

EXPERIMENT NUMBER - 7

EXPERIMENT NAME - PROBABILITY OF ERROR (PAM)

DATE - 01/12/2022, THURSDAY

* AIM:

To find probability of errors in Pulse Amplitude Modulation (PAM) and plot the graphs.

* SOFTWARE REQUIRED:

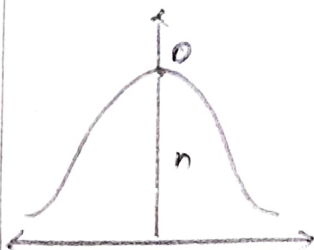
- ① Anaconda3 2021.11 (Python 3.9.7 64-bit)
- ② Spyder 5.1.5 Integrated Development Environment (IDE)

* THEORY:

Pulse Amplitude Modulation (PAM) are also called as antipodal signals and they can be represented using one basis function. When you transmit either '0' or '1', it is called as binary PAM.

Bit	Transmitted	$x(t)$	$y(t)$ [demod]
0	$\sqrt{E_b} \psi(t)$	$\sqrt{E_b} \psi(t) + n(t)$	$\sqrt{E_b} + n_1$
1	$-\sqrt{E_b} \psi(t)$	$-\sqrt{E_b} \psi(t) + n(t)$	$-\sqrt{E_b} + n_1$

It follows Gaussian distribution and it is random since, it is Gaussian for noise too, $E(n) = 0$.



mean, median, mode lies at same point and it is symmetric.

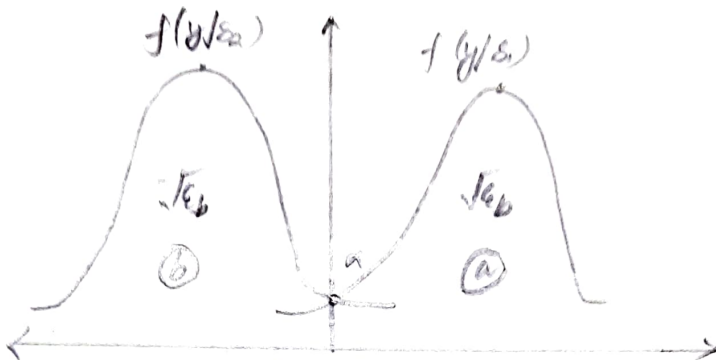
$$\text{PDF of Gaussian} = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(y-\sqrt{E_b})^2}{2\sigma^2}}$$

[Transmitted bit = 0]
for PAM, $\mu = \sqrt{E_b}$

If $\mu = -\sqrt{E_b}$.

$$PDF = \frac{1}{\sqrt{2\pi}\sigma^2} e^{-\frac{(y - (-\sqrt{E_b}))^2}{2N_0/2}}$$

[Transmitted bit = 1]



Region a: $f(y/s_1)$ dominating

Region b: $f(y/s_2)$ dominating

Possible Errors -

- Value that belongs to $f(y/s_2)$ in $f(y/s_1)$
- Value that belongs to $f(y/s_1)$ in $f(y/s_2)$

$P(\alpha) = T \times S_1$, favouring S_2 + $T \times S_2$, favouring S_1

$$P(\alpha) = P(S_1) \int_{-\infty}^{\alpha} \frac{1}{\sqrt{2\pi}\sigma^2} e^{-\frac{(y - (-\sqrt{E_b}))^2}{2N_0/2}}$$

$$+ P(S_2) \int_{\alpha}^{\infty} \frac{1}{\sqrt{2\pi}\sigma^2} e^{-\frac{(y - \sqrt{E_b})^2}{2N_0/2}}$$

Differentiating w.r.t α , we get -

$$P(S_1) f\left(\frac{\alpha}{S_1}\right) - P(S_2) f\left(\frac{\alpha}{S_2}\right) = 0$$

$$\frac{P(S_1)}{P(S_2)} = \frac{f(\alpha/S_2)}{f(\alpha/S_1)}$$

Solving by substituting the values, we get -

$$\frac{P(S_2)}{P(S_1)} = e^{4\alpha \sqrt{\frac{E_b}{N_0}}}$$

$$\Rightarrow 4\alpha \sqrt{\frac{\epsilon_b}{N_0}} = \ln \frac{P(S_1)}{P(S_2)}$$

$$\Rightarrow \boxed{\alpha = \frac{4}{4\sqrt{\epsilon_b}} \ln \frac{P(S_1)}{P(S_2)}}$$

$$P(S_1) > P(S_2) \rightarrow \alpha (+ve)$$

$$P(S_2) > P(S_1) \rightarrow \alpha (-ve)$$

$$P(S_1) = P(S_2) \rightarrow \alpha = 0$$

Substituting $\alpha = 0$, we get -

$$P_2(0) = P(S_1) \int_{-\infty}^0 f(y/S_1) dy + P(S_2) \int_0^{\infty} f(y/S_2) dy$$

$$= \frac{1}{2} \times 2 \int_0^{\infty} f(y/S_1) dy$$

$$= \frac{1}{\sqrt{2\pi N_0/2}} e^{-(y-\sqrt{\epsilon_b})^2/N_0}$$

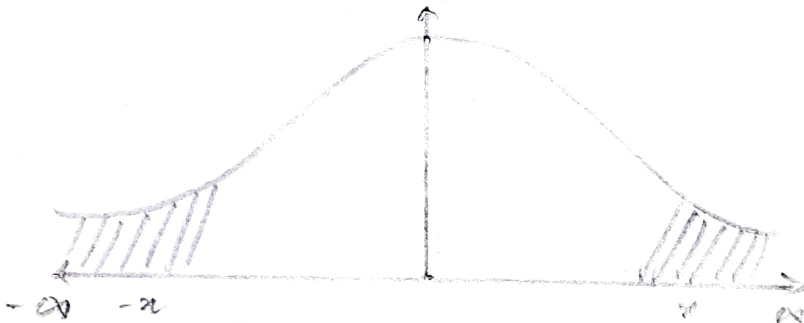
$$\boxed{z = (y - \sqrt{\epsilon_b}) / \sqrt{N_0/2}}$$

$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{-\sqrt{2\epsilon_b}/N_0} e^{-x^2/2} dx$$

$$= Q\left(\sqrt{\frac{2\epsilon_b}{N_0}}\right)$$

$$\boxed{Q(z) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^z e^{-x^2/2} dx = \frac{1}{\sqrt{2\pi}} \int_z^{\infty} e^{-x^2/2} dx}$$

$$\boxed{Q\left(\sqrt{\frac{2\epsilon_b}{N_0}}\right) = P(S_2)}$$



* PYTHON CODE :

`import matplotlib.pyplot as plt` # Provides an implicit way of plotting

`import numpy as np` # support for large, multi-dimensional arrays and matrices

`import math` # Provides access to the mathematical functions defined by the C standard

PTD

```
x = []
```

```
xt = np.random.normal(0, 1, 10000)
```

```
for i in range(len(xt)):
```

```
    if i > 0.5:
```

```
        x.append(i)
```

```
    elif i < 0.5:
```

```
        x.append(0)
```

```
y = []
```

```
for i in x:
```

```
    if i == 0:
```

```
        y.append(1)
```

```
    else:
```

```
        y.append(1-i)
```

```
x = np.arange(0, len(y), 1); plt.step(x, y)
```

```
plt.xlim(0, 100); plt.show()
```

```
l = p = []
```

```
for num in range(1, 11):
```

```
    final = error = xt = []
```

```
    sigma = (10 * 10 ** (-num/10)) / 2
```

```
    print("10" + str(num) + "-sigma Value: " + str(sigma))
```

```
    noise = np.random.normal(0, sigma, 10000)
```

```
    count, bins, ignored = plt.hist(noise, 1000, density=True)
```

```
    plt.plot(bins, np.ones_like(bins), linewidth=2, color='r')
```

```
    plt.show()
```

```
    xt = y + noise
```

```
    plt.plot(xt); plt.title("Signal with Noise")
```

```
    plt.xlim(0, 100); plt.show()
```

```

for i in range(len(xt)):
    if xt[i] > 0:
        final.append(1)
    else:
        final.append(-1)

```

```

error_count = 0

```

```

for i in range(len(xt)):
    if final[i] != yt[i]:
        error_count += 1

```

```

print("Error count: " + str(error_count));
error.append(error_count)

```

```

plt.stem(final)
plt.xlabel("Sequence"); plt.ylabel("Amplitude")
plt.title("Demodulated Signal")
plt.xlim(0, 100); plt.show()

```

```

for i in range(len(error)):
    e.append(error[i]/10000)

```

```

p.append(math.erfc(np.sqrt(num)* np.sqrt(2))) #

```

returns the complementary error function of a number

```

plt.semilogx(p); plt.semilogy(e)
plt.xlabel("SNR in dB"); plt.ylabel("Probability of error (Perr)")
plt.show()

```

* METHODOLOGY:

we know that,

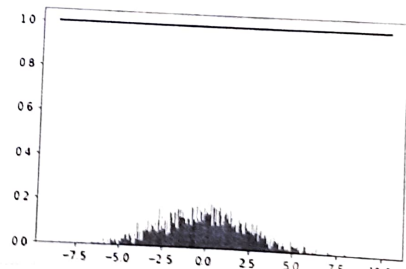
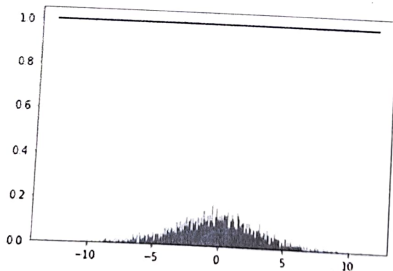
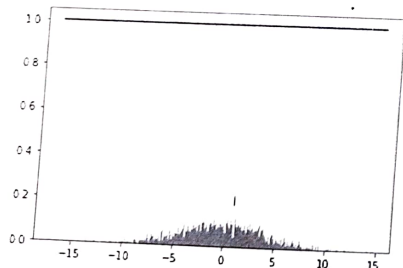
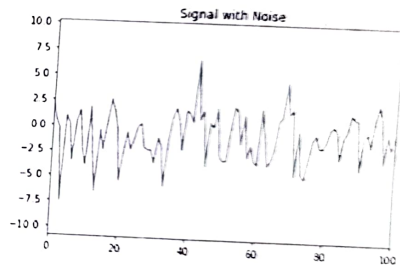
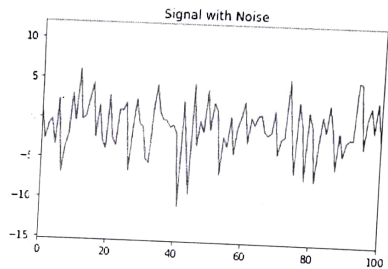
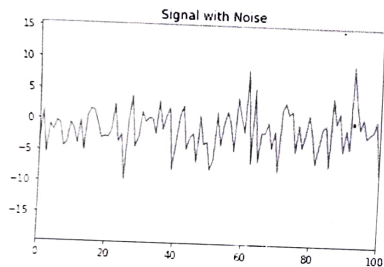
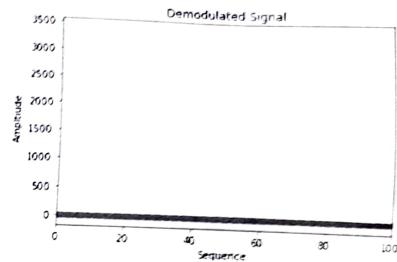
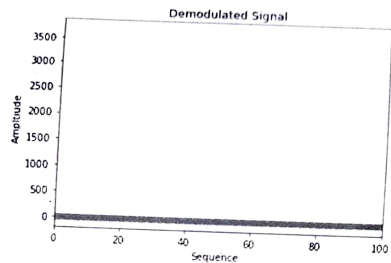
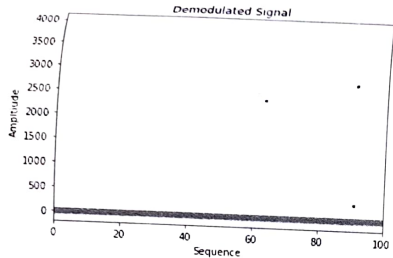
$$(SNR)_{dB} = \log_{10} \left(\frac{S}{N} \right)$$

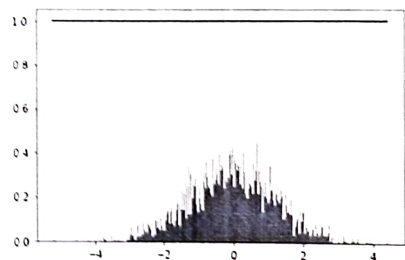
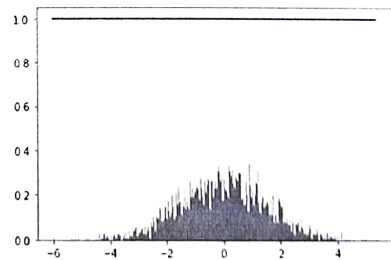
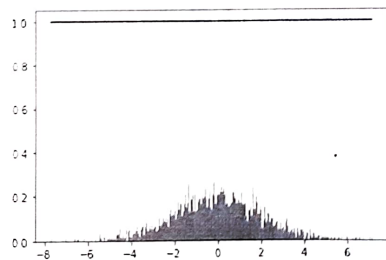
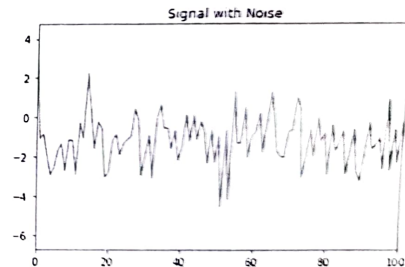
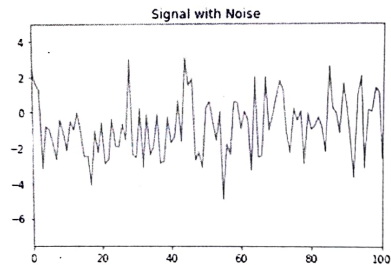
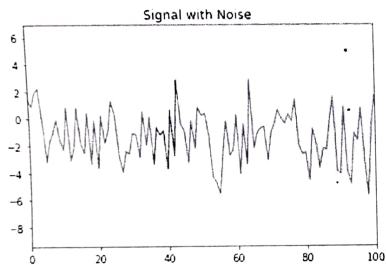
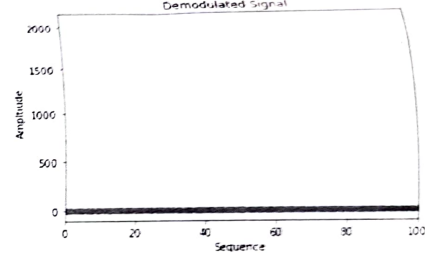
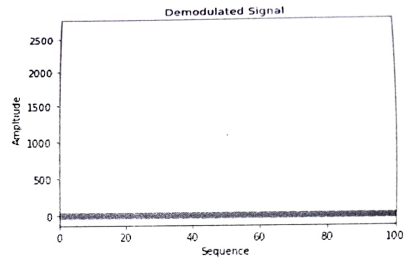
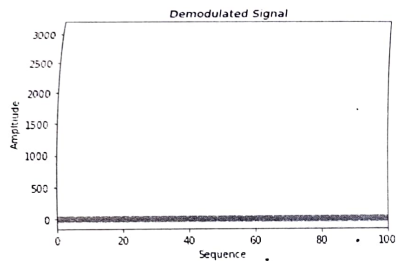
$$\Rightarrow \frac{SNR}{10} = \log_{10} \left(\frac{S}{N} \right)$$

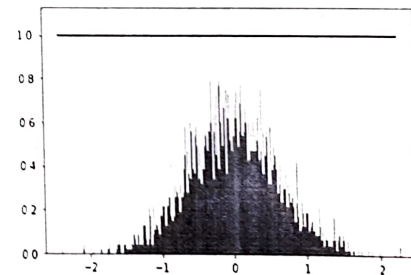
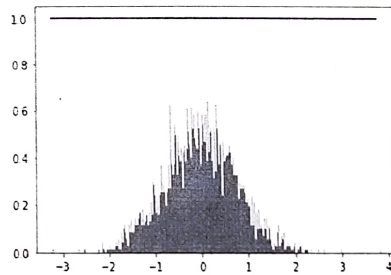
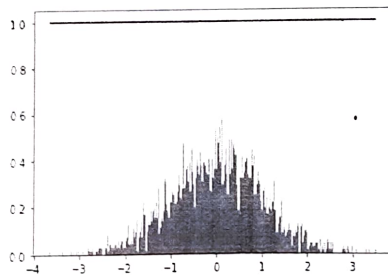
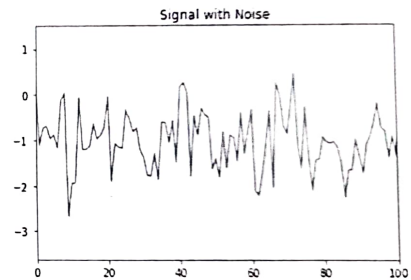
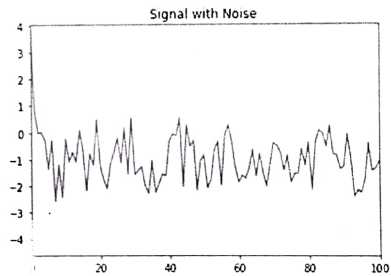
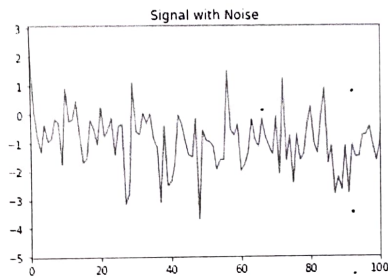
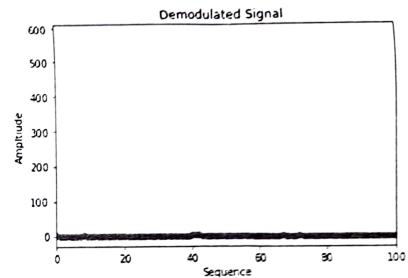
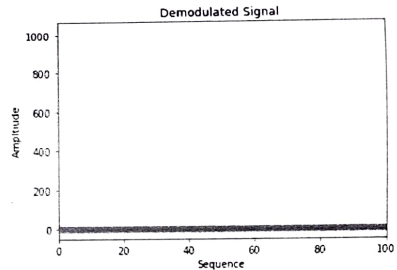
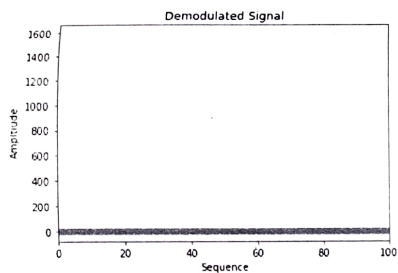
$$\Rightarrow \frac{N}{S} = 10 \left(\frac{SNR}{10} \right)$$

$$\Rightarrow \sigma^2 = S \times 10 \left(\frac{-SNR}{10} \right)$$

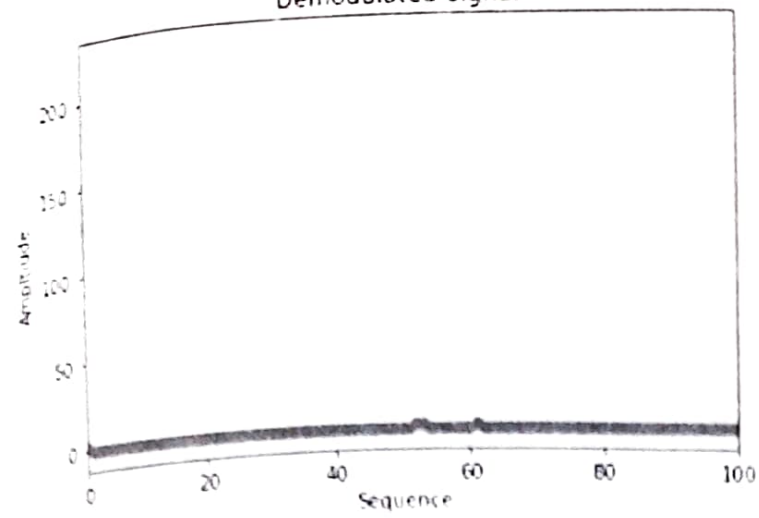
$$\Rightarrow \boxed{\sigma = S^{1/2} \times 10 \left(\frac{-SNR}{20} \right)}$$



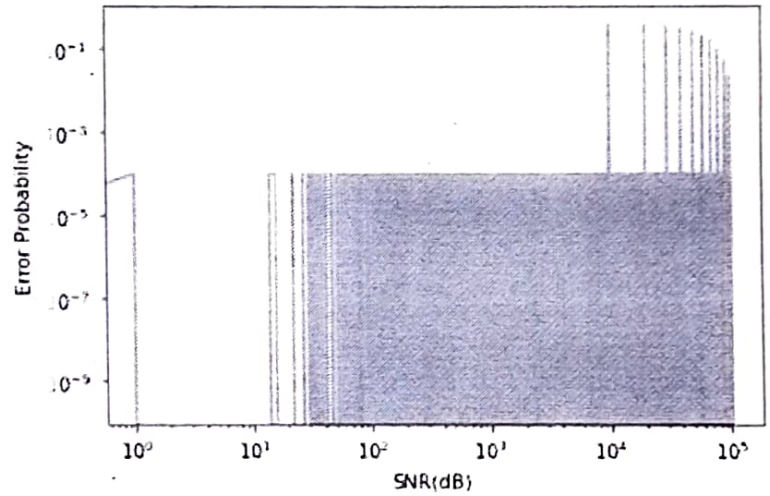
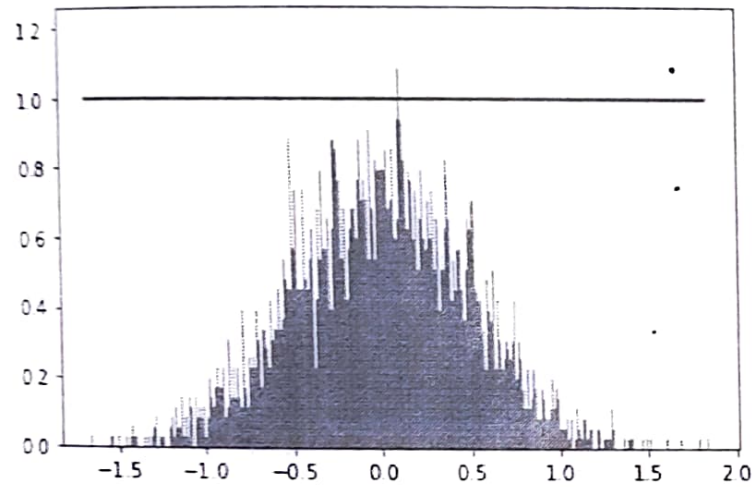
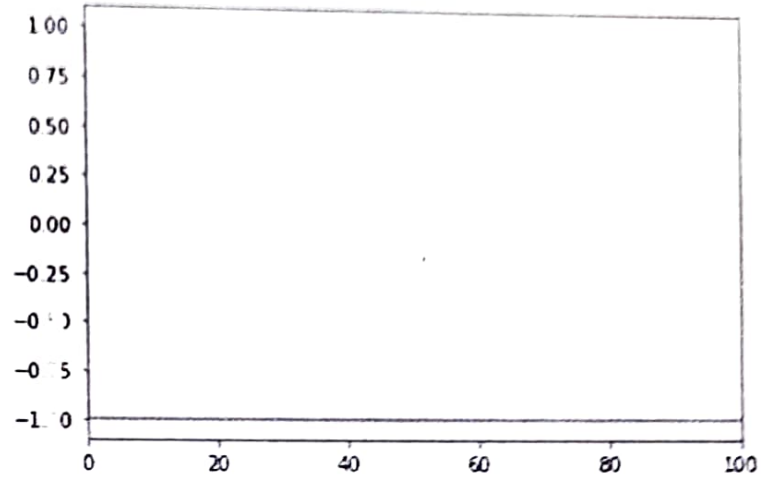
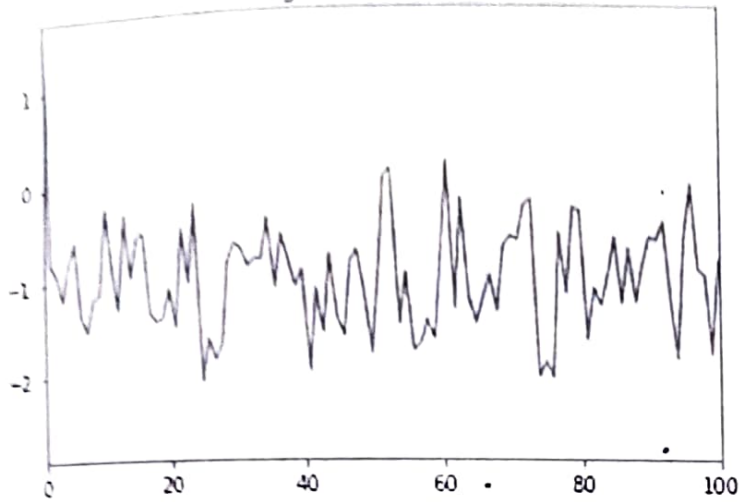




Demodulated Signal

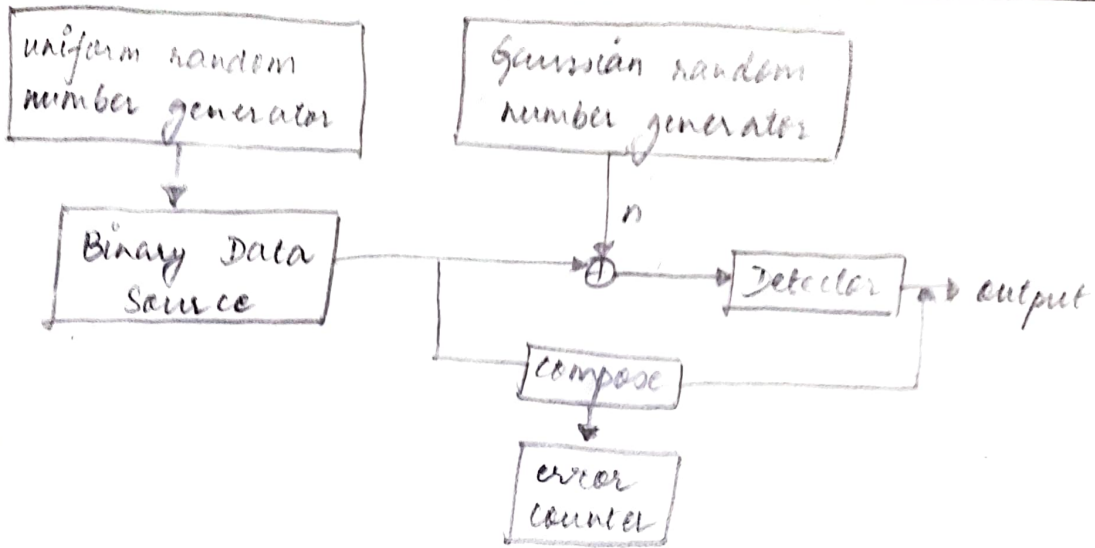


Signal with Noise



* OUTPUTS:

- 1 - Sigma Value : 3.9716411736214075
Error Count : 3982
- 2 - Sigma Value : 3.1547867224009662
Error Count : 3740
- 3 - Sigma Value : 2.505936168138361
Error Count : 3495
- 4 - Sigma Value : 1.990535852467466
Error Count : 3180
- 5 - Sigma Value : 1.5811388300841598
Error Count : 2557
- 6 - Sigma Value : 1.25894321375479
Error Count : 2137
- 7 - Sigma Value : 0.9976311574844397
Error Count : 1590
- 8 - Sigma Value : 0.7921465962305567
Error Count : 1053
- 9 - Sigma Value : 0.6294627058970836
Error Count : 515
- 10 - Sigma Value : 0.5
Error Count : 226



* RESULT:

Thus, we have plotted the probability of error by varying SNR values and observed that with higher SNR probability of error leads to zero. All the simulation results were verified successfully.