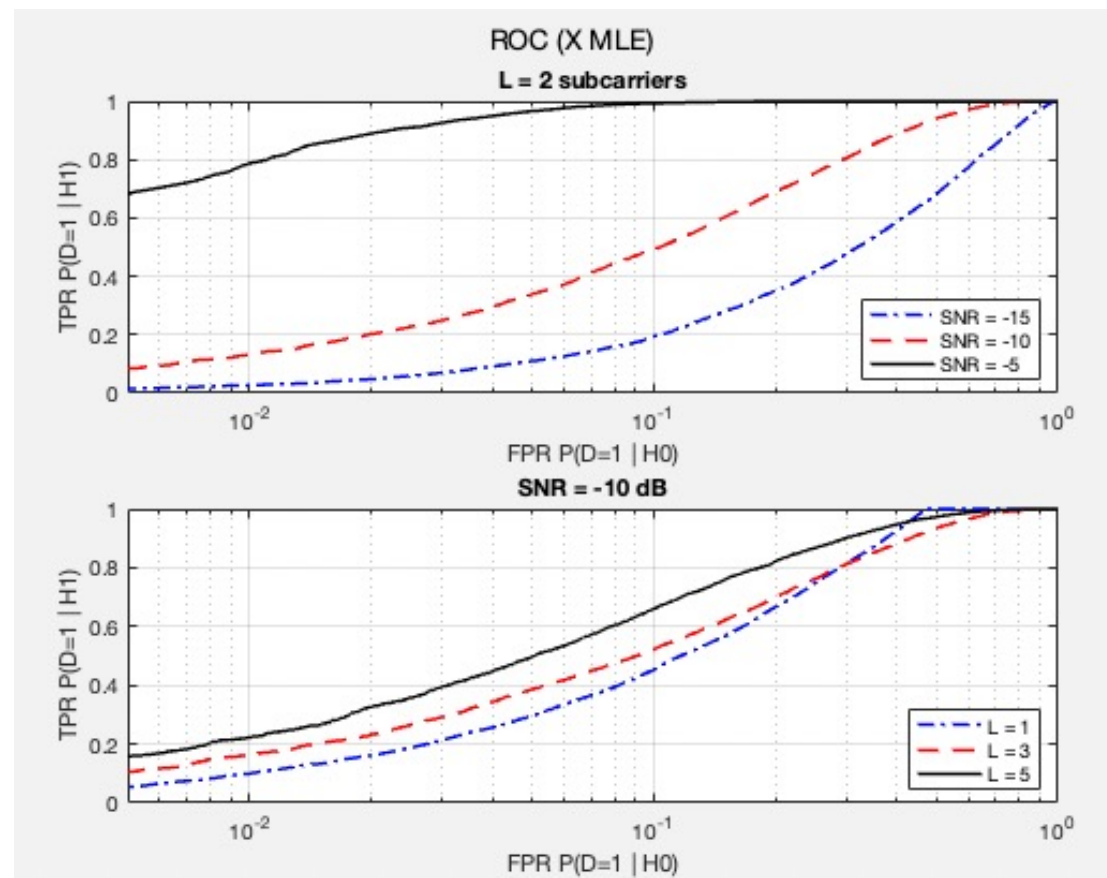
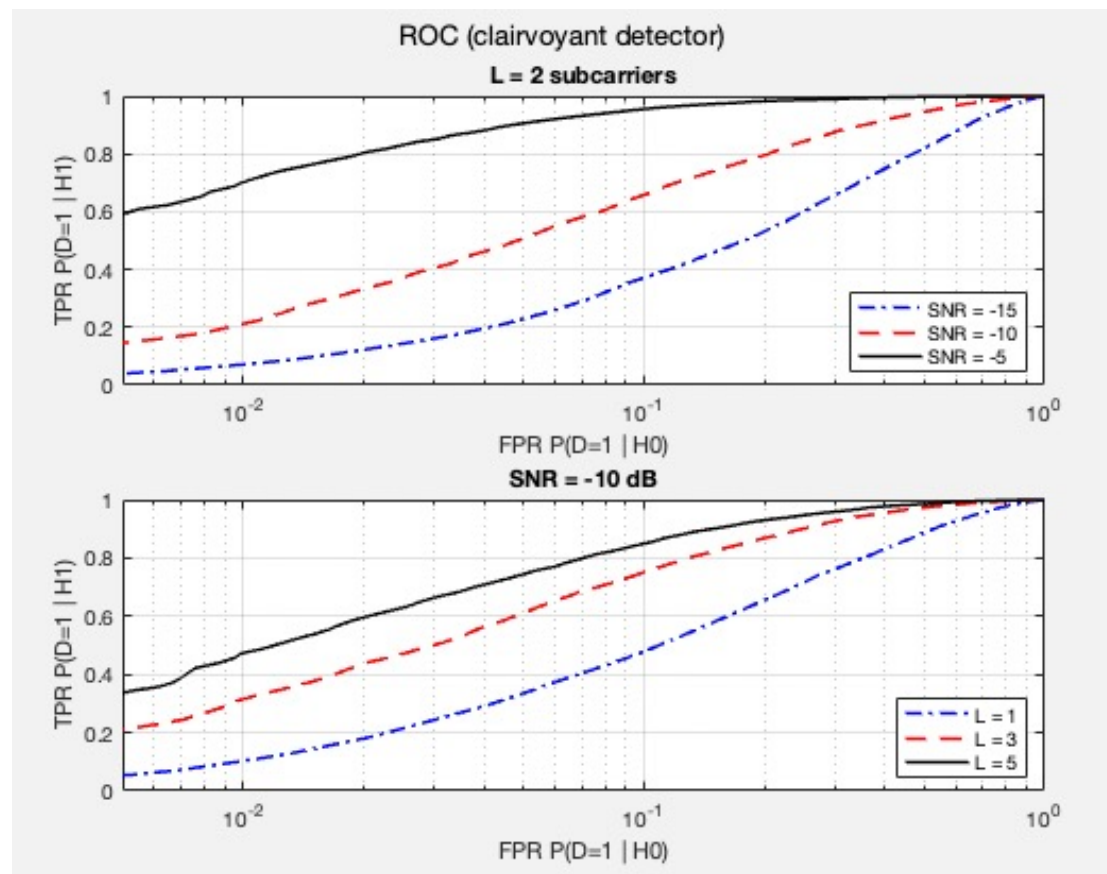


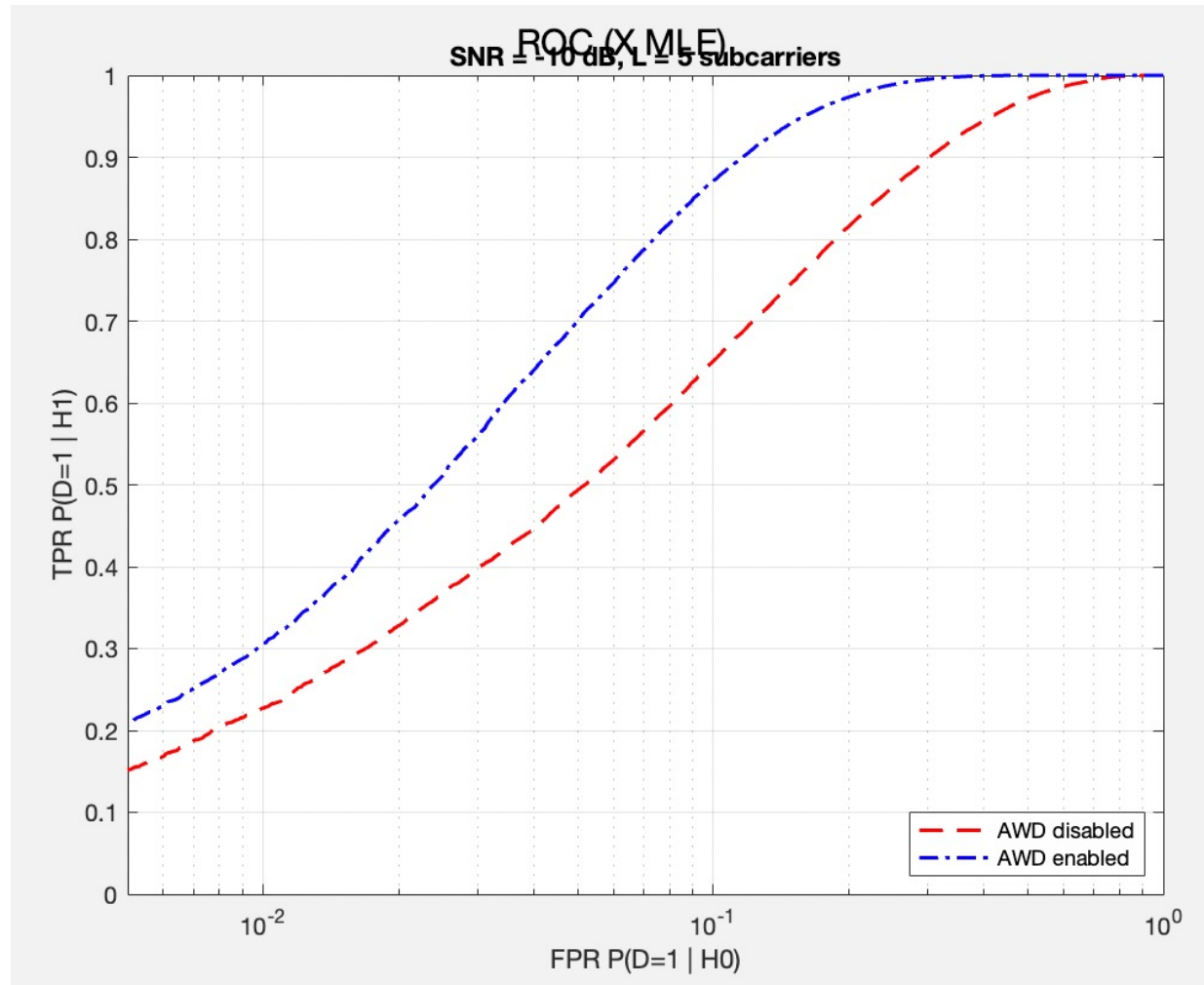
Cognitive IRR

Update Tue 15th Aug

Detection results



Waveform optimization (AWD module)



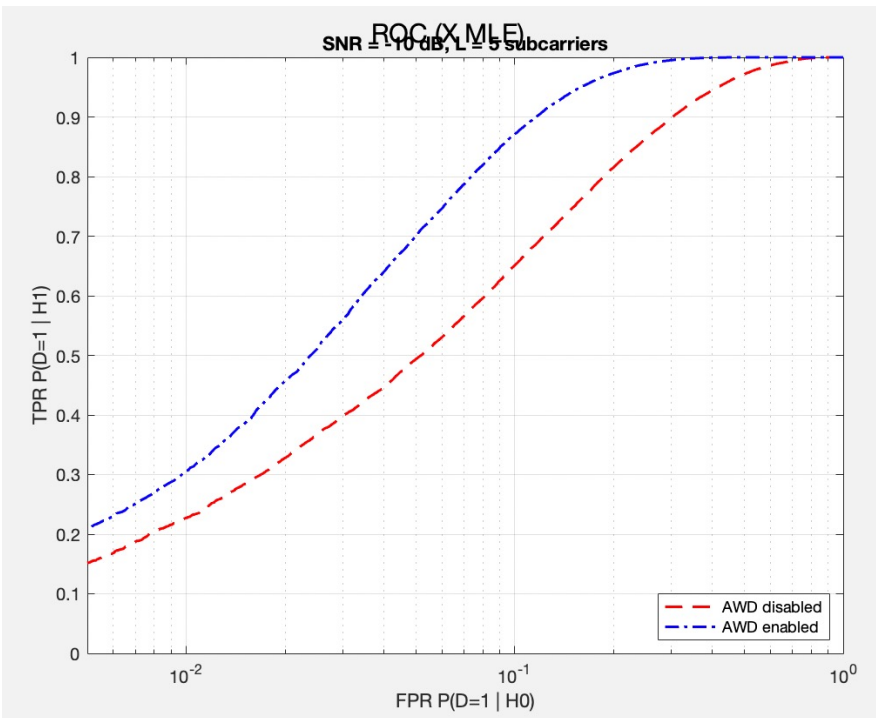
```
Command Window
K>> A_nonopt =
4.472135954999579e-01    0    0    0    0
0    4.472135954999579e-01    0    0    0
0    0    4.472135954999579e-01    0    0
0    0    0    4.472135954999579e-01    0
0    0    0    0    4.472135954999579e-01

K>> A_opt =
4.094307129291800e-01    0    0    0    0
0    5.246537681258305e-01    0    0    0
0    0    6.057606806448812e-01    0    0
0    0    0    8.284657941164966e-01    0
0    0    0    0    6.942260073469741e-01

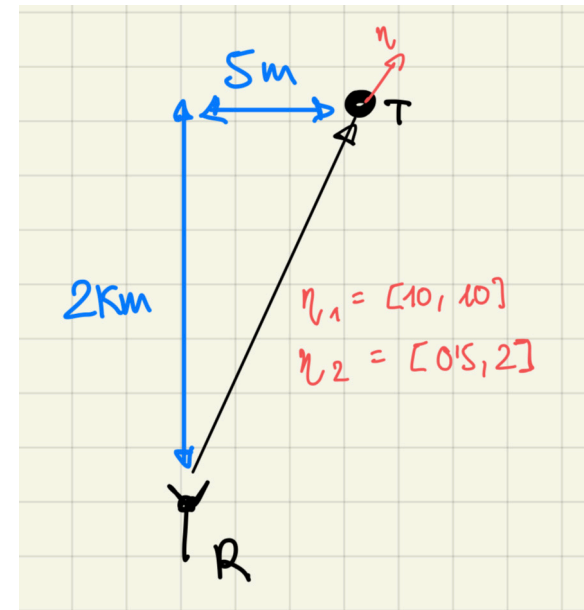
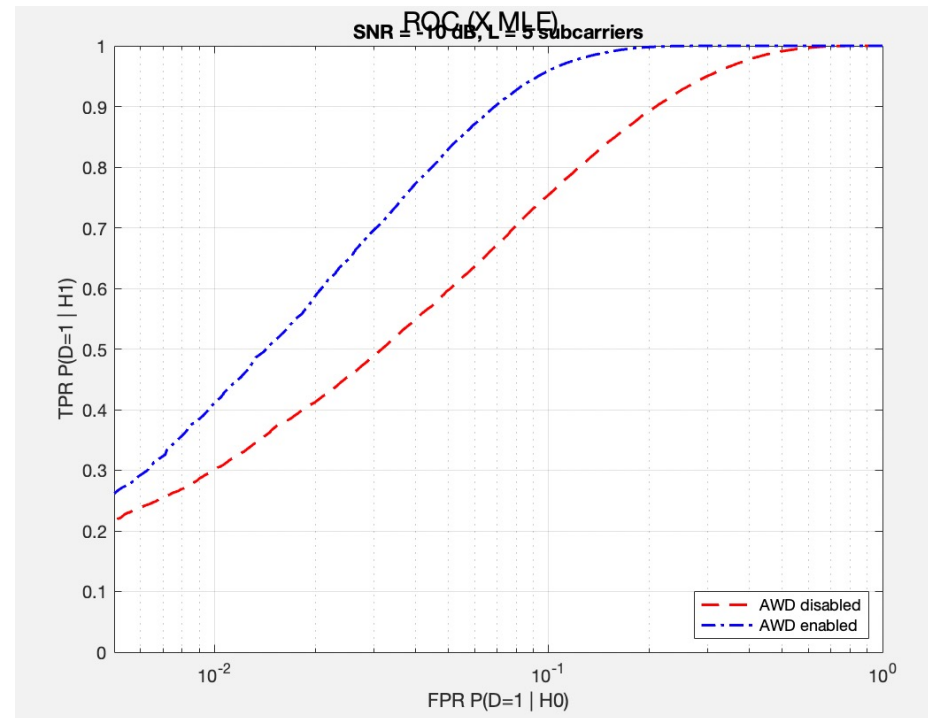
K>>
```

Reducing angle between target velocity vector and radar LOS

```
config.target_velocity = [10; 10];
```



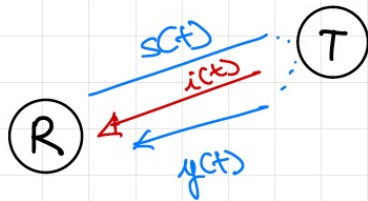
```
config.target_velocity = [0.5; 2];
```



Interference module (1)

Helena August 2023

Interference Module



Type 1: Delay Interference \rightarrow T sends $i_1(t)$ to lie about its POSITION

Description: uniform distribution around correct delay value.

Type 2: Doppler Interference \rightarrow T sends $i_2(t)$ to lie about its VELOCITY

Description: uniform distribution around the correct doppler freq.

Adversarial effect on $\Phi(\eta_T)$ matrix, where $\eta_T = [v_x, v_y]^T$ is the target velocity

Interference module (2)

Description: uniform distribution around the correct doppler freq.

Adversarial effect on $\Phi(\eta_T)$ matrix, where $\eta_T = [v_x, v_y]^T$ is the target velocity

$$\Phi = [\phi(t_1, \eta) \dots \phi(t_2, \eta)] \quad L \times N$$

$$\phi(t, \eta) = [e^{j\omega_{1,0}t} \dots e^{j\omega_{L,0}t}]^T \quad L \times 1$$

$$\omega_{l,0} = 2\pi\beta f_l, \quad \beta = 2 \cdot \langle \vec{v}, \vec{u} \rangle / c$$

$$f_l = f_c + l \Delta f$$

• Example Φ for $L=2$ subcarriers and $N=3$:

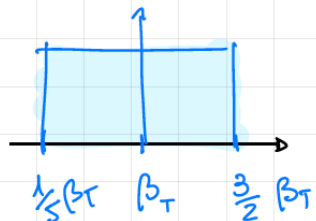
$$t_n = (z_0 + n \cdot T_p) \quad \forall n=0, \dots, N-1$$

$$\Phi = \begin{bmatrix} e^{j2\pi\beta f_1 t_1} & e^{j2\pi\beta f_1 t_2} & e^{j2\pi\beta f_1 t_3} \\ e^{j2\pi\beta f_2 t_1} & e^{j2\pi\beta f_2 t_2} & e^{j2\pi\beta f_2 t_3} \end{bmatrix}$$

→ same subcarrier freq

↙ same time instant

β is the RELATIVE DOPPLER SHIFT along the signal path



$$\beta_i = \beta_T [1 + \alpha \cdot \mathcal{U}(-1, 1)]$$

$$\beta_T = \frac{2}{c} \langle \vec{v}, \vec{v} \rangle$$

Interference module (3)

For each time instant t , we have $\phi(t, \eta_T)$ and $\phi(t, \eta_i)$

$\phi(t, \eta_T)$ is labeled **TRUE** (we have knowledge of it)
 $\phi(t, \eta_i)$ is labeled **interfered**

```

1 function varargout = get_ROC_from_config(config, LOS_vector, T_rangecell)
2
3 % Given the input configuration, obtain the OFDM model parameters
4 [A_nonopt, X_H1, Phi_t, Sigma_c] = get_OFDM_model(config, LOS_vector, T_rangecell);
5
6 % Initialize GLRT metrics
7 GLRT_H0 = zeros(config.N_mc, 1);
8 GLRT_H1 = zeros(config.N_mc, 1);
9
10 % Initialize GLRT metrics for AWD, if enabled
11 if config.enable_awd
12     GLRT_H0_opt = zeros(config.N_mc, 1);
13     GLRT_H1_opt = zeros(config.N_mc, 1);
14 end
15
16 % For each Monte Carlo realization, generate OFDM measurements
17 finalA = zeros(size(A_nonopt,1), size(A_nonopt,2), config.N_mc);
18 for idx_mc = 1:config.N_mc
19
20     [Y_H0, Y_H1] = build_OFDM_meas(config, Sigma_c, A_nonopt, X_H1, Phi_t);
21
22 % MLE of target coefficients X under hypothesis H1
23 if config.clairvoyant
24     X_hat = X_H1;
25 else
26     X_hat = inv(A_nonopt) * (Y_H1*Phi_t') * pinv(Phi_t*Phi_t'); % expression after eq 8 paper 0
27 % X_hat = inv(A_nonopt)*Y_H1*Phi_t'*inv(Phi_t*Phi_t');
28 % X_hat = diag(diag(X_hat));
29 end
30
31 % GLRT (non-optimal)
32 GLRT_H0(idx_mc, 1) = det(Y_H0*Y_H0') / det((Y_H0-A_nonopt*X_hat*Phi_t)*(Y_H0-A_nonopt*X_hat*Phi_t)');
33 GLRT_H1(idx_mc, 1) = det(Y_H1*Y_H1') / det((Y_H1-A_nonopt*X_hat*Phi_t)*(Y_H1-A_nonopt*X_hat*Phi_t)');

```

```

35 % Adaptive Waveform Design (AWD) module:
36 if config.enable_awd
37     A_ini = ones(1,config.L);
38     A_ini = diag(A_nonopt);
39 % sgm_sqr = 1; % set to 1 (best!) or sqrt(Sigma_c(1,1)) for good AWD performance
40 [A_opt, ~] = fminsearch(@(A) opt_waveform(A, config.L, X_H1, Phi_t, chol(Sigma_c), sgm_sqr), A_ini);
41 A_opt = diag(A_opt);
42 finalA(:, :, idx_mc) = A_opt;
43
44 % GLRT (optimal)
45 X_hat_opt = X_hat;
46 % X_hat_opt = inv(A_opt)*Y_H1*Phi_t'*inv(Phi_t*Phi_t');
47 GLRT_H0_opt(idx_mc, 1) = det(Y_H0*Y_H0') / det((Y_H0-A_opt*X_hat_opt*Phi_t)*(Y_H0-A_opt*X_hat_opt*Phi_t)');
48 GLRT_H1_opt(idx_mc, 1) = det(Y_H1*Y_H1') / det((Y_H1-A_opt*X_hat_opt*Phi_t)*(Y_H1-A_opt*X_hat_opt*Phi_t)');
49
50 end
51
52 if mod(idx_mc, 50) == 0
53     disp(['Monte carlo it ', num2str(idx_mc)])
54 end
55
56 % opt_A = mean(finalA); % we average A matrices from all MC realizations
57
58 [Ppn, Ppp] = get_ROC_from_GLRTs(config, GLRT_H0, GLRT_H1);
59 varargout{1} = Ppn;
60 varargout{2} = Ppp;
61
62 if config.enable_awd
63     [Ppn_opt, Ppp_opt] = get_ROC_from_GLRTs(config, GLRT_H0_opt, GLRT_H1_opt);
64     varargout{3} = Ppn_opt;
65     varargout{4} = Ppp_opt;
66 end
67

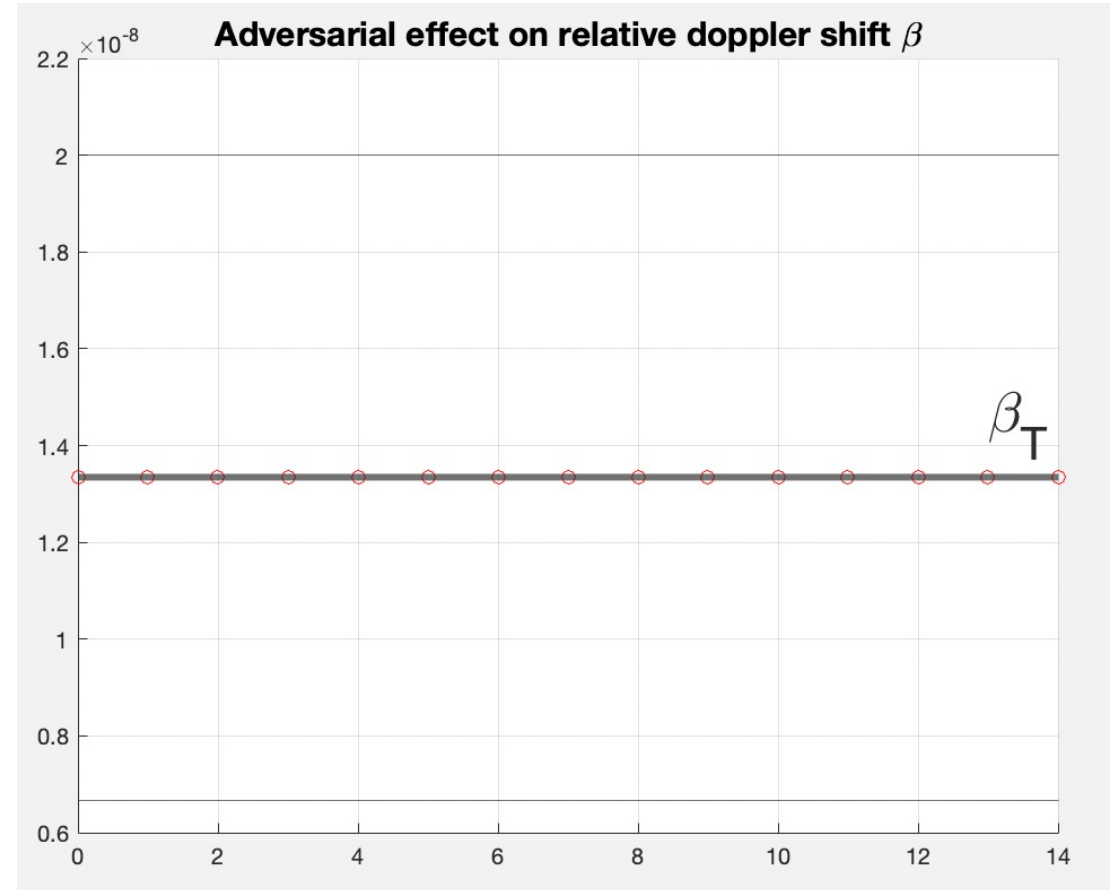
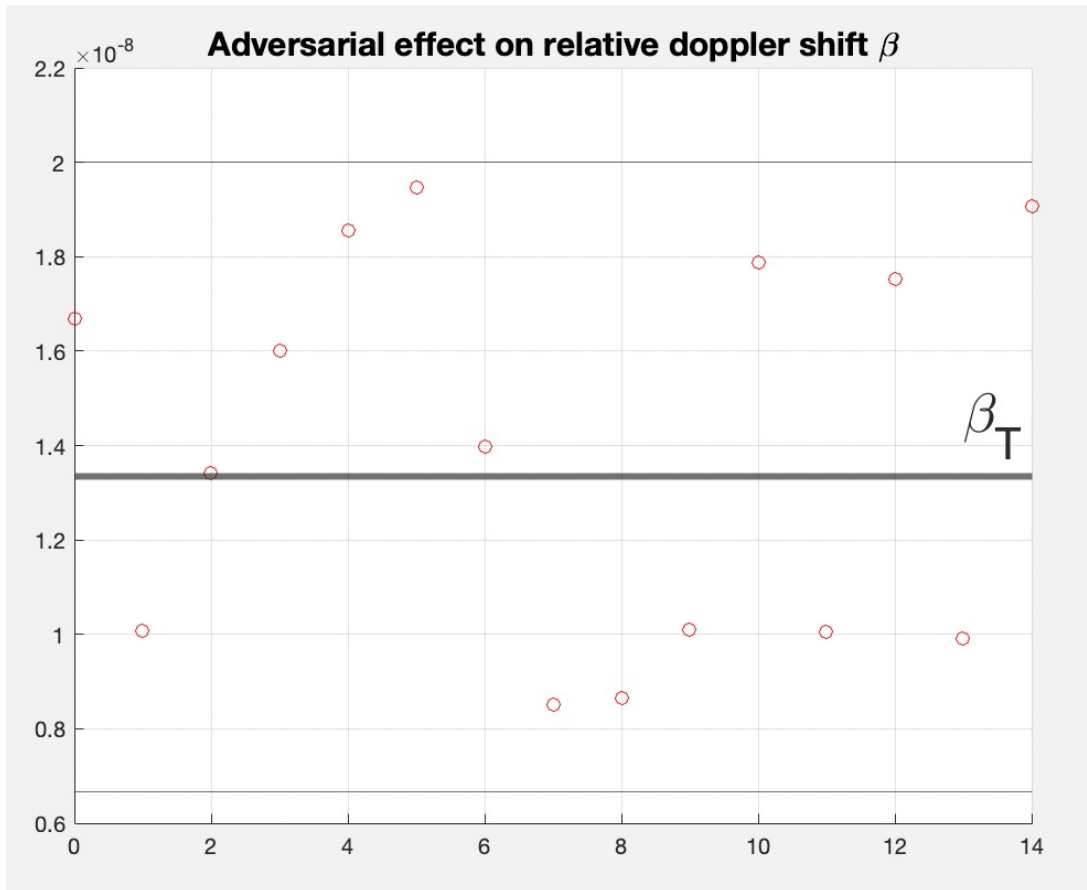
```

↗ This damages the AWD module!

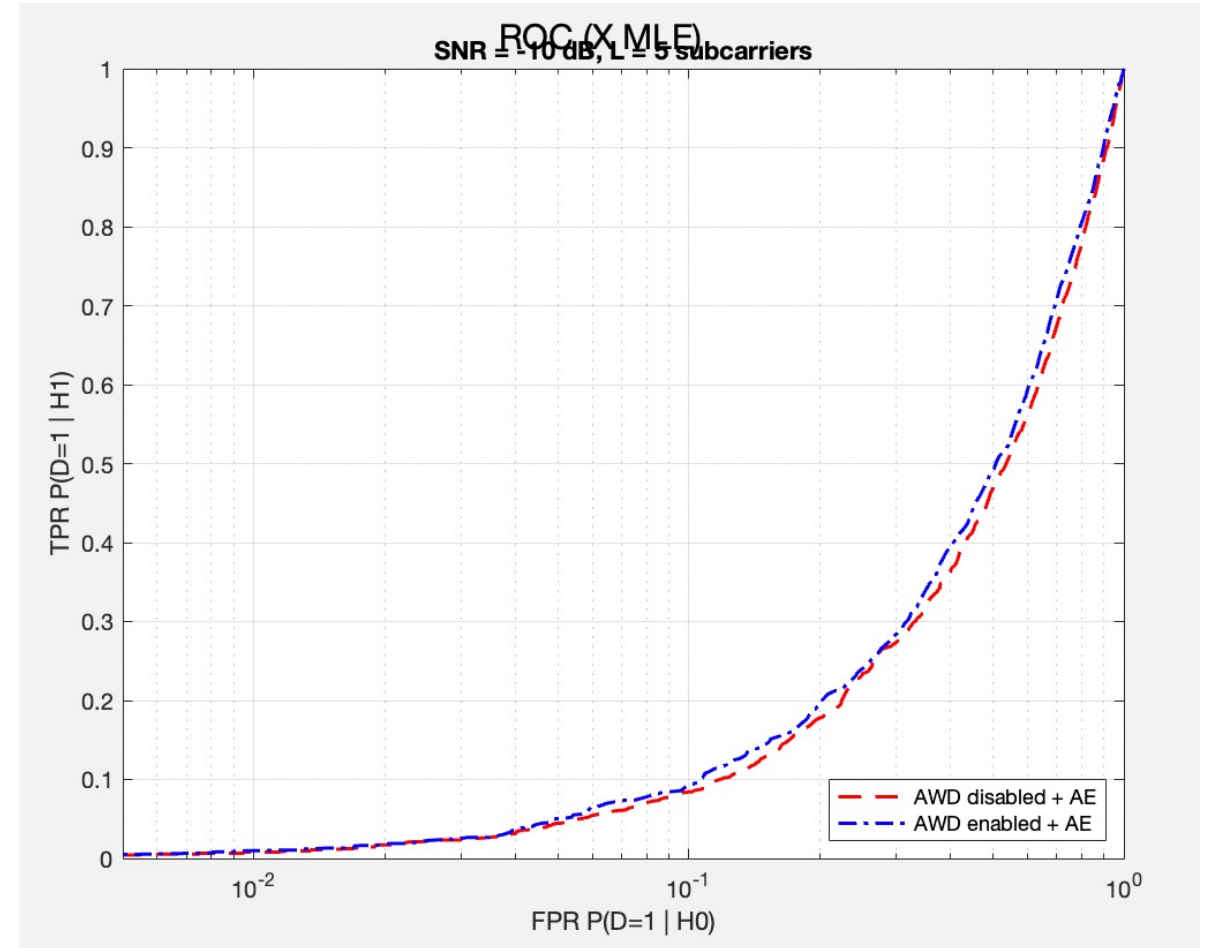
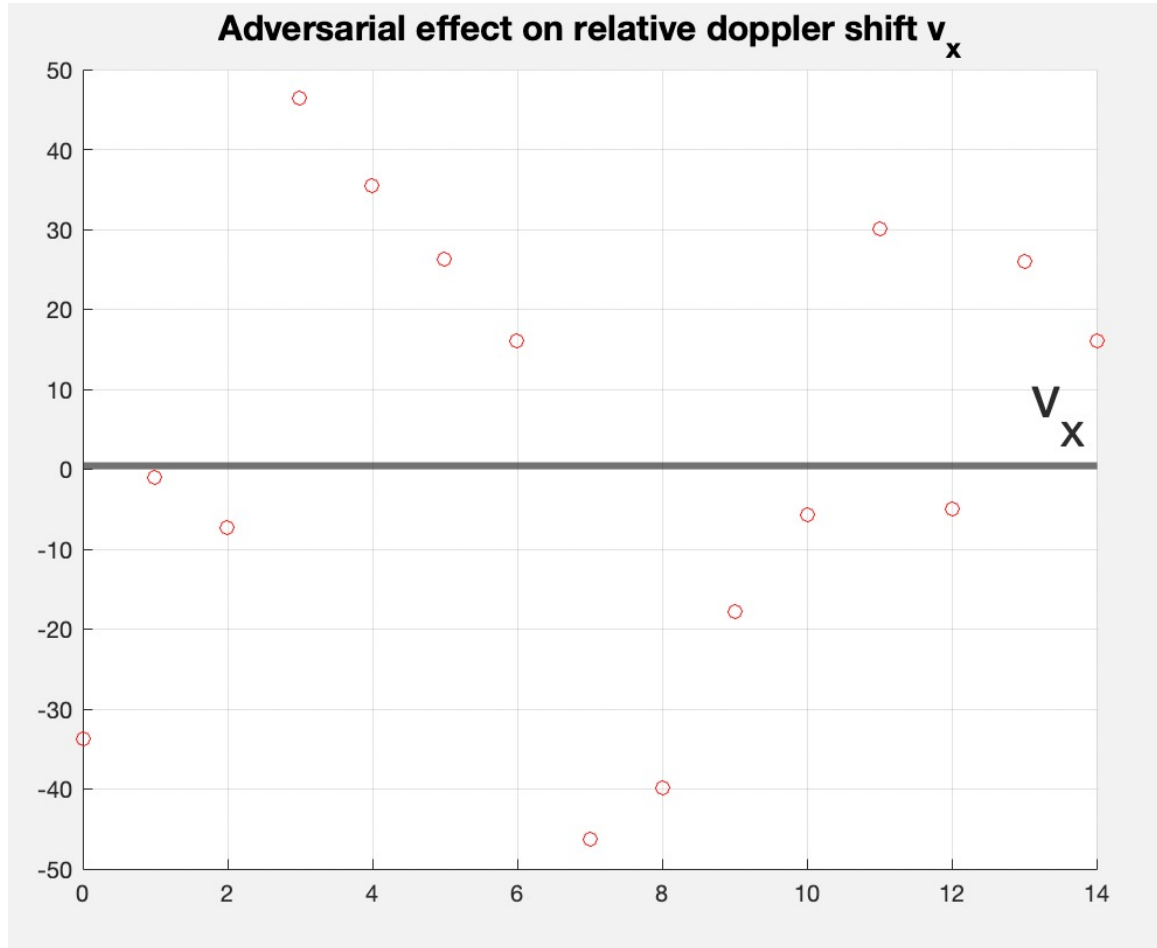
Adversarial effect on rel. doppler shift (1)

$$\beta_i = \beta_i + U(-1, 1) * \alpha * \beta_t$$

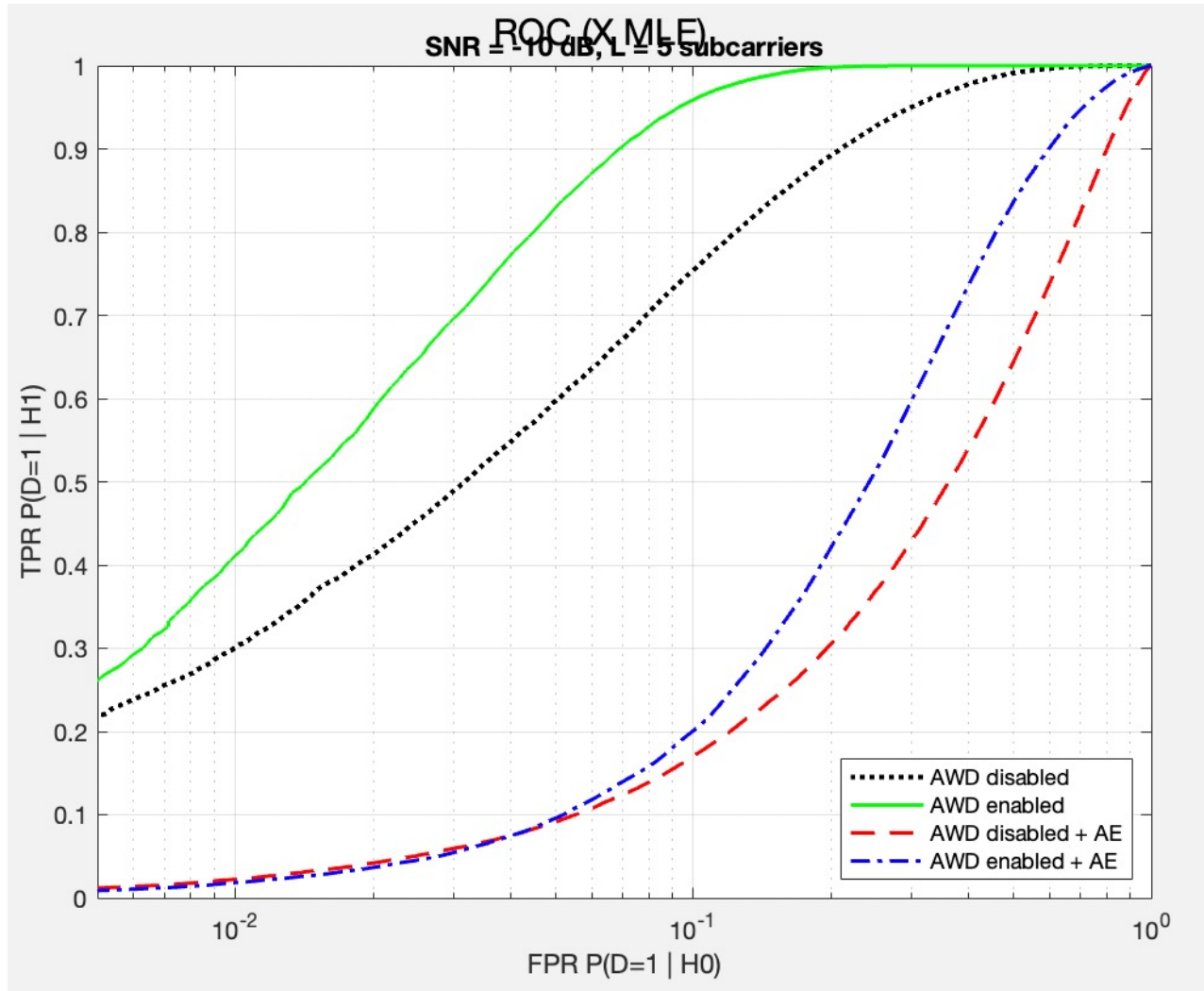
non-interference case



Changing target velocity



Adversarial effect on rel. doppler shift (2)



When adversarial effects are added, the AWD helps. Nevertheless, the performance is clearly disturbed even when enabling the AWD.

Our objective is to propose a method that can do that.

Conclusion/how to proceed:

- I need to make the parameter α too large to observe a strong adversarial effect. How does this translate in terms of velocity? Asking this to see if $\alpha=1000$ is realistic.
- To check: AWD should not work because all subcarriers are equally disturbed.
- To think about: AWD would not work even if few subcarriers were disturbed because the AWD module uses Φ_{TRUE} to optimize A , while the actual Φ is $\Phi_{\text{INTERFERENCE}}$.

