ELC 325B - Spring 2023

Digital Communications

Assignment #2

Matched Filters

Submitted to

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Part I Requirements:

a) s(t) for $b_0 = '0'$, $b_1 = '1'$, and $b_2 = '1'$

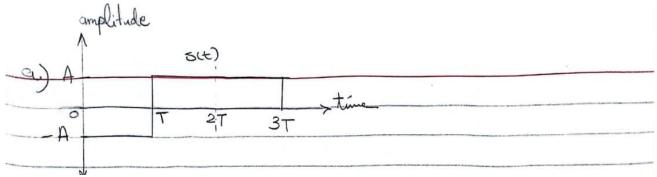
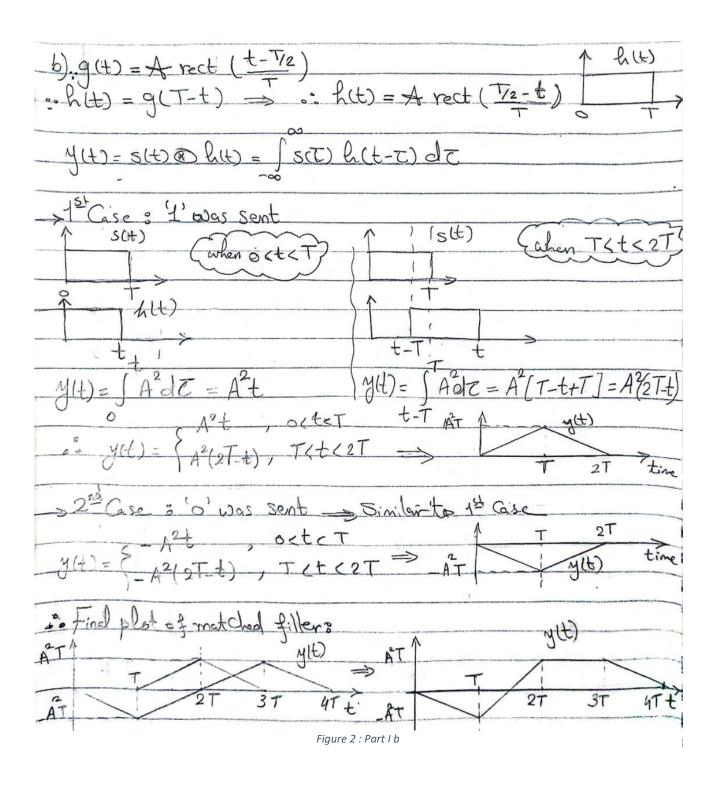


Figure 1 : Part I a

b) Matched filter output due to signal only



c) Sampling instants markings (We sampled @ T)

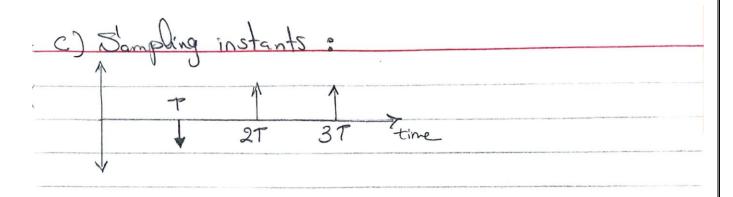


Figure 3 : Part I c

d) Block diagram of transmitter

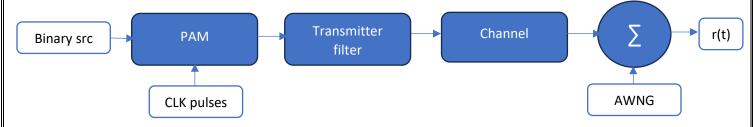


Figure 4 : Part I d

e) Block diagram of receiver



Figure 5 : Part I e

Part II Requirements:

1. Probability of error in the three mentioned cases

h (+) is unit energy:

h(+) = g(T-1) = A red (T-t-12) - A Red (T/2-t)

r(+) = g(+)+ w(+)

(+) + (+) w + (+) = g(+) + h(+) + h(+)

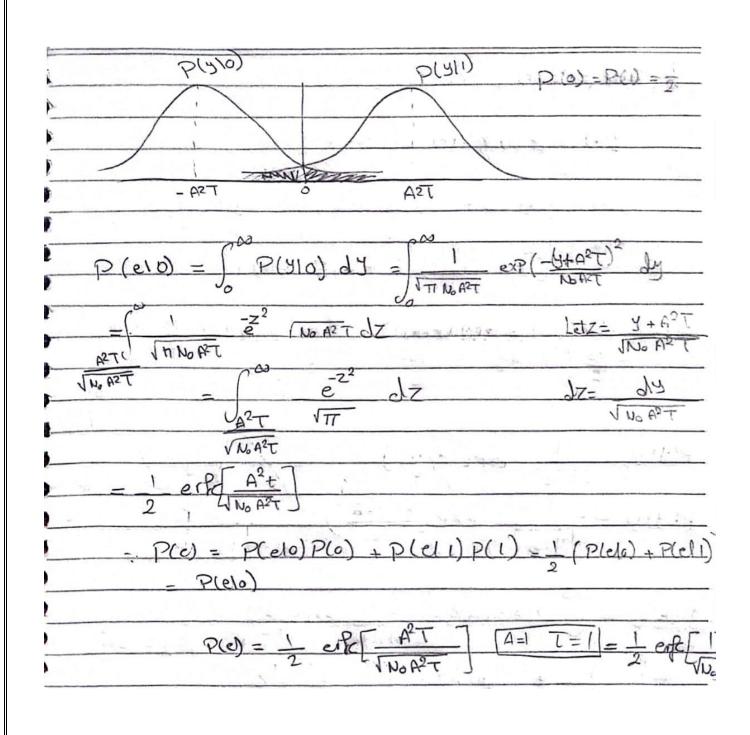
$$My = E(Y(T)) = E(Q(T)) + E(n(T))$$

$$= E(\pm A^{2}T) + E(n(T))$$

$$= E(h(T)) = E[h(T)] + E(n(T))$$

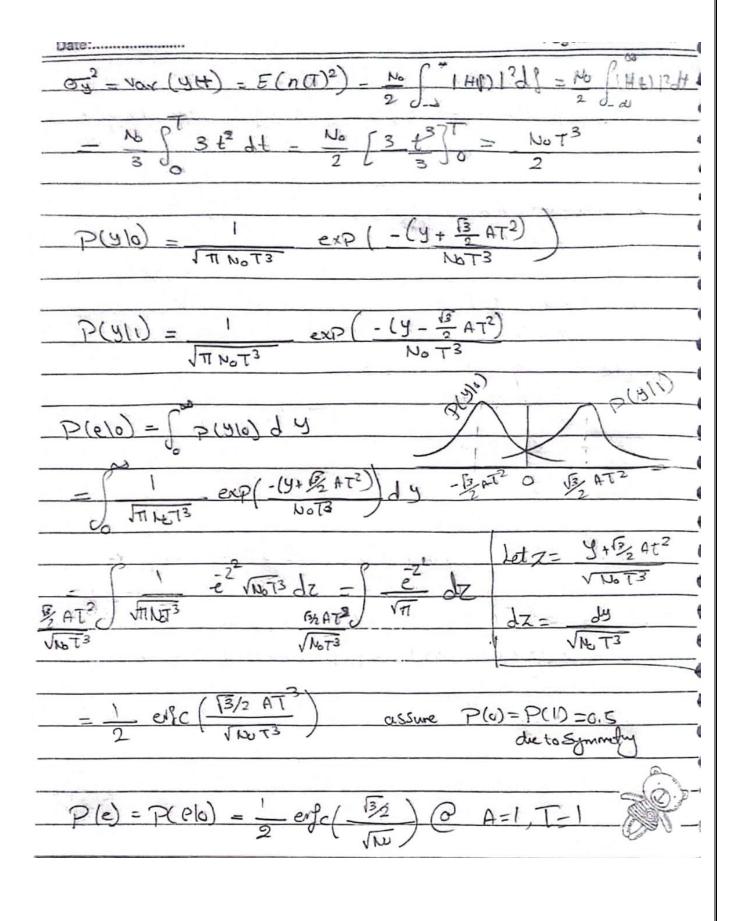
$$= E[h(T)] + E(n(T))$$

$$= E[h(T)] + E[h(T)]$$



b) h(+) =
$$g(+)$$
 + $h(+)$ = $r(+)$ * $S(+)$

$$y(+) = g(+) + h(+) = + A + h(+) + y(+) = -A + h(+) + y(+) =$$



2. Python code

```
import numpy as np
import matplotlib.pyplot as plt
import math
BITS NUM = 5
SAMPLES_PER_BIT_NUM = 5
E=1
def generate random bits(size = BITS NUM):
    return np.random.choice(a=[0,1], size= size)
def generate_signal(random_bits,samples_per_bit_num,duration):
    signal = np.ones((len(random_bits), samples_per_bit_num))
    for i in range (len(random_bits)):
        if random bits[i] == 1:
            signal[i] = np.ones(int(samples per bit num * duration))
        elif random_bits[i] == 0:
            signal[i] = np.full(int(samples per bit num * duration), -1)
    return signal/math.sqrt(samples_per_bit_num)
def generate gaussian noise (sigma, size):
    return np.random.normal(loc = 0 ,scale=sigma, size=size)
def add_noise(signal,sigma,samples_per_bit_num = SAMPLES_PER_BIT_NUM):
    noise = generate_gaussian_noise(sigma = sigma,
size=len(signal)*samples per bit num)
    noisy_signal=np.copy(signal)
    for i in range(len(signal)):
        noisy_signal[i, :] +=noise[i*samples_per_bit_num:(i+1)*samples_per_bit_num]
    return noisy_signal
def calc_convolution(noisy_signal, filter):
    convolved = None
    convolved sampled=np.zeros(noisy_signal.shape[0])
    if(filter is None):
        convolved = noisy_signal.flatten()
    else:
        convolved = np.convolve(noisy_signal.flatten(), filter)
    for i in range(noisy_signal.shape[0]):
        convolved_sampled[i] =convolved[(noisy_signal.shape[1] - 1) +
noisy_signal.shape[1] *i]
    convolved_sampled = (convolved_sampled > 0).astype(int)
    return convolved, convolved sampled
```

```
def calc_sim_error(expected, received):
    return (np.sum(received != expected)) / len(expected)
def calc_theo_error(N, filter_type):
    if(filter_type==0 ): # matched
        return (0.5 * math.erfc(1/(N ** 0.5)))
    if(filter_type==1): #ramp
        return 0.5 * math.erfc((3 /N) ** 0.5 / 2)
    else: # none
        return (0.5 * math.erfc((1/math.sqrt(SAMPLES PER BIT NUM))/(N ** 0.5)))
def plot filter out(ax,noisy signal,random bits,filter,filter type,N):
    convolved,convolved_sampled = calc_convolution(noisy_signal=noisy_signal , filter=
filter)
    sim_error = calc_sim_error(random_bits,convolved_sampled)
    print("Prob of error for filter type ",filter_type," is ",sim_error)
    ax[filter type].plot(range(∅, convolved.flatten().shape[∅]), convolved.flatten(),
label = "bit value")
    for i in range(SAMPLES PER BIT NUM - 1, convolved.flatten().shape[∅],
SAMPLES_PER_BIT_NUM):
        ax[filter type].stem([i], [convolved.flatten()[i]], linefmt='magenta')
    ax[filter type].set xlabel('Time (s)')
    ax[filter_type].set_ylabel('Amplitude')
    if filter type == 0:
        ax[filter_type].set_title(f'Matched Filter Output')
    elif filter type == 1:
        ax[filter_type].set_title(f'Ramp Filter Output')
    else:
        ax[filter_type].set_title(f'No Filter Output')
    ax[filter_type].grid(True)
random_bits = np.array([0,1,0,0,0])
print("Signal ",random_bits)
E_over_N=10**(20/10)
N=E/E_over_N
signal=generate_signal(random_bits,SAMPLES_PER_BIT_NUM,1)
noisy signal=add noise(signal,np.sqrt(N/2),SAMPLES PER BIT NUM)
fig, ax = plt.subplots(1, 3, figsize=(15, 5))
```

```
filter matched=np.ones(SAMPLES PER BIT NUM)
plot filter out (ax,noisy signal , random bits , filter matched,⊘,N )
filter ramp= np.linspace(0, 3**0.5, SAMPLES PER BIT NUM)
plot_filter_out (ax,noisy_signal , random_bits , filter_ramp,1,N )
plot_filter_out (ax,noisy_signal , random_bits , None,2,N )
plt.show()
random bits = generate random bits( size= 10**5)
matched_sim_BER,matched_theo_BER= [] , []
ramp_sim_BER,ramp_theo_BER= [] , []
none_sim_BER,none_theo_BER= [] , []
# Define a dictionary to store lists for each filter type
sim_BER_dict = {0: matched_sim_BER, 1: ramp_sim_BER, 2: none_sim_BER}
theo_BER_dict = {0: matched_theo_BER, 1: ramp_theo_BER, 2: none_theo_BER}
#BER part
random_bits = generate_random_bits( size= 10**5)
matched_sim_BER,matched_theo_BER= [] , []
ramp_sim_BER,ramp_theo_BER= [] , []
none_sim_BER,none_theo_BER= [] , []
# Define a dictionary to store lists for each filter type
sim_BER_dict = {0: matched_sim_BER, 1: ramp_sim_BER, 2: none_sim_BER}
theo_BER_dict = {0: matched_theo_BER, 1: ramp_theo_BER, 2: none_theo_BER}
def calc_filter_error (random_bits , convolved_sampled ,filter_type,N):
    # filter type represents the type of filter (0 for matched, 1 for ramp, 2 for none)
    sim_error = calc_sim_error(random_bits, convolved_sampled)
    theo_error = calc_theo_error(N, filter_type)
    sim BER dict[filter type].append(sim error)
    theo_BER_dict[filter_type].append(theo_error)
for E over N db in range (-10,21,1):
    E_over_N=10**(E_over_N_db/10)
    N=E/E_over_N
```

```
signal=generate_signal(random_bits,SAMPLES_PER_BIT_NUM,1)
    noisy signal=add noise(signal,np.sqrt(N/2),SAMPLES PER BIT NUM)
    convolved,convolved_sampled=calc_convolution(noisy_signal,filter_matched)
    calc_filter_error(random_bits,convolved_sampled,0,N)
    convolved,convolved_sampled = calc_convolution(noisy_signal,filter_ramp)
    calc_filter_error(random_bits,convolved_sampled,1,N)
    convolved,convolved sampled=calc convolution(noisy signal,None)
    calc_filter_error(random_bits,convolved_sampled,2,N)
sim BER label dict = {0: 'matched sim BER', 1: 'matched sim BER', 2: 'matched sim BER'}
theo_BER_label_dict = {0: 'matched_sim_BER', 1: 'matched_sim_BER', 2:
'matched sim BER'}
def plot_filter_BER(ax,filter_type):
    ax[filter type].semilogy(range(-10,21), sim BER dict[filter type]
,label=sim_BER_label_dict[filter_type])
    ax[filter_type].semilogy(range(-10,21), theo_BER_dict[filter_type]
,label=theo_BER_label_dict[filter_type])
    ax[filter_type].legend()
    ax[filter_type].set_xlabel('E/N0 (dB)') # X-axis label
    ax[filter_type].set_ylabel('BER')
    if(filter type == 0 ):
        ax[filter_type].set_title('BER vs. E/N0 (Matched Filter)')
    if(filter type == 1 ):
        ax[filter_type].set_title('BER vs. E/N0 (Ramp Filter)')
    else:
        ax[filter_type].set_title('BER vs. E/N0 (None Filter)')
    ax[filter type].set ylim(10**(-4))
    ax[filter_type].grid(True)
fig, ax = plt.subplots(1, 3, figsize=(15, 5))
plot filter BER(ax, ∅)
plot filter BER(ax,1)
plot_filter_BER(ax,2)
plt.show()
#plot all
plt.semilogy(range(-10,21), matched sim BER , label='matched sim BER')
```

```
plt.semilogy(range(-10,21), matched_theo_BER, label='matched_theo_BER')
plt.semilogy(range(-10,21), ramp_sim_BER , label='ramp_sim_BER')
plt.semilogy(range(-10,21), ramp_theo_BER, label='ramp_theo_BER')
plt.semilogy(range(-10,21), none_sim_BER , label='none_sim_BER')
plt.semilogy(range(-10,21), none_theo_BER, label='none_theo_BER')
plt.legend()
plt.xlabel('E/N0 (dB)') # X-axis label
plt.ylabel('BER')
plt.title('BER vs. E/N0 for different Filters ')
plt.grid(True)
plt.grid(True)
plt.tight_layout()
plt.show()
```

3. And 4. Output of the receive filter for the three mentioned cases

Original Signal = [0 1 0 0 0]

With probability of error of 0 for all filters

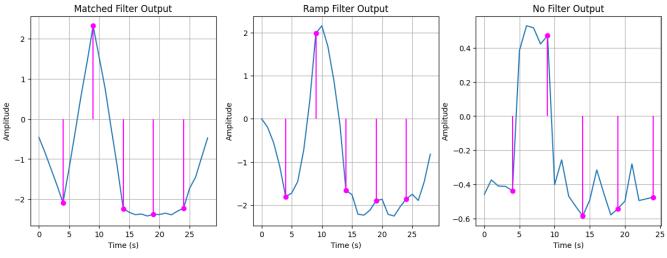


Figure 6: Filters output

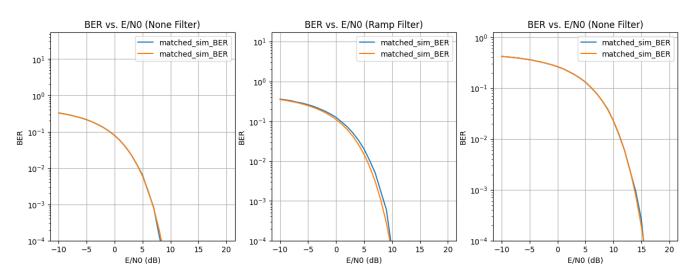


Figure 7: BER vs E/NO for each filter.

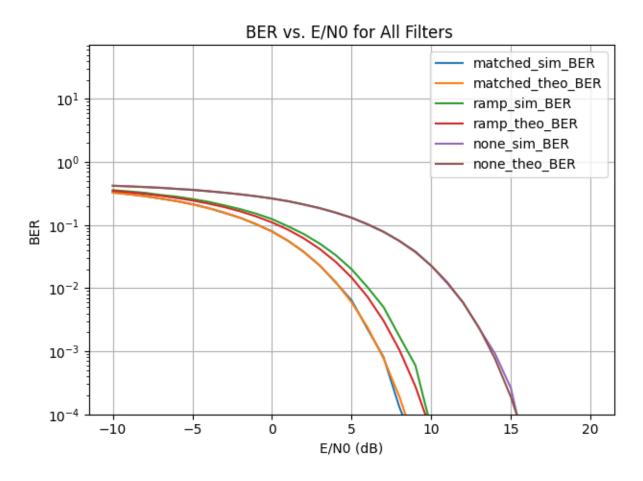


Figure 8: BER vs E/NO for all filters.

г	DED is decreasing function	۰ŧ	Е
5.	BER is decreasing function	ΟI	N

Increasing the E/No ratio reduces the bit error rate (BER). This entails boosting E/No, decreasing noise (NO), narrowing the Gaussian noise distribution, and enhancing SNR, thus minimizing noise interference, facilitating signal detection, and leading to an exponential decrease in BER as E/No rises.

6. Matched filter has the lowest BER

Because it's designed to match the transmitted signal's characteristics. This allows the filter to provide the best detection strategy by comparing the received signal with a copy of the transmitted one. As a result, this reduces errors and maximizes the signal