

Signal and Image Processing by Computer (236327)

Spring 2018

Homework #1

Publication: April 16th, 2018

Submission: by May 2nd at 13:30

Guidelines:

- Submission is only in pairs.
- Submit your solution to the course box at Taub floor 1.
- Matlab code should be electronically submitted via the course website.
- You may answer in Hebrew or English, in a clear-handwrite or a printed form.
- Follow the FAQ at course website, as clarifications and corrections may be published only there.

הנחיות:

- **תאריך ושעת הגשה:** עד ה-2.5.2018 בשעה 13:30.
- ההגשה בזוגות בלבד.
- ההגשה ידנית לתא הקורס בטאוב קומה 1.
- את קוד המטלב יש להגיש אלקטרונית באתר הקורס.
- ניתן לענות בעברית או באנגלית, בכתב ברור או מודפס.
- חובה לעקוב אחרי FAQ באתר. הבהרות ותיקונים בקשר לתרגיל יפורסמו שם בלבד.

Question #1 (25 points) –**Optimal Quantization for Minimum Expected Absolute Deviation**

In this question we consider the design of a scalar quantizer for the criterion of minimum expected absolute-deviation.

The input x is a realization of the random variable X that follows the probability density function $p(x)$. The PDF $p(x)$ is assumed to be positive only for $x \in [\varphi_L, \varphi_H]$.

The expected absolute-deviation is defined as

$$E\varepsilon_Q^1 \triangleq \int_{\varphi_L}^{\varphi_H} |x - Q(x)| p(x) dx$$

The considered quantizer uses b bits for representing values, providing $K = 2^b$ representation levels. The representation levels are denoted as $\{r_i\}_{i=1}^J$ and the decision levels are $\{d_i\}_{i=0}^K$.

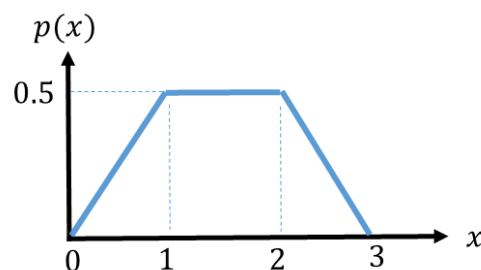
- a. Formulate the **optimization problem** for designing a b -bit quantizer that minimizes the expected absolute-deviation for a given input PDF $p(x)$.
- b. Develop the conditions for optimal **representation** levels given the **decision** levels. Which mathematical operation do the optimality conditions resemble?
- c. Develop the conditions for optimal **decision** levels given the **representation** levels.
- d. Formulate the **Max-Lloyd procedure** for designing a b -bit quantizer that minimizes the expected absolute-deviation for a given input PDF $p(x)$.

Question #2 (25 points)

Consider a random variable that corresponds to the following probability density function:

$$p(x) = \begin{cases} \frac{x}{2} & , \text{ for } x \in [0,1) \\ \frac{1}{2} & , \text{ for } x \in [1,2) \\ \frac{1}{2} - \frac{1}{2}(x-2) & , \text{ for } x \in [2,3) \\ 0 & , \text{ otherwise} \end{cases}$$

that is graphically described as



We consider here the design of a quantizer having 3 representation values: r_1, r_2, r_3 .

The corresponding decision levels are: $d_0 = 0, d_1, d_2, d_3 = 3$.

The optimizations in this question are in the sense of **minimum expected squared error**.

Consider the Max-Lloyd optimization for the initialization of the decision levels to:

$$d_0^{(0)} = 0, \quad d_1^{(0)} = 1, \quad d_2^{(0)} = 2, \quad d_3^{(0)} = 3$$

- a. What are the optimal representation values,

$$r_1^{(1)}, r_2^{(1)}, r_3^{(1)}$$

calculated in the first iteration of the Max-Lloyd procedure ?

- b. What are the optimal decision levels,

$$d_0^{(1)}, d_1^{(1)}, d_2^{(1)}, d_3^{(1)}$$

computed in the first iteration of the Max-Lloyd procedure ?

Question #3 (15 points) – The Miraculous Expressions of Optimal Quantization in the Sense of Minimum Expected Squared-Error

In this question we consider the design of a scalar quantizer for the criterion of minimum expected squared-error.

The quantizer consists of K representation values, $\{r_i\}_{i=1}^K$, and $K + 1$ decision levels $\{d_i\}_{i=0}^K$. Accordingly, the mapping function is $Q(x) = r_i$ for $x \in [d_{i-1}, d_i)$.

The input x is a realization of the random variable X that follows the probability density function $p_X(x)$.

The expected squared-error for a quantizer of K representation values is

$$E\varepsilon_Q^2 \triangleq \sum_{i=1}^K \int_{d_{i-1}}^{d_i} (x - r_i)^2 p_X(x) dx$$

- a. Let us define $Y \triangleq Q(X)$.
Explain why Y is a discrete random variable, and write its probability mass function (PMF), denoted as $P_Y(y)$, using the PDF $P_X(x)$.
- b. Prove (in detail) that the **optimal** expected squared-error holds

$$\sum_{i=1}^K \int_{d_{i-1}^{opt}}^{d_i^{opt}} (x - r_i^{opt})^2 p_X(x) dx = \text{Var}(X) - \text{Var}(Y)$$

where $\text{Var}(X)$ and $\text{Var}(Y)$ are the variances of X and Y , respectively.

In particular, consider the case where X has a non-zero mean.

Matlab Part (35 points)

Instructions:

- In this part use a grayscale image (of 256 graylevels) and of a size of at least 256x256 pixels.
You can transform RGB-color image to a grayscale image using the matlab function `rgb2gray`.
 - Along this part show the image in the same size and dynamic range (pass [0,255] to the `imshow` function).
 - Figures should be appropriately titled.
 - Your written report should be submitted with the dry part, whereas the code should be electronically submitted.
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1. We would like to estimate the probability-density-function of the graylevels in the image using the image histogram. This can be calculated using the Matlab function `imhist` (Note that `imhist` is sensitive to the graylevel range, e.g. [0,1], [0,255], etc.). If the histogram seems too uniform, please pick another image with a non-uniform distribution.
 2. Apply uniform quantization on the image using b bits per pixel.
 - a. Show the MSE as a function of the bit-budget b for $b = 1, \dots, 8$.
 - b. Plot the decision and representation levels for representative b values.
 3. Implement the Max-Lloyd quantizer.
 - a. Show the MSE as a function of the bit-budget b (for $b = 1, \dots, 8$).
 - b. Plot the decision and representation levels for representative b values.
 - c. Compare the results to those of the uniform quantizer. Explain the differences.