

# Continuous-Time Band-limited Gaussian Channel

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## I. CT BAND-LIMITED CHANNEL

Consider a system where input  $x(t)$  and  $z(t)$  are both continuous. The relationship between the two is described as

$$y(t) = x(t) + z(t), \quad (1)$$

where  $x(t)$  is power constrained  $\frac{1}{2T} \int_{-T}^T x^2(t)dt \leq P$ , and  $z(t)$  is continuous Gaussian white noise process, characterized by auto-correlation function,

$$r(\tau) = \mathbb{E}[z(t)z(t-\tau)] = \frac{N_o}{2} \delta(\tau). \quad (2)$$

This states that  $z(t)$  and  $z(t-\tau)$  are uncorrelated for  $\tau \neq 0$ . In frequency domain,  $z(t)$  is characterized by its power spectral density (PSD) described as

$$S(f) = \mathcal{F}\{r(\tau)\} = \frac{N_o}{2}, \quad (3)$$

which states that all the frequency components has constant  $\frac{N_o}{2}$ . How then, do we transmit a continuous signal through this continuous time channel? The idea is to look at the band-width limited version of the channel. We do this by applying  $h(t)$ , a low-pass filter, to  $y(t)$  to produce  $\tilde{y}(t)$ . Since  $\tilde{y}(t)$  is band-limited, we only need to consider band-limited  $x(t)$ ; which allows us to use samples of  $x(t)$  to represent  $x(t)$ . Sampling theorem states that if  $x(t)$  is such that  $X(f)$  is zero outside of  $[-\omega, \omega]$ , then sampled  $x(t)$  at  $2\omega$  samples per second is sufficient for reconstructing  $x(t)$  exactly.

Consider a system with continuous signal  $x(t)$ . We now want to sample this signal by multiplying it with impulse train described as

$$p(t) = \sum_{k=-\infty}^{\infty} \delta(t - kT). \quad (4)$$

Delta function is only non-zero  $t = kT$  and zero else where ( $t \neq kT$ ). In effect this creates a train of impulse (delta) functions at times  $kT$ , where  $k$  is integer from  $-\infty$  to  $\infty$ . The sampling period  $T = \frac{1}{2\omega}$ , where  $\omega$  is the highest frequency to represent  $X(f)$ . Therefore,  $x_s(t) = x(t)p(t)$  effectively samples  $x(t)$  at every sampling period  $T$ . Then, we can describe the sampled signal as

$$x_s(t) = x(t)p(t) \quad (5)$$

$$x_s(t) = x(kT) \sum_{k=-\infty}^{\infty} \delta(t - kT). \quad (6)$$

Now, Fourier transform of  $x_s(t)$  should give

$$X_s(f) = X(f) * P(f), \quad (7)$$

where  $*$  is convolution. We know that

$$\mathcal{F}[p(t)] = P(f) = \frac{1}{T} \sum_{k=-\infty}^{\infty} \delta(f - \frac{k}{T}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} \delta(f - 2\omega k). \quad (8)$$

This leads to

$$X_s(f) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X(f - 2\omega k). \quad (9)$$

This means that  $X(f)$  will repeat every  $2\omega$  in frequency domain. Note that it is crucial to sample at 2 times the maximum frequency component of  $X(f)$  because the components will overlap if the sampling frequency is lower. Now, we have repeating  $X(f)$  every  $2\omega$  is frequency domain. We want to extract out one from each sample; therefore, we pass this with low-pass filter. This filter blocks out frequency components and only save the content in the interval  $[-\omega, \omega]$ . Description of this low pass filter is described as

$$Q(f) = \frac{1}{2\omega}, \text{ in } [-\omega, \omega]. \quad (10)$$

To filter, you simply multiply  $X_s(f)$  by (10) and recover

$$X(f) = X_s(f)Q(f). \quad (11)$$

Now, what is the time version of this filter? We can compute this by using Fourier transform

$$x(t) = \int_{-\infty}^{\infty} X(f) e^{j2\pi ft} df. \quad (12)$$

Now, we find the time version of  $Q(f)$  to be,

$$q(t) = \int_{-\omega}^{\omega} \frac{1}{2\omega} e^{j2\pi ft} df = \frac{e^{j2\pi\omega t} - e^{-j2\pi\omega t}}{j2\pi t 2\omega} = \frac{\sin(2\pi\omega t)}{2\pi\omega t} = \text{sinc}(t). \quad (13)$$

Putting it all together in time domain, you convolve (5) with (13) and get

$$\tilde{x}_s(t) = \sum_{k=-\infty}^{\infty} x(kT) \text{sinc}(t - kT). \quad (14)$$

In effect, we have converted the continuous-time signal  $x(t)$  to a discrete-time signal  $x(kT)$ .

Now, our input has become (14). For notation purpose, denote  $x(kT)$  as  $x_n$ , which denotes  $n^{th}$  sample of  $x(t)$ . We pass this into the channel  $z(t)$  and receive  $y(t)$ . However, how can we get back the samples of  $x_n$  from output now? We use matched filtering on  $y(t)$  to get back  $x_n$ . We do this under the following knowledge that  $\text{sinc}(t - kT)$  is a set of orthogonal basis functions; this is described by

$$\int_{-\infty}^{\infty} \text{sinc}\left(t - \frac{n}{2\omega}\right) \text{sinc}\left(t - \frac{m}{2\omega}\right) dt = \begin{cases} \frac{1}{2\omega} & \text{if } m = n \\ 0 & \text{if } m \neq n \end{cases}. \quad (15)$$

The match filtering then is described as

$$y_n = 2\omega \int_{-\infty}^{\infty} y(t) \text{sinc}\left(t - \frac{n}{2\omega}\right) dt \quad (16)$$

$$= 2\omega \int_{-\infty}^{\infty} \left[ \sum_{k=-\infty}^{\infty} x(kT) \text{sinc}(t - kT) \right] \text{sinc}\left(t - \frac{n}{2\omega}\right) dt. \quad (17)$$

Hence, we get a discrete-time channel:

$$y_n = x_n + z_n, \quad (18)$$

where  $z_n = 2\omega \int_{-\infty}^{\infty} z(t) \text{sinc}\left(t - \frac{n}{2\omega}\right) dt$ .