

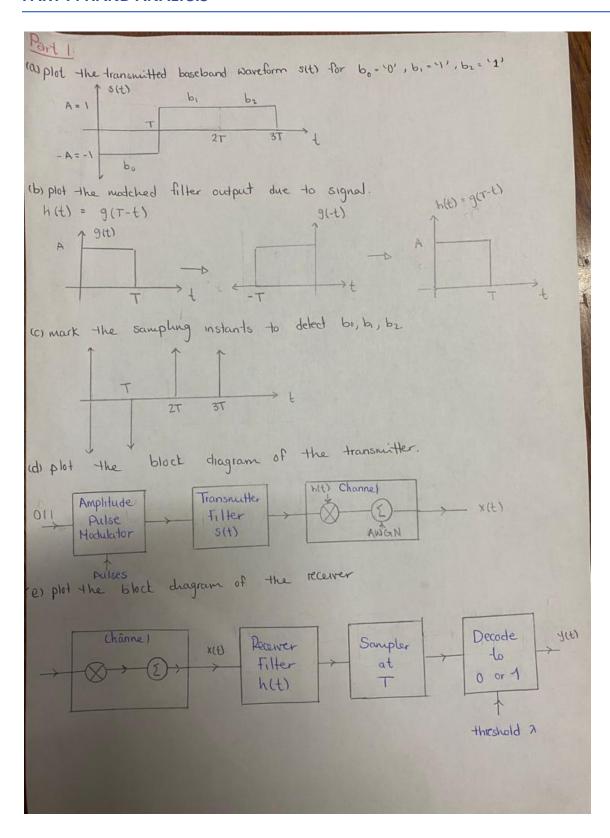


Digital Communication

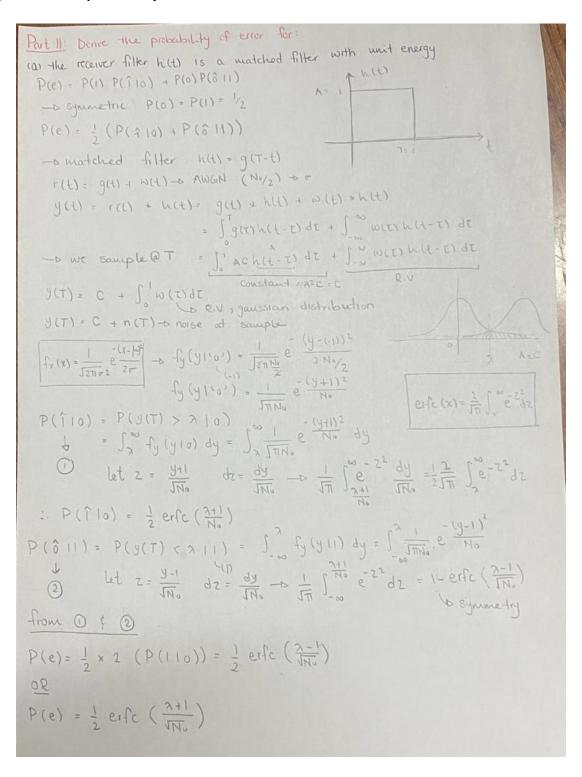
Assignment Two
Matched Filter

STUDENTS INFORMATION

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1) Derive the probability of error in the three mentioned cases.



(b) the receiver filter with is now existent (with = 8(t))

$$P(e) = \frac{1}{2} (P(110) + P(011))$$

filter $h(t) = S(t)$
 $F(t) = g(t) + w(t) = b$ AWGN

 $f(t) = r(t) + h(t) = g(t) + h(t) + w(t) + h(t)$
 $f(t) = r(t) + h(t) = g(t) + h(t) + w(t) + h(t)$
 $f(t) = g(t) + h(t) + \int_{0}^{1} w(t)$
 $f(t) = g(t) + h(t) + \int_{0}^{1} w(t)$
 $f(t) = g(t) + h(t) + \int_{0}^{1} w(t)$
 $f(t) = \frac{1}{2} erfc(\frac{2-h}{N_0}) + oR P(e) = \frac{1}{2} erfc(\frac{2+h}{N_0})$

(c) the receiver filter h(t) has the impulse response

 $f(t) = \frac{1}{2} (P(110) + P(011))$

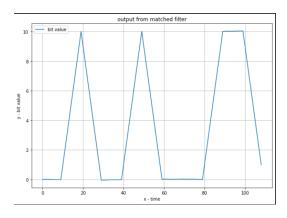
filter $h(t) = J3t$
 $f(t) = g(t) + h(t) = g(t) + h(t) + w(t) + h(t)$
 $f(t) = g(t) + h(t) = g(t) + h(t) + w(t) + w(t)$
 $f(t) = \frac{1}{2} (P(110) + P(011))$
 $f(t) = \frac{1}{2} (P(110) +$

Write the code that simulates the system In CODE section

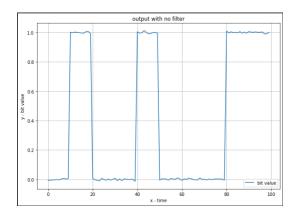
3) Plot the output of the receive filter for the three mentioned cases

These outputs were a result of a small sample of bits (10 bits) as the large number of bits wasn't showing in the graphs.

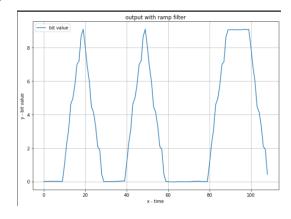
a) Matched Filter



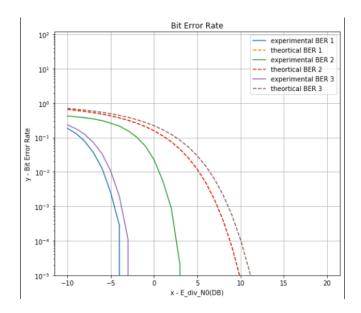
b) No Filter



c) Ramp Filter



4) Plot the theoretical and simulated Bit Error Rate (BER) Vs E/No for the three mentioned cases.



5) Is the BER an increasing or a decreasing function of E/No? Why?

The BER is a decreasing function as shown in the graph above as when E/No decreases the No increases which increases the noise power but SNR which is the signal to the noise ratio decreases which will result in a higher probability of flipping the bit from -A to higher than zero and A to less than zero as signal after noise.

6) Which case has the lowest BER? Why?

Matched filter has the lowest BER which shows that it is the optimum receiving filter having the max signal to noise ratio with impulse response h(t) = kg (T - t) which achieves maximum efficiency and hence lowest bit error rate.

```
def generatRandomBits():
   return np.random.choice([0, 1], size=(number of bits))
def convertToPulses(generated bits):
   generated bits[generated bits == 0] = -1
   signal = np.repeat(generated bits, fs)
   return signal
def addWhiteGaussianNoise(generated bits, sigma):
             generated noise = np.random.normal(loc=0, scale=sigma,
size=fs*number of bits)
   scaled samples with noise = np.ones((generated bits.shape[0], fs))
   scaled samples = np.ones((generated bits.shape[0], fs))
   for i in range(generated bits.shape[0]):
       scaled samples[i, :] *= generated bits[i]
       scaled_samples_with_noise[i, :] *= generated bits[i]
       scaled samples with noise[i, :] += generated noise[i*fs:(i+1)*fs]
   return scaled samples, scaled samples with noise
def convolute(scaled samples with noise, received filter):
   convolution_result_sampled_Tp = np.zeros(number_of_bits)
   if(received filter is None):
       convolution result = scaled samples with noise.flatten()
                                                convolution result
np.convolve(scaled samples with noise.flatten(), received filter)
   for i in range(number of bits):
         convolution result sampled Tp[i] = convolution result[(fs - 1) +
fs * i]
```

```
return convolution_result, convolution_result_sampled_Tp

def calculateProbabilityOfError(true_bits, convolution_result, Z):

# applying threshold to sample output with lambda = 0
received_samples = np.ones(true_bits.shape[0])
received_samples += (-2 * (convolution_result < 0))

# calculate simulation probability of error
simulation = np.sum(received_samples != true_bits)
simulation /= true_bits.shape[0]

# calculate theoretical probability of error
theoretical = math.erfc(Z)</pre>
```

```
received filter matched = np.ones(fs)
for E div N0 db in range(-10, 21):
    transmitted samples, received samples =
addWhiteGaussianNoise(generated bits, 1/(2*E div N0))
    filtered samples, filtered bits = convolute(received samples,
received filter matched)
plt.figure(figsize=(10,7))
plt.plot(range(0, filtered samples.flatten().shape[0]),
filtered samples.flatten(), label = "bit value")
plt.xlabel('x - time')
plt.ylabel('y - bit value')
plt.title('output from matched filter')
plt.legend()
plt.grid()
plt.show()
\# CASE TWO: The receive filter h\left(t
ight) is not existent (i.e. h\left(t
ight) = oldsymbol{\delta}\left(t
ight))
```

```
received_filter_matched = None

for E_div_N0_db in range(-10, 21):
    E_div_N0 = 10 ** (E_div_N0_db/10)
    transmitted_samples, received_samples =
addWhiteGaussianNoise(generated_bits, 1/(2*E_div_N0))
    filtered_samples, filtered_bits = convolute(received_samples,
received_filter_matched)

# plot output of the received filter
plt.figure(figsize=(10,7))
plt.plot(range(0, filtered_samples.flatten().shape[0]),
filtered_samples.flatten(), label = "bit value")

plt.xlabel('x - time')
plt.ylabel('y - bit value')
plt.title('output with no filter')
plt.legend()
plt.grid()
plt.show()
```

```
# CASE THREE: The receive filter is a ramp with h(t) = √3t
received_filter_ramp = np.random.uniform(low=0, high=3**0.5,
size=number_of_bits)
E = 1
for E_div_N0_db in range(-10, 21):
    E_div_N0 = 10 ** (E_div_N0_db/10)
    transmitted_samples, received_samples =
addWhiteGaussianNoise(generated_bits, E/(2*E_div_N0))
    filtered_samples, filtered_bits = convolute(received_samples,
received_filter_ramp)

#plotting
plt.figure(figsize=(10,7))
plt.plot(range(0, filtered_samples.flatten().shape[0]),
filtered_samples.flatten(), label = "bit value")

plt.xlabel('x - time')
plt.ylabel('y - bit value')
```

```
plt.title('output with ramp filter')
plt.legend()
plt.grid()
plt.show()
```

```
# CONSTANTS TO CALCULATE THE BIT ERROR RATE

fs = 20

number_of_bits = 10**5
```

```
CALCULATE BIT ERROR RATE FOR THE THREE CASES
generated bits = generatRandomBits()
received filter matched = np.ones(fs)
BER simulation 1 = []
BER theortical 1 = []
# receive with no filter
received filter empty = None
BER simulation 2 = []
BER theortical 2 = []
# receive with ramp filter
received filter ramp = np.random.uniform(low=0, high=3**0.5, size=fs)
BER simulation 3 = []
BER theortical 3 = []
for E div N0 db in range (-10, 21):
    transmitted samples, received samples =
addWhiteGaussianNoise(generated bits, 1/(2*E div N0))
    filtered samples1, filtered bits1 = convolute(received samples,
received filter matched)
    filtered samples2, filtered bits2 = convolute(received_samples,
received filter empty)
    filtered samples3, filtered bits3 = convolute(received samples,
received filter ramp)
    theoretical BER, simulation BER =
calculateProbabilityOfError(generated bits, filtered bits1, E div N0 **
0.5)
```

```
BER_simulation_1.append(simulation_BER)

BER_theortical_1.append(theoretical_BER)

theoretical_BER, simulation_BER =

calculateProbabilityOfError(generated_bits, filtered_bits2, E_div_N0 **

0.5)

BER_simulation_2.append(simulation_BER)

BER_theortical_2.append(theoretical_BER)

theoretical_BER, simulation_BER =

calculateProbabilityOfError(generated_bits, filtered_bits3, (3**0.5/2) *

E_div_N0 ** 0.5)

BER_simulation_3.append(simulation_BER)

BER_theortical_3.append(theoretical_BER)
```

```
plt.figure(figsize=(8,7))
plt.plot(range(-10, 21), BER simulation 1, label = "simulation BER 1")
plt.plot(range(-10, 21), BER theortical 1, "--", label = "theoretical BER
1")
plt.plot(range(-10, 21), BER simulation 2, label = "simulation BER 2")
plt.plot(range(-10, 21), BER theortical 2, "--", label = "theoretical BER
2")
plt.plot(range(-10, 21), BER simulation 3, label = "simulation BER 3")
plt.plot(range(-10, 21), BER theortical 3, "--", label = "theoretical BER
3")
plt.xlabel('x - E div N0(DB)')
plt.ylabel('y - Bit Error Rate')
plt.yscale('log')
plt.ylim(10**(-5))
plt.title('Bit Error Rate')
plt.legend()
plt.grid()
plt.show()
```