

## Lecture 3: Digital Signal Processing

### Probability Density $p(x)$ :

probability that at any moment of time the value of the process  $x(t)$  falls within an interval  $[x - \frac{1}{2}dx, x + \frac{1}{2}dx]$ .

### Expected Value $E\{x\}$ :

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

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

Example: mean  $\mu = E\{x\}$

$$\text{dispersion } \sigma^2 = E\{x^2\} - E^2\{x\}$$

## Time Average $\langle f(x) \rangle$ :

$$\langle f(x) \rangle = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{+T/2} f(x(t)) dt$$

$$\langle f(x) \rangle_T = \frac{1}{T} \int_{-T/2}^{+T/2} f(x(t)) dt$$

## Fourier Transform:

$$\text{if } f(t) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} F(\nu) e^{i2\pi\nu t} d\nu,$$

$$\text{then } F(\nu) = \int_{-\infty}^{+\infty} f(t) e^{-i2\pi\nu t} dt$$

$$X_T(\nu) = \int_{-T/2}^{+T/2} x(t) e^{-i2\pi\nu t} dt$$

## Example: F.T. of a Finite Pulse

$$\begin{aligned} f(t) &= 1 \text{ for } |t| < T_0 \\ &= 0 \text{ for } |t| \geq T_0 \end{aligned}$$

$$\begin{aligned} F(\omega) &= \int_{-T_0}^{+T_0} e^{i\omega t} dt = \frac{1}{-i\omega} e^{i\omega t} \Big|_{-T_0}^{+T_0} \\ &= 2T_0 \frac{\sin \omega T_0}{\omega T_0} \end{aligned}$$

## Auto-correlation Function (ACF):

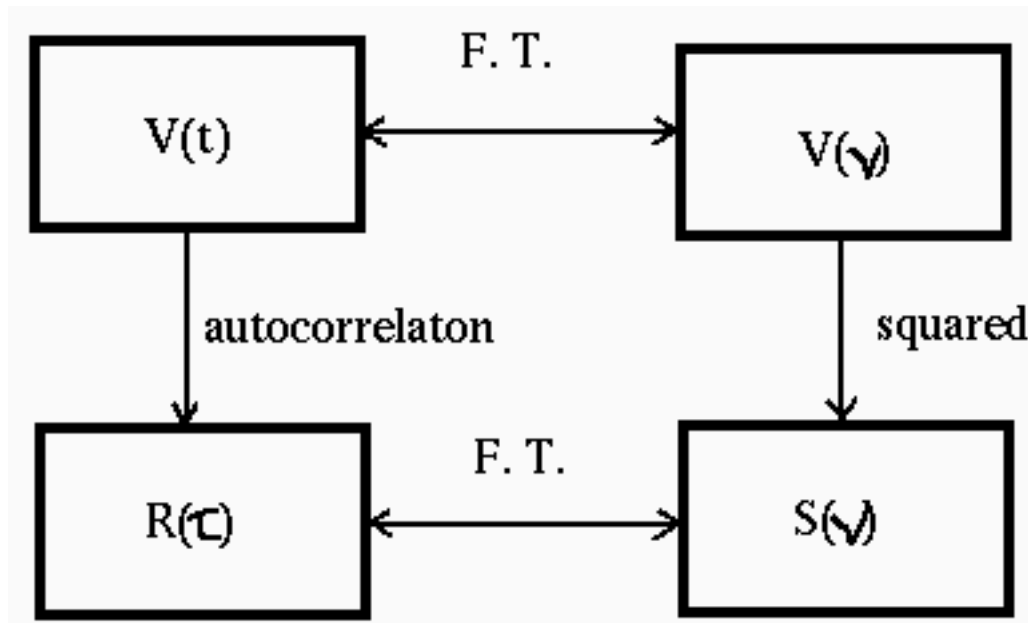
$$\begin{aligned} R_T(\tau) &= E_T\{x(s)x(s + \tau)\} \\ &= \int_{-T/2}^{+T/2} x(s)x(s + \tau)ds \end{aligned}$$

### Example: ACF of a Finite Pulse

$$\begin{aligned} f(t) &= 1 \text{ for } |t| < T_0 \\ &= 0 \text{ for } |t| \geq T_0 \end{aligned}$$

$$R(\tau) = \int_{-T_0}^{+T_0} f(t)f(t + \tau)dt$$

## Wiener-Khinchin Theorem



$$V_T(\nu) = \int_{-T/2}^{+T/2} V(t) e^{-i2\pi\nu t} dt$$

$$S(\nu) = \lim_{T \rightarrow \infty} \frac{1}{T} E_T \{ V^2(\nu) \}$$

$$E_T \{ V^2(\nu) \} = E \left\{ \int_{-T/2}^{+T/2} V(s) V(t) e^{-i2\pi\nu(t-s)} ds dt \right\}$$

$$\text{since } R_T(\tau) = \int_{-T/2}^{+T/2} V(t) V(t - \tau) dt$$

$$S(\nu) = \int_{-\infty}^{+\infty} R(\tau) e^{-i2\pi\nu\tau} d\tau$$

$$\text{and } R(\tau) = \int_{-\infty}^{+\infty} S(\nu) e^{i2\pi\nu\tau} d\nu$$

## Digital Spectrometers

- $V(t)$  and  $V(\nu)$  are Fourier pairs (Fourier transform spectrometer, FTS)
- $R(\tau)$  and  $S(\nu)$  are Fourier pairs (W-K theorem – autocorrelation spectrometer)
- Nyquist's sampling:  $\Delta t \leq \frac{1}{2B}$
- digital encoding of  $V(t)$  for FFT means  $B$  restricted by sampling speed
- finite sampling length limitations

## 2-Level Autocorrelation Spectrometer

$$\begin{aligned}y(t) &= +1 \text{ for } x(t) \geq 0 \\ &= -1 \text{ for } x(t) < 0\end{aligned}$$

$$R_y(\tau) = E\{y(t + \tau)y(t)\}$$

For a Gaussian process, this can be derived as

$$R_y(\tau) = \frac{2}{\pi} \arcsin \frac{R_x(\tau)}{R_x(0)}$$

$\text{and, } R_x(\tau) = R_x(0) \sin[\pi/2 R_y(\tau)]$

## Finite Sampling Length:

→ filter or apply weighting

$$\begin{aligned} \text{e.g. } w(\tau) &= 1 \text{ for } |\tau| \leq \tau_m \\ &= 0 \text{ otherwise} \end{aligned}$$

$$S = S(\nu) \otimes W(\nu)$$

$W(\nu)$  determines spectral resolution,

$$\Delta\nu = \frac{0.605}{\tau_m}$$

## Higher Bit Sampling

number of bits sampling	Efficiency	
	at $1/2B$	at $1/4B$
1	0.64	0.74
2	0.81	0.89
3	0.88	0.94
$\infty$	1.00	1.00