



Faculty of Engineering  
Computer Department  
Communications (ELC 325B) – Spring 2023



## Assignment 3

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## 1. Part One

### 1.1 Gram-Schmidt Orthogonalization

Here we compute the orthonormal bases function of inputs  $s_1$ ,  $s_2$

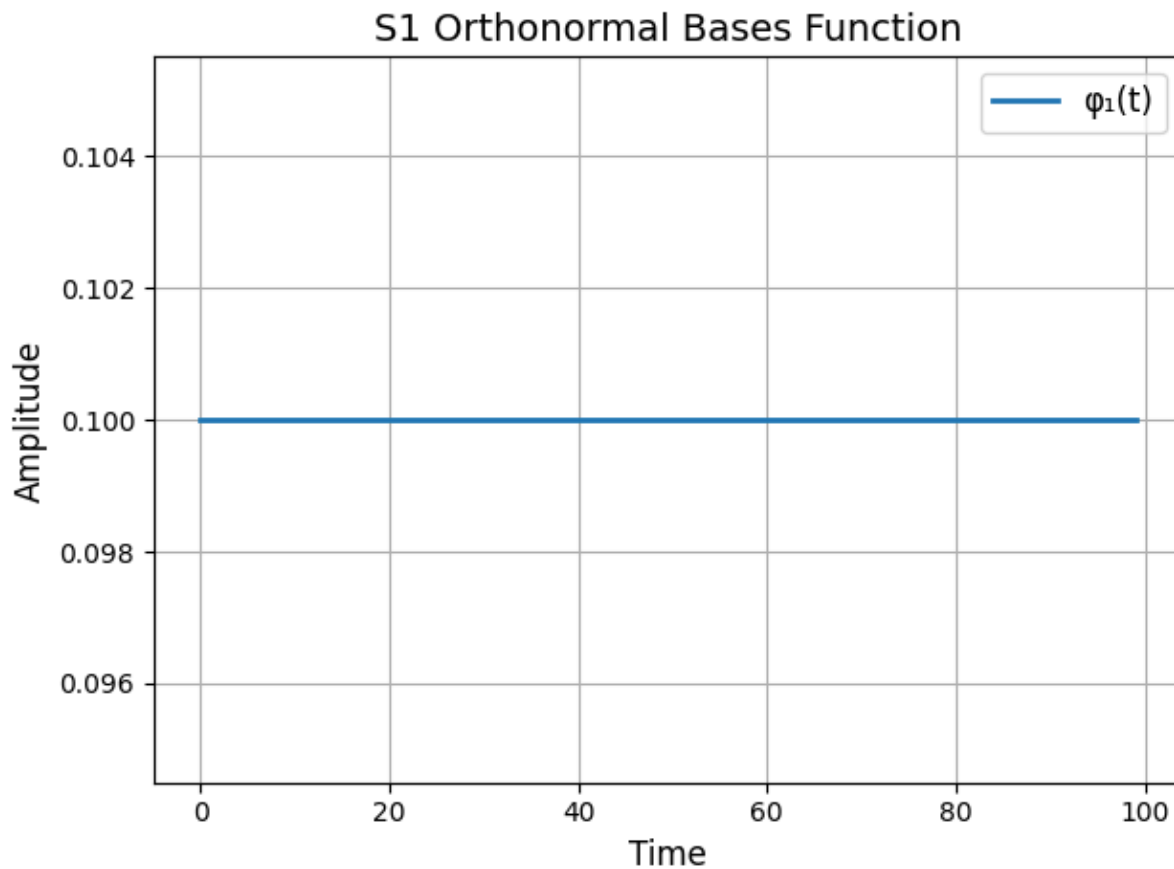


Figure 1  $\Phi_1$  VS time after using the GM\_Bases function

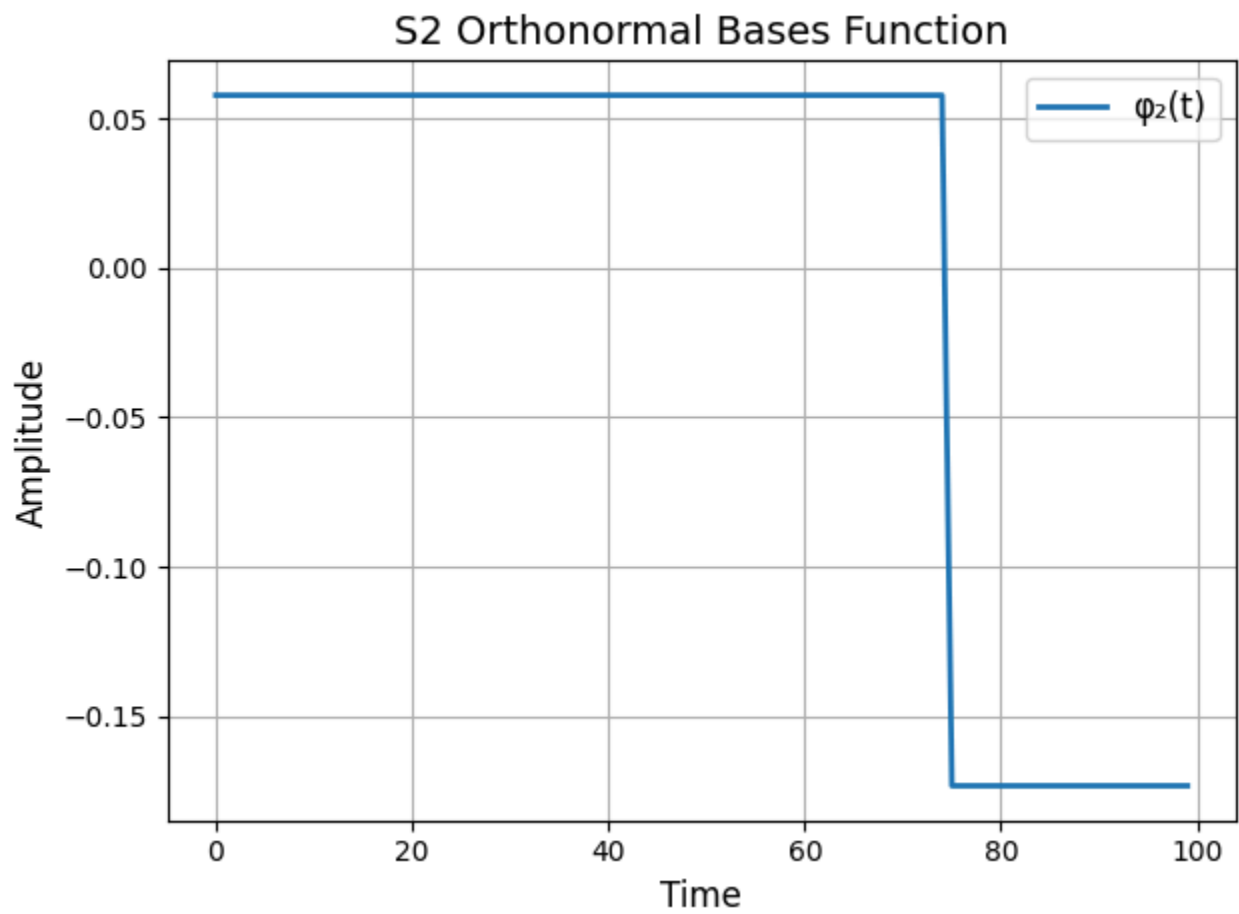


Figure 2  $\Phi_2$  VS time after using the GM\_Bases function



## 1.2 Signal Space Representation

Here we represent the signals using the base functions.

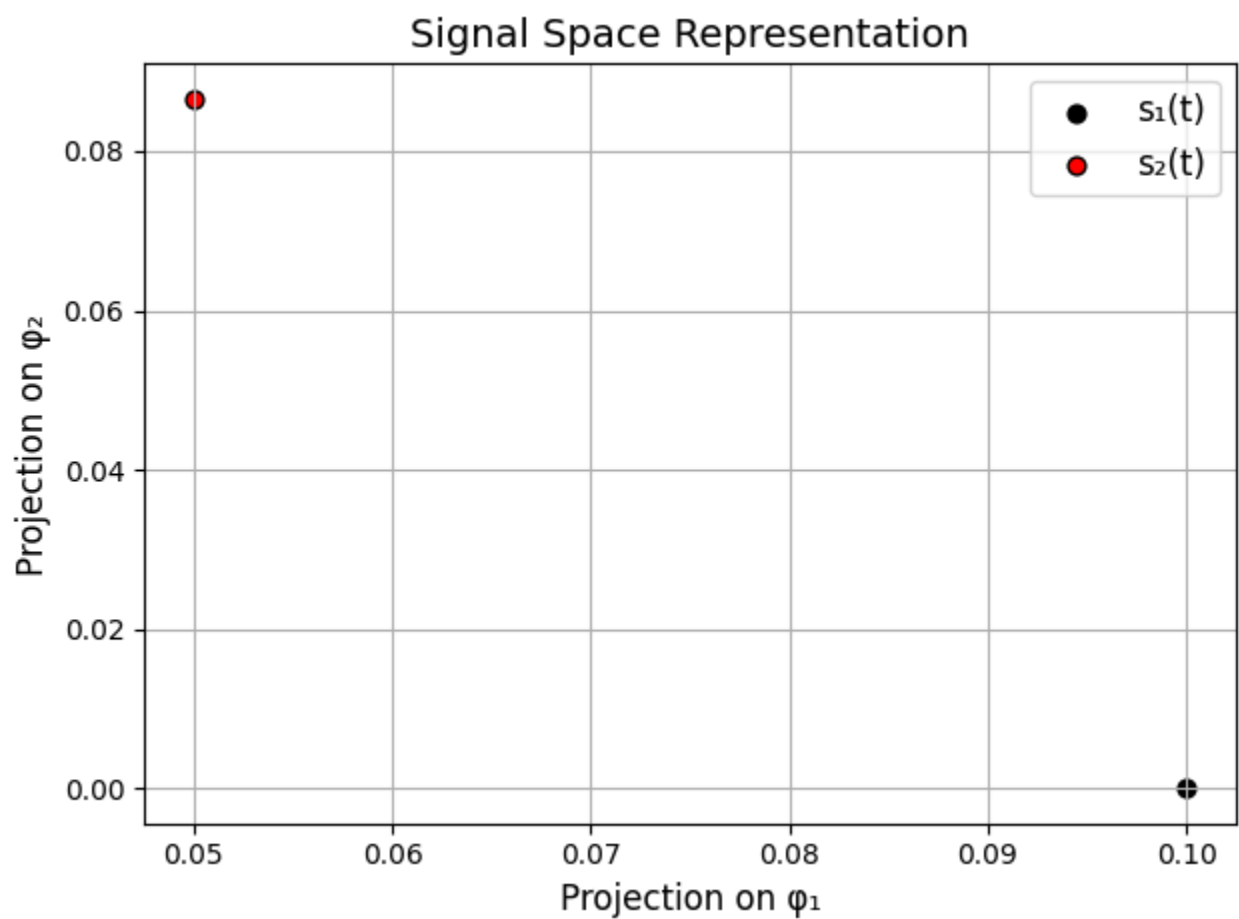


Figure 3 Signal Space representation of signals  $s_1, s_2$



### 1.3 Signal Space Representation with adding AWGN

the expected real points will be solid and the received will be hollow

**Case 1:**  $10 \log(E/\sigma^2) = -5 \text{ dB}$

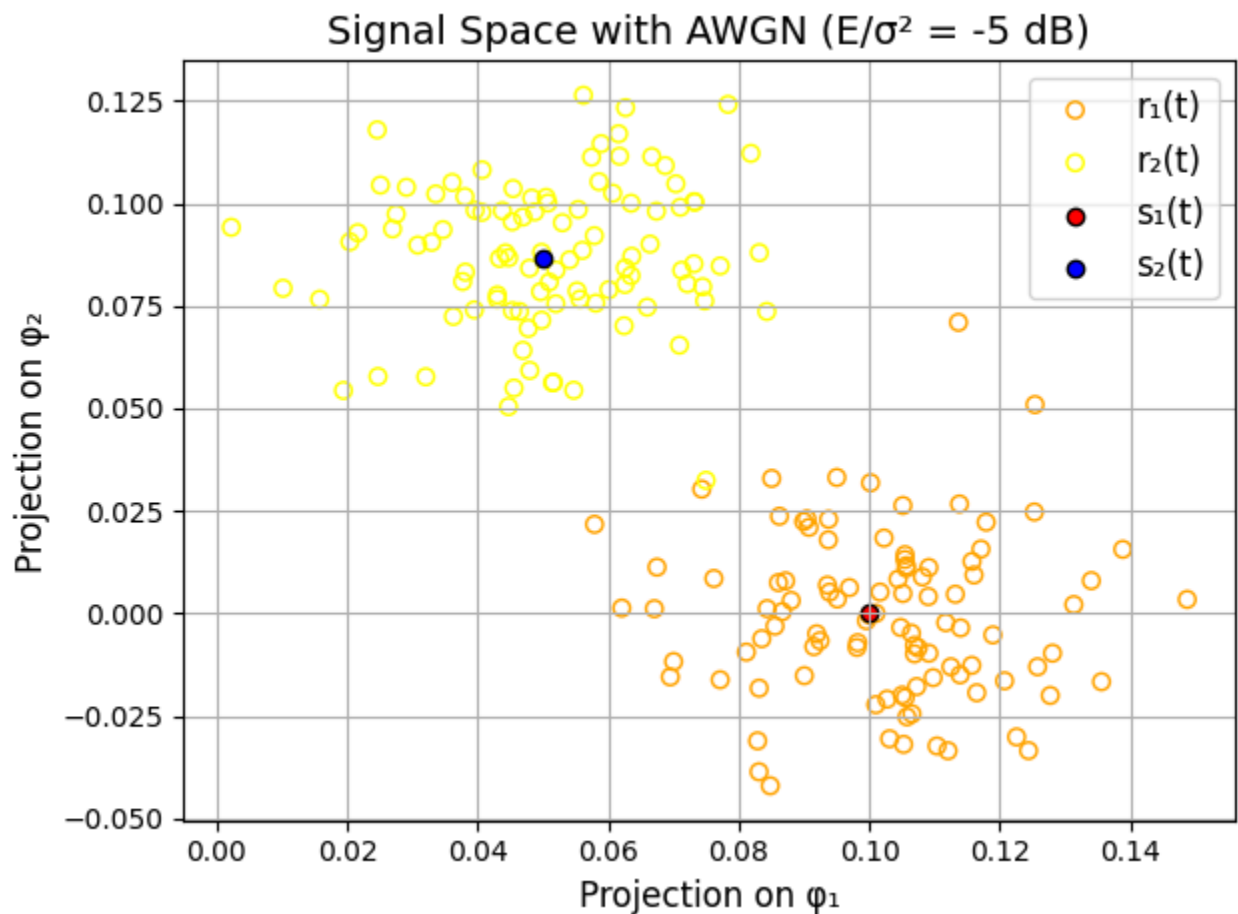
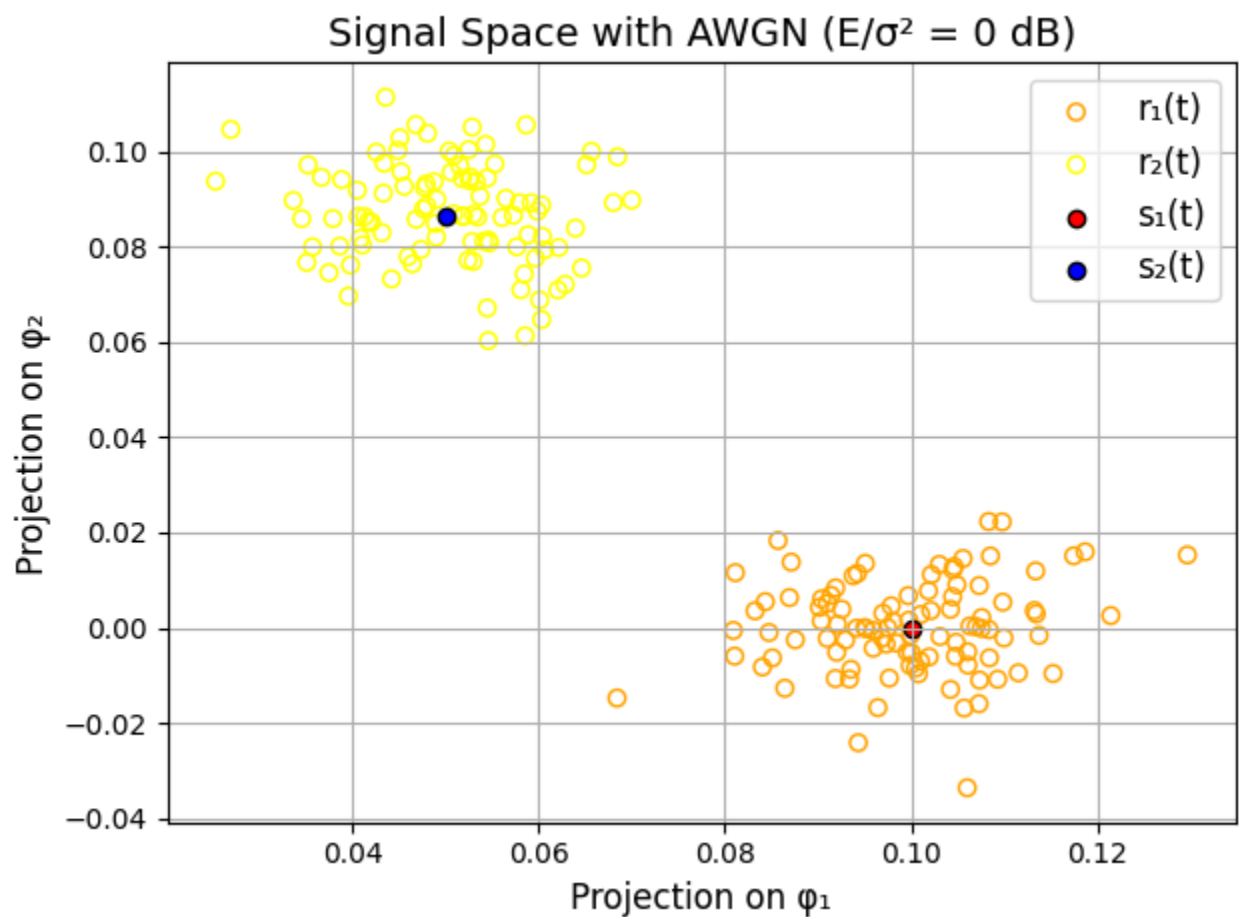


Figure 4 Signal Space representation of signals  $s_1, s_2$  with  $E/\sigma^2 = -5 \text{ dB}$



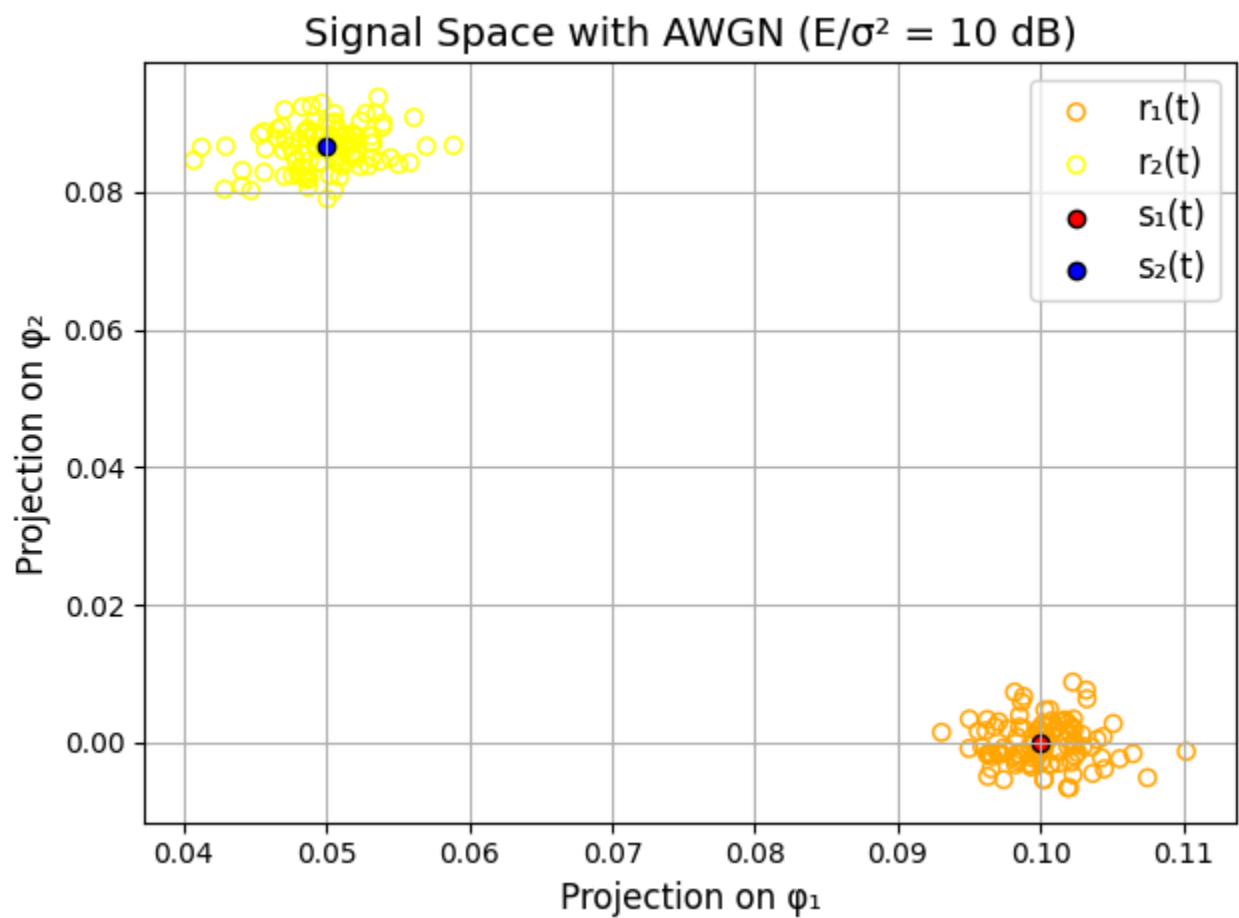
**Case 2:**  $10 \log(E/\sigma^2) = 0 \text{ dB}$



**Figure 5** Signal Space representation of signals  $s_1, s_2$  with  $E/\sigma^2 = 0 \text{ dB}$



**Case 3:**  $10 \log(E/\sigma^2) = 10 \text{ dB}$



**Figure 6** Signal Space representation of signals  $s_1, s_2$  with  $E/\sigma^2 = 10 \text{ dB}$





## 1.4 Noise Effect on Signal Space

The noise affects the signal space point by starting to have variations around the base point.

When increasing  $\sigma^2$ , the noise effect increases. We will not be able to determine the signal because of all the error.

### Appendix A: Codes:

#### A.1 Imports & Global Variables

```
1 import numpy as np
2 import matplotlib.pyplot as plt
3
4 SAMPLES = 100
5
6 s1 = np.ones(SAMPLES)
7 s2 = np.ones(SAMPLES)
8 s2[int(0.75 * SAMPLES):] = -1
9 E = np.sum(s1 ** 2) / SAMPLES
```



## A.2 Functions

```
1  def GM_Bases(s1, s2):
2      phi1 = s1 / np.linalg.norm(s1)
3      s2_proj_onto_phi1 = np.dot(s2, phi1) * phi1
4
5      u = s2 - s2_proj_onto_phi1
6      norm = np.linalg.norm(u)
7      phi2 = np.zeros_like(s2) if norm == 0 else u / norm
8
9      return phi1, phi2
10
11 def signal_space(s, phi1, phi2):
12     v1 = np.dot(s, phi1) / SAMPLES
13     v2 = np.dot(s, phi2) / SAMPLES
14     return v1, v2
15
16 def add_noise(s, sigma2):
17     noise = np.random.normal(0, np.sqrt(sigma2), s.shape)
18     return s + noise
```



## A.3 Plotting

```
1 def plot(title, signals, labels, xlabel="Time", ylabel="Amplitude"):  
2     plt.figure()  
3     for i in range(len(signals)):  
4         plt.plot(signals[i], label=labels[i], linewidth=2)  
5     plt.xlabel(xlabel, fontsize=12)  
6     plt.ylabel(ylabel, fontsize=12)  
7     plt.legend(fontsize=12)  
8     plt.title(title, fontsize=14)  
9     plt.grid(True)  
10    plt.tight_layout()  
11    plt.show()  
12  
13 def plot_scatter(title, xlabel, ylabel, signals, labels):  
14     plt.figure()  
15     colors = ['orange', 'yellow', 'red', 'blue']  
16     for i in range(len(signals)):  
17         if i >= len(signals) - 2:  
18             plt.scatter(signals[i][0], signals[i][1], label=labels[i],  
19                         color=colors[i], edgecolors='k')  
20         else:  
21             plt.scatter(signals[i][0], signals[i][1], label=labels[i],  
22                         facecolors='none', edgecolors=colors[i])  
23     plt.xlabel(xlabel, fontsize=12)  
24     plt.ylabel(ylabel, fontsize=12)  
25     plt.legend(fontsize=12)  
26     plt.title(title, fontsize=14)  
27     plt.grid(True)  
28     plt.tight_layout()  
29     plt.show()  
30  
31 phi1, phi2 = GM_Bases(s1, s2)  
32 plot("S1 Orthonormal Bases Function", [phi1], [" $\phi_1(t)$ "])  
33 plot("S2 Orthonormal Bases Function", [phi2], [" $\phi_2(t)$ "])  
34  
35 v1_s1, v2_s1 = signal_space(s1, phi1, phi2)  
36 v1_s2, v2_s2 = signal_space(s2, phi1, phi2)  
37 plot_scatter("Signal Space Representation", "Projection on  $\phi_1$ ", "Projection on  $\phi_2$ ",  
38             [(v1_s1, v2_s1), (v1_s2, v2_s2)], [" $s_1(t)$ ", " $s_2(t)$ "])
```



## A.4 Effect of noise on the Signal space Representations

```
1  for snr_db in [-5, 0, 10]:
2      sigma2 = E / (10 ** (snr_db / 10))
3      r1_points = []
4      r2_points = []
5
6      for _ in range(SAMPLES):
7          r1 = add_noise(s1, sigma2)
8          r2 = add_noise(s2, sigma2)
9          r1_points.append(signal_space(r1, phi1, phi2))
10         r2_points.append(signal_space(r2, phi1, phi2))
11
12     r1_points = np.array(r1_points)
13     r2_points = np.array(r2_points)
14     plot_scatter(f"Signal Space with AWGN ( $E/\sigma^2 = \{snr\_db\}$  dB)",
15                 "Projection on  $\phi_1$ ", "Projection on  $\phi_2$ ",
16                 [(r1_points[:,0], r1_points[:,1]), (r2_points[:,0], r2_points[:,1])],
17                 (v1_s1, v2_s1), (v1_s2, v2_s2)], ["r1(t)", "r2(t)", "s1(t)", "s2(t)"])
```