



Cairo University - Faculty Of Engineering  
Computer Engineering Department  
Digital Communication - Spring 2025



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# Digital Communications: Assignment 3 (Signal Space)

Submitted to

**Dr. Mai Badawi**

**Dr. Hala**

**Eng. Mohamed Khaled**

Submitted by

**Akram Hany Sec 1 BN 14**

**Amir Anwar Sec 1 BN 15**

# Contents

<b>1</b>	<b>Gram-Schmidt Orthogonalization:</b>	<b>2</b>
1.1	Original Signals . . . . .	2
1.2	Basis Functions . . . . .	3
<b>2</b>	<b>Part 2: Signal Space Representation</b>	<b>4</b>
2.1	Comment on the signal space representation . . . . .	4
<b>3</b>	<b>Part 3: Effect of AWGN on Signal Space</b>	<b>4</b>
3.1	SNR = -5 dB . . . . .	5
3.2	SNR = 0 dB . . . . .	5
3.3	SNR = 10 dB . . . . .	6
3.4	SNR = 15 dB . . . . .	6
<b>4</b>	<b>Part 4: Discussion</b>	<b>7</b>
<b>5</b>	<b>Appendix</b>	<b>8</b>
5.1	Python Code . . . . .	8

# List of Figures

1	Signal $s_1(t)$ . . . . .	2
2	Signal $s_2(t)$ . . . . .	2
3	Basis Function $\phi_1(t)$ . . . . .	3
4	Basis Function $\phi_2(t)$ . . . . .	3
5	Signal space projection of $s_1(t)$ and $s_2(t)$ . . . . .	4
6	AWGN Scatter Plot for $E/\sigma^2 = -5$ dB . . . . .	5
7	AWGN Scatter Plot for $E/\sigma^2 = 0$ dB . . . . .	5
8	AWGN Scatter Plot for $E/\sigma^2 = 10$ dB . . . . .	6
9	AWGN Scatter Plot for $E/\sigma^2 = 15$ dB . . . . .	6

# 1 Gram-Schmidt Orthogonalization:

We apply the Gram-Schmidt process to two input signals  $s_1(t)$  and  $s_2(t)$ .

## 1.1 Original Signals

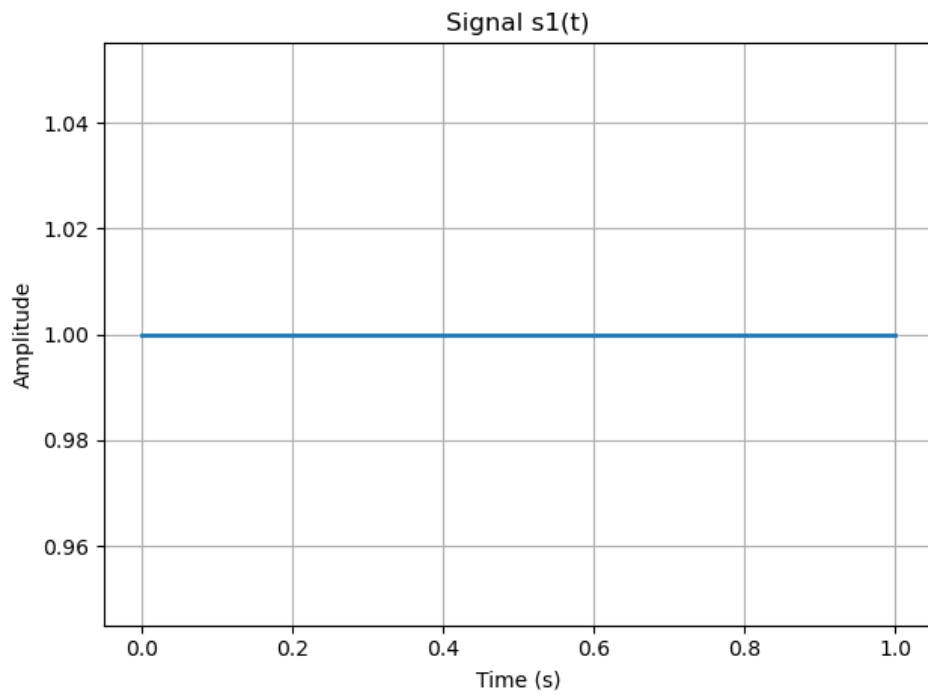


Figure 1: Signal  $s_1(t)$

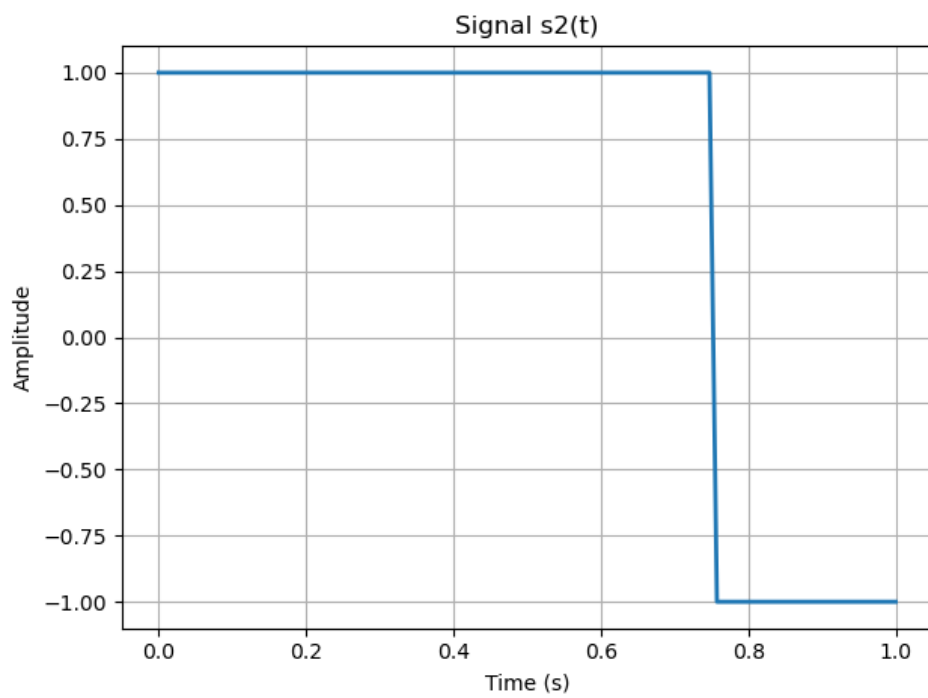


Figure 2: Signal  $s_2(t)$

## 1.2 Basis Functions

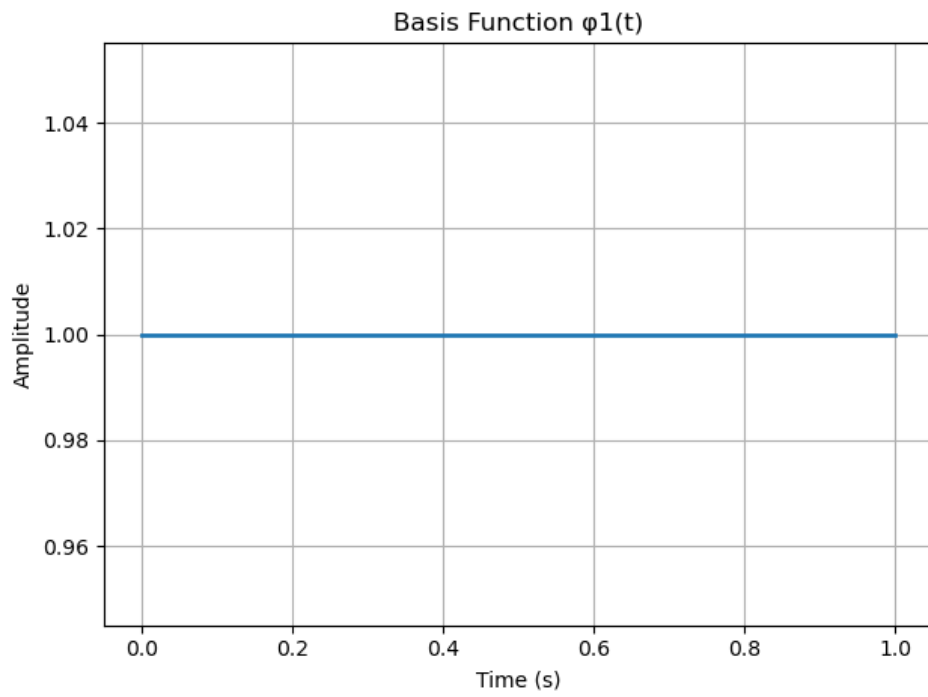


Figure 3: Basis Function  $\phi_1(t)$

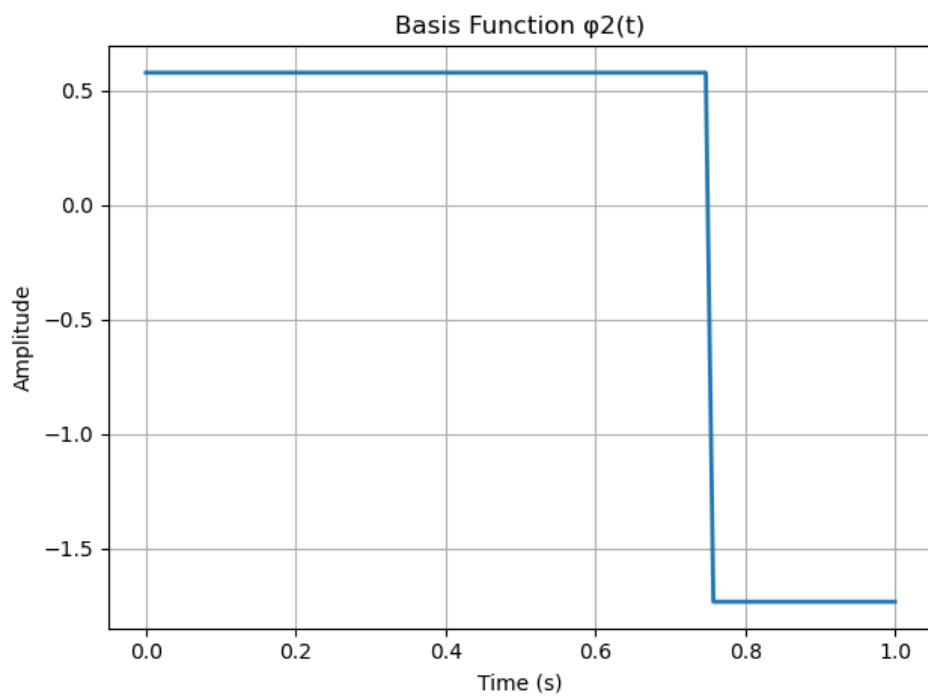


Figure 4: Basis Function  $\phi_2(t)$

## 2 Part 2: Signal Space Representation

Using the orthonormal basis, we project  $s_1(t)$  and  $s_2(t)$  into the signal space.

- $s_1$ :  $(v_1, v_2) \approx (1, 0)$
- $s_2$ :  $(v_1, v_2) \approx (0.5, 0.8)$

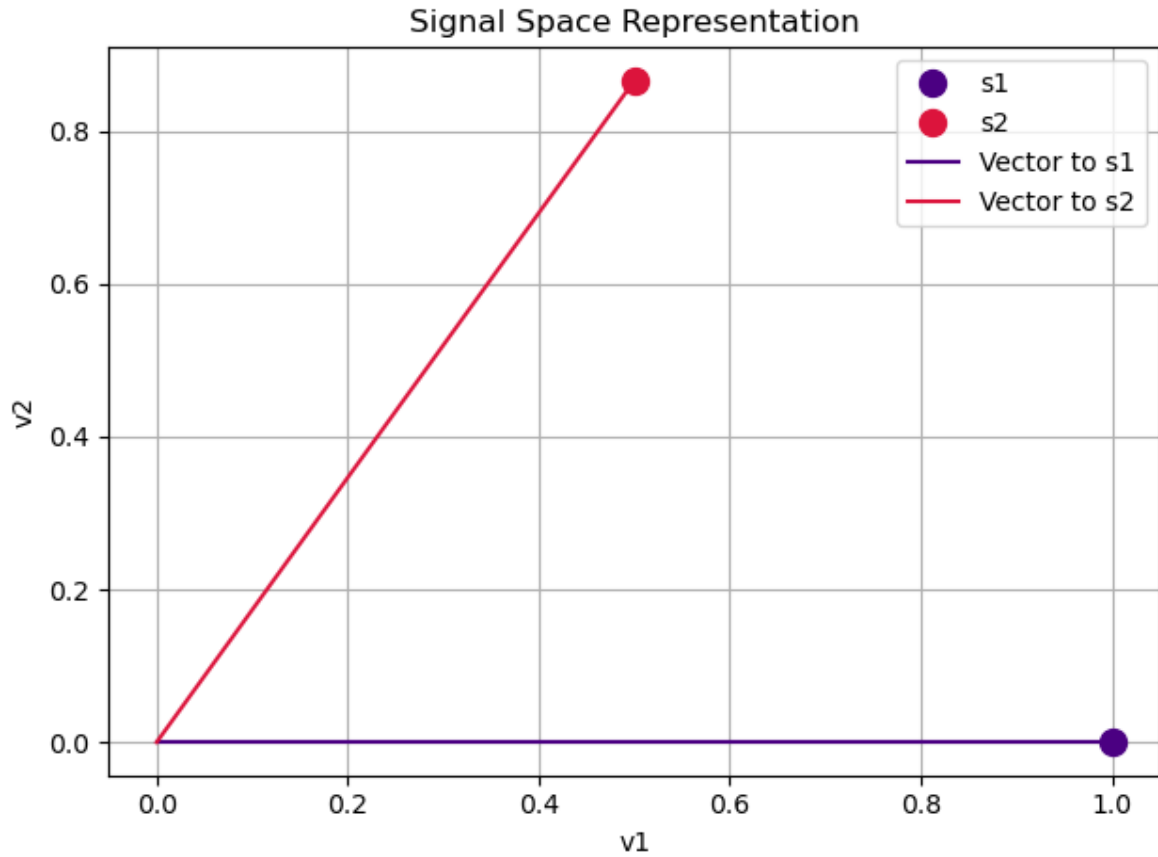


Figure 5: Signal space projection of  $s_1(t)$  and  $s_2(t)$

### 2.1 Comment on the signal space representation

As it shows  $S_1$  is aligned with basis that it looks like it completely with no projection on the other axis on the contrary  $S_2$  has projection on the 2 basis.

## 3 Part 3: Effect of AWGN on Signal Space

We add AWGN (additive white gaussian noise) to both signals and project 100 noisy versions each to signal space.

### 3.1 SNR = -5 dB

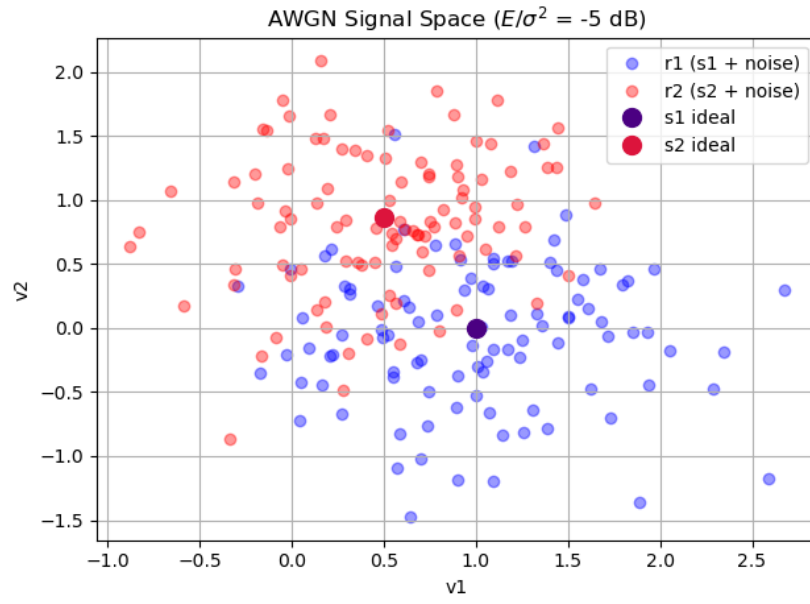


Figure 6: AWGN Scatter Plot for  $E/\sigma^2 = -5$  dB

### 3.2 SNR = 0 dB

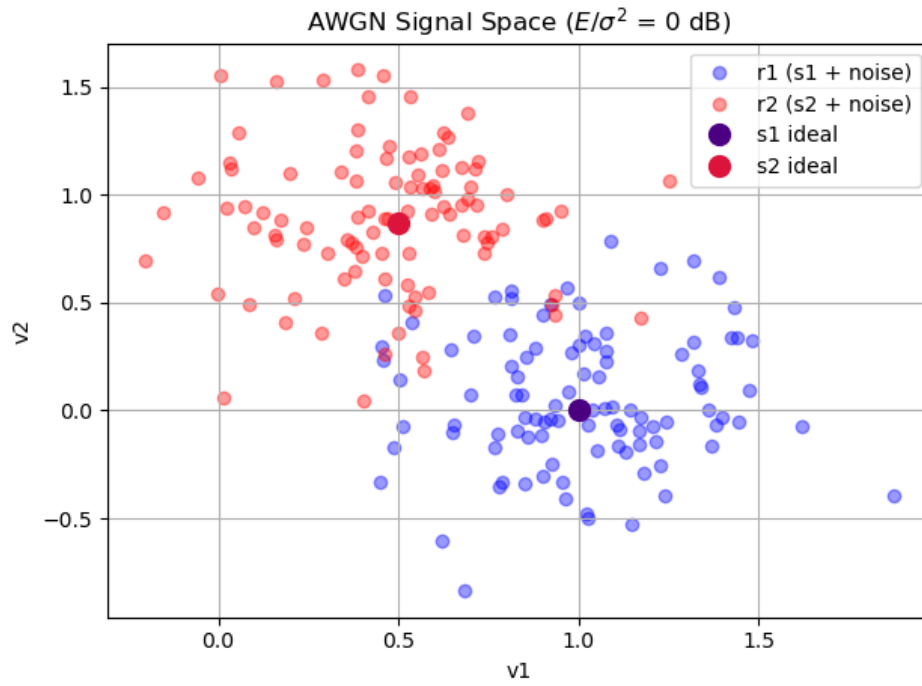


Figure 7: AWGN Scatter Plot for  $E/\sigma^2 = 0$  dB

### 3.3 SNR = 10 dB

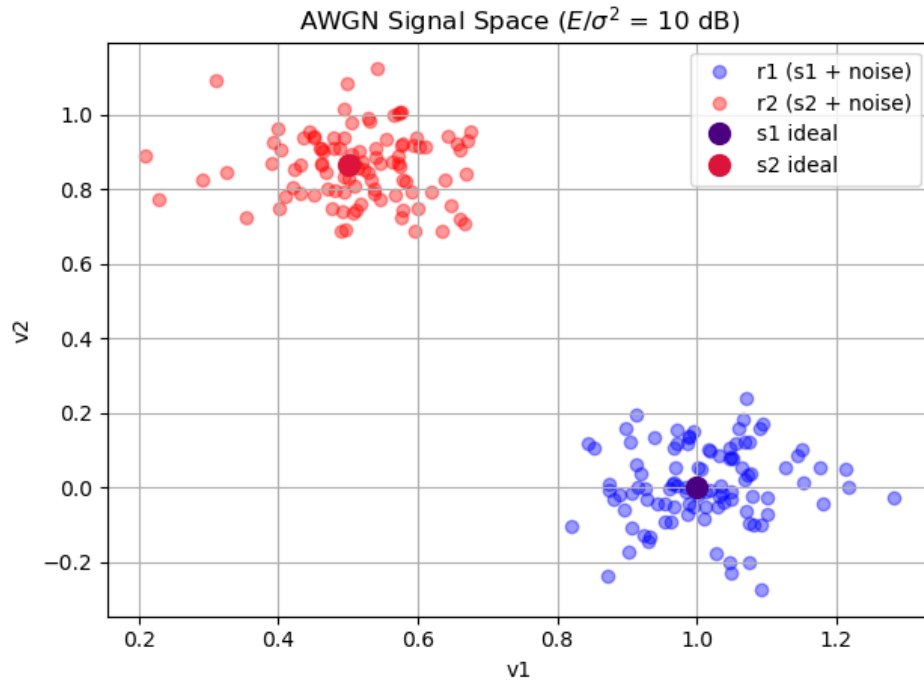


Figure 8: AWGN Scatter Plot for  $E/\sigma^2 = 10$  dB

### 3.4 SNR = 15 dB

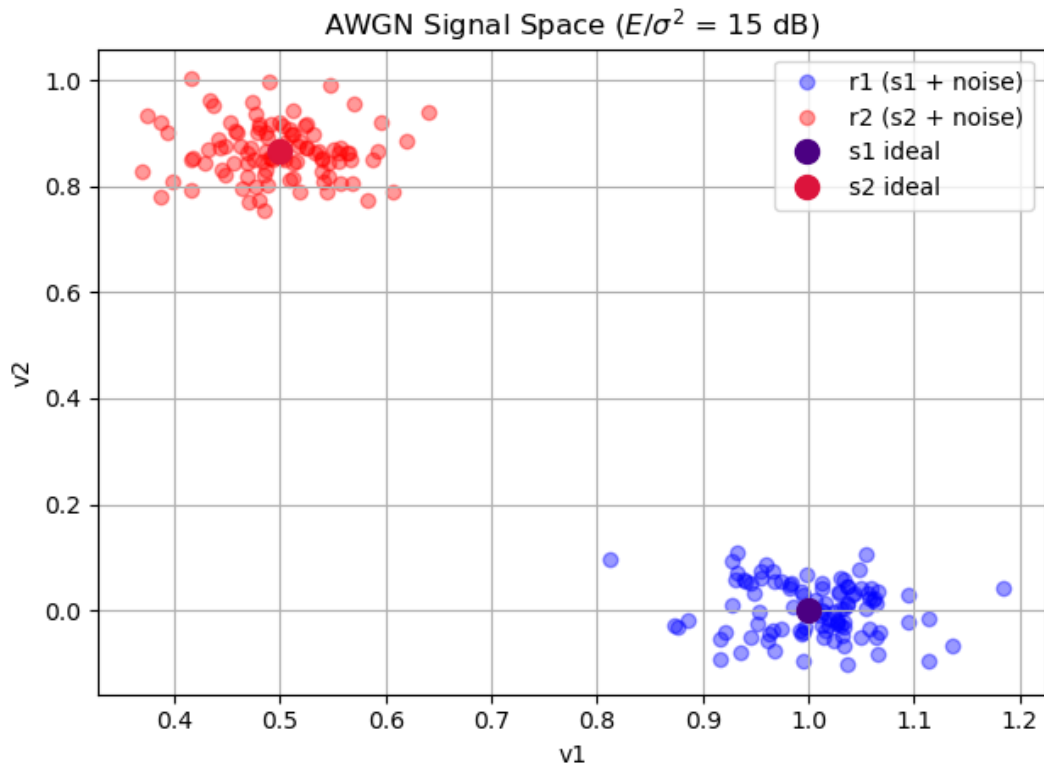


Figure 9: AWGN Scatter Plot for  $E/\sigma^2 = 15$  dB

Note I added  $\text{SNR} = 15$  dB because 10 dB was not enough to show the separation.

## 4 Part 4: Discussion

**How does noise affect the signal space?**

Noise introduces variation around the signal space points. The clusters of noisy projections around  $s_1$  and  $s_2$  become wider as noise increases.

**Does noise effect increase or decrease with increasing  $\sigma^2$ ?**

As  $\sigma^2$  increases (i.e., lower SNR), noise spreads the points more, making signals less distinguishable and increasing error probability. So, **the noise effect increases with increasing  $\sigma^2$ .**



## 5 Appendix

### 5.1 Python Code

```
1 import matplotlib.pyplot as plt
2 import numpy as np
```

Listing 1: Imports and external dependencies

```
1 def gm_bases(s1, s2):
2     # applying 2 steps of Gram-Schmidt process
3     norm_s1 = np.linalg.norm(s1)
4     phi1 = s1 / norm_s1
5
6     proj = np.dot(s2, phi1) * phi1
7     ortho = s2 - proj
8
9     norm_ortho = np.linalg.norm(ortho)
10    phi2 = ortho / norm_ortho if norm_ortho != 0 else np.zeros_like(s1)
11
12    return phi1, phi2
```

Listing 2: Gram-Schmidt Orthogonalization (part 1)

```
1 def signal_space(s, phi1, phi2):
2     # NOTE: Simply project into phi1, phi2 axes (linear algebra)
3     v1 = np.dot(s, phi1) / NUMBER_SAMPLES
4     v2 = np.dot(s, phi2) / NUMBER_SAMPLES
5     return v1, v2
```

Listing 3: Signal Space representation: (part 2)

```
1 def add_awgn(signal, sigma2):
2     # NOTE: awgn: Additive White Gaussian Noise
3     noise = np.random.normal(0, np.sqrt(sigma2), signal.shape)
4     return signal + noise
```

Listing 4: AWGN: (part 3)

```
1 def plot_signal(t, signal, title, filename):
2     # Simple utility to not repeat this 4 times below :)
3     plt.figure()
4     plt.plot(t, signal, linewidth=2)
5     plt.title(title)
6     plt.xlabel("Time (s)")
7     plt.ylabel("Amplitude")
8     plt.grid(True)
9     plt.tight_layout()
10    plt.savefig(filename)
11    plt.close()
```

Listing 5: Utility function to plot a signal

```
1 def run_awgn_experiment(
2     s1, s2, phi1, phi2, num_samples=100, eb_sigma2_db_list=[-5, 0, 10, 15]
3 ):
4     # NOTE: s1, s2 are the signals, phi1, phi2 are the basis functions
5     # NOTE: eb_sigma2_db_list is the list of SNR values in dB ( $E_b/N$ )
6
7     Es = np.sum(s1**2)
8     Es /= np.sqrt(NUMBER_SAMPLES)
9
10    v1_s1, v2_s1 = signal_space(s1, phi1, phi2)
11    v1_s2, v2_s2 = signal_space(s2, phi1, phi2)
12    results = {}
13
14    for db in eb_sigma2_db_list:
15        sigma2 = Es / (10 ** (db / 10))
16        r1_points = []
17        r2_points = []
18
19        for _ in range(num_samples):
```

```

20     r1 = add_awgn(s1, sigma2)
21     r2 = add_awgn(s2, sigma2)
22     v1_r1, v2_r1 = signal_space(r1, phi1, phi2)
23     v1_r2, v2_r2 = signal_space(r2, phi1, phi2)
24     r1_points.append((v1_r1, v2_r1))
25     r2_points.append((v1_r2, v2_r2))
26
27     results[db] = (np.array(r1_points), np.array(r2_points))
28
29     # Plot
30     plt.figure()
31     # Plotting the noisy signal space
32     plt.scatter(*zip(*r1_points), label="r1 (s1 + noise)", alpha=0.7, color="blue")
33     plt.scatter(*zip(*r2_points), label="r2 (s2 + noise)", alpha=0.7, color="red")
34     # Plotting the ideal signal space
35     plt.plot(v1_s1, v2_s1, "o", label="s1 ideal", markersize=10, color="indigo")
36     plt.plot(v1_s2, v2_s2, "o", label="s2 ideal", markersize=10, color="crimson")
37
38     plt.xlabel("v1")
39     plt.ylabel("v2")
40     plt.title(f"AWGN Signal Space ( $E/\sigma^2 = \{db\}$  dB)")
41     plt.grid(True)
42     plt.legend()
43     plt.tight_layout()
44     plt.savefig(f"noisy_signal_space_{db}dB.png")
45     plt.close()
46
47     return results

```

Listing 6: Runs AWGN as described above in the beginning of part 3

```

1  if __name__ == "__main__":
2      # NOTE: our time space 1 is 100 samples for simplicity
3
4      t = np.linspace(0, 1, NUMBER_SAMPLES)
5
6      # signal one is all ones
7      s1 = np.ones(NUMBER_SAMPLES)
8
9      # signal two is 75 ones and 25 -1
10     s2 = np.concatenate(
11         [np.ones(int(0.75 * NUMBER_SAMPLES)), -1 * np.ones(int(0.25 * NUMBER_SAMPLES))]
12     )
13
14
15     plot_signal(t, s1, "Signal s1(t)", "signal_s1.png")
16     plot_signal(t, s2, "Signal s2(t)", "signal_s2.png")
17
18     phi1, phi2 = gm_bases(s1, s2)
19     phi1 = phi1 * scale
20     phi2 = phi2 * scale
21
22     plot_signal(t, phi1, "Basis Function 1 (t)", "basis_phi1.png")
23     plot_signal(t, phi2, "Basis Function 2 (t)", "basis_phi2.png")
24
25     # Signal space representation
26     v1_s1, v2_s1 = signal_space(s1, phi1, phi2)
27     v1_s2, v2_s2 = signal_space(s2, phi1, phi2)
28     plt.figure()
29
30     plt.plot(v1_s1, v2_s1, "o", label="s1", markersize=10, color="indigo")
31     plt.plot(v1_s2, v2_s2, "o", label="s2", markersize=10, color="crimson")
32
33     plt.plot([0, v1_s1], [0, v2_s1], "-", color="indigo", label="Vector to s1")
34     plt.plot([0, v1_s2], [0, v2_s2], "-", color="crimson", label="Vector to s2")
35
36
37     plt.xlabel("v1")
38     plt.ylabel("v2")
39     plt.title("Signal Space Representation")
40     plt.grid(True)
41     plt.legend()
42     plt.tight_layout()
43     plt.savefig("signal_space_clean.png")

```

```
44 plt.close()  
45  
46 run_awgn_experiment(s1, s2, phi1, phi2)
```

Listing 7: Main functions calles function above and plot the signales