



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



Digital Communications: Assignment 3 (Signal Space)

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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

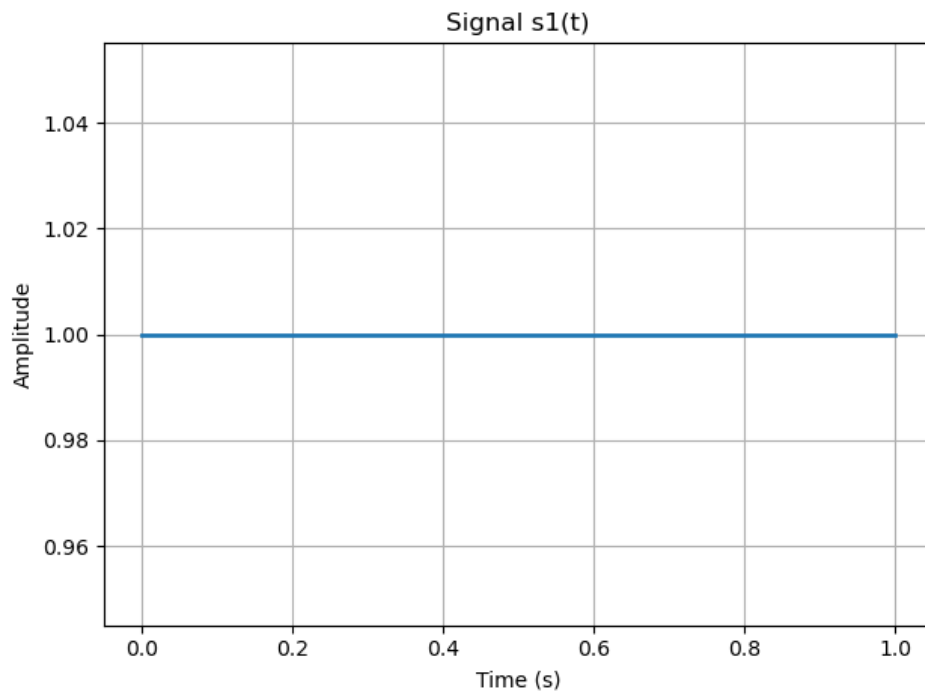


Figure 1: Signal $s_1(t)$

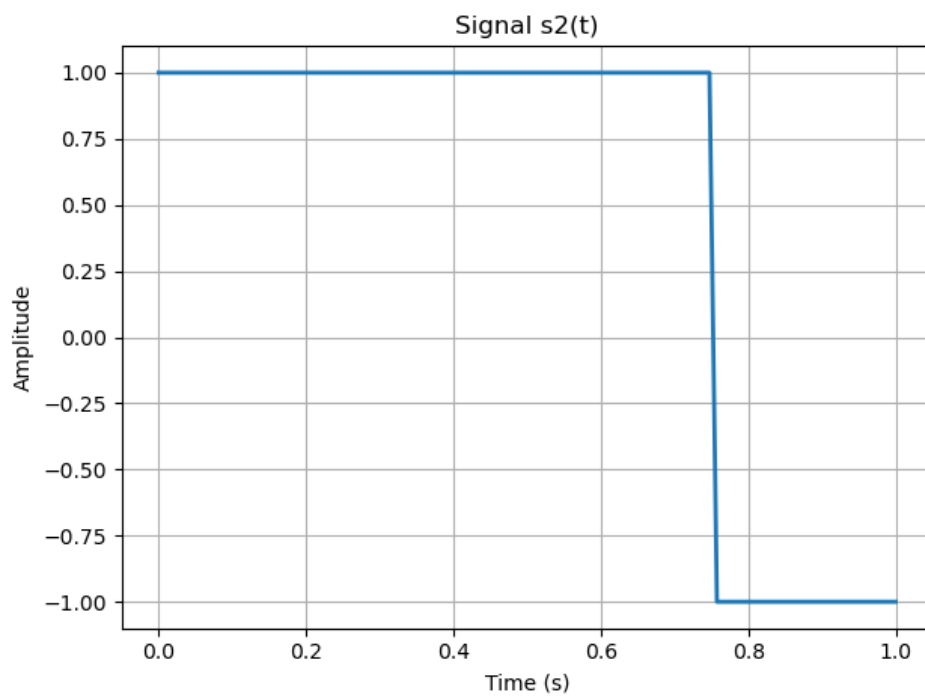


Figure 2: Signal $s_2(t)$

1.2 Basis Functions

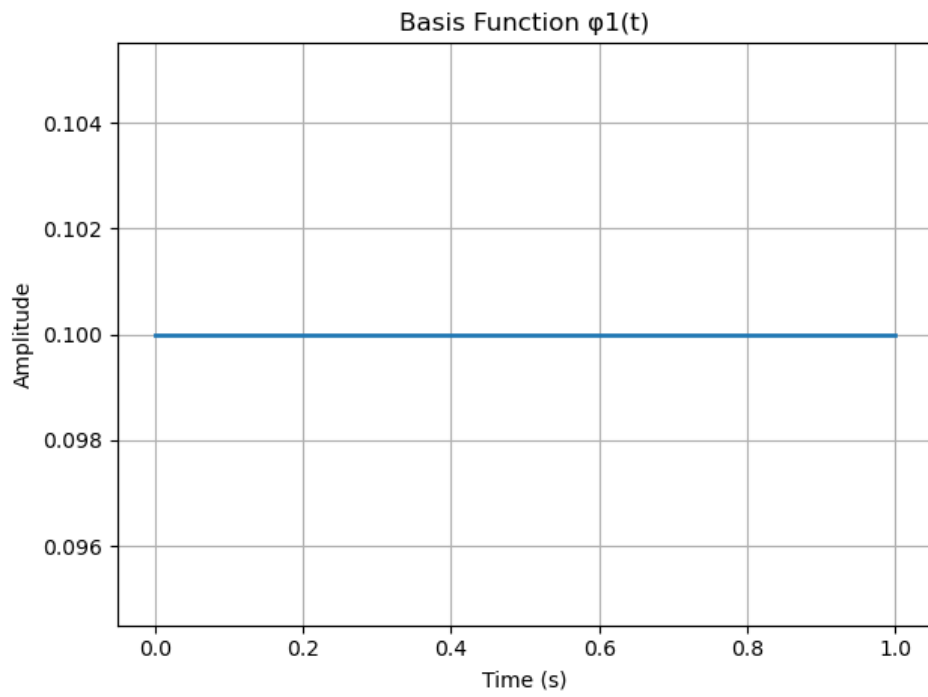


Figure 3: Basis Function $\phi_1(t)$

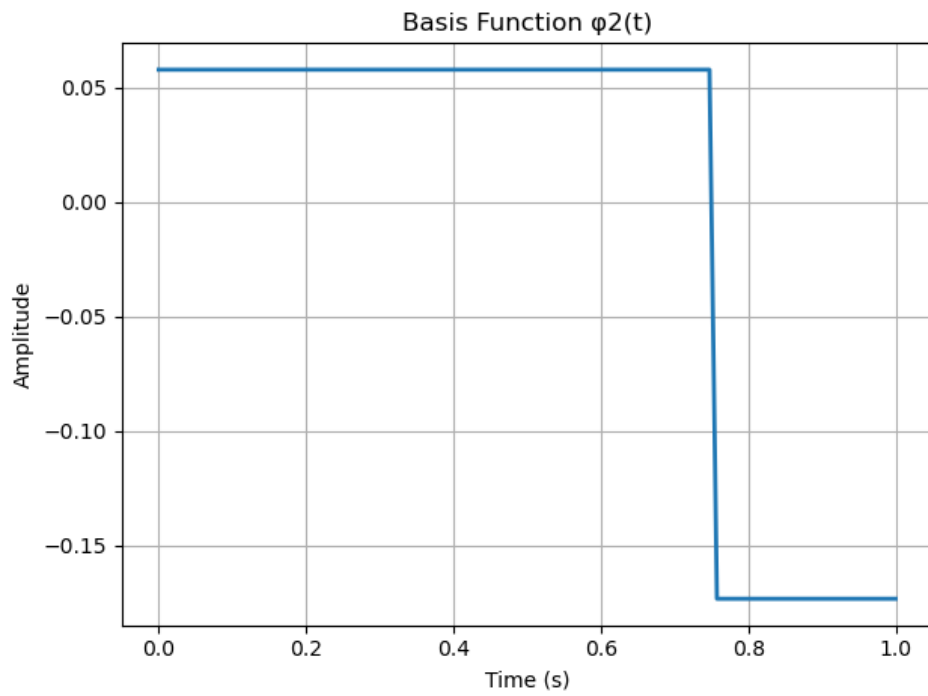


Figure 4: Basis Function $\phi_2(t)$

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 : $(v_1, v_2) \approx (10, 0)$
- s_2 : $(v_1, v_2) \approx (5, 8.5)$

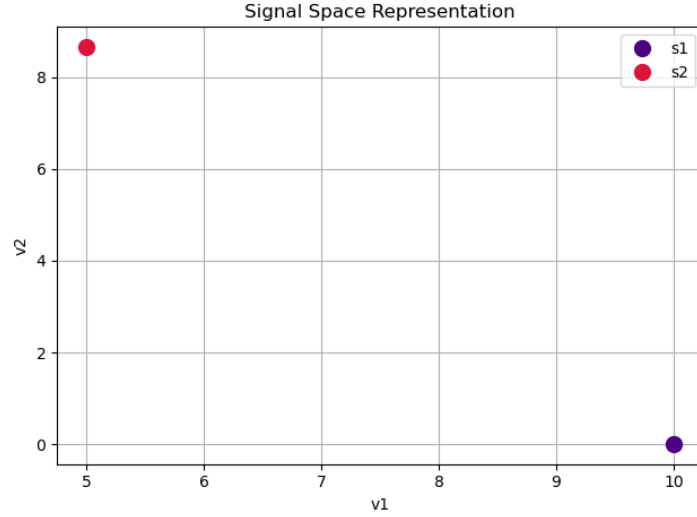


Figure 5: Signal space projection of $s_1(t)$ and $s_2(t)$

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

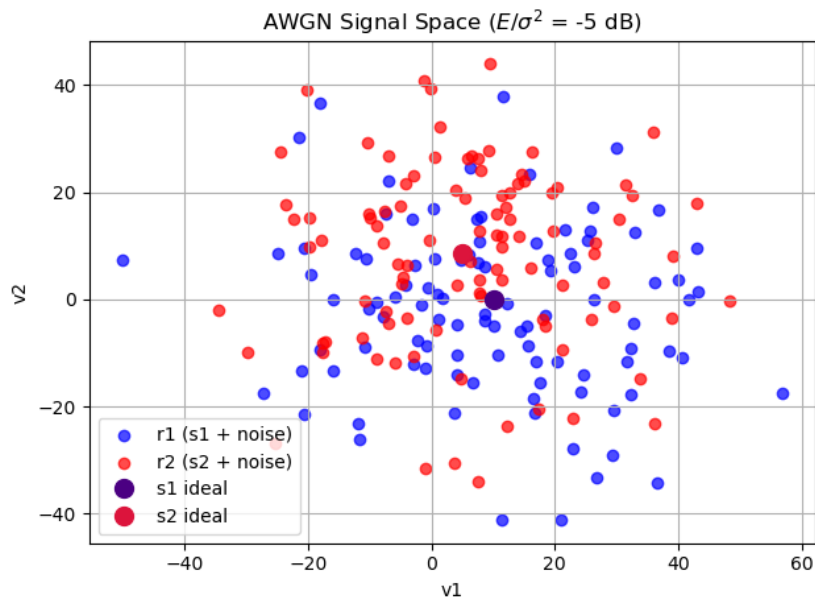


Figure 6: AWGN Scatter Plot for $E/\sigma^2 = -5$ dB

3.2 SNR = 0 dB

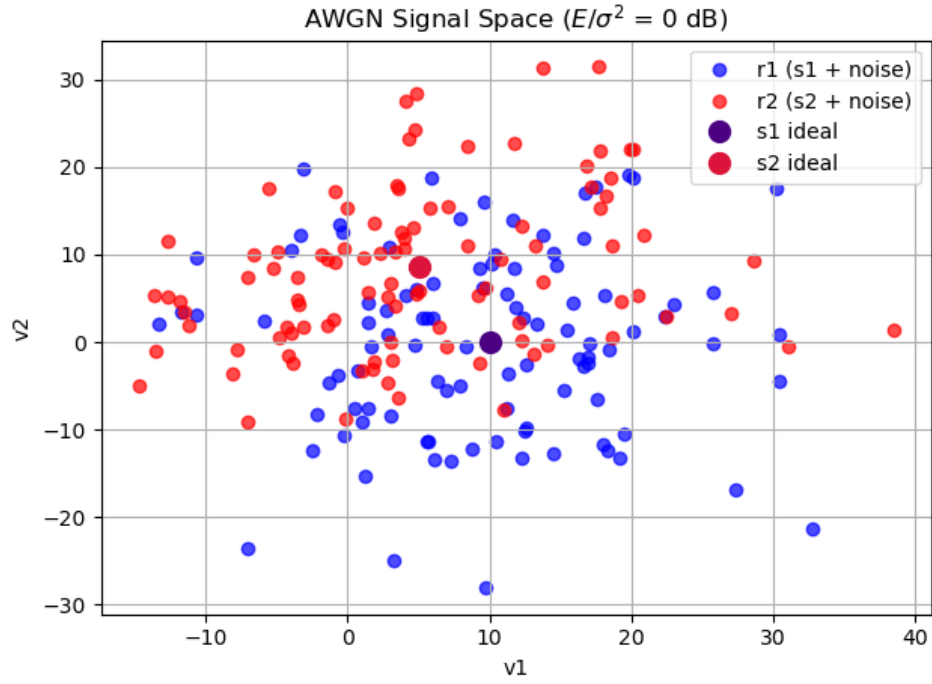


Figure 7: AWGN Scatter Plot for $E/\sigma^2 = 0$ dB

3.3 SNR = 10 dB

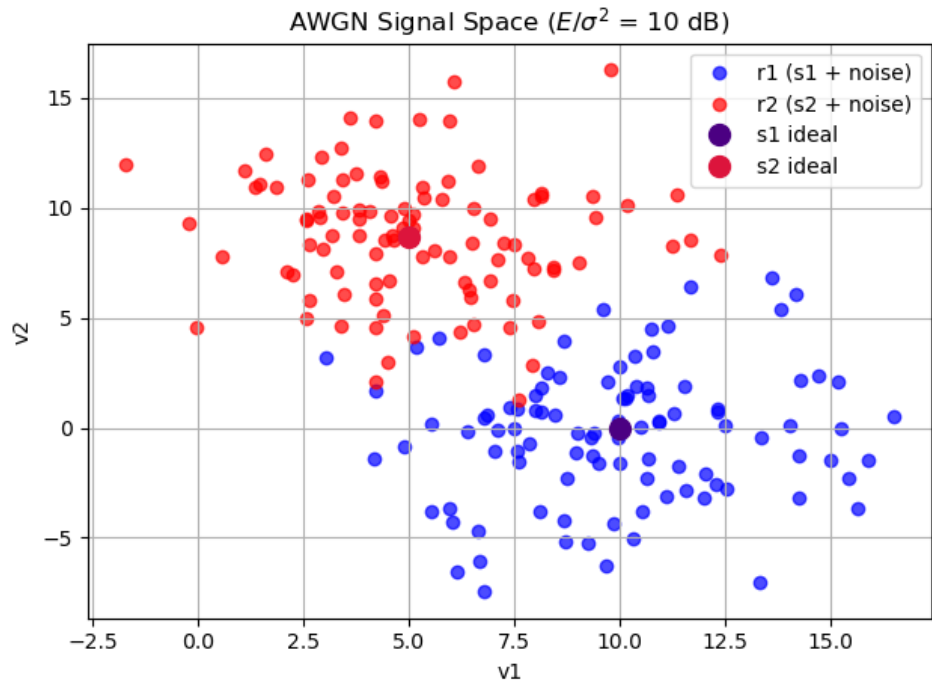


Figure 8: AWGN Scatter Plot for $E/\sigma^2 = 10$ dB

3.4 SNR = 15 dB

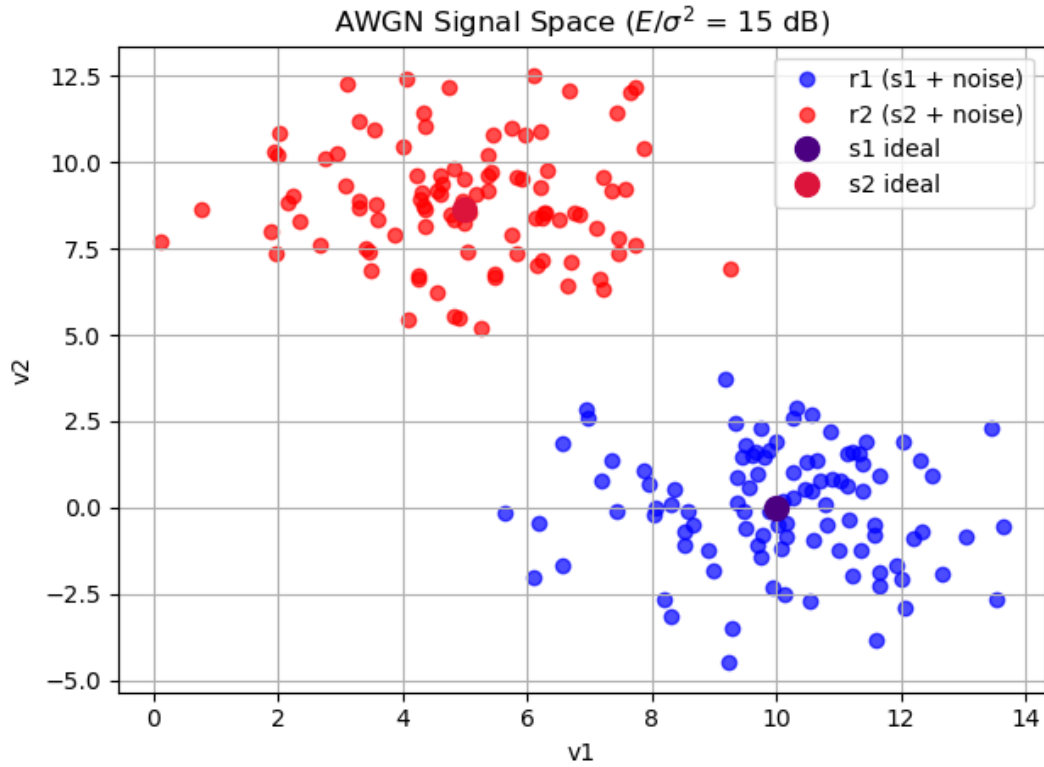


Figure 9: AWGN Scatter Plot for $E/\sigma^2 = 15$ dB

Note I added 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 σ^2 ?

As σ^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 σ^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)
4     v2 = np.dot(s, phi2)
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     v1_s1, v2_s1 = signal_space(s1, phi1, phi2)
9     v1_s2, v2_s2 = signal_space(s2, phi1, phi2)
10    results = {}
11
12    for db in eb_sigma2_db_list:
13        sigma2 = Es / (10 ** (db / 10))
14        r1_points = []
15        r2_points = []
16
17        for _ in range(num_samples):
18            r1 = add_awgn(s1, sigma2)
19            r2 = add_awgn(s2, sigma2)
```

```

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

```

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

```

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

```

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