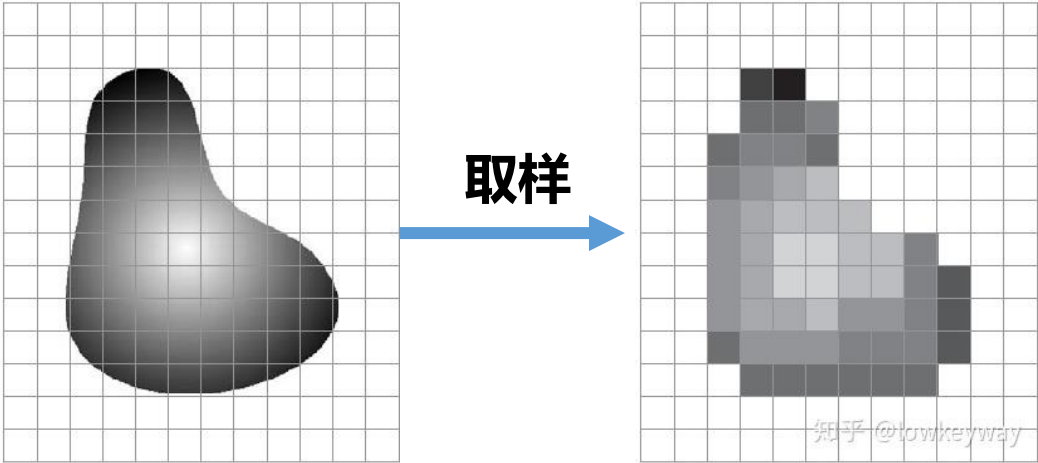
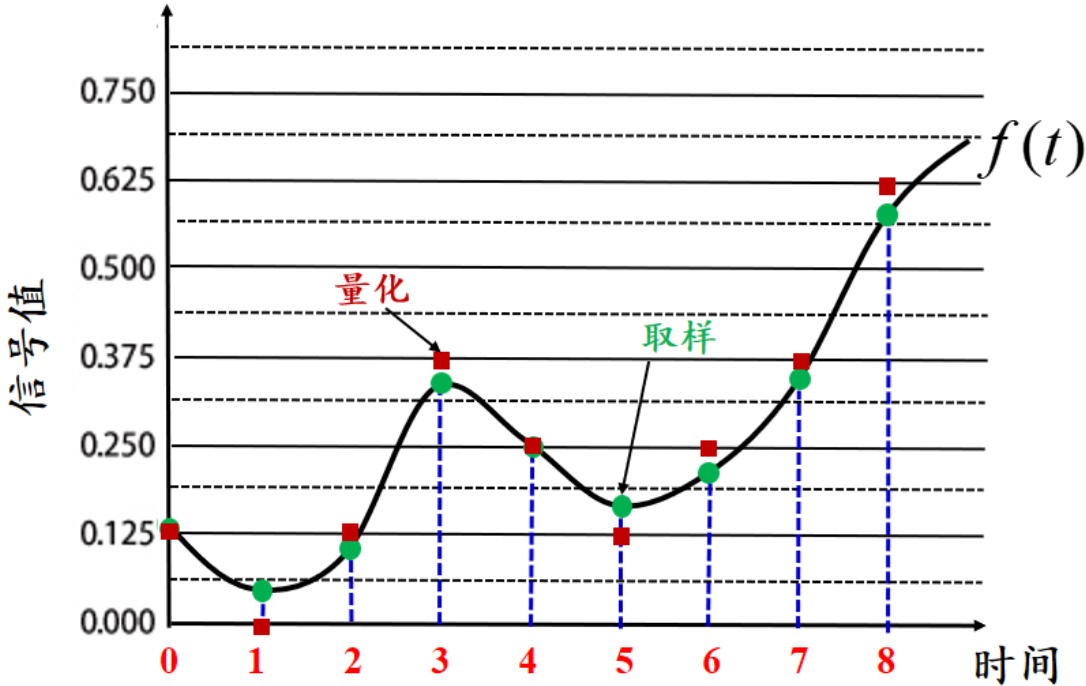


频率域滤波

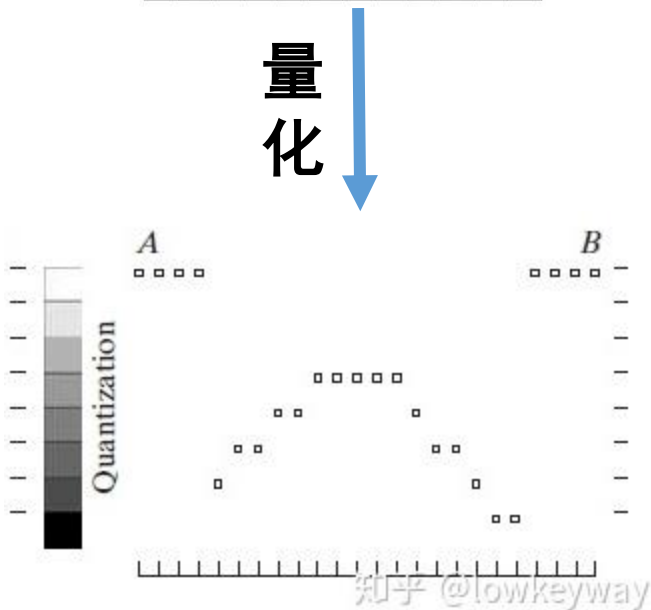
- 4.1 背景
- 4.2 基本概念
- 4.3 取样和取样函数的傅立叶变换
- 4.4 单变量的离散傅立叶变换
- 4.5 两个变量的扩展
- 4.6 二维离散傅立叶变换的一些性质
- 4.7 频率域滤波基础
- 4.8 频率域滤波器平滑图像
- 4.9 频率域滤波器锐化图像
- 4.10 选择性滤波

§4.3 取样和取样函数的傅立叶变换

模拟信号与数字信号



模拟图像与数字图像



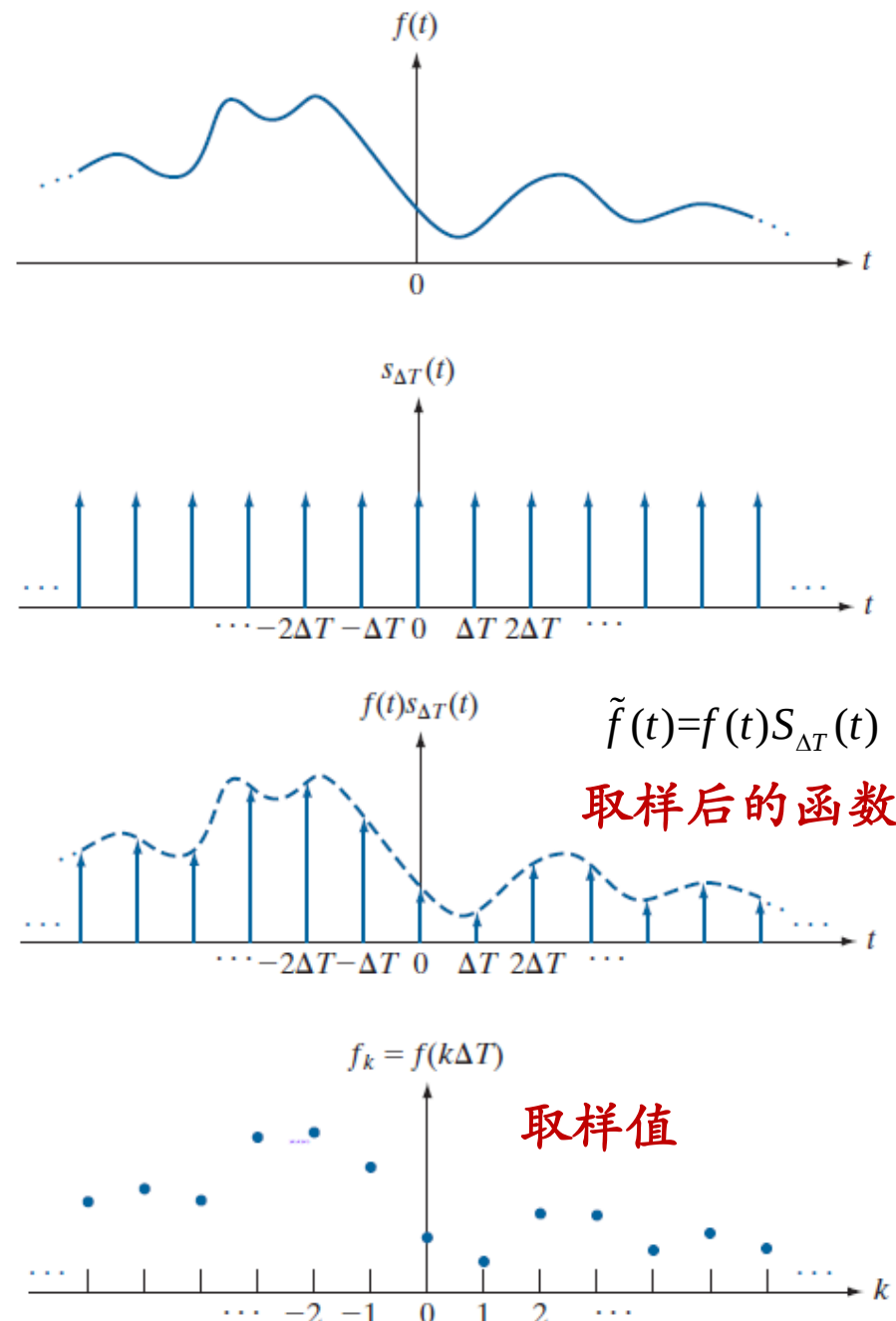
§4.3.1 取样

取样函数用以均匀间隔(ΔT)的冲击串代替, 因此模拟取样的结果是用取样函数同连续函数相乘

$$\begin{aligned}\tilde{f}(t) &= f(t)S_{\Delta T}(t) \\ &= \sum_{n=-\infty}^{\infty} f(t)\delta(t - n\Delta T)\end{aligned}$$

序列中的任意取样值 f_k 为

$$f_k = \int_{-\infty}^{\infty} f(t)\delta(t - k\Delta T)dt = f(k\Delta T)$$



§4.3.2 取样函数的傅立叶变换

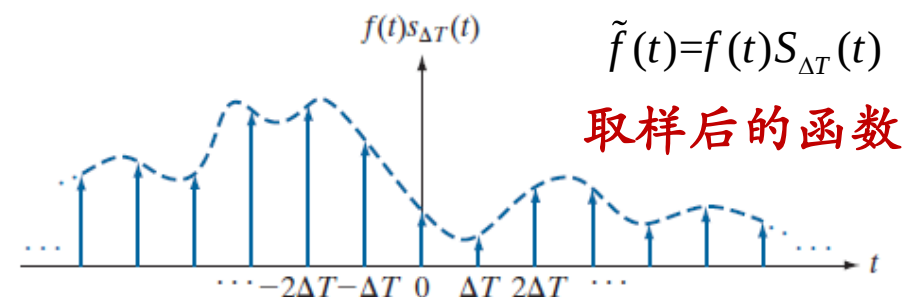
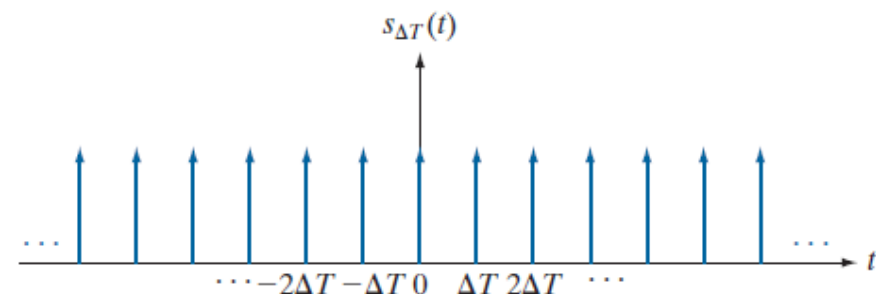
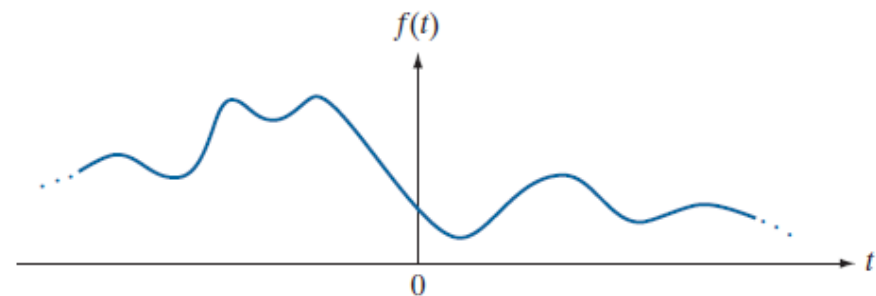
$$\tilde{f}(t) = f(t)s_{\Delta T}(t)$$

$$\tilde{F}(\mu) = \mathfrak{F}\{\tilde{f}(t)\} = \mathfrak{F}\{f(t)s_{\Delta T}(t)\} = F(\mu) \star S(\mu)$$

$$\tilde{F}(\mu) = F(\mu) \star S(\mu)$$

$$= \int_{-\infty}^{\infty} F(\tau)S(\mu - \tau)d\tau$$

$$= \frac{1}{\Delta T} \sum_{n=-\infty}^{\infty} F(\mu - \frac{n}{\Delta T})$$

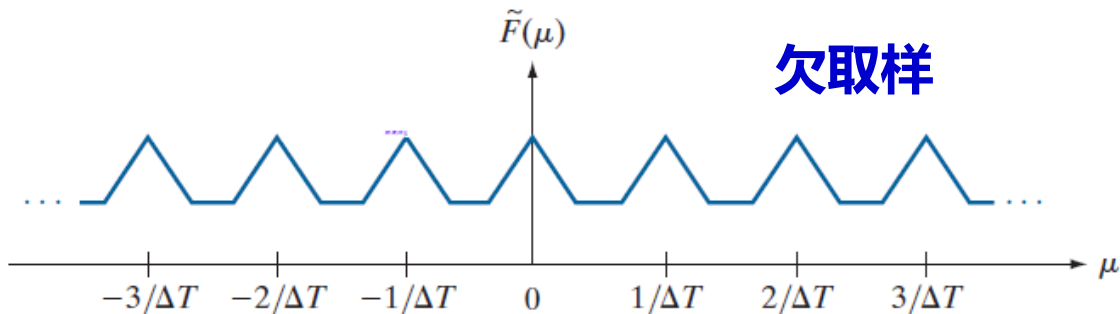
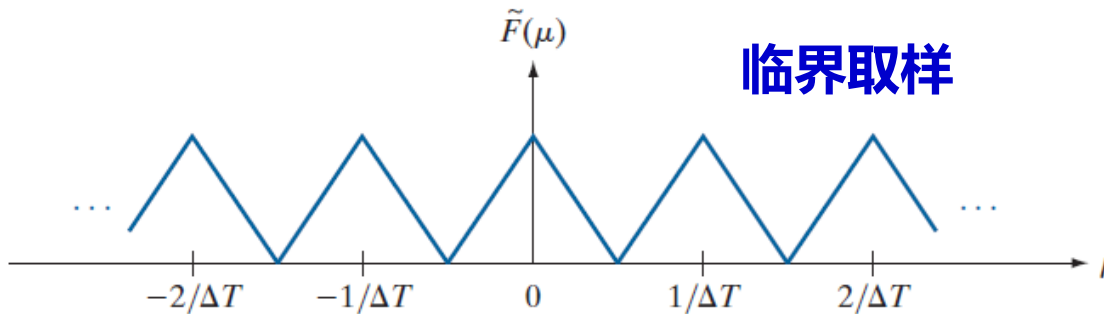
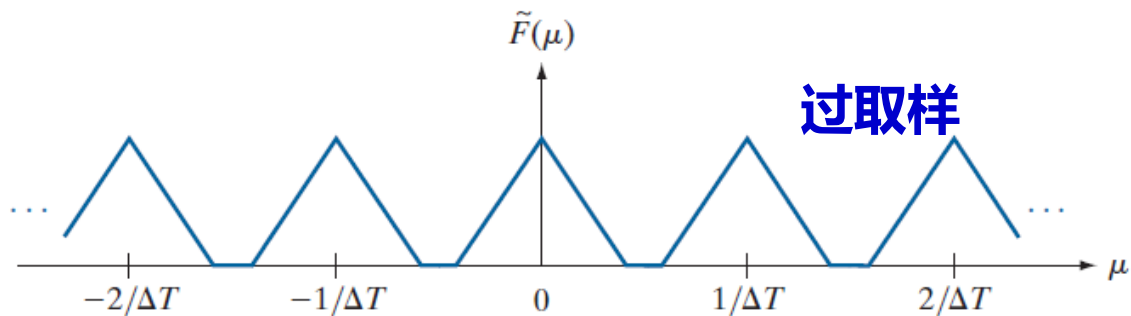
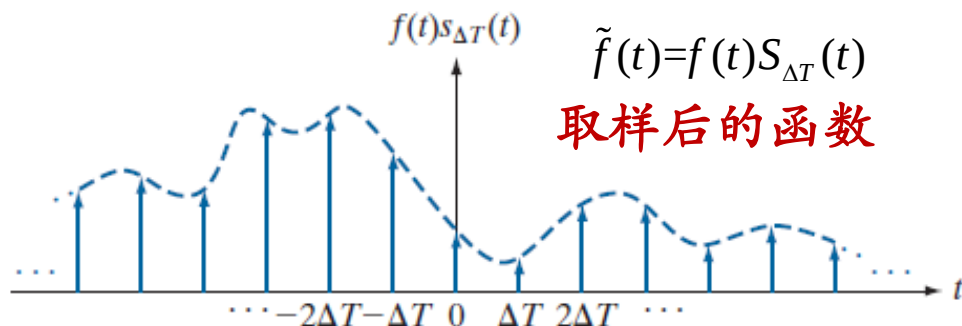
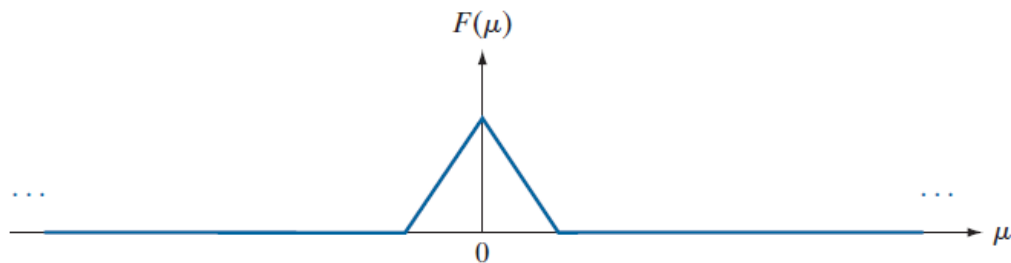
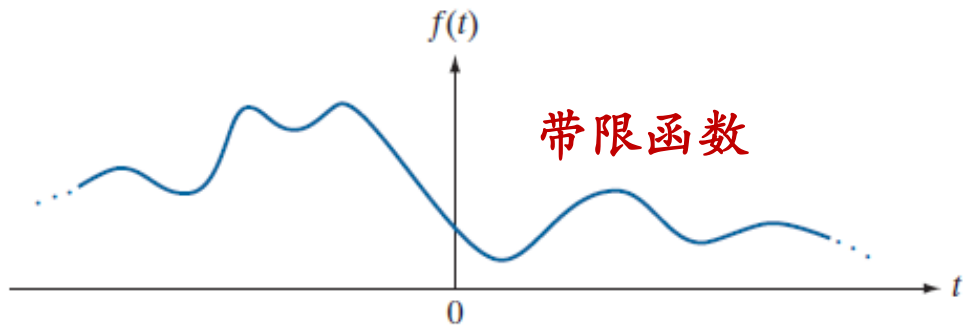


冲激串的傅里叶变换

$$s_{\Delta T}(t) = \sum_{n=-\infty}^{\infty} \delta(t - n\Delta T) \quad \longrightarrow \quad S(\mu) = \frac{1}{\Delta T} \sum_{n=-\infty}^{\infty} \delta(\mu - \frac{n}{\Delta T})$$

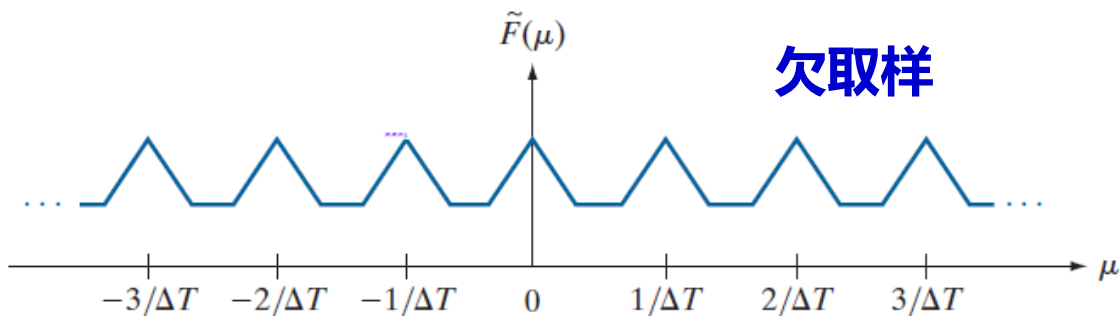
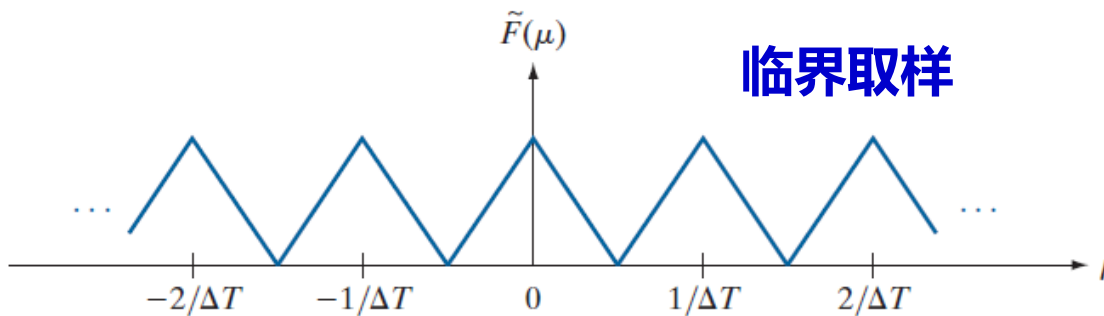
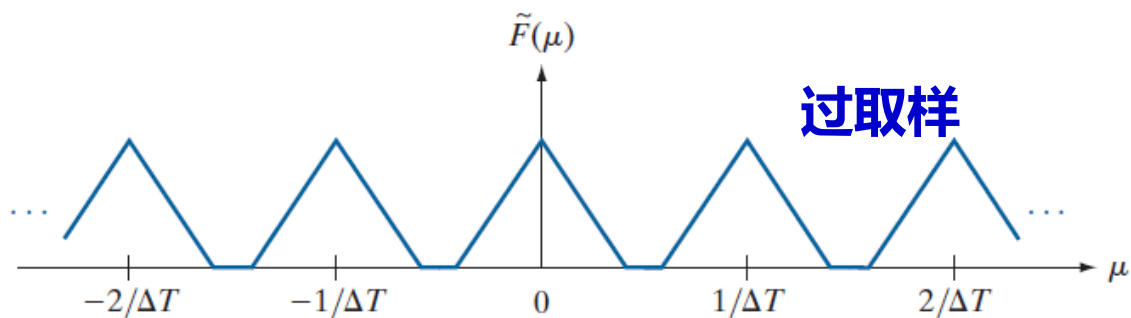
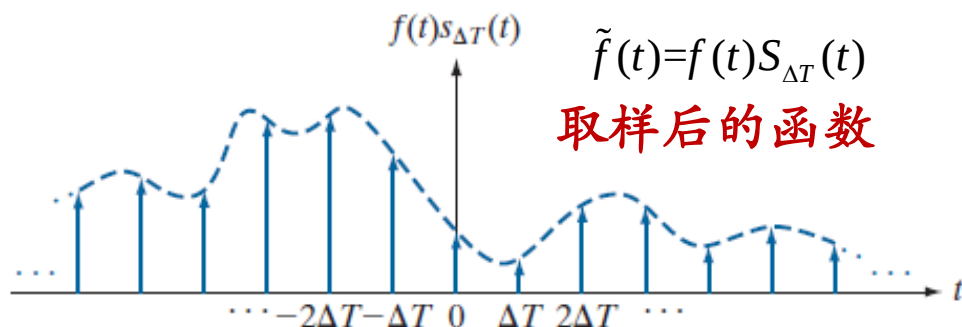
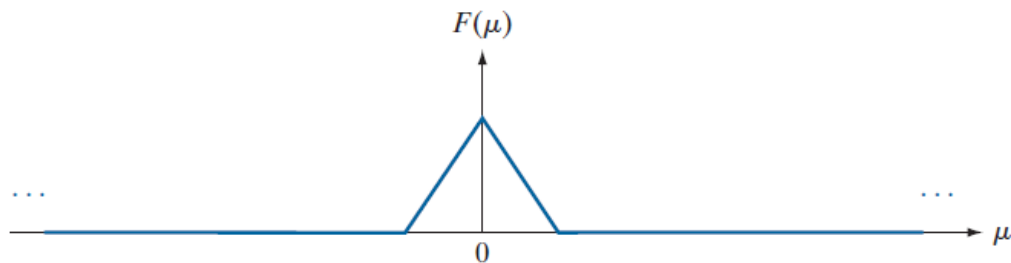
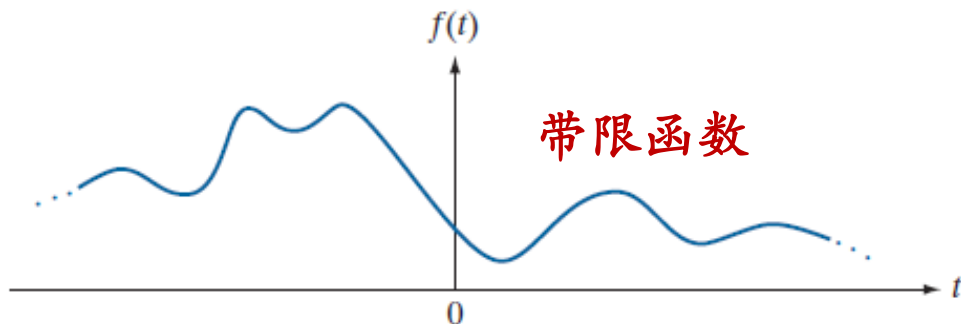
§4.3.2 取样函数的傅立叶变换

$$\tilde{F}(\mu) = \frac{1}{\Delta T} \sum_{n=-\infty}^{\infty} F(\mu - \frac{n}{\Delta T})$$



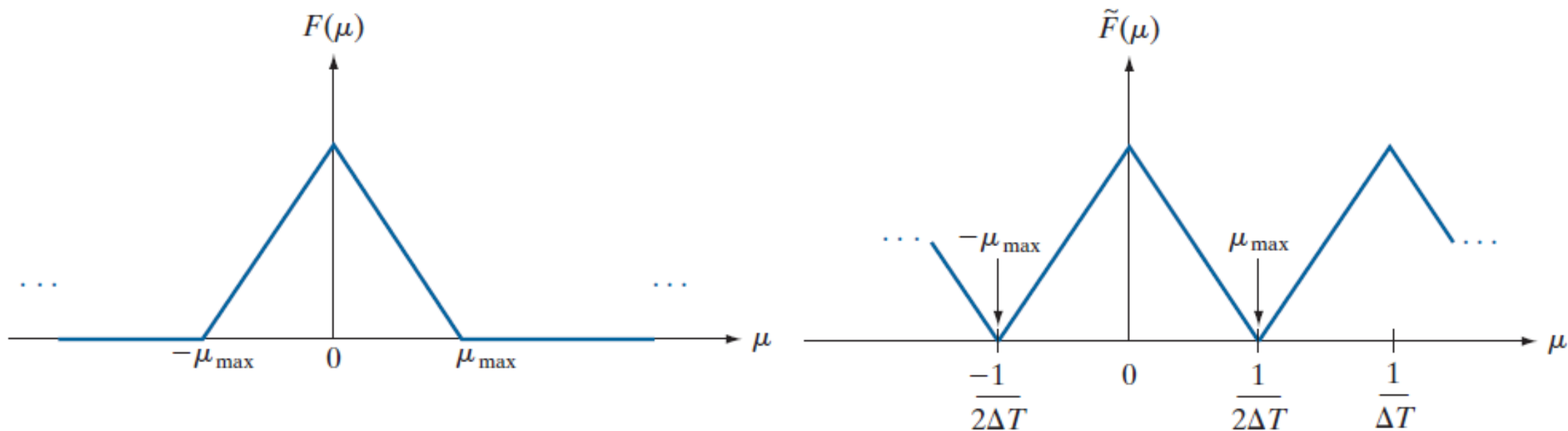
§4.3.2 取样函数的傅立叶变换

$1/\Delta T$ 是取样函数的取样率。取样率必须足够高，才能使得不同周期之间有有效的间隔，



§4.3.3 取样定理

如何给出适当条件，使得一个连续函数完全可由其取样集合来恢复？



取样定理
奈奎斯特准则

$$\frac{1}{\Delta T} > 2\mu_{\max}$$

取样率非常关键

§4.3.3 取样定理

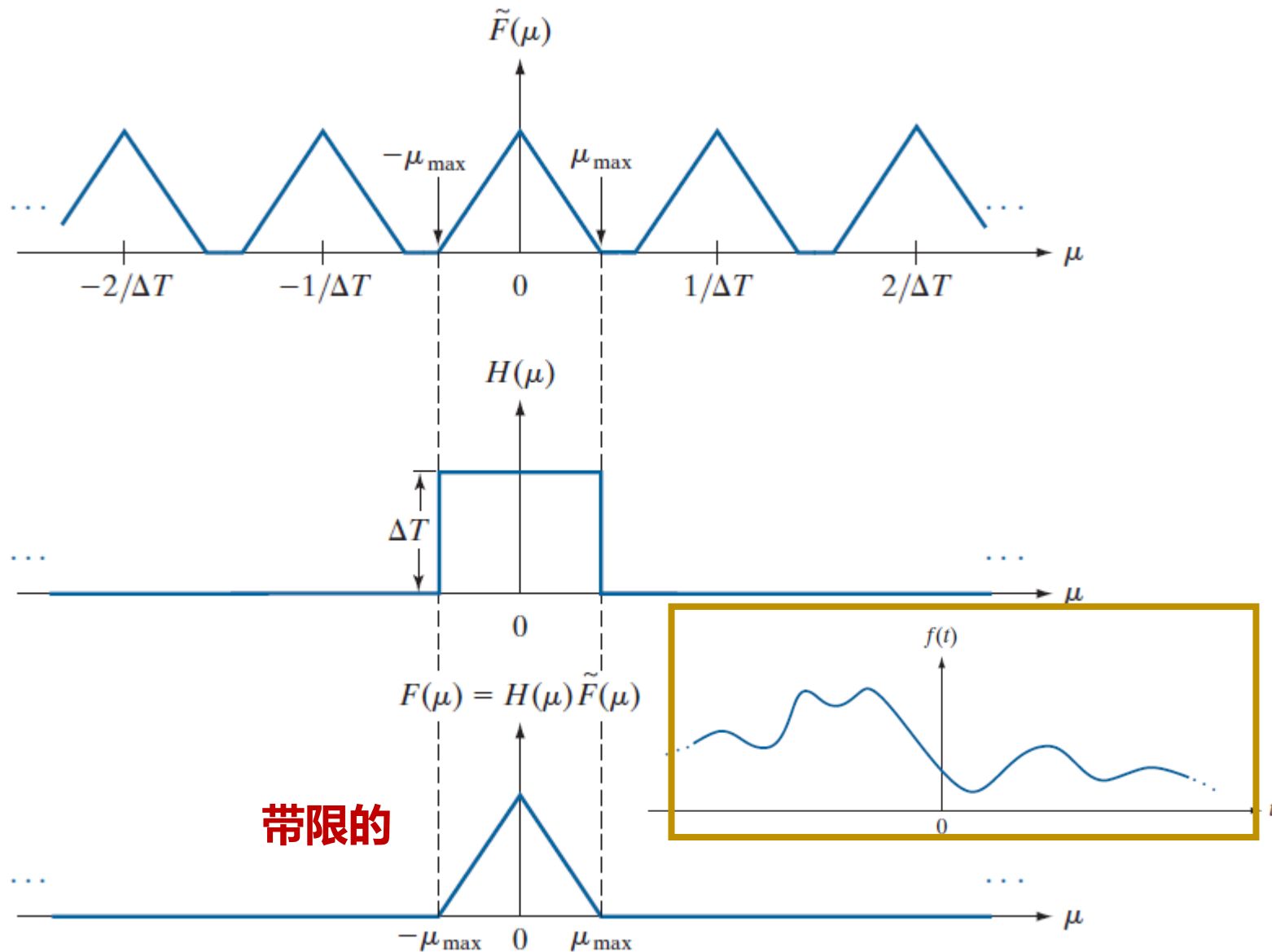
• 恢复/复原过程

$$F(\mu) = H(\mu)\tilde{F}(\mu)$$

$$f(t) = \int_{-\infty}^{\infty} F(\mu) e^{j2\pi\mu t} d\mu$$

- $f(t)$ 必须是带限的，即必须从 $-\infty$ 扩展到 ∞

- 限制一个函数的持续时间会妨碍该函数的完美复原



§4.3.4 混淆

频率混淆是由函数欠采样导致

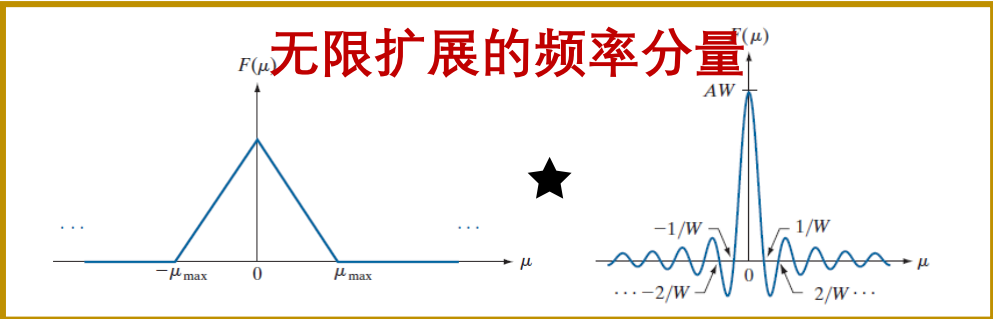
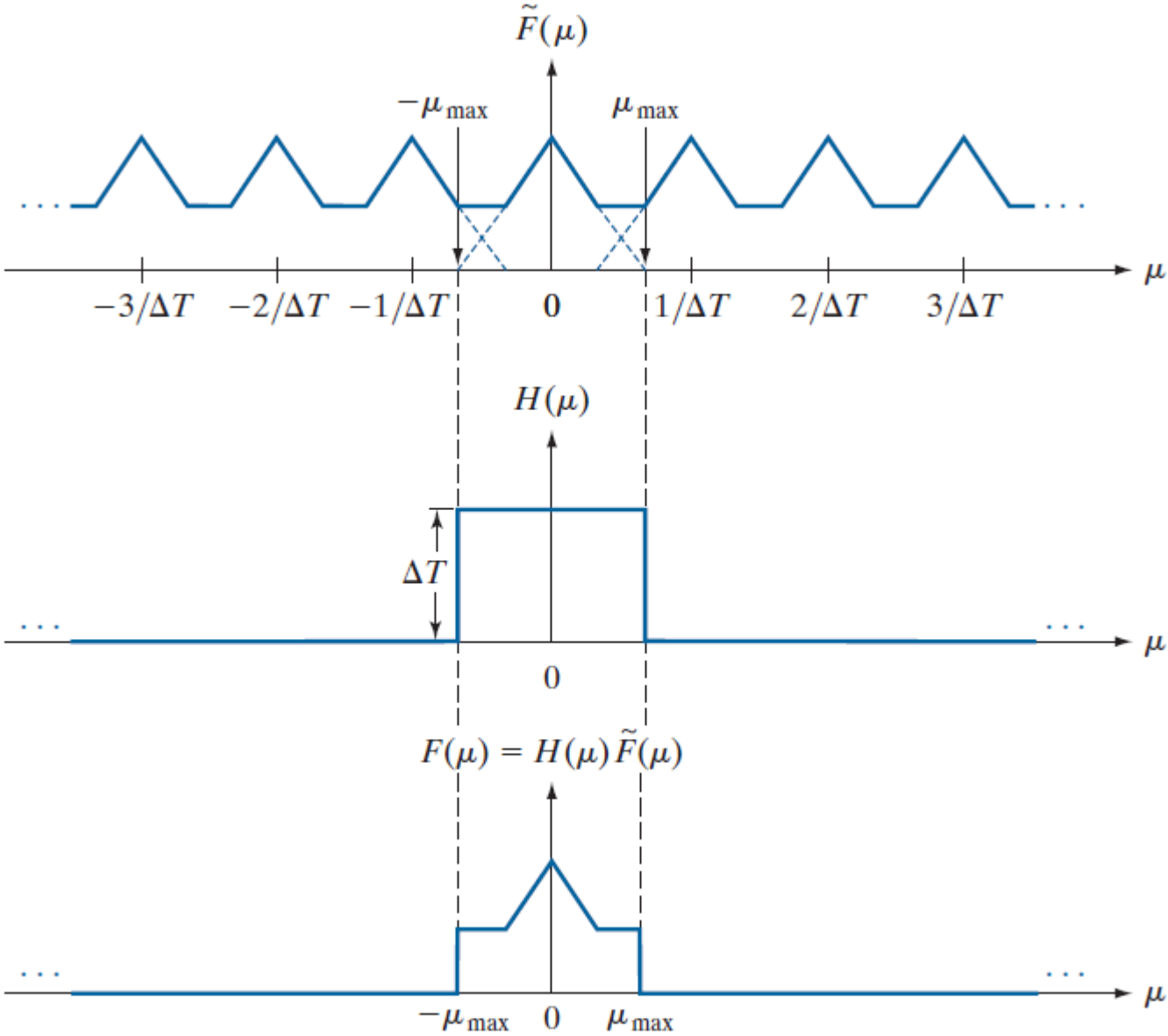
混淆在所难免

实际中限制函数的时间，通过将持续时间限制在区间 $[0,T]$ 内, $f(t)h(t)$

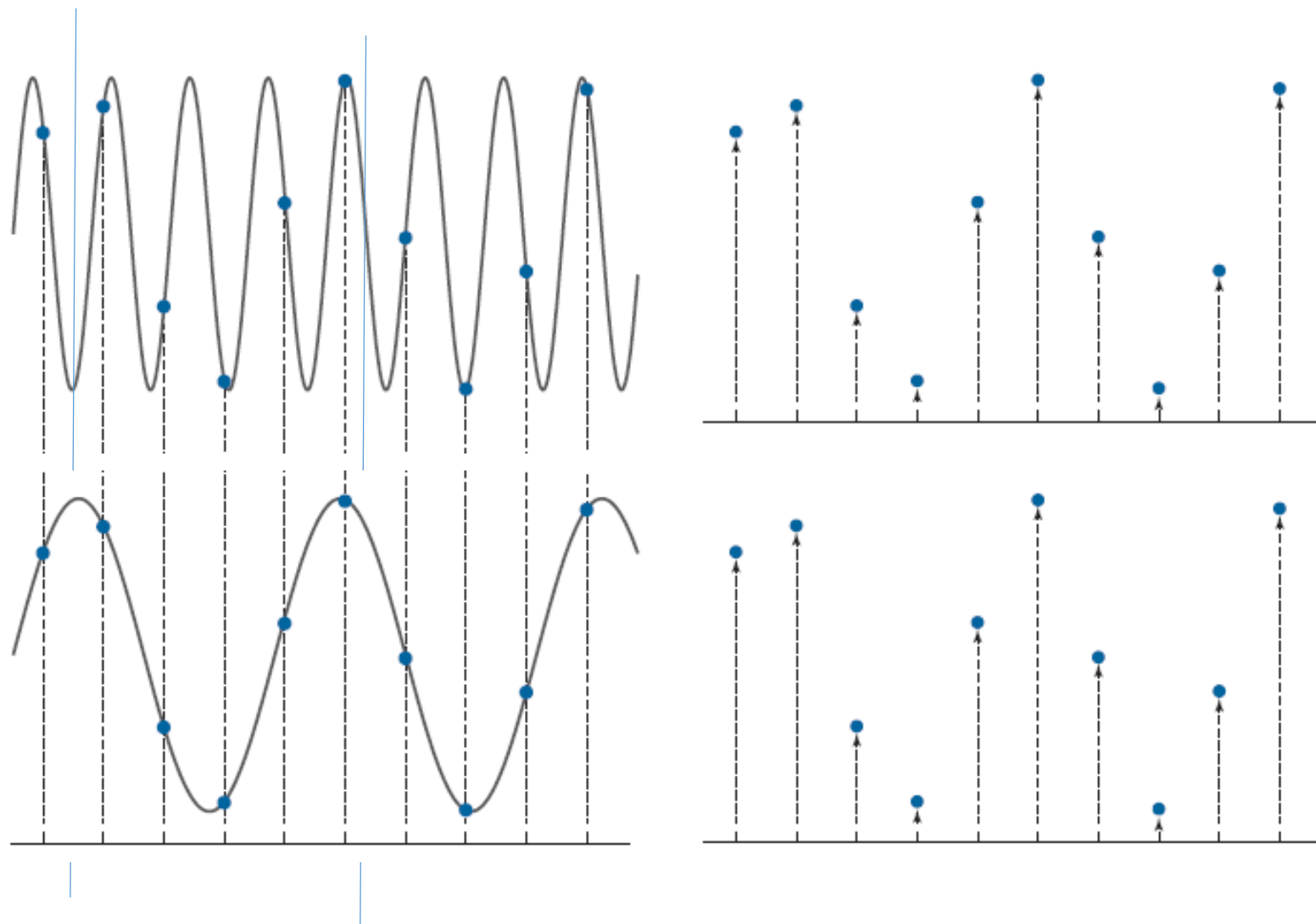
$$h(t) = \begin{cases} 1, & 0 \leq t \leq T \\ 0, & \text{others} \end{cases}$$

$f(t)h(t)$ 的变换等于两个函数变换的卷积

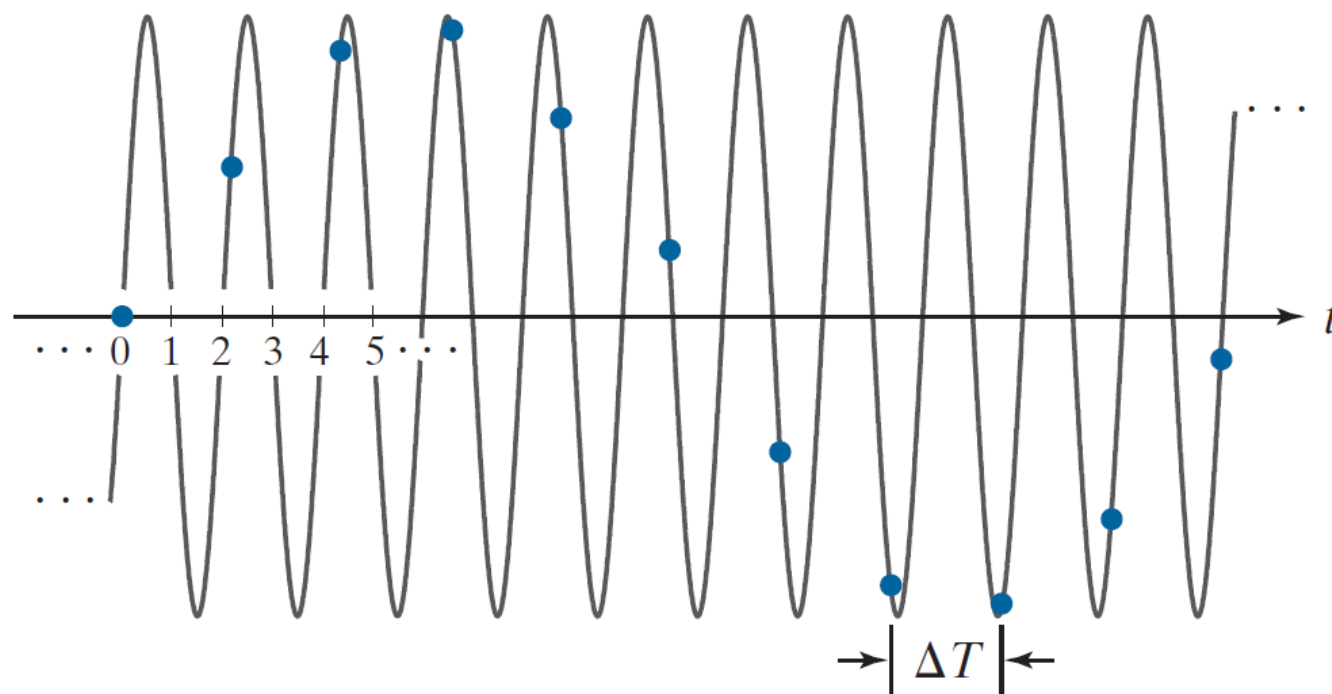
$f(t)h(t) \rightarrow F(\mu) \star H(\mu)$



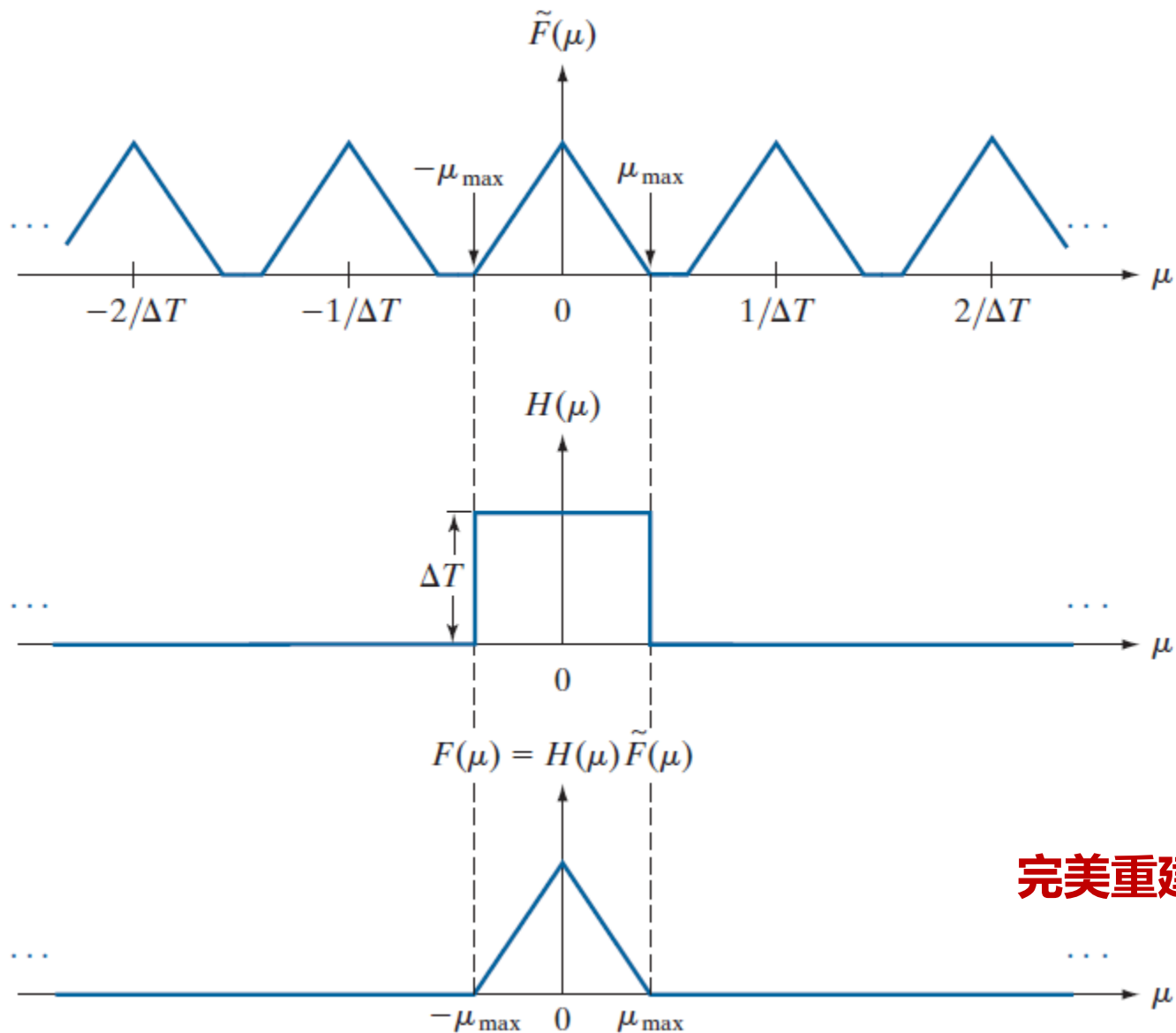
§4.3.4 混淆



§4.3.4 混淆



§4.3.5 由取样后的数据重建（复原）函数



$$F(\mu) = H(\mu)\tilde{F}(\mu)$$

$$f(t) = \mathfrak{F}^{-1}\{F(\mu)\}$$

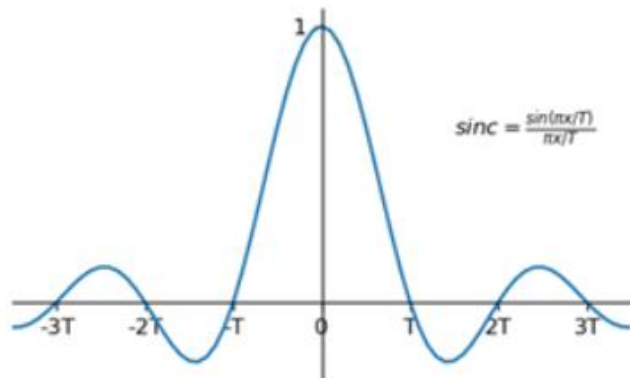
$$= \mathfrak{F}^{-1}\{H(\mu)\tilde{F}(\mu)\}$$

$$= h(t) \star \tilde{f}(t)$$

$$= \sum_{n=-\infty}^{\infty} f(n\Delta T) \operatorname{sinc}\left(\frac{t - n\Delta T}{\Delta T}\right)$$

完美重建函数是用取样值加权的sinc函数的无限和

§4.3.5 由取样后的数据重建（复原）函数



$$f(t) = \sum_{n=-\infty}^{\infty} f(n\Delta T) \text{sinc}\left(\frac{t - n\Delta T}{\Delta T}\right)$$

