

# ELECENG 3TQ3 Project 2

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## Task 1:

Task 1: Find the pdf of the received signal  $r(t_j)$  and the pmf of the estimated sent signal  $\hat{s}_j$  assuming  $\tau = 0.5$  and  $\sigma^2 = 0.25$ .

$$\therefore r(t_j) = s(t_j) + e(t_j)$$

$\therefore e(t_j)$  is Gaussian distributed  $\therefore s(t_j)$  is binomial distributed

$\therefore$  pdf of  $e(t_j)$

$$f_e(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$

$\therefore$  pmf of  $s(t_j)$

$$f_s(x) = p\delta(x) + (1-p)\delta(x-1)$$

$x \in \mathbb{K}$

$\therefore$  pdf of  $r(t_j)$ :

Yichen Lu 400247738 10/19/19

$$f_r(x) = f_s(x) \otimes f_e(x)$$

" $\otimes$ " means convolution

$$\therefore f_r(x) = \frac{1}{\sigma\sqrt{2\pi}} \left[ p \cdot e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} + (1-p) e^{-\frac{1}{2}\left(\frac{x-1}{\sigma}\right)^2} \right]$$

$$\therefore \sigma^2 = 0.25; \tau = 0.5; \mu = 0$$

$$\therefore \sigma = 0.5$$

$$\therefore f_r(x) = \sqrt{\frac{2}{\pi}} \left[ p \cdot e^{-2x^2} + (1-p) e^{-2(x-1)^2} \right]_{x \in \mathbb{R}}$$

pmf of the estimated sent signal:

$$\begin{aligned} \textcircled{1} P(\hat{s}_j = 0) &= P(r_{t_j} < \tau) \because \tau = 0.5 \\ &= P(r_{t_j} < 0.5) \end{aligned}$$

$$= \int_{-\infty}^{0.5} \sqrt{\frac{2}{\pi}} [p e^{-2x^2} + (1-p) e^{-2(x-1)^2}] dx$$

$$\textcircled{2} p(\hat{s}_j = 1) = P(r_{tj} > \tau) \because \tau = 0.5$$

$$= P(r_{tj} > 0.5)$$

$$= \int_{0.5}^{\infty} \sqrt{\frac{2}{\pi}} [p e^{-2x^2} + (1-p) e^{-2(x-1)^2}] dx$$

ANS:

o the pdf of the received signal  $r(t_j)$

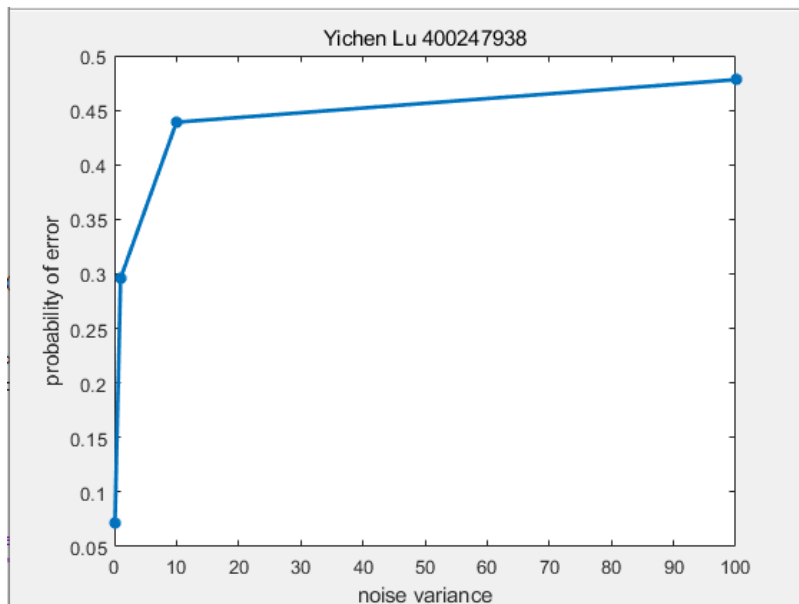
$$f_r(x) = \sqrt{\frac{2}{\pi}} [p e^{-2x^2} + (1-p) e^{-2(x-1)^2}] \quad x \in \mathbb{R}$$

o the pmf of the estimate sent signal  $\hat{s}_j$

$$P(\hat{s}_j) = \begin{cases} \sqrt{\frac{2}{\pi}} \int_{-\infty}^{0.5} [p e^{-2x^2} + (1-p) e^{-2(x-1)^2}] dx & \hat{s}_j = 0 \\ \sqrt{\frac{2}{\pi}} \int_{0.5}^{\infty} [p e^{-2x^2} + (1-p) e^{-2(x-1)^2}] dx & \hat{s}_j = 1 \end{cases}$$

## Task 2:

```
1 %Yichen Lu luy191 400247938
2 clc;
3 clear;
4
5 p=0.5;
6 tau=0.5;
7 variance=[0.1,1,10,100];
8 N=1000;
9 error=zeros(1,4);
10 s=randi([0,1],[1,1000]);
11
12 for i= 1:4
13     e = randn(1,1000)*sqrt(variance(i));
14     r = s+e;
15     for j = 1:N
16         if(s(j)==0 && r(j)>=tau) || (s(j)==1 && r(j)<tau)
17             error(i) = error(i)+1;
18         end
19     end
20 end
21
22 figure
23 plot(variance,error/N,'LineWidth',2,'Marker','*')
24 title('Yichen Lu 400247938')
25 xlabel('noise variance')
26 ylabel('probability of error')
27
28
```



## Task 3:

```
1 %Yichen Lu luy191 400247938
2 clc;
3 clear;
4 word='YichenLuisthebest';
5 l=length(word);
6 binary=dec2bin(word,8); %transfer the words into ASSCLL form
7 tempholder=zeros(1,8);
8 %constant
9 t = [0.25 0.5 0.75];
10 variance=[0.1 1 10];
11 error= zeros(length(t),length(variance));
12 times =5000;
13
14 %transfer string into integer
15 for i=1:17
16     for j=1:8
17         if binary(i,j)=='0'
18             tempholder(i,j)=0;
19         end
20         if binary(i,j)=='1'
21             tempholder(i,j)=1;
22         end
23     end
24 end
25
26
27
28 %main part
29 for i=1:length(t) %go through tau
30     for j=1:length(variance) %go through variance
31         wrong=0; %incorrect numbers holder
32         for x=1:times
33             e=randn(1,8)*sqrt(variance(j)); %noise function
34             for k = 1:17
35                 eachletter=tempholder(k,:); %for each row
36                 r=eachletter+e;
37                 wrongbit=false;
38                 for n=1:8 %run through every bit of the letters
39                     if(eachletter(n)==0&&round(r)>t(i)||eachletter(n)==1&&round(r)<t(i))
40                         wrongbit=true;
41                     end
42                 end
43                 if wrongbit
44                     wrong=wrong+1;
45                 end
46             end
47         end
48         error(i,j)=wrong/times; %get average error
49         fprintf('when tau=%2f and variance=%1f, the average number of incorrect words is %1f out of %d. \n',t(i),variance(j),error(i,j),17);
50     end
51 end
52
```

```
when tau=0.25 and variance=0.1, the average number of incorrect words is 10.7 out of 17.
when tau=0.25 and variance=1.0, the average number of incorrect words is 16.2 out of 17.
when tau=0.25 and variance=10.0, the average number of incorrect words is 16.8 out of 17.
when tau=0.50 and variance=0.1, the average number of incorrect words is 6.3 out of 17.
when tau=0.50 and variance=1.0, the average number of incorrect words is 16.1 out of 17.
when tau=0.50 and variance=10.0, the average number of incorrect words is 16.8 out of 17.
when tau=0.75 and variance=0.1, the average number of incorrect words is 10.7 out of 17.
when tau=0.75 and variance=1.0, the average number of incorrect words is 16.2 out of 17.
when tau=0.75 and variance=10.0, the average number of incorrect words is 16.8 out of 17.
```

In conclusion, the tau is the same for all numbers in the same row of the resulting matrix, and the variance is the same for all numbers in the same column.

I ran the test 5000 times in total by using my code, and the resulting matrix is the average of all the test results. We can observe that both tau and the variance of the noise have an impact on the number of errors that occur from the output result.

In terms of tau, we can see that when tau equals 0.5, it has the lowest the average number of incorrect words. However, when tau is 0.25 or 0.75, the average number of incorrect words is almost identical, although it is greater than when tau is 0.5.

When it comes to variance, there is a clear tendency that the larger the sigma, the greater the chance of inaccuracy. When variance = 0.1 with the same tau, 0.25, average error = 10.7, when variance = 1, average error = 16.2, and when variance = 10, average error = 16.8.

Thus, the number of correctly decoded letters did depend on tau and sigma.