

Lecture 7: Analogue to Digital Conversion

EE213 - Introduction to Signal Processing

Semester 1, 2020

- Introduce A/D conversion.
- Sampling:
 - Characterize the frequency domain representation of the sampling process.
 - Explain the aliasing phenomena and identify the sampling rate.
- Quantization:
 - Describe uniform quantisation and determine the signal to quantisation noise ratio.
- Coding:
 - Represent the quantization value using binary digits.

A/D : Analogue to Digital

- Most signal processing these days is done digitally.
- For this reason, analogue signals must first be converted into digital signals first. i.e., discrete-time & discrete-value.
- Two Step:
 - Sampling: Continuous-time -> Discrete-time
 - Quantization: Continuous-value -> Discrete-value
 - Coding: Discrete-value -> Binary representation

Sampling: Continuous Time to Discrete Time

- Under certain conditions, a continuous-time signal can be completely represented by and recoverable from knowledge of its values, or samples, at points equally spaced in time.
- The first step in analogue-to-digital conversion is to sample the signal.
- We are interested in equally spaced sampling, which is also known as **periodic sampling**.

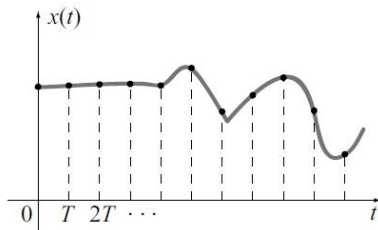
Sampling...

- **Periodic sampling:** **equally spaced sampling**

Given the sampling period $T > 0$, we convert a continuous-time signal $x(t)$ into a discrete-time signal $x[n]$, where

$$x[n] = x(nT), n = \dots, -2, -1, 0, 1, 2, \dots \quad (1)$$

$f_s = 1/T$, as the **sampling frequency**.



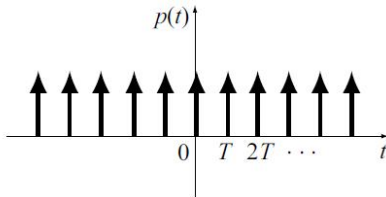
Sampling...

- **Ideal uniform sampling:** using **impulse signal**.

- The impulse train

$$p(t) = \sum_{k=-\infty}^{\infty} \delta(t - kT) \quad (2)$$

is referred to as the sampling function.



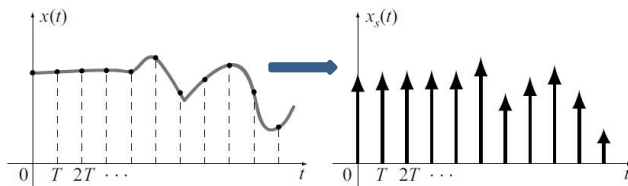
Sampling...

- We can represent **the sampled signal** in continuous time by

$$x_s(t) = x(t)p(t) \quad (3)$$

$$= \sum_{k=-\infty}^{\infty} x(t)\delta(t - kT)$$

In Time Domain:



- **What happens in spectrum domain?**

Sampling...

- The impulse train $p(t)$ is periodic signal, so we can write it as a complex exponential Fourier series

$$p(t) = \sum_{n=-\infty}^{\infty} p_n e^{jn\omega_s t} \quad (4)$$

where $\omega_s = 2\pi/T$ and

$$p_n = \frac{1}{T} \int_{-T/2}^{T/2} p(t) e^{-jn\omega_s t} dt = \frac{1}{T} (!!!) \quad (5)$$

- Thus, the impulse train equals

$$p(t) = \frac{1}{T} \sum_{n=-\infty}^{\infty} e^{jn\omega_s t} \quad (6)$$

Sampling...

- We have

$$x_s(t) = x(t)p(t) = \frac{1}{T} \sum_{n=-\infty}^{\infty} x(t)e^{jn\omega_s t} \quad (7)$$

Therefore

$$X_s(\omega) = \mathcal{F}\left(\frac{1}{T} \sum_{k=-\infty}^{\infty} x(t)e^{jn\omega_s t}\right) \quad (8)$$

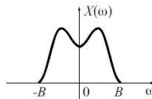
$$= \frac{1}{T} \sum_{n=-\infty}^{\infty} \mathcal{F}(x(t)e^{jn\omega_s t}) \quad (9)$$

$$= \frac{1}{T} \sum_{n=-\infty}^{\infty} X(\omega - n\omega_s) \quad (10)$$

- the Fourier transform of the sampled signal $x_s(t)$ is **a sum of scaled and shifted replicas** of the Fourier transform of the original signal $x(t)$.

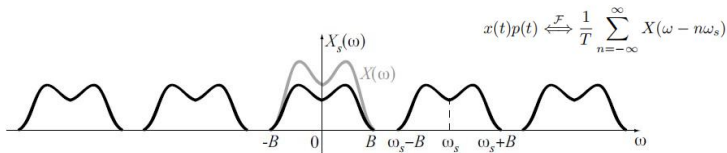
Sampling...

- Assume $X(\omega)$ looks like



that it is band-limited, i.e., $|X(\omega)| = 0$ for $|\omega| > B$. (refer to B as the signal bandwidth.)

- Thus, we have $X_s(\omega)$, **sum of scaled and shifted replicas of $X(\omega)$**



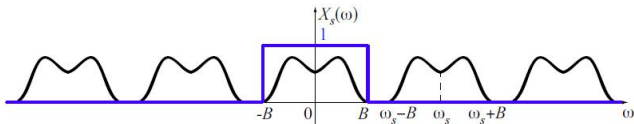
What happens if $\omega_s > 2B$ ----- No overlapping!
What happens if not?

Sampling...

- When the condition $\omega_s > 2B$ is satisfied, we can **recover the original signal $x(t)$ from $x_s(t)$** by lowpass filtering and scaling

$$X(\omega) = T \underbrace{H_{LP}(\omega)}_{\text{lowpass filter}} X_s(\omega) \quad (12)$$

- Illustration of lowpass filtering of $x_s(t)$



Sampling...

- We have the following important result, discovered in various forms by Shannon, Nyquist, Whittaker and Kotelnikov:

The Sampling Theorem.

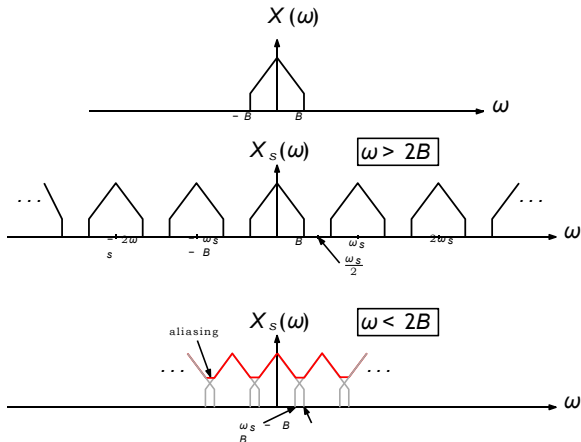
Consider a bandlimited signal $x(t)$ with bandwidth B . Then, provided the sampling frequency $\omega_s > 2B$, the signal $x(t)$ can be recovered exactly from its sampled version $x_s(t)$ by scaling and lowpass filtering.

- The minimum required sampling frequency $\omega_s \triangleq 2B$ is called the **Nyquist frequency** or the **Nyquist sampling rate**.

- If $w_s < 2B$, $X_s(\omega)$ will overlap, called **aliasing**.
- In this case the original signal **cannot be recovered**.

Aliasing...

Illustration of aliasing



Preventing Aliasing

In practice, **two approaches** to prevent aliasing:

- Apply an analog (continuous-time) lowpass filter before sampling to ensure the signal is bandlimited
 - ▶ These lowpass filters are often called anti-aliasing filters
- Use a higher sampling rate than required by **the sampling theorem** (e.g. $w_s = 2.2B$)

Quantisation

- Note that the sampling process produces discrete time, but **continuous-amplitude** signals.
- Converting the continuous-amplitude into discrete-amplitude is called **quantisation**.
- The error between the sampled continuous-valued amplitude and the approximated discrete amplitude is called **quantisation error** or **quantisation noise**.

The output of quantizer is

•

$$x_q[n] = Q(x_a[n]) \quad (14)$$

The quantizer error is

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$$e_q[n] = x_a[n] - x_q[n] \quad (15)$$

Quantization...

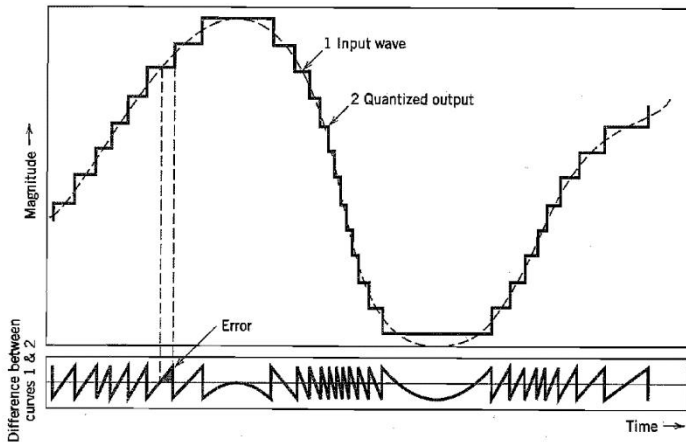


FIGURE 3.11 Illustration of the quantization process. (Adapted from Bennett, 1948, with permission of AT&T.)

Uniform Quantisation

Given the signal is in a finite range (x_{\min} , x_{\max})

- The entire data range is divided into L equal intervals

quantisation Interval: $\Delta = \frac{x_{\max} - x_{\min}}{L}$ (16)

If L increases, Δ decreases. Hence, the quantisation error decreases and the accuracy of the quantizer increases.

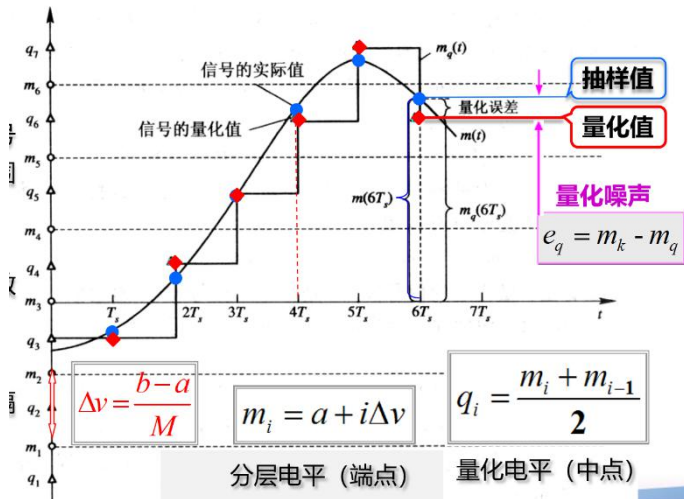
- The values allowed in the digital signal are called **quantisation levels**.

$$\text{Value}_k = x_{\min} + k^* \Delta - \frac{\Delta}{2}$$

- Distance between two quantization levels is called **quantisation step size** or **resolution**, for uniform quantisation, also Δ .

Uniform Quantisation

Sampled value is mapped to the **middle value** of its interval.



Uniform Quantisation...

- The quantisation error is the range of

$$\frac{-\Delta}{2} \leq e_q(n) \leq \frac{\Delta}{2} \iff |e_q(n)| \leq \frac{\Delta}{2} \quad (17)$$

and thus the maximum error is

$$e_q^{\max}(n) = \frac{\Delta}{2} \quad (18)$$

Example

For the following sequence $\{1.2, -0.2, -0.5, 0.4, 0.89, 1.3, \dots\}$, quantise it using a uniform quantiser in the range of $(-1.5, 1.5)$ with 4 levels, and write the quantised sequence.

($\Delta=0.75$, $\{1.125, -0.375, -0.375, 0.375, 1.125, 1.125, \dots\}$)

Uniform Quantisation

- An important indicator for the performance of a quantiser is the ratio of the signal average power to the noise power is the **signal-quantization noise ratio** (SQNR) defined as

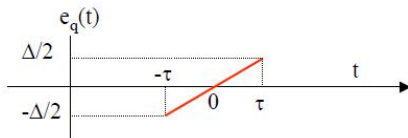
$$\text{SQNR} = \frac{P_x}{P_q} \quad (19)$$

where

- ▶ P_q is the quantisation noise power.
- ▶ P_x is the average power of the signal.

Uniform Quantisation - Quantization Error...

- Example: case when the analogue signal $x_a(t)$ is (almost) linear between quantization levels.
- Consider one interval



- Quantization error:

$$e_q(t) = x_a(t) - x_q(t) \quad (20)$$

$$= \frac{\Delta}{2\tau} t \quad (21)$$

for $-\tau \leq t \leq \tau$.

The power of quantisation error

$$P_q = \frac{1}{\tau} \int_{-\tau}^{\tau} e_q^2(t) dt = \frac{1}{\tau} \int_{-\tau}^{\tau} \left(\frac{\Delta}{2\tau} t \right)^2 dt = \frac{\Delta^2}{12} \quad (22)$$

Uniform Quantisation - Uniformly Distributed Signals

- Consider the case where signals are uniformly distributed over an interval $[-A, A]$
 - ▶ Examples include full amplitude triangle waves and sawtooth waves.
- The quantisation error is uniformly distributed between $\frac{\Delta}{2}$ and $\frac{\Delta}{2}$ and the power of the quantisation error is $P_q = \Delta^2/12$.
- Let b be the number of bits to represent all quantisation levels between $-A$ and A . The quantisation interval is

$$\Delta = \frac{2A}{2^b} \quad (27)$$

- The power of the signal (triangle waves) in this case is given by

$$P_x = 2 \times \frac{1}{T} \int_{-\frac{T}{2}}^0 \left(\frac{4A}{T} t + A \right)^2 dt = \frac{A^2}{3} \quad (28)$$

Uniform Quantisation - Quantization Error...

- The SQNR in this case is given by

$$\text{SQNR} = \frac{A^2/3}{\Delta^2/12} = 2^{2b} \quad (22)$$

- In dB scale

$$\text{SQNR}_{\text{dB}} = 10 \log_{10} 2^{2b} = 6.02 \times b \quad (29)$$

Uniform Quantisation - Sinusoidal Signals

- Example: a sinusoid signal. Its quantisation error $e_q(t)$ is NOT uniformly distributed between $\frac{\Delta}{2}$ and $\frac{\Delta}{2}$.
- The quantisation error is said to be almost linear between $\frac{\Delta}{2}$ and $\frac{\Delta}{2}$ and **the power of quantisation error** is approximated as

$$P_q \approx \frac{\Delta^2}{12} \quad (23)$$

- Assume the signal is a sine wave between $-A$ and A . **The power of the signal** is

$$P_x = \frac{1}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} (A \sin t)^2 dt = \frac{1}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} \frac{1}{2} A^2 (1 - \cos 2t) dt = \frac{A^2}{2}. \quad (24)$$

Uniform Quantisation - Sinusoidal Signals...

- The SQNR in this case is given by

$$\text{SQNR} = \frac{A^2/2}{\Delta^2/12} = \frac{3}{2} 2^{2b} \quad (25)$$

In dB scale

$$\text{SQNR}_{\text{dB}} = 10 \log_{10} \frac{3}{2} 2^{2b} = 1.76 + 6.02b \quad (26)$$

Quantization Example

