

UNIVERSITÉ DE GENÈVE

ANALYSE ET TRAITEMENT DE L'INFORMATION

14X026

TP 1: Linear Algebra

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**UNIVERSITÉ
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Exercise 1. Chain rules for probabilities

1. 1

(a) $P(X, Y, Z) = P(X) \cdot p(Y|X) \cdot p(Z|X, Y)$ \rightarrow (apply rule)

(b) $P(X, Z) = P(X) \cdot P(Z|X)$

(c) $P(X, Y) = P(X) \cdot P(Y|X)$

(d) $P(Y, Z) = P(Y) \cdot P(Z|Y)$

2. 2

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

(b) $P(X_1, X_3, X_5, X_7) = P(X_1) \cdot P(X_3|X_1) \cdot P(X_5) \cdot P(X_7|X_5, X_3)$

(c) $P(X_2, X_4, X_6, X_7) = P(X_2) \cdot P(X_4) \cdot P(X_6|X_4, X_2) \cdot P(X_7|X_6)$

(d) $P(X_3, X_6, X_7) = P(X_3) \cdot P(X_6|X_3) \cdot P(X_7|X_6)$

(e) $P(X_1, X_2, X_4, X_5) = P(X_1) \cdot P(X_2) \cdot P(X_4) \cdot P(X_5)$

3. 3

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

(b) $P(X_1, X_3, X_5, X_7) = P(X_1) \cdot P(X_3|X_1) \cdot P(X_5) \cdot P(X_7|X_5, X_3)$

(c) $P(X_2, X_4, X_6, X_7) = P(X_2) \cdot P(X_4) \cdot P(X_6|X_4, X_2) \cdot P(X_7|X_6, X_4)$

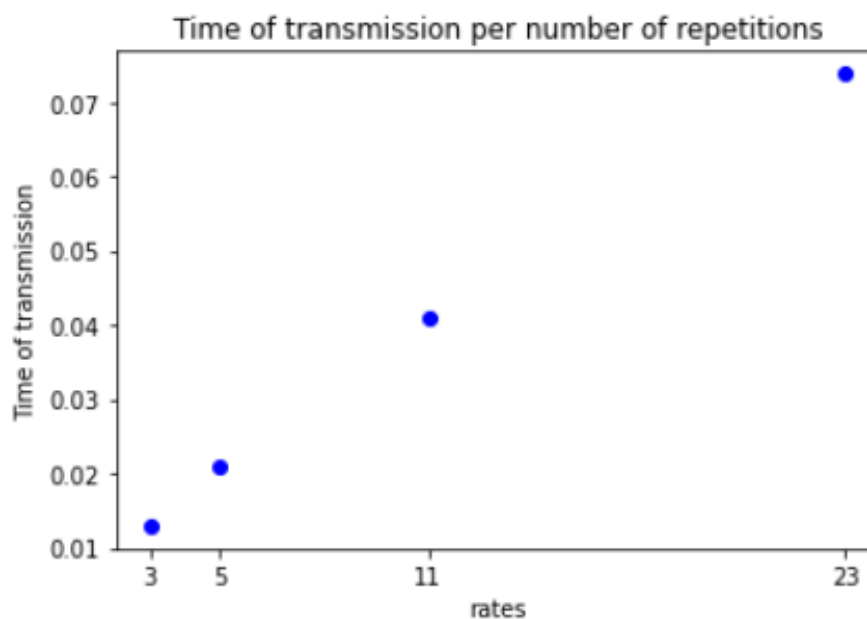
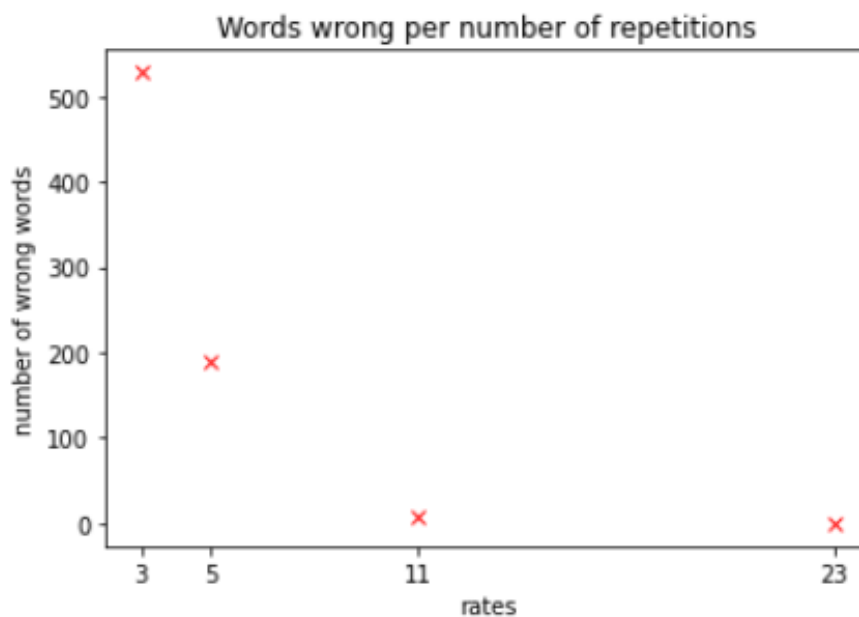
(d) $P(X_3, X_6, X_7) = P(X_3) \cdot P(X_6|X_3) \cdot P(X_7|X_6)$

(e) $P(X_1, X_2, X_4, X_5) = P(X_1) \cdot P(X_2) \cdot P(X_4) \cdot P(X_5|X_2)$

Exercise 2. Communication through noisy channels

See file `exo_2.ipynb` for code

1. There are on average 50% wrong words, which is expected, as each word is composed of 5 bits, so 2 words make up a total of 10 bits, since there is a 10% bit error rate, 1 out of every 2 words should be wrong.
2. As expected the error rate is much lower.
3. By adding repetitions, the error rate becomes lower
4. Obviously by making the message longer, we increase the transfer time

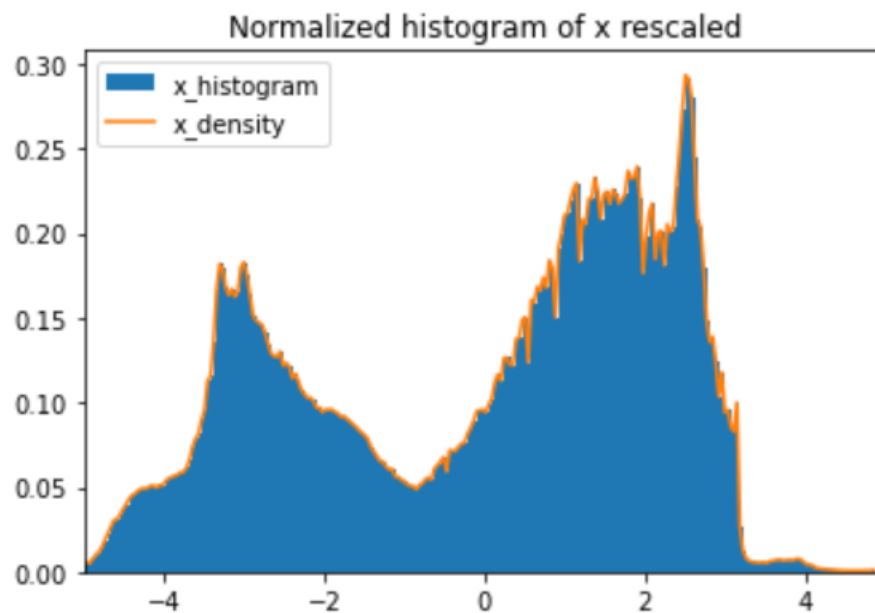


Exercise 3. Moments

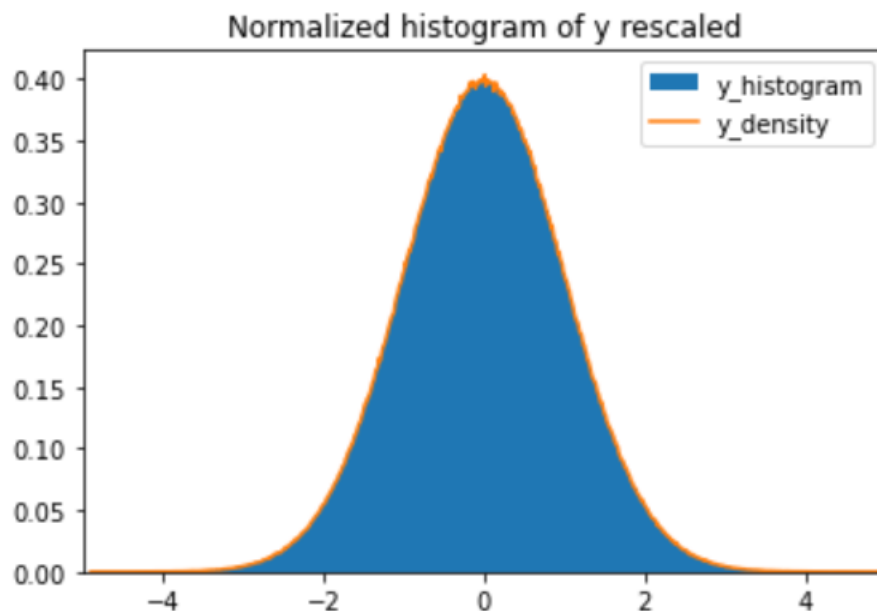
See file exo_3.ipynb for code

```
x = 5 * (x - np.mean(x)) / np.max(np.abs(x - np.mean(x)))
# rescaling between -5 and 5
y = np.random.randn(len(x))
y = y[-5 < y]
y = y[y < 5]
```

1.



2.



```
def mfx(t, x_X, y_X):
    return np.dot(y_X * np.exp(t * x_X[: -1]), x_X[1:] - x_X[: -1])

def mfy(t):
    return integrate.quad(lambda x: np.exp(-x ** 2 / 2) * np.exp(x * t) / np.sqrt(2 * np.pi), -5, 5)[0]
```

3.

```
def moments(data, n):  
    return [np.sum((data ** k) / len(data) for k in range(n))  
  
def amf(t, m):  
    u = [t ** k / np.math.factorial(k) for k in range(len(m))]  
    return np.dot(u, m)
```

```
n_m = 7  
m_X = moments(x, n_m)  
print('Moments of X : ', m_X)  
m_Y = moments(y, n_m)  
print('Moments of Y : ', m_Y)
```

Moments of X : [1.0, 5.48443567728444e-15, 4.969513782353428, -4.623405633388906, 45.058973938097395, -86.27514065057433, 542.9180664370957]

Moments of Y : [1.0, 0.0003855931791905327, 1.0000755504445094, 0.0009402074440729503, 2.998895800364789, 0.012145451654388146, 14.974178113627703]

4.