

# Analyse et traitement de l'information

## TP5 : Probabilities and communications

### Exercice 1 : Chain rules for probabilities

The idea of the chain rules is that to compute a given probability we check the dependencies (ie an arrow (direct or not) from X to Y means that Y depends of X).

a.

$$p(X, Y, Z) = p(X) * p(Y|X) * p(Z|X,Y) \quad (\text{by applying the rule})$$

$$p(X, Z) = p(X) * p(Z|X)$$

$$p(X, Y) = p(X) * p(Y|X)$$

$$p(Y, Z) = p(Y) * p(Z|Y)$$

b.

$$P(X_1, X_2, \dots, X_7) = p(X_1)p(X_2)p(X_3|X_1, X_2)p(X_4)p(X_5)p(X_6|X_3, X_4, X_5)p(X_7|X_6)$$

$$P(X_1, X_3, X_5, X_7) = p(X_1)p(X_3|X_1)p(X_5)p(X_7|X_3, X_5)$$

$$P(X_2, X_4, X_6, X_7) = p(X_2)p(X_4)p(X_6|X_2, X_4)p(X_7|X_6)$$

$$P(X_3, X_6, X_7) = p(X_3)p(X_6|X_3)p(X_7|X_6)$$

$$P(X_1, X_2, X_4, X_5) = p(X_1)p(X_2)p(X_4)p(X_5)$$

c.

$$P(X_1, X_2, \dots, X_7) = p(X_1)p(X_2)p(X_3|X_1, X_2)p(X_4)p(X_5|X_2)p(X_6|X_3, X_4, X_5)p(X_7|X_6, X_4)$$

$$P(X_1, X_3, X_5, X_7) = p(X_1)p(X_3|X_1)p(X_5)p(X_7|X_3, X_5)$$

$$P(X_2, X_4, X_6, X_7) = p(X_2)p(X_4)p(X_6|X_2, X_4)p(X_7|X_6, X_4)$$

$$P(X_3, X_6, X_7) = p(X_3)p(X_6|X_3)p(X_7|X_6)$$

$$P(X_1, X_2, X_4, X_5) = p(X_1)p(X_2)p(X_4)p(X_5|X_2)$$

## Exercise 2 : Communication through noisy channels

### a. Main part of the code:

```
def startCommunication(self):

    start = time.time()
    # Add the redundancy, for example if msg is [0110] and rate=3 => [000111111000]
    preprocessed_message = np.repeat(self.input_message, self.rate)
    output_bsc = self.bsc(preprocessed_message)

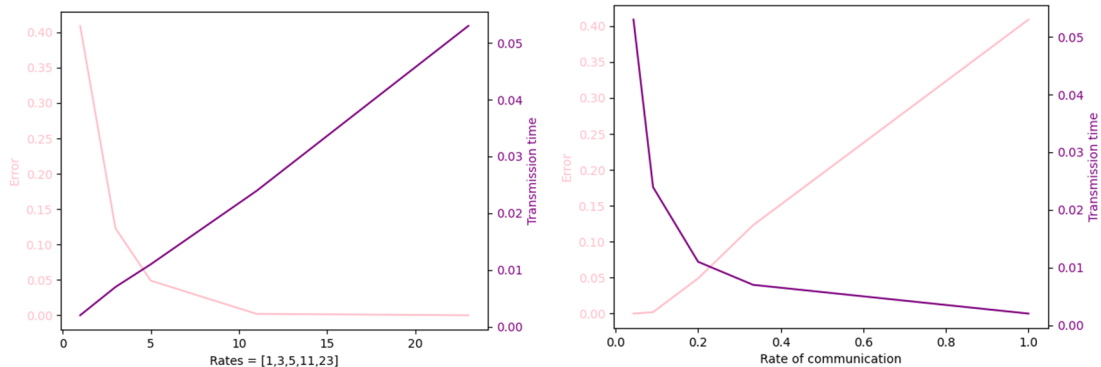
    # Decoding the output message
    output_message = np.ndarray.astype(np.sum(output_bsc.reshape((-1, self.rate)), 1) > rate / 2, 'int')
    transmission_time = time.time() - start

    # Convert bits to words for the input and the output message
    input_words = np.dot(self.input_message.reshape((-1, self.M)), 2 ** np.arange(self.M)[::-1])
    output_words = np.dot(output_message.reshape((-1, self.M)), 2 ** np.arange(self.M)[::-1])

    # Fraction of Error
    nbrError = np.sum(input_words != output_words) / len(input_words)

    return nbrError, transmission_time, self.communicationRate()
```

### b. Results:



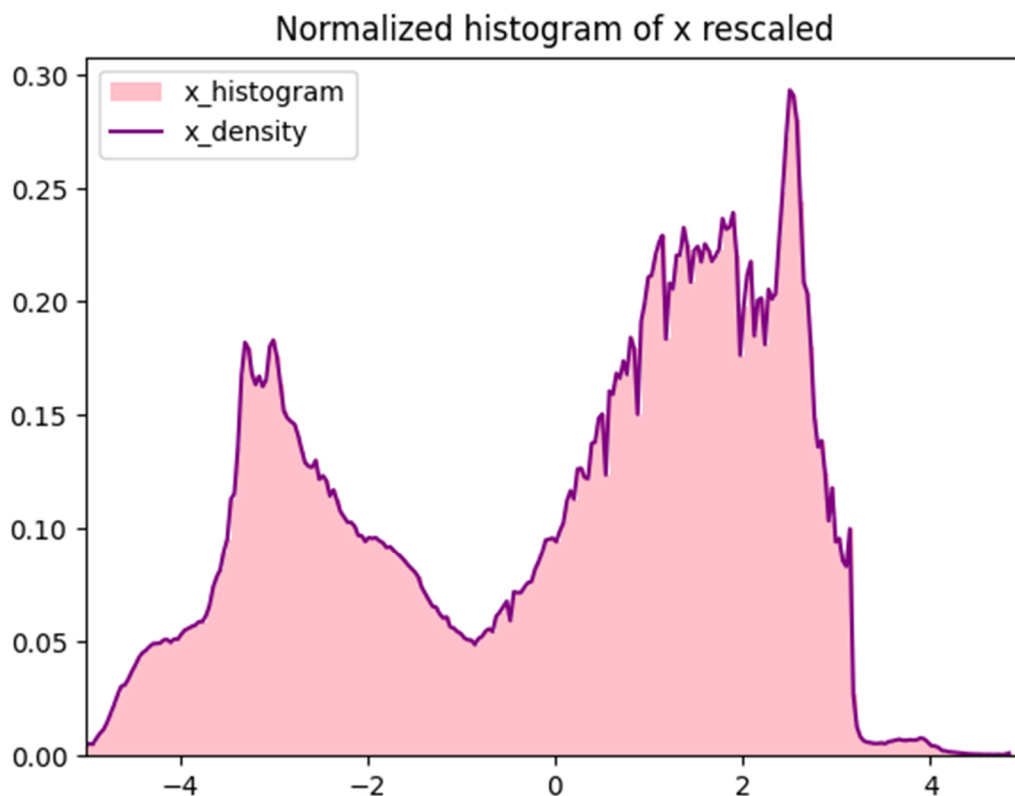
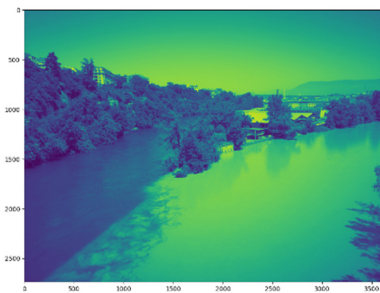
	Rates = [1,3,5,11,23]	Error	Transmission time	Rate of communication
0	1	0.409	0.002021	1.000000
1	3	0.123	0.007001	0.333333
2	5	0.049	0.011000	0.200000
3	11	0.002	0.023942	0.090909
4	23	0.000	0.052997	0.043478

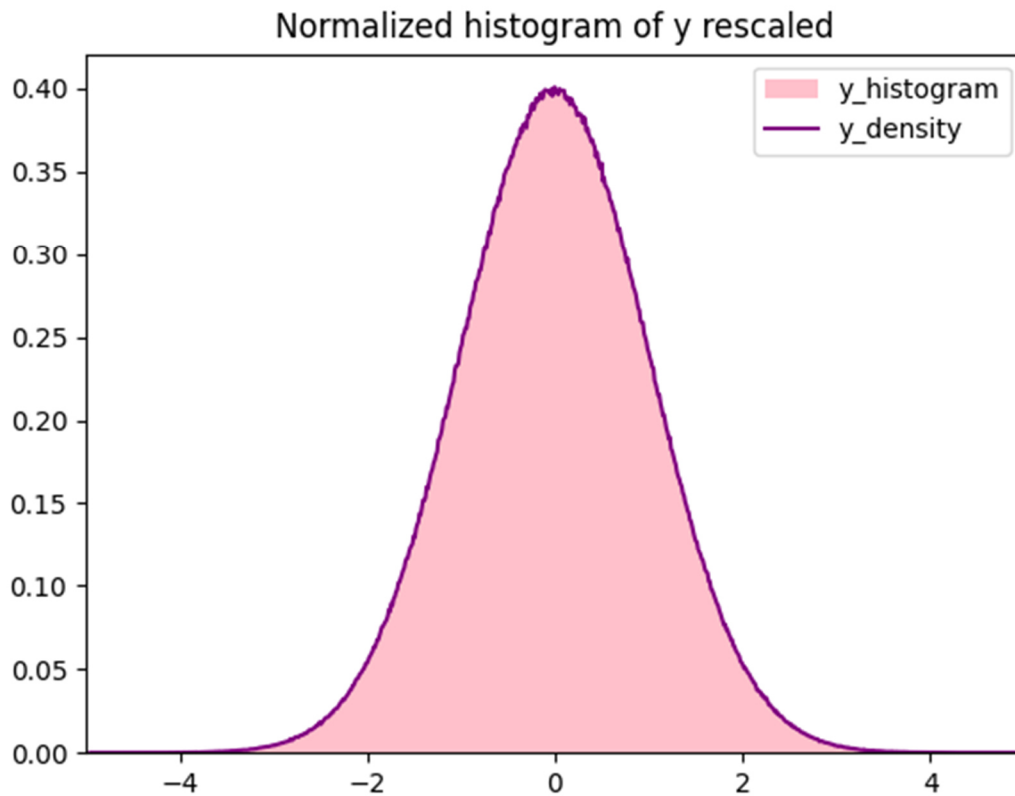
### c. Interpretations:

- We can see that as the rate increase, the error decrease till reaching an error of 0, the rate of communication also decreases.
- On the other hand, the transmission time increase which is logical because the length of the message being sent get bigger.
- The error decreases because the probability of each bit being wrong get lower because there are more bits.
- The communication rate decreases because we send more bits for each message with the length of the alphabet being fixed. (Redundant bits)

## Exercise 3 : Moments

2. The normalized histograms of the data:





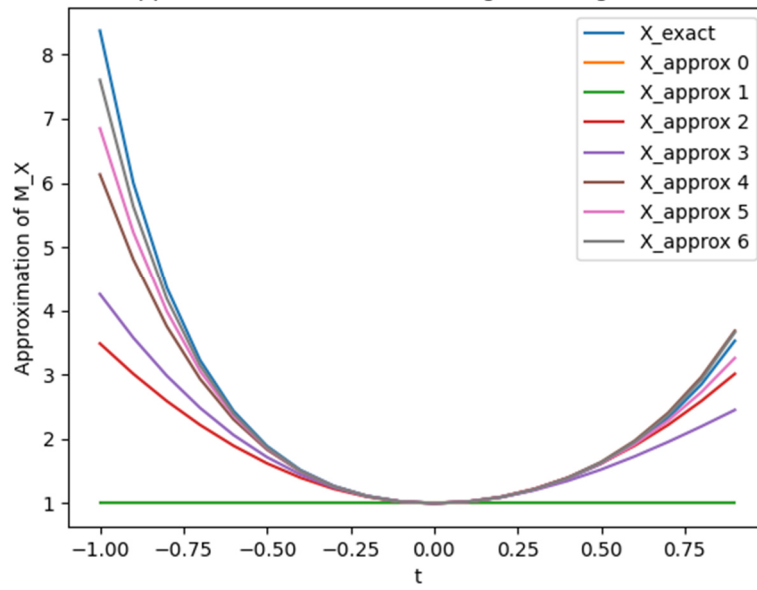
Since  $Y$  values were randomly sampled from the gaussian distribution of mean 0 and variance 5, the plot looks like a gaussian centered at 0 having a variance of 5.

4. The 7 first moments of the centered distribution  $X$  and  $Y$ .

```
Moments of X : [1.0, 5.48443567728444e-15, 4.969513782353428, -4.623405633388906, 45.058973938097395, -86.27514065057433, 542.9180664370957]
Moments of Y : [1.0, 0.00011921914798030496, 1.0001850805456336, 0.0005921135005593106, 2.9990887216169533, -0.002201655712986162, 14.978629542287985]
```

5. For each  $N = [0,1,2,3,4,5,6]$  here are the plots of the moment-generating function between  $[-1,1]$ . We can see in the following pictures that, for the gaussian distribution  $Y$ , the moments of order 1 and 2 are sufficient to capture the distribution, but for a distribution more “real” like  $X$  we need much more moments to capture the distribution. In particular, we see that even with the 7 first moments we do not obtain an approximation as good as the 2 moments for the gaussian.

Approximation of the moment-generating function



Approximation of the moment-generating function

