

Sampling

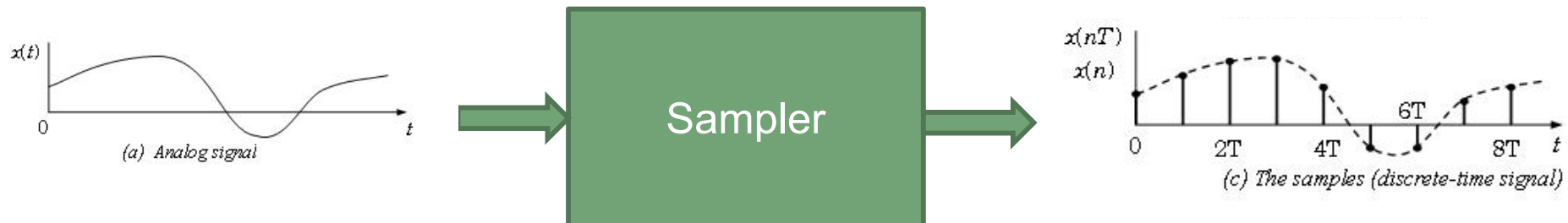
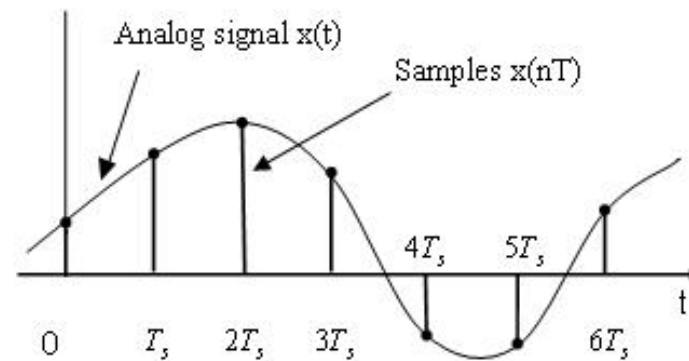
From continuous to discrete time

Lathi Chapt. 5



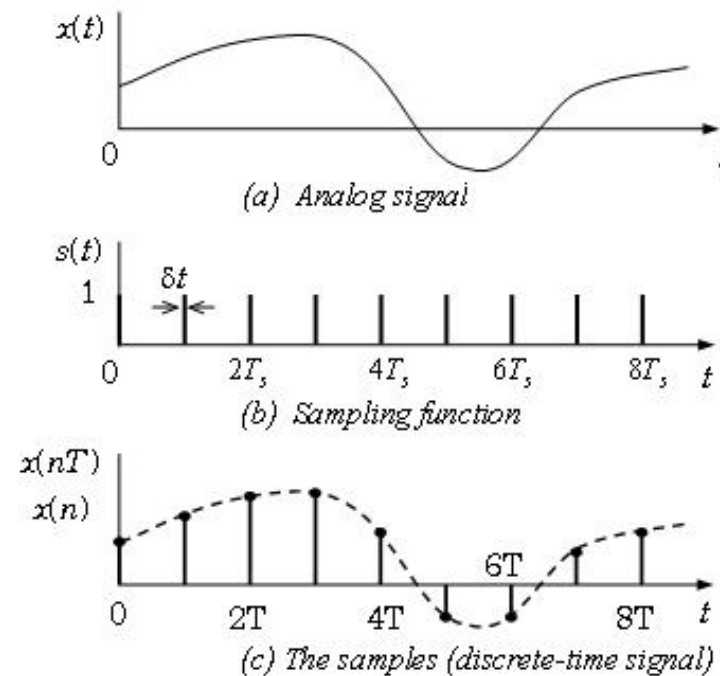
From continuous to discrete time

- Sampling is the process that consists in taking a continuous time signal and transforming it into a discrete time signal



How to obtain it?

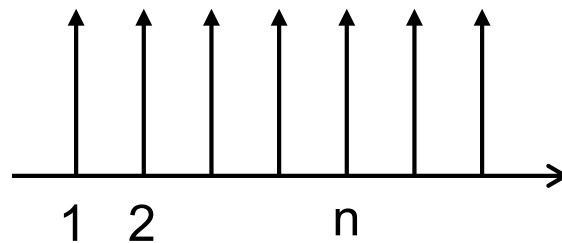
- The main actors are
 - The impulse train
 - The convolution operator



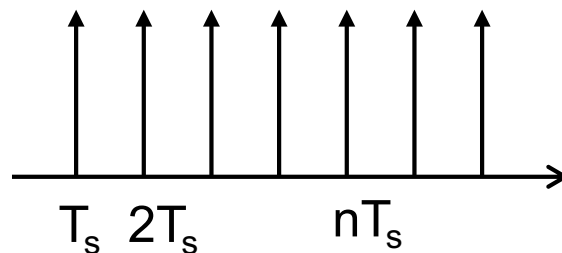
The sampled signal can be seen as the product of the CT signal times the impulse train

What else do we know?

- The product in signal domain corresponds to the convolution in Fourier domain
- The DTFT of a discrete time signals is periodic of period 2π
- So far, we have considered sample indices that were integer numbers which amounts to assume unitary spacing among samples



while for sampled signals the spacing among samples is T_s



*Stretching ($T_s > 1$) or
shrinking ($T_s < 1$) of
the time axis*

What do we need

- F-transform of the impulse train but we know that
 - The impulse train is a periodic function of period T_s

$$S_{T_s}(t) = \sum_{n=-\infty}^{+\infty} \delta(t - nT_s)$$

- Thus its transform is

$$S_{T_s}(t) = \sum_{n=-\infty}^{+\infty} \delta(t - nT_s) \rightarrow S_{T_s}(\omega) = \frac{1}{T_s} \sum_{k=-\infty}^{+\infty} \delta(\omega - n\omega_s)$$

- Thus it is a scaled version of the train of delta where the spacing between deltas is $1/T_s$



Sampling

- Exploiting the convolution theorem

$$f_s(t) = f(nT_s) = f(t)S_{T_s}(t) = \sum_n f(nT_s)\delta(t - nT_s)$$

$$F_s(\omega) = f(\omega) \star S_{T_s}(\omega) = \frac{1}{T_s} \sum_n F(\omega - n\omega_s)$$



Sampling periodizes the spectrum

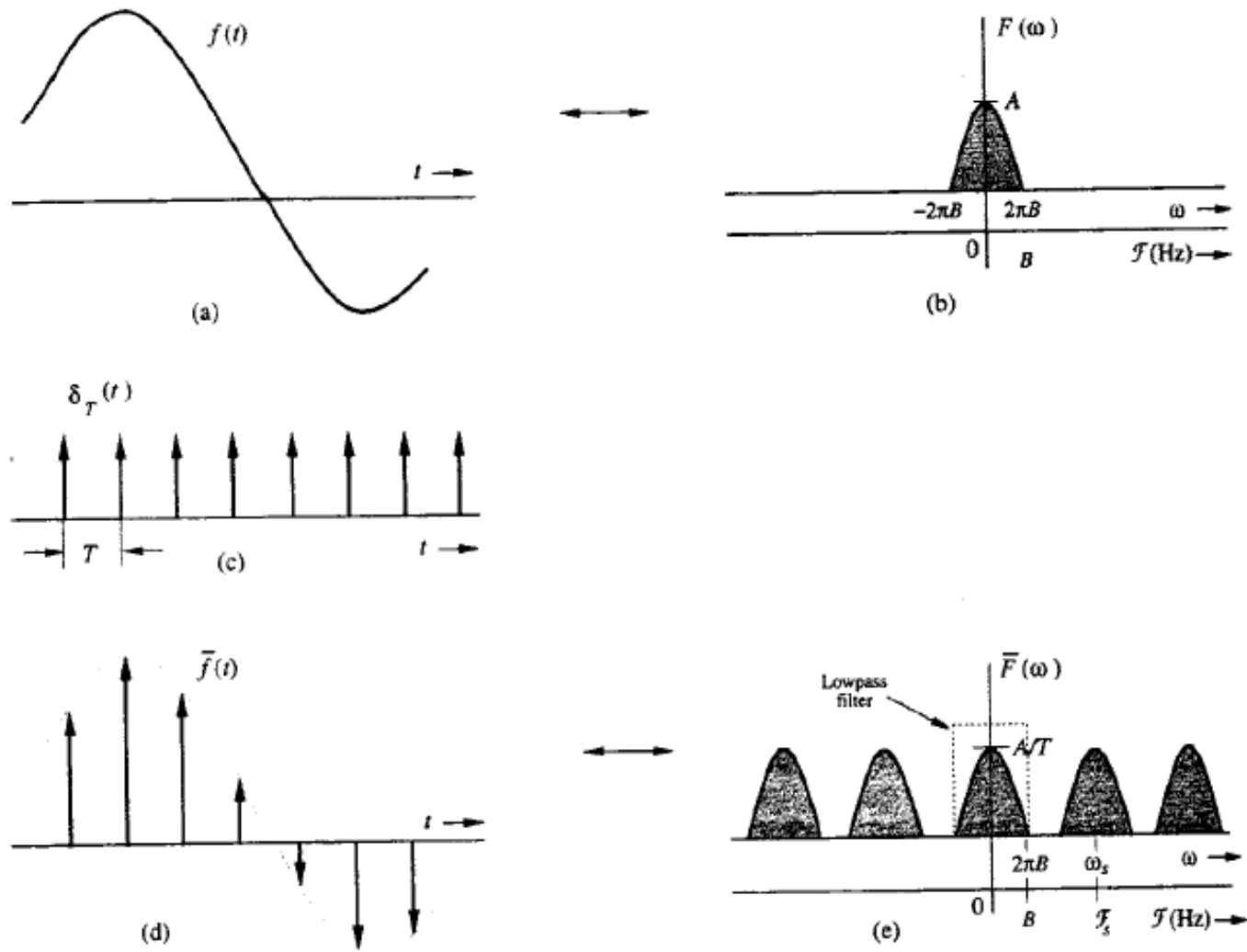


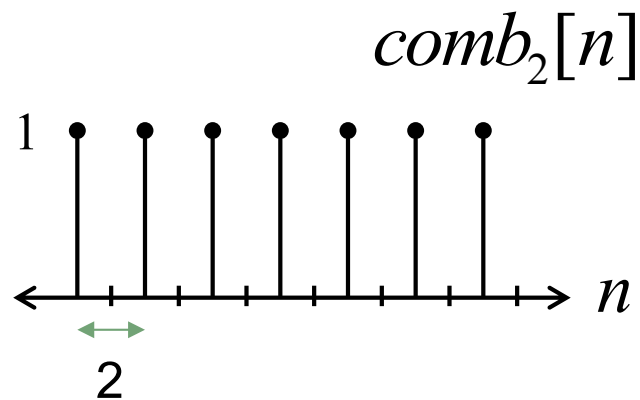
Fig. 5.1 Sampled signal and its Fourier spectrum.

Discrete impulse train

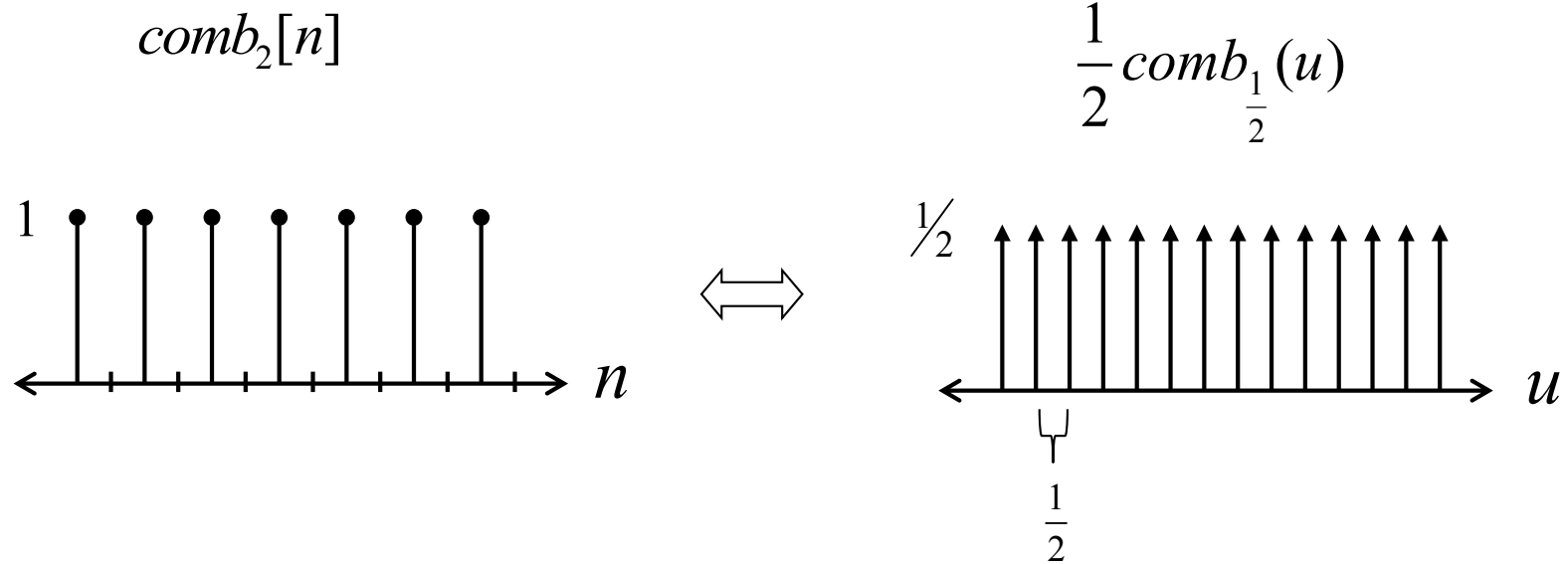
- Define a *comb* function (impulse train) as follows

$$\text{comb}_N[n] = \sum_{l=-\infty}^{\infty} \delta[n - lN]$$

where M and N are integers



Impulse Train

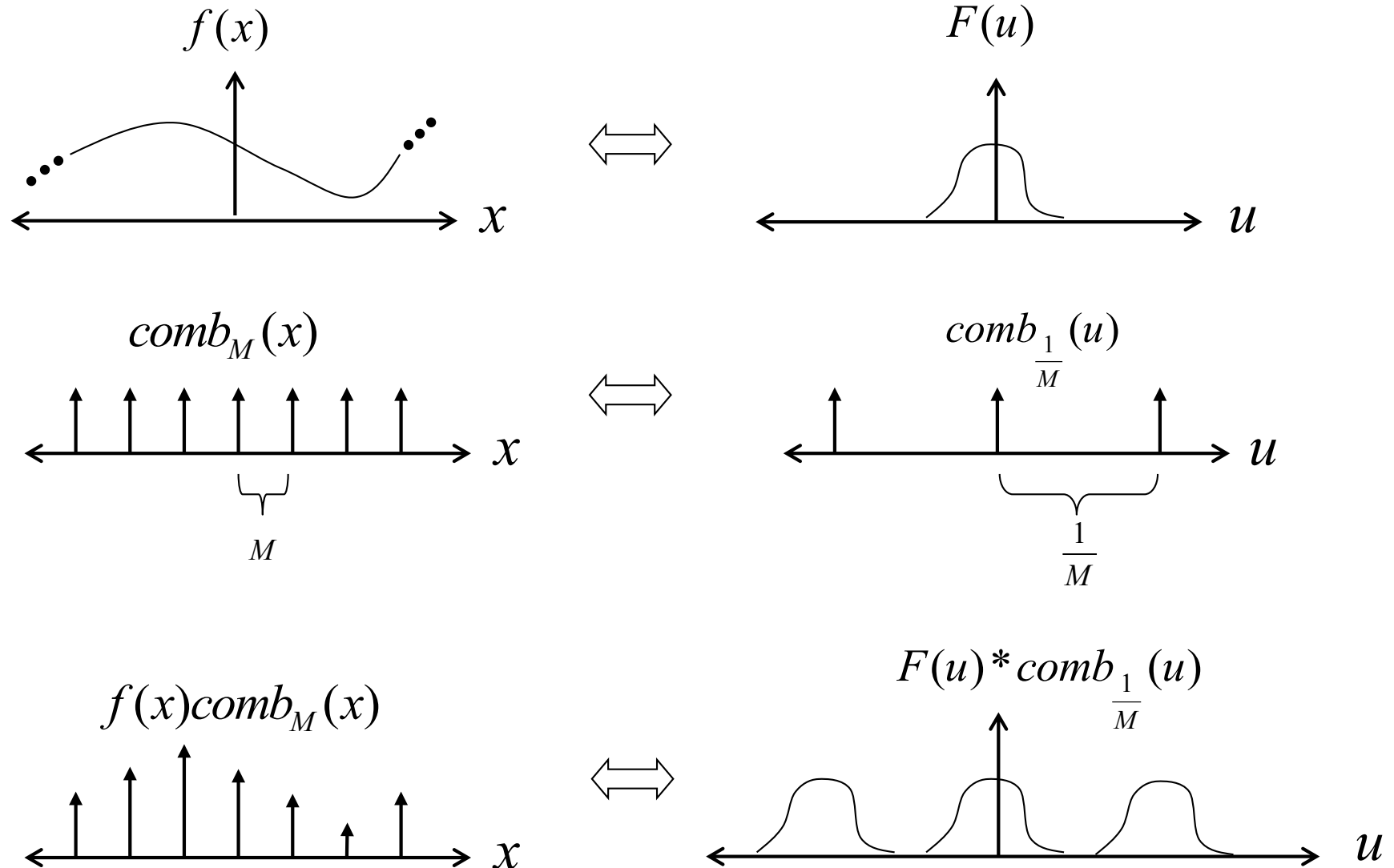


Consequences

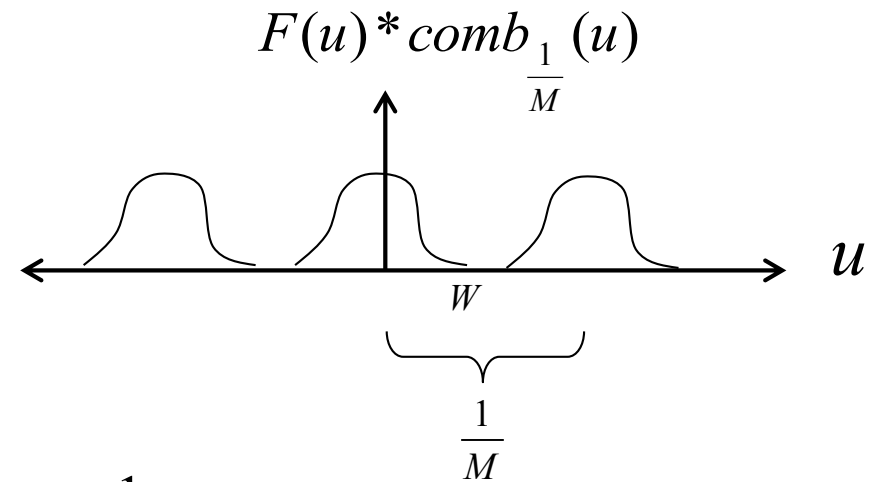
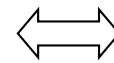
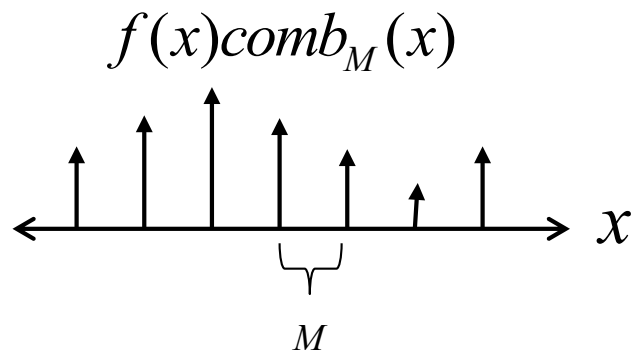
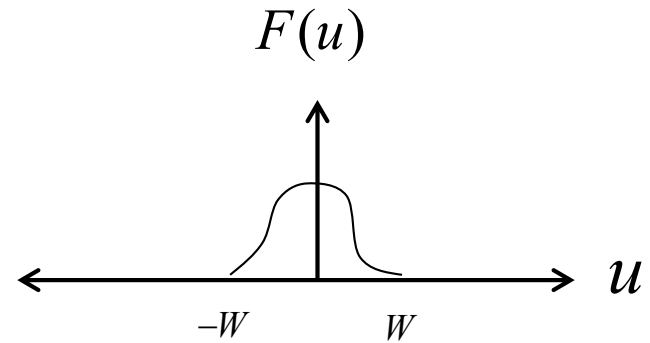
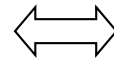
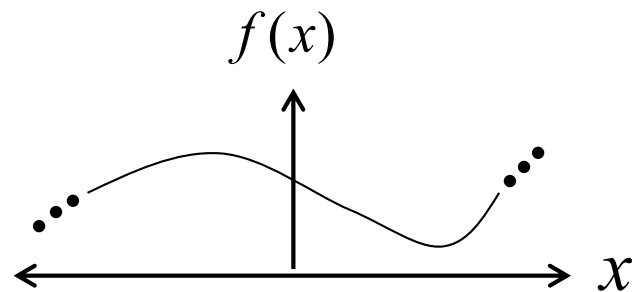
Sampling (Nyquist) theorem



Sampling



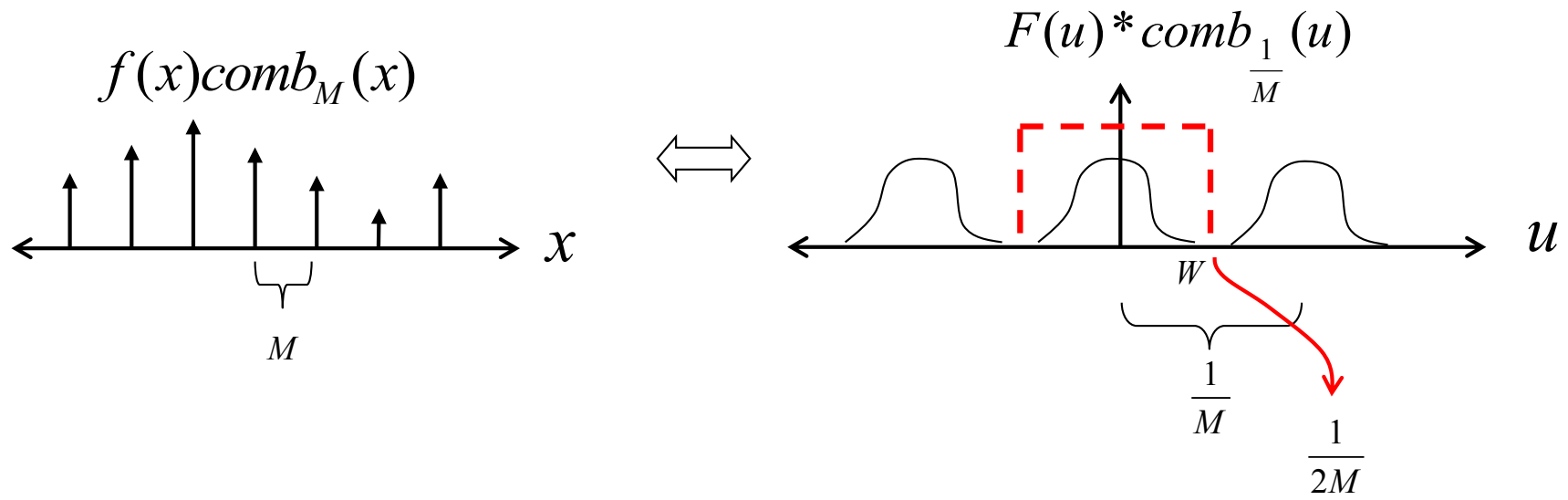
Sampling



Nyquist theorem: No aliasing if

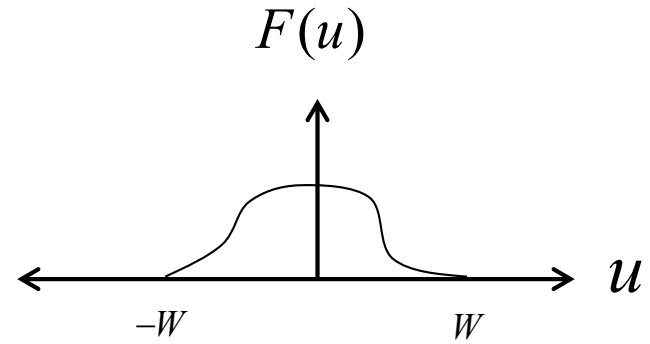
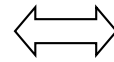
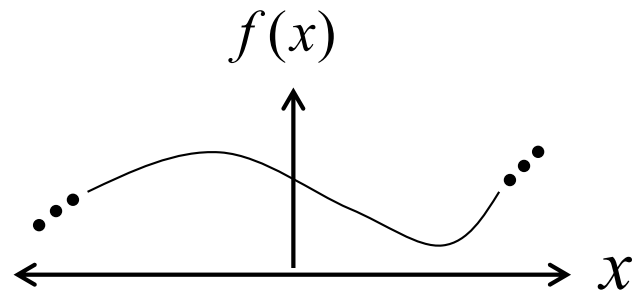
$$\frac{1}{M} > 2W$$

Sampling

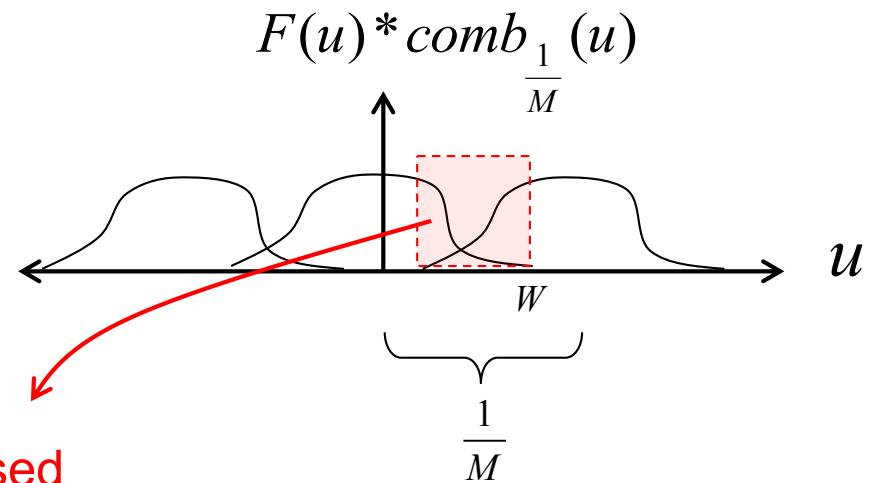
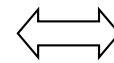


If there is no aliasing, the original signal can be recovered from its samples by low-pass filtering.

Sampling

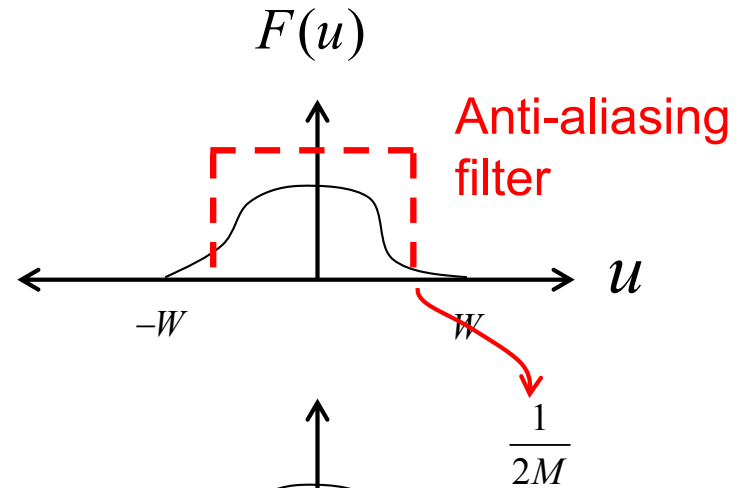
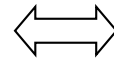
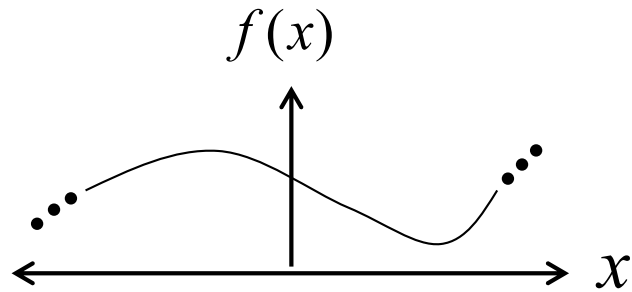


$$f(x) \text{comb}_M(x)$$

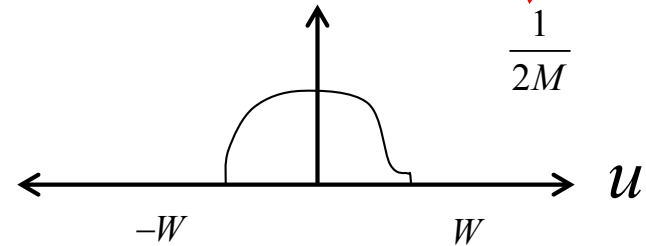
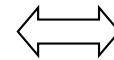


Aliased

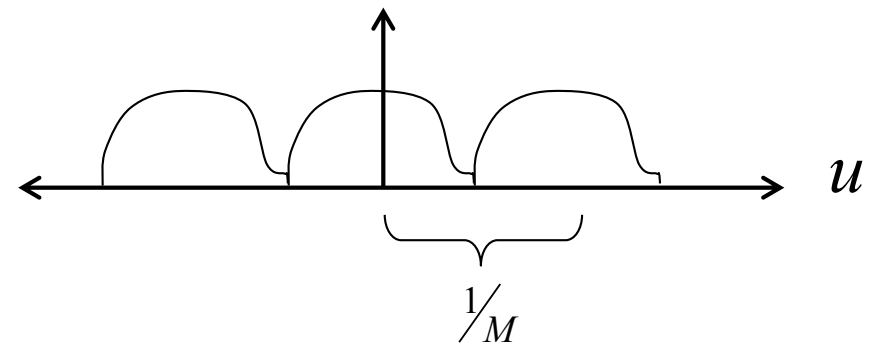
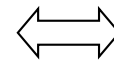
Sampling



$$f(x) * h(x)$$



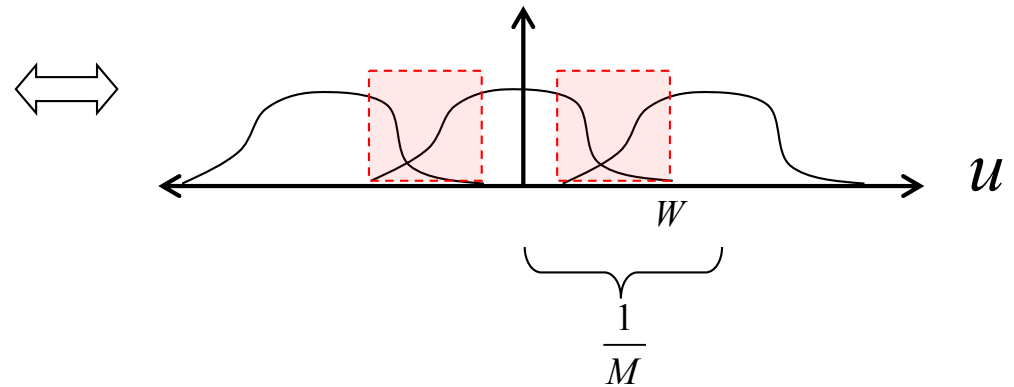
$$[f(x) * h(x)] \text{comb}_M(x)$$



Sampling

- Without anti-aliasing filter:

$$f(x)comb_M(x)$$



- With anti-aliasing filter:

$$[f(x) * h(x)]comb_M(x)$$

