

# Project II

## High-Resolution Beamforming on farfield monochromatic signals

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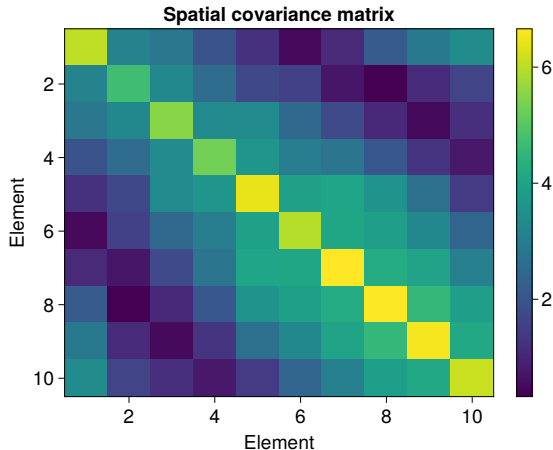
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# Estimating the spatial correlation matrix

We want to estimate the spatial correlation matrix from the data generated by `generate_data.jl`

- Find the spatial correlation matrix by definition

```
1 R = x*x' / N  
2 heatmap!(ax, abs.(R))
```



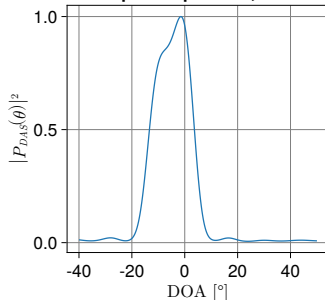
# Estimate spatial spectrum

We now want to estimate the classical spatial spectrum

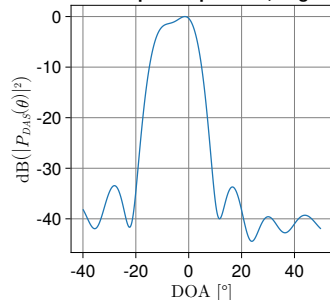
- We implement functions for the phase factor, steering vector and DAS.
- Apply  $P_{DAS}$  to every angle in DOA

```
1 # Define functions
2 DOA = -40:0.25:50
3  $\phi(\theta) = -k*d*\sin.(\theta)$ 
4  $a(\theta) = @. \exp(1im*\phi(\theta)*(0:M-1))$ 
5  $P_{DAS}(\theta) = (a(\theta)'*R*a(\theta)) / M$ 
6
7 # Apply on DAS
8  $P_{BF} = @. \text{abs}(P_{DAS}(DOA))^2$ 
9 # Normalize
10  $P_{BF} ./= \text{maximum}(P_{BF});$ 
```

Classical Spatial Spectrum, Linear scale



Classical Spatial Spectrum, Log scale

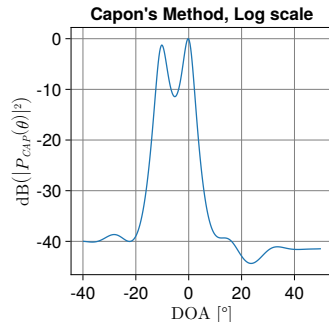
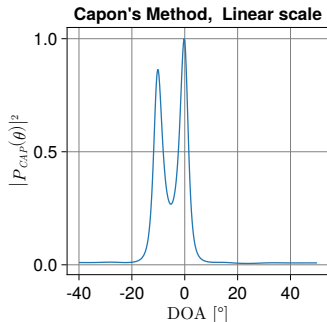


# Estimate spatial spectrum with Capon's method

Then we calculate it using minimum variance

- Implementing the function is straight forward, following the definition of Capon's method

```
1 P_CAP(theta) = 1 / (a(theta)'*inv(R)*a(theta))  
2  
3 P_BF2 = @. abs(P_CAP(DOA))^2  
4 P_BF2 ./= maximum(P_BF2);
```

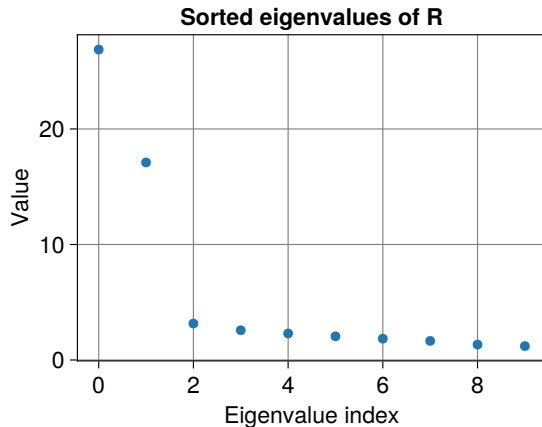


# Eigenvalue distribution

We want to find the distribution of eigenvalues and plot them in descending order.

- We decompose **R** using `eigvals` and `eigvecs`.
- They are sorted in ascending order, so we reverse them inplace.

```
1 dd, V = eigvals(R), eigvecs(R);  
2 reverse!(dd); reverse!(V, dims=2);
```

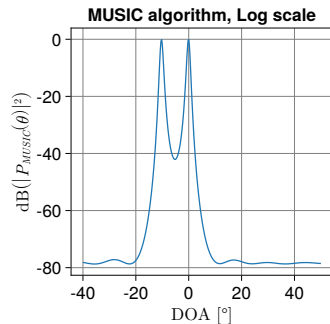
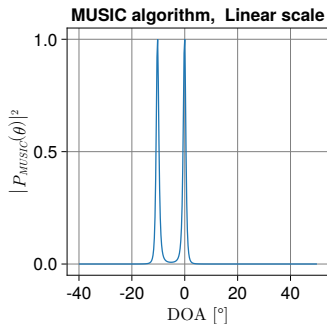


# Spatial spectrum using MUSIC

We assume to know the number of sources is know (2) and we find the spatial spectrum using the MUSIC method.

- Take all eigenvectors corresponding to noise space
- Make MUSIC

```
1 U = V[:, 3:end]
2 Π = U*U'
3 P_M(θ) = 1/(a(θ)'*Π*a(θ))
4
5 P_BF3 = @. abs(P_M(DOA))^2
6 P_BF3 ./= maximum(P_BF3);
```

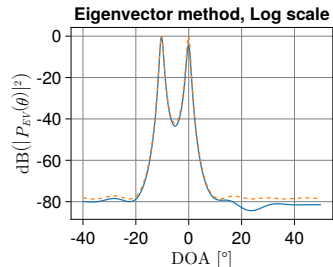
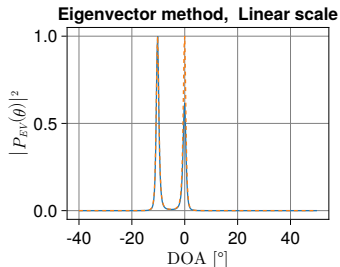


# Spatial spectrum using Eigenvector method

Now we use the eigenvalues as weights in the MUSIC method. This gives the Eigenvector method.

- We construct a diagonal matrix of the Eigenvalues  $\Lambda$  which serve as weights for the Eigenvectors.

```
1 U = V[:, 3:end]
2  $\Lambda^{-1}$  = inv(diag(dd[3:end]))
3  $P_{EV}(\theta) = 1 / (a(\theta)' * U * \Lambda^{-1} * U' * a(\theta))$ 
4
5 P_BF4 = @. abs(P_EV(DOA))^2
6 P_BF4 ./= maximum(P_BF4);
```



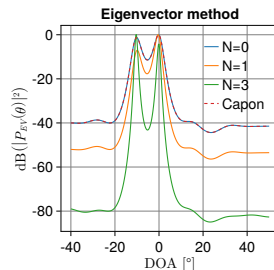
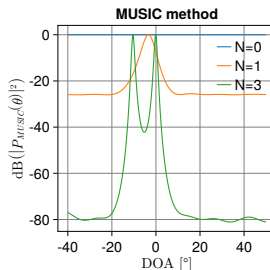
— Eigenvector    - - - MUSIC

# Incorrect estimation of number of sources

We now vary the number of sources we assume and plot both the MUSIC and the Eigenvector method.

- We define a function that returns the spatial spectrum given a number of sources
- We see that Eigenvector method given  $N = 0$  is identical to Capon's method

```
1 function assume_sources(N)
2   U = V[:, N+1:end]
3    $\Lambda^{-1}$  = inv(diagm(dd[N+1:end]))
4   P0 =  $\theta \rightarrow 1/(a(\theta)' * U * U' * a(\theta))$ 
5   P1 =  $\theta \rightarrow 1/(a(\theta)' * U * \Lambda^{-1} * U' * a(\theta))$ 
6   return (
7     (@. abs(P0(DOA))^2) |> x -> x ./ maximum(x),
8     (@. abs(P1(DOA))^2) |> x -> x ./ maximum(x)
9   )
10 end
```





# Coherent sources

Now we modify the `generate_data.jl` to create coherent signals.

- Code is identical to previous tasks, but uses different data.
- There is 30 deg separation between the sources compared to only 10 deg in the previous data.
- We see the classical DAS is now better than all other methods.

