Analog to Digital Conversion: Quantization

from the Perspective of Pattern Recognition

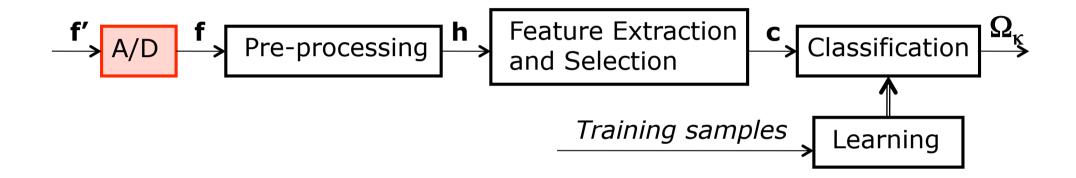


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Pattern Recognition Pipeline

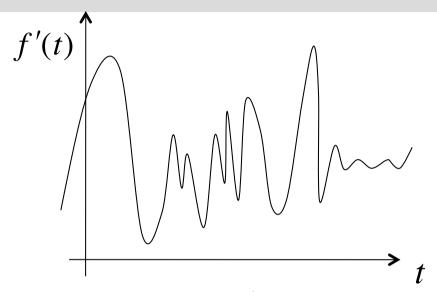




■ The goal of analog to digital conversion is to gather sensed data f' and change it to a representation that is amenable to further digital processing.

Need for A/D Conversion



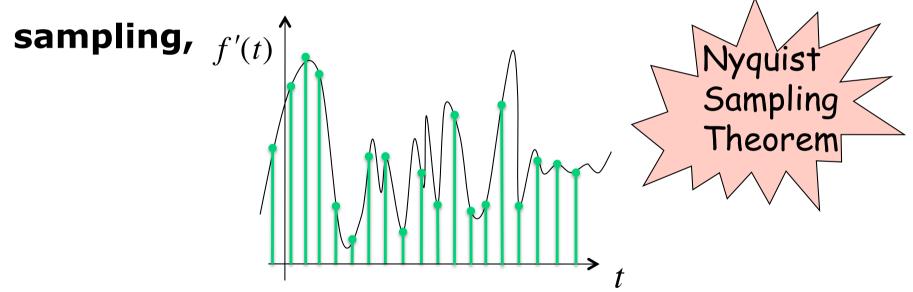


- \blacksquare Continuous range of t values
- lacksquare Continuous range of amplitude f'(t) values.
- We can only store a finite amount of values
- in a finite number of bits (discrete values).
- Goal: Find a discrete representation such that the original analog signal can be accurately reconstructed.

A/D Conversion Steps



- The A/D conversion (coding) involves:
- measuring the amplitude values (or function values) at a finite number of positions:

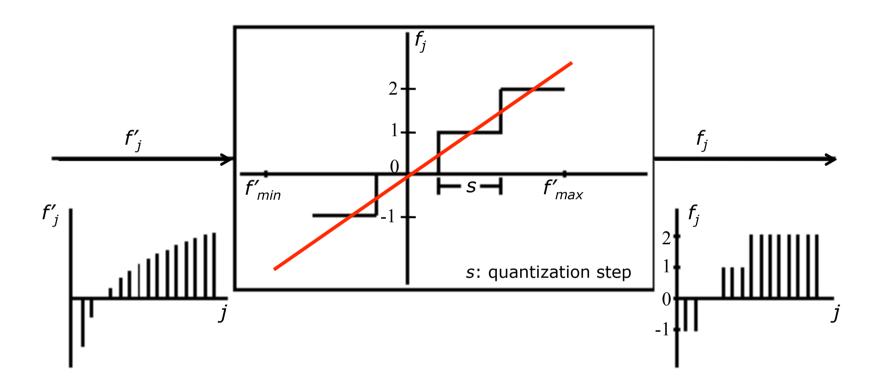


2. representing the amplitude values by a finite number of natural numbers:

quantization

Quantization





■ The number of quantization steps is defined by the number of bits we use to represent the value of the function.

Bits



- Two key questions:
 - 1. How many bits?
 - 2. How do we use these bits?
- When we use B bits, we get 2^B quantized levels.
- Examples:
 - intensity images: B = 8-12, 256 4,096 different gray values.
 - medical images: B = 10 16, 1024 65,536 different gray values.
 - color images: B = 24-36, 8-12 for each color channel, at least 16 million colors.
- Typical data sizes for a 1024 x 1024 (1 MP) image:
 - at 8 bits => 1MB/img => a movie at 30fps creates 30MB/sec
 - at 12 bits => almost 1.6 MB/img => at 30 fps we get 47MB/sec
 - at 24 bits => 3.1 MB/img => at 30 fps we get 93MB/sec
 - => a 5 minute movie needs 27GB.

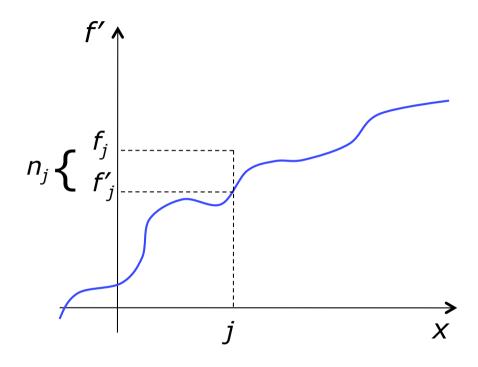
Audio vs. Video Data Rates



| Туре | Specifications | Data Rate |
|-----------------------|------------------------------|--------------|
| Audio, understandable | 1 channel, 8kHz @ 8 bits | 64 kbit/sec |
| Audio, MPEG encoded | CD equivalence | 384 kbit/sec |
| Audio, CD quality | 2 channels, 44.1kHz @16 bits | 1.4 Mbit/sec |
| Video, MPEG-2 | 640 × 480, 24 bits/pixel | 0.42 MB/sec |
| Video, NTSC | 640 × 480, 24 bits/pixel | 27 MB/sec |
| Video, HDTV | 1280 × 720, 24 bits/pixel | 81 MB/sec |

Quantization Error





• Quantization Error: The error we make when we approximate a real value f_j' by a discrete value f_j :

$$n_j = f_j' - f_j$$

Signal-to-Noise Ratio (SNR)



- There exists a standardized way of expressing the noise in a system or sensor that is associated with quantization. It is called the *Signal-to-Noise Ratio*.
- SNR is a general measure that is used for different types (sources) of noise.

 D
- In Engineering SNR is a power ratio: $SNR = \frac{P_{signal}}{P_{noise}}$
- Within the context of pattern recognition, because of the uncertainty involved in the input signal, SNR is the ratio of the expected signal over the expected quantization noise. $E\{f'^2\}$

$$SNR = \frac{E\{f'^2\}}{E\{n^2\}}$$

Signal-to-Noise Ratio (SNR) - continued



■ The Signal-to-Noise Ratio is defined as:

$$SNR = r' = \frac{E\{f'^2\}}{E\{n^2\}}$$

where the quantization noise n is $n_j = f_j' - f_j$.

■ The expected value $E\{\}$ is defined as:

$$E\{x\} = \int_{-\infty}^{\infty} x p(x) dx$$

where x is a random variable, and p(x) is the probability density function (pdf) of x, which tells us how often different values of x occur.

■ So, similar information on f' can guide us on how many bits to use.

SNR and logarithmic scale



Because input signals can have a wide dynamic range, SNR is usually expressed in terms of the logarithmic decibel scale:

$$SNR_{dB} = r = 10\log_{10}\frac{E\{f'^2\}}{E\{n^2\}} = 10\log_{10}(r')$$

Do we want a small or a large SNR? Why? Large is better.

We want over 30dB SNR. Systems with 60dB are considered very good.

Does One Bit Make a Difference?



- Important question: How many bits should one use when quantizing a particular family of functions/ signals (i.e. medical images, or remote sensing data etc.)?
- Does one additional bit make a difference?
- Under certain assumptions (see next slide), the SNR is directly proportional to the number of bits used for quantization:

$$SNR_{db} = r = 6B - 7.2$$

This means that 1 extra bit can increase the SNR by 6dB.

Assumptions



1. On average we have white noise.

$$E\{\vec{n}\} = 0 \text{ and } E\{\vec{n}\vec{n}^T\} = \sigma I$$

- 2. We have a signal with $E\{f'\}=0$.
- 3. The error (noise) is uniformly distributed.
- 4. The signal values lie in a limited range:

$$-4\sigma_{f'} \leq f' < 4\sigma_{f'}$$

If we have a normal distribution, then about 68% of the values lie within 1 σ of the mean, about 95% of the values lie within 2 σ of the mean, about 99.7% of the values lie within 3 σ of the mean, about 99.99% of the values lie within 4 σ of the mean.

So **if** the values of f' follow a normal distribution, assumption 4 is reasonable.

Assumptions 1 and 3



■ We have uniformly distributed white noise, $E\{n\} = 0$. Let s be the quantization step (quantization interval). Then p(n) will be of the form:

$$p(n) = \begin{cases} 1/\text{ for } -\left(\frac{s}{2}\right) \le n \le \frac{s}{2} \\ 0 \text{ otherwise} \end{cases}$$

The width of the pdf has to be s and centered around the value 0 (since $E\{n\}=0$), and the integral of the pdf has to sum up to 1 by definition.

SNR Denominator



- Recall that $SNR = r' = \frac{E\{f'^2\}}{E\{n^2\}}$
- What is $E\{n^2\}$?
- The definition of expected value is $E\{x\} = \int xp(x)dx$.

■ Thus,
$$E\{n^2\} = \int_0^\infty n^2 p(n) dn$$

$$= \int_{-\frac{s}{2}}^{-\frac{s}{2}} n^2 p(n) dn = \frac{1}{s} \int_{-\frac{s}{2}}^{\frac{s}{2}} n^2 dn$$

$$= \frac{1}{s} \frac{1}{3} \left[n^3 \right]_{-\frac{s}{2}}^{\frac{s}{2}} = \frac{1}{s} \frac{1}{3} \left(\frac{s^3}{8} - \left(-\frac{s^3}{8} \right) \right) = \frac{s^2}{12}$$

$$E\{n^2\} = \frac{s^2}{12}$$
 (1)

Assumption 2



- We have a signal with $E\{f'\}=0$.
- According to the definition of standard deviation:

$$\sigma_{f'} = \sqrt{E\{f'^2\} - (E\{f'\})^2}$$

■ However, by assumption 2, we get

$$\sigma_{f'} = \sqrt{E\{f'^2\}}$$

$$\sigma_{f'}^2 = E\{f'^2\}$$
 (2)

Assumption 4



- The signal values lie in the range: $-4\sigma_{f'} \le f' < 4\sigma_{f'}$
- lacksquare So the length of the interval of the f' values is $8\sigma_{f'}$
- When we use B bits to store these $8\sigma_{f'}$ values, we have 2^B quantization levels.
- Assuming equidistant quantization, each quantization step, s, is

$$s = \frac{8\sigma_{f'}}{2^B} \tag{3}$$

Assumption Combination



So far, by exploiting the 4 assumptions we have shown:

$$E\{n^2\} = \frac{s^2}{12}$$
 (1)

$$\sigma_{f'}^2 = E\{f'^2\}$$
 (2)

$$s = \frac{8\sigma_{f'}}{2^B} \tag{3}$$

■ From (1) and (3):

$$E\{n^2\} = \frac{2^6 \sigma_{f'}^2}{12 \cdot 2^B} \quad (4)$$

$$E\{f'^2\}$$

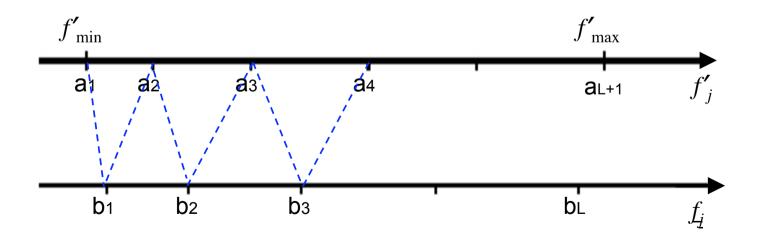
From (1) and (3): $E\{n^{2}\} = \frac{2^{6}\sigma_{f'}^{2}}{12\cdot 2^{B}}$ (3) Recall that SNR is defined as $r' = \frac{E\{f'^{2}\}}{E\{n^{2}\}}$

$$r' = \sigma_{f'}^2 / \left(\frac{2^6 \sigma_{f'}^2}{12 \cdot 2^B}\right) = \frac{12 \cdot 2^B}{2^6} = 12 \cdot 2^{B-6} \qquad r = 10 \log_{10} r' = 6B - 7.27$$

Mapping



- Using SNR as a criterion, we know how many bits to use, but how do we use them?
- To which discrete value do we map a continuous interval?



Good Mapping



- How can I tell whether my mapping is good?
- What is a possible objective function, a criterion to judge the quality of the mapping?
- \blacksquare Error measure (error that occurs when mapping f' to b_v)

$$\varepsilon = \sum_{v=1}^{L} \int_{a_{v}}^{a_{v+1}} (f' - b_{v})^{2} p(f') df'$$

- By weighing the error by the probability density of f', values that have a higher probability of occurring have a higher impact on the error term.
- The optimal quantization characteristics are defined by the values a_v , b_v which minimize the error ε .

Optimal Quantization Characteristics



Optimal discrete value:

$$\frac{\partial \mathcal{E}}{\partial b_{v}} = \sum_{v=1}^{L} \int_{a_{v}}^{a_{v+1}} 2(f' - b_{v}) p(f') df' = 0 \qquad v = 1, 2, ..., L$$

$$\int_{a_{v+1}}^{a_{v+1}} f' p(f') df' = b_{v} \int_{a_{v}}^{a_{v+1}} p(f') df' \Leftrightarrow b_{v} = \frac{a_{v}}{a_{v+1}}$$

$$\int_{a_{v}}^{a_{v+1}} f' p(f') df'$$

Optimal threshold level:

$$\frac{\partial \mathcal{E}}{\partial a_{v}} = \partial \left[\sum_{v=1}^{L} \int_{a_{v}}^{a_{v+1}} (f' - b_{v})^{2} p(f') df' \right] / \partial a_{v} = 0$$

$$(a_{v} - b_{v-1})^{2} p(a_{v}) - (a_{v} - b_{v})^{2} p(a_{v}) = 0 \Leftrightarrow a_{v} = \frac{b_{v} + b_{v-1}}{2}$$

Pulse Code Modulation



- A linear quantization characteristic function (with equally spaced quantization levels) is an optimal quantization if and only if the signal amplitudes are equally distributed.
- Coding using the methods introduced so far is called Pulse Code Modulation.
- Other coding methods, depending on the application are:
 - Coding with a minimal number of bits
 - Error detection and correction
 - Run-length encoding
 - Chain code

Vector Quantization

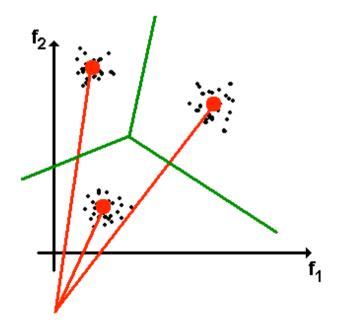


- So far, we have considered the quantization of real valued functions, i.e. $f \in R$.
- There exist signals where we have to deal with vector valued functions, $\vec{f} \in R^N$ (e.g. color images with RGB values).
- The quantization of vectors to discrete vectors is called vector quantization.
- **Vector quantization** is the process of mapping N-dimensional vectors in the vector space R^N into a finite set of vectors $Y = \{\vec{y}_i | i = 1 \cdots k\}$, where k < N.
- **Each** vector \vec{y}_i is called a **code** vector or a **codeword**.
- \blacksquare The set of all the codewords, Y, is called a **codebook**.

Codebook Design



- There exist many vector quantization methods.
- We are just going to present one method which is based on mean values.
- Another one is based on computing nearest neighbor regions, aka Voronoi regions.



Using the Mean Vectors



- For each cluster in the training data compute the mean vector $\vec{\mu}_i$.
- Each mean vector $\vec{\mu}_i$ becomes the code vector or codeword, \vec{y}_i .
- All the mean vectors define the so-called code book, Y.
- Given an arbitrary input vector \vec{f}_j' find the nearest code vector \vec{y}_u , s.t. $u = \min_i d(\vec{f}_j', \vec{y}_i)$.
- Store the offset to the closest mean \vec{y}_u . There is a finite number of bits that can be used for the offset.
- Use your favorite distance metric, e.g. Euclidean, Manhattan, etc. We often use the Euclidean distance.

Computing the Codebook



- k-means algorithm
- k:# of code vectors
- Input: M data vectors $\vec{f}_1, \vec{f}_2, ..., \vec{f}_M, \vec{f}_i \in R^N$
- 1. Randomly assign the vectors $\vec{f}_1, \vec{f}_2, \dots, \vec{f}_M$ to k clusters.
- 2. Compute the mean vector $\vec{\mu}_i$ for each cluster.
- 3. Reassign each vector $\vec{f}_1, \vec{f}_2, \dots, \vec{f}_M$ to the cluster with the nearest mean vector $\vec{\mu}_i$.
- 4. Repeat 2. and 3. until no further changes occur
- Output: code book

Linde-Buzo-Gray Algorithm



- The Linde-Buzo-Gray (LBG) algorithm is a widely-used vector quantization algorithm which is very similar to the k-means algorithm.
- Main idea. Start with a single code vector. At each iteration, each code vector is split into two new vectors.
- 1. Initial state: compute the mean of the training data.
- 2. Initial estimation #1: code book of size 2.
- 3. Final estimation for code book of size 2, after training data reassignment.
- 4. Initial estimation #2: code book of size 4.
- 5. Final estimation for code book of size 4, after training data reassignment. ...