) = w_mvdr'*Xn(:,aa);
218
            end
219
          Rzz(mm,mm) = y_em*y_em';
       end
221
   end
222
   P_{em} = abs(diag(Rzz));
   P_EM=10*log10((P_em)/max((P_em)));
224
   for oo = 1:length(fo)
   if oo>=aa1 && oo<ab1</pre>
227
       r11(oo) = P_{-}EM(oo);
   else
229
       r11(oo) = -80;
230
   end
   end
232
   [mm,nn]=max(r11);f1_est = fo(nn);
   for oo = 1:length(fo)
   if oo>=aa2 && oo<ab2
       r11(oo) = P_EM(oo);
   else
237
       r11(oo) = -80;
   end
239
   end
240
   [mm,nn]=max(r11);f2\_est = fo(nn);
   for oo = 1:length(fo)
   if oo>=aa3 && oo<ab3
       r11(oo) = P_{-}EM(oo);
244
   else
       r11(oo) = -80;
   end
247
   end
   [mm,nn]=max(r11);f3_est = fo(nn);
   for oo = 1:length(fo)
   if oo>=aa4 && oo<ab4</pre>
       r11(oo) = P_EM(oo);
252
253 else
```

```
r11(00) = -80;
254
   end
255
   end
256
   [mm,nn]=max(r11);f4\_est = fo(nn);
   for oo = 1:length(fo)
259
   if oo>=aa5 && oo<ab5
      r11(oo) = P_EM(oo);
261
   else
262
      r11(00) = -80;
   end
264
   end
265
   [mm,nn]=max(r11);f5_est = fo(nn);
   f_estimated = [f1_est f2_est f3_est f4_est f5_est];
aoa = asin(2*f_estimated);
   theta11 = sort(theta)';
270 aoa11 = sort(aoa);
mmse_PEM(nsim,ii) = (1/K)*sum((thetall'-aoall).^2);
   %% ESPRIT
[U,S,V] = svd(Xn);
_{274} MM = diag(S);
[rr,cc]=find(abs(MM)-sigma^2<0.00001);
kapa = sum(cc);
277 KK = N-kapa;
U_{est} = U(:,[1:K]); val = 1;
   J_x = [eye(N-val) zeros(N-val,1)];
   J_y = [zeros(N-val,1) eye(N-val)];
   Ux = J_x*U_est;
  Uy = J_y*U_est;
   Z = pinv(Ux)*Uy;
   [T,thetas] = eig(Z);
   ang = asin(angle(thetas)/pi)*180/pi;
  ang = diag(ang)*pi/180;
288 W_1 =T*U_est';
theta11 = sort(thetass)'*pi/180;
  aoa11 = sort(ang);
   mmse_eprit(nsim,ii) = (1/K)*sum((thetall-aoall).^2);
```

```
%% MF
   y = xn'*W1;
   y_mf=10*log10(abs(y)/max(abs(y)));
   for oo = 1:length(fo)
   if oo>=aa1 && oo<ab1
       r11(oo) = y_mf(oo);
297
   else
298
       r11(00) = -80;
299
   end
300
   end
   [mm,nn]=max(r11);f1_est = fo(nn);
302
   for oo = 1:length(fo)
   if oo>=aa2 && oo<ab2
       r11(oo) = y_{-}mf(oo);
305
   else
306
       r11(oo) = -80;
307
   end
308
   end
   [mm,nn]=max(r11);f2\_est = fo(nn);
   for oo = 1:length(fo)
311
   if oo>=aa3 && oo<ab3
       r11(oo) = y_mf(oo);
313
   else
       r11(oo) = -80;
315
   end
316
   end
317
   [mm,nn]=max(r11);f3_est = fo(nn);
   for oo = 1:length(fo)
   if oo>=aa4 && oo<ab4
320
       r11(oo) = y_mf(oo);
321
   else
       r11(00) = -80;
323
   end
324
   end
325
   [mm,nn]=max(r11);f4\_est = fo(nn);
326
   for oo = 1:length(fo)
   if oo>=aa5 && oo<ab5
```

```
r11(oo) = y_mf(oo);
330
   else
331
       r11(00) = -80;
332
   end
333
   end
   [mm,nn]=max(r11);f5_est = fo(nn);
335
   f_estimated = [f1_est f2_est f3_est f4_est f5_est];
   aoa = asin(2*f_estimated);
   theta11 = sort(theta)';
   aoal1 = sort(aoa);
   mmse_MF(nsim,ii) = (1/K)*sum((thetall'-aoall).^2);
340
341
   %% Window
_{343} wc = kaiser(N,2.4);
xn = xn.*wc;
   y = xn'*W1;
   y_win=10*log10(abs(y)/max(abs(y)));
   for oo = 1:length(fo)
   if oo>=aa1 && oo<ab1
       r11(oo) = y_win(oo);
349
   else
       r11(oo) = -80;
351
   end
352
   end
   [mm,nn]=max(r11);f1_est = fo(nn);
   for oo = 1:length(fo)
   if oo>=aa2 && oo<ab2
       r11(oo) = y_win(oo);
357
   else
358
       r11(oo) = -80;
   end
   end
361
   [mm,nn]=max(r11);f2\_est = fo(nn);
   for oo = 1:length(fo)
   if oo>=aa3 && oo<ab3</pre>
       r11(oo) = y_win(oo);
   else
366
       r11(00) = -80;
367
```

```
end
368
   end
369
   [mm,nn]=max(r11);f3_est = fo(nn);
   for oo = 1:length(fo)
   if oo>=aa4 && oo<ab4
       r11(oo) = y_win(oo);
373
   else
374
       r11(00) = -80;
375
   end
376
   end
   [mm,nn]=max(r11);f4\_est = fo(nn);
378
379
   for oo = 1:length(fo)
380
   if oo>=aa5 && oo<ab5
381
       r11(oo) = y_win(oo);
382
   else
383
       r11(oo) = -80;
384
   end
385
   end
386
   [mm,nn]=max(r11);f5_est = fo(nn);
387
   f_estimated = [f1_est f2_est f3_est f4_est f5_est];
   aoa = asin(2*f_estimated);
   theta11 = sort(theta)';
   aoal1 = sort(aoa);
391
   mmse\_Window(nsim,ii) = (1/K)*sum((thetall'-aoall).^2);
392
393
   end
394
   end
   %%
396
   % mmse = mmse/mnop;
   mmse_psd = sum(mmse_psd)/nsim;
   mmse_PEM = sum(mmse_PEM)/nsim;
399
   mmse_mvdr = sum(mmse_mvdr)/nsim;
   mmse_music = sum(mmse_music)/nsim;
401
   mmse_esprit = sum(mmse_esprit)/nsim;
402
   mmse_Window = sum(mmse_Window)/nsim;
   mmse_MF = sum(mmse_MF)/nsim;
404
405
```

```
mmse_psd = 10*log10(mmse_psd/max(mmse_psd));
   mmse_PEM = 10*log10(mmse_PEM/max(mmse_PEM));
   mmse_mvdr = 10*log10(mmse_mvdr/max(mmse_mvdr));
408
   mmse_music = 10*log10(mmse_music/max(mmse_music));
   mmse_esprit = 10*log10(mmse_esprit/max(mmse_esprit));
   figure(1);
411
  plot(snr,(mmse_psd),'linewidth',2);hold on;
  % xlabel('SNR in dB');ylabel('MSE');
   % title('MSE using FFT Approach');
  % figure(2);
   plot(snr,(mmse_PEM),'linewidth',2);hold on;
  % xlabel('SNR in dB');ylabel('MSE');
   % title('MSE using EM Approach');
  % figure(3);
plot(snr,(mmse_mvdr),'linewidth',2);hold on;
  % xlabel('SNR in dB');ylabel('MSE');
% title('MSE using MVDR Approach');
423 % figure(4);
plot(snr,(mmse_music),'linewidth',2);hold on;
  % xlabel('SNR in dB');ylabel('MSE');
  % title('MSE using MUSIC Approach');
427 % figure(5);
plot(snr,(mmse_esprit),'linewidth',2);hold on;
   % xlabel('SNR in dB');ylabel('MSE');
  % title('MSE using ESPRIT Approach');
   plot(snr,(mmse_Window),'linewidth',2);hold on;
  % xlabel('SNR in dB');ylabel('MSE');
% title('MSE using Window Approach');
   plot(snr,(mmse_MF),'linewidth',2);hold on;
  % xlabel('SNR in dB');ylabel('MSE');
  % title('MSE using MF Approach');
   legend('FFT','EM','MVDR','MUSIC','ESPRIT','Window','MF')
  % saveas(figure(1), 'FFT.png'); saveas(figure(2), 'EM.png');
  % saveas(figure(4),'MUSIC.png');
saveas(figure(3),'MVDR.png');
   saveas(figure(5), 'ESPRIT.png');
```