

# Sampled Time Series

Mitch Withers, Res. Assoc. Prof., Univ. of Memphis

See Aster and Borchers, Time Series Analysis, chapter 3.

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where,  $\mu = \text{mean}$   
 $\sigma = \text{standard deviation } (\sigma^2 = \text{variance})$

This distribution is very common (hence the name normal distribution).

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Where,

$$a_n = \frac{1}{\pi} \int_{-\pi}^{\pi} f(x) \cos(nx) dx$$

$$b_n = \frac{1}{\pi} \int_{-\pi}^{\pi} f(x) \sin(nx) dx$$

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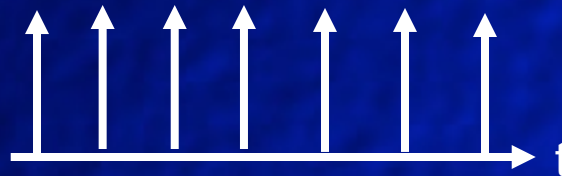
$$\Phi_k = \sum_{n=0}^{N-1} \phi_n e^{-i2\pi kn/N}$$

} Discrete, band limited, finite length series.

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The Shah function,  $\text{III}(t)$ , is a series of delta functions,



It is also sometimes called a dirac comb (after the dirac delta).

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The Fourier Transform of a gaussian is another gaussian!

Now consider the function,

$$f(t) = \frac{1}{\tau} e^{-\pi\tau^2 t^2} \sum_{n=-\infty}^{\infty} e^{-\pi(t-n)^2 / \tau^2}$$

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The sum is a series of gaussian “spikes” each of width  $\tau$ .

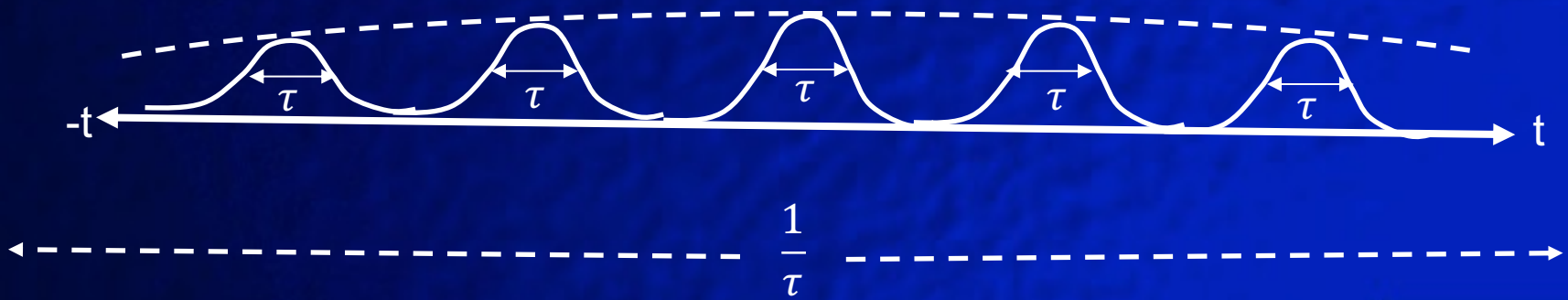


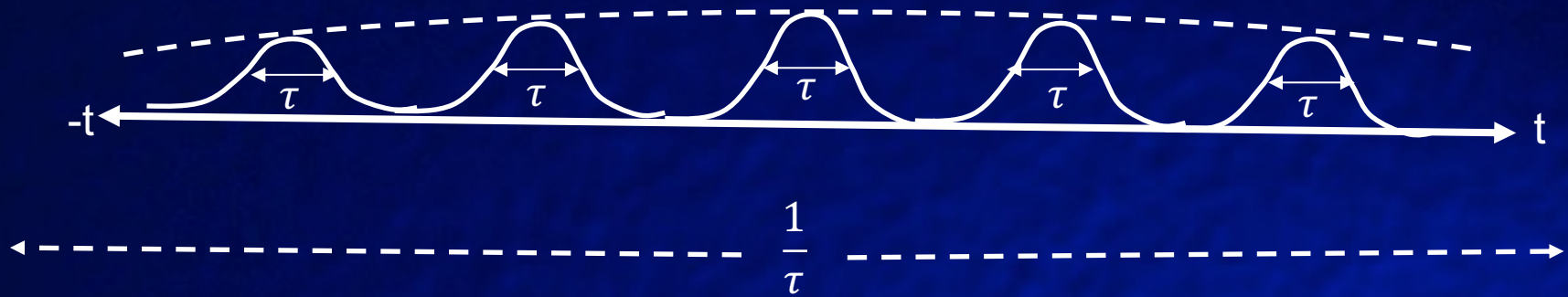
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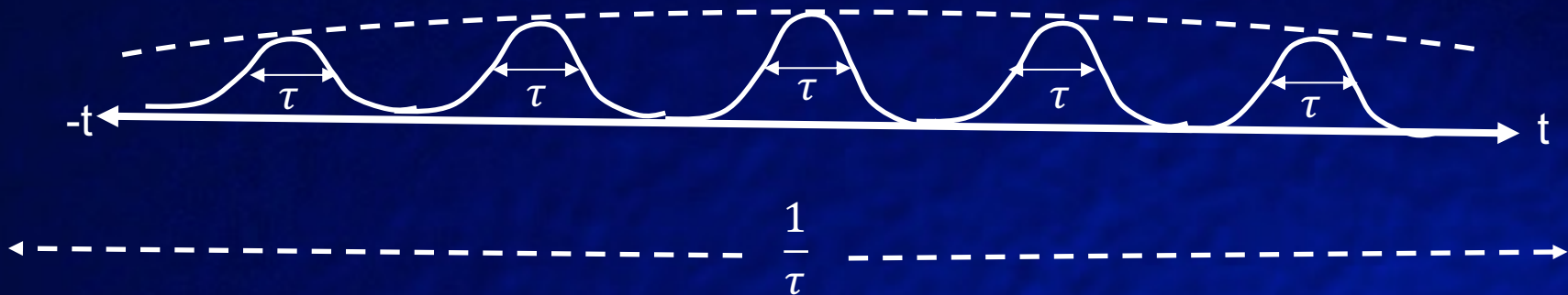
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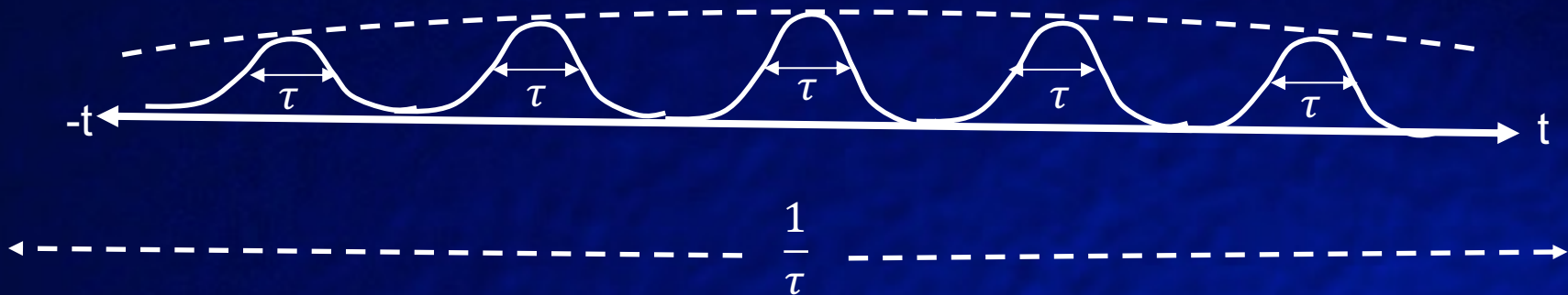


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$$\text{III}(t) = \lim_{\tau \rightarrow 0} f(t) = \sum_{n=-\infty}^{\infty} \delta(t - n)$$



We leave as an exercise for the student to show that the Fourier Series of our gaussian spikes is

$$\frac{1}{\tau} \sum_{n=-\infty}^{\infty} e^{-\pi(t-n)^2/\tau^2} = \sum_{n=-\infty}^{\infty} e^{-\pi\tau^2 n^2} e^{i2\pi nt}$$

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$$\mathbb{I}(t) = \lim_{\tau \rightarrow 0} e^{-\pi\tau^2 t^2} \sum_{n=-\infty}^{\infty} e^{-\pi\tau^2 n^2} e^{i2\pi n t}$$

$$F[\mathbb{I}(t)] = \lim_{\tau \rightarrow 0} \sum_{n=-\infty}^{\infty} e^{-\pi\tau^2 n^2} F[e^{-\pi\tau^2 t^2} e^{i2\pi n t}]$$

Recall the shifting property,

$$F[\phi(t - a)] = e^{i2\pi af} \Phi(f)$$



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Now let  $\psi(t) = \phi(t) \cdot r \cdot \Pi(rt)$  where  $r$  is the sampling rate.

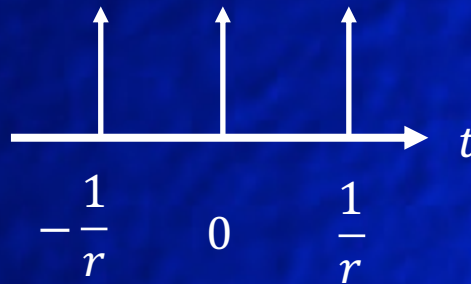


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If the rate is in samples per second, then samples are separated in time by  $\frac{1}{r}$  seconds.



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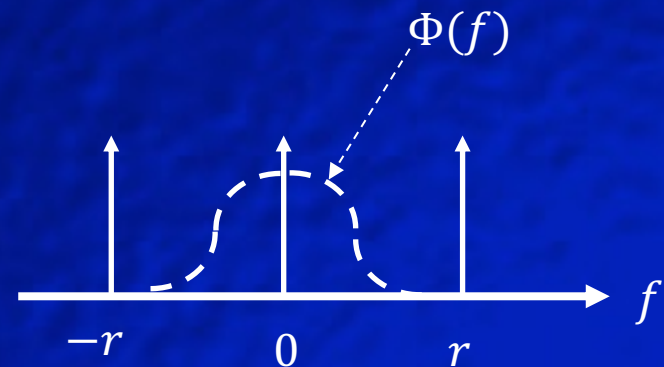
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$\Pi\left(\frac{f}{r}\right)$  replicates  $\Phi(f)$  at intervals of  $r$ .

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Digitizing in  $t$  makes  $\Phi(f)$  periodic (replicates).

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From Aster's notes:

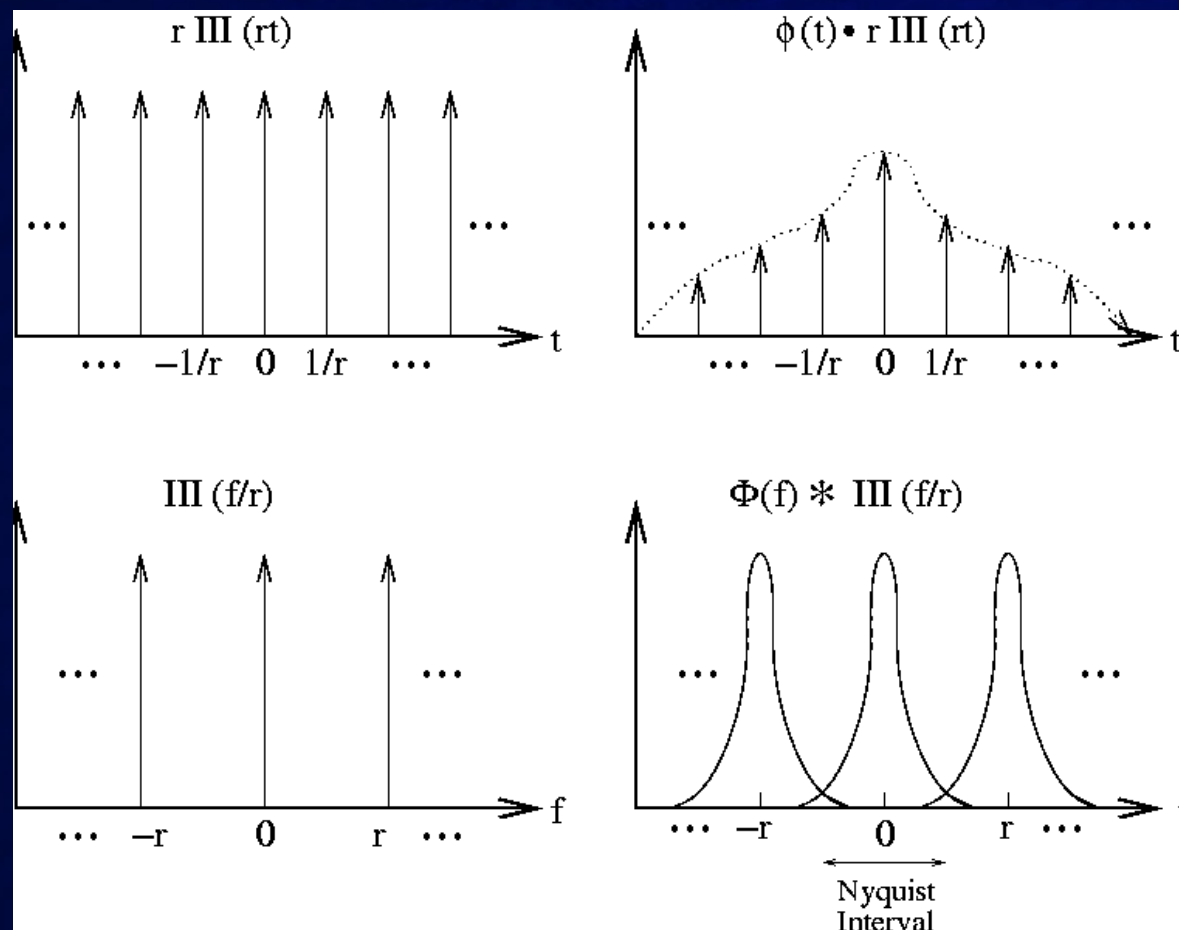
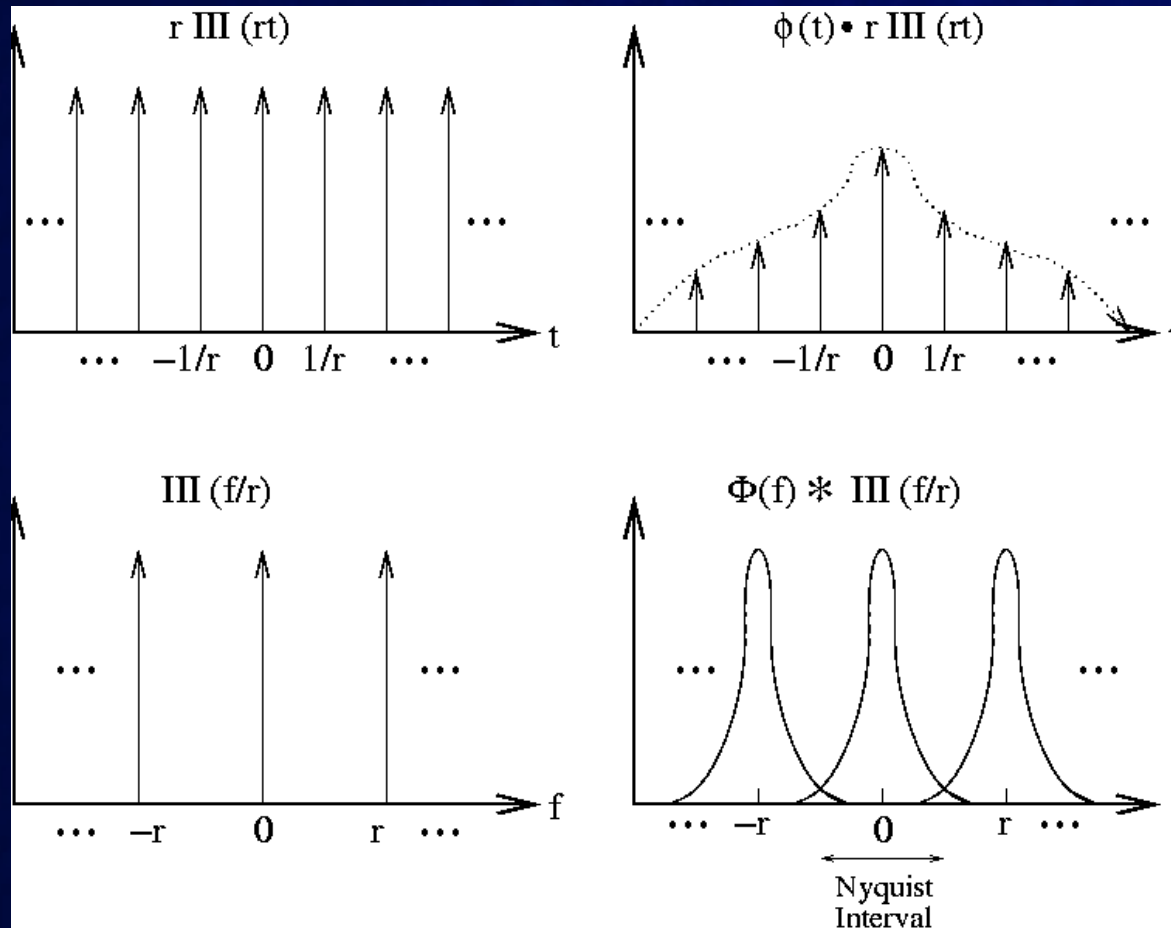


Figure 1: The Shah function and its Fourier Transform;  
Fourier Transform of a Sampled Function (slightly aliased)

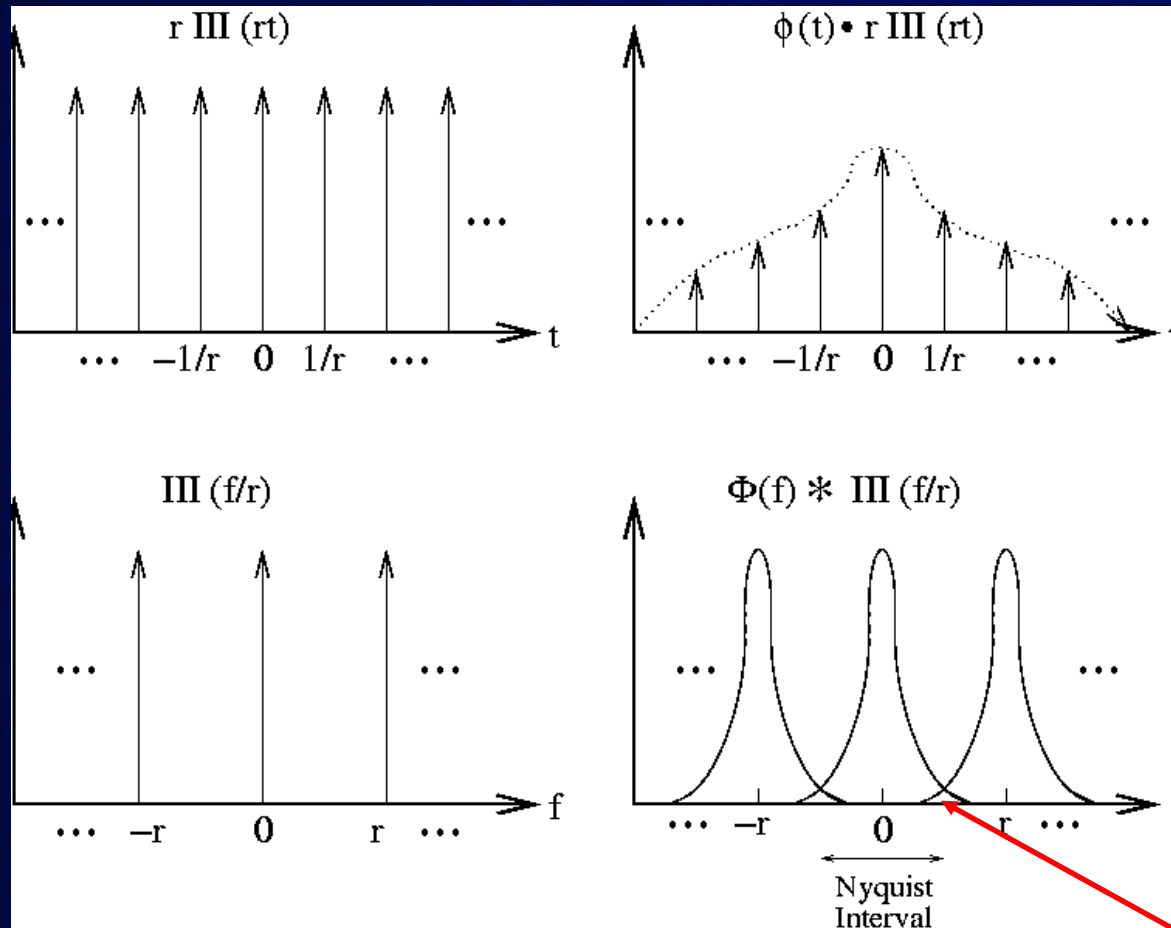
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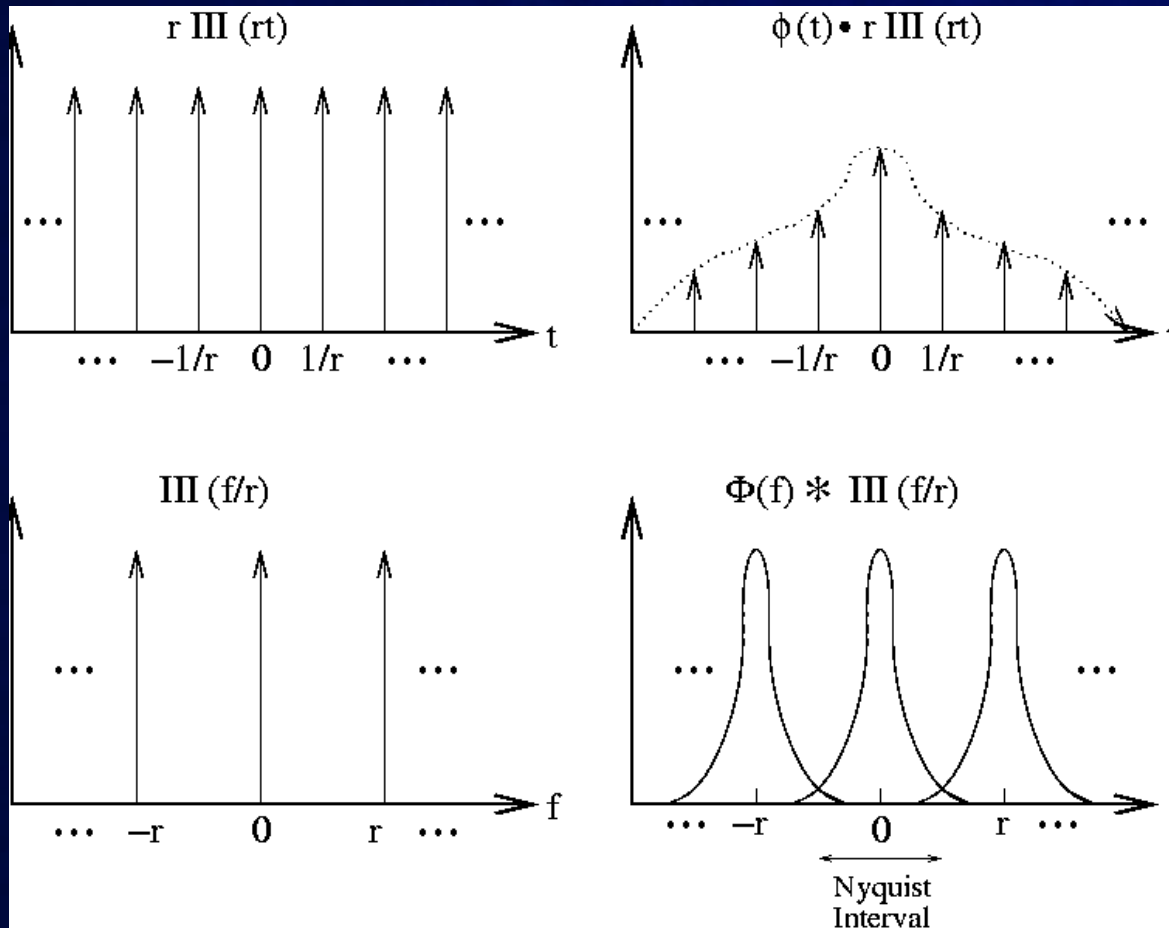
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Aster and Borchers Figure 3.1: The Shah function and its Fourier Transform; Fourier Transform of a Sampled Function (slightly aliased)

Aliased overlap.

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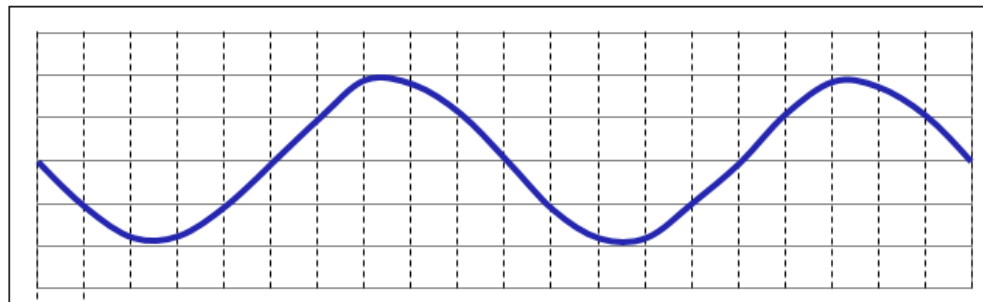
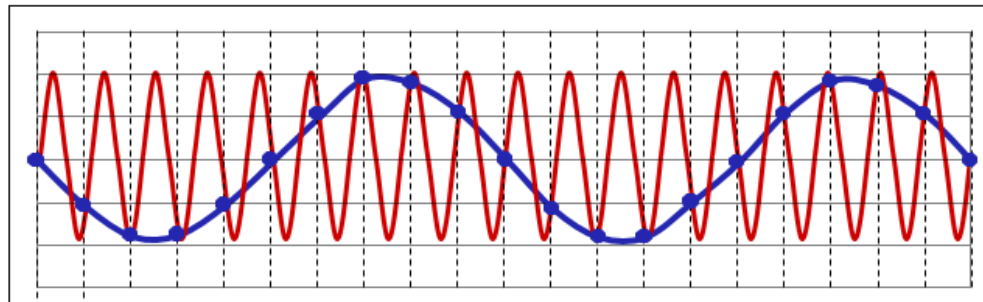
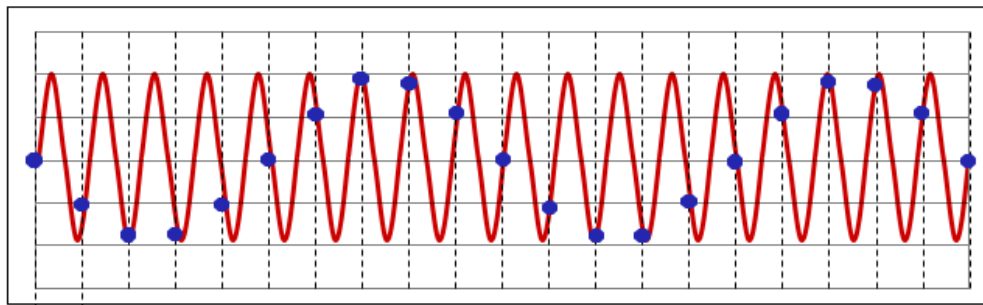
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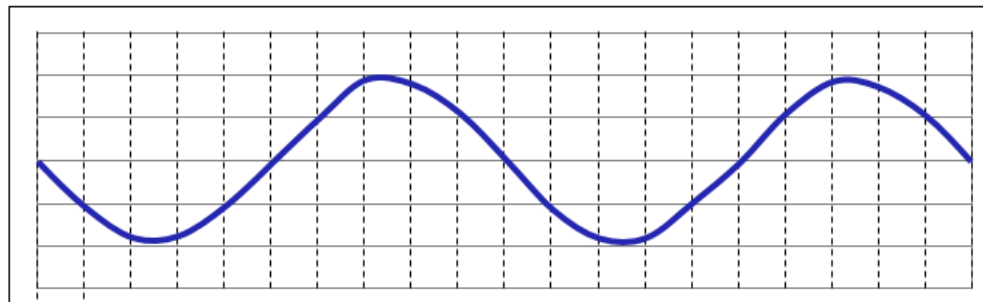
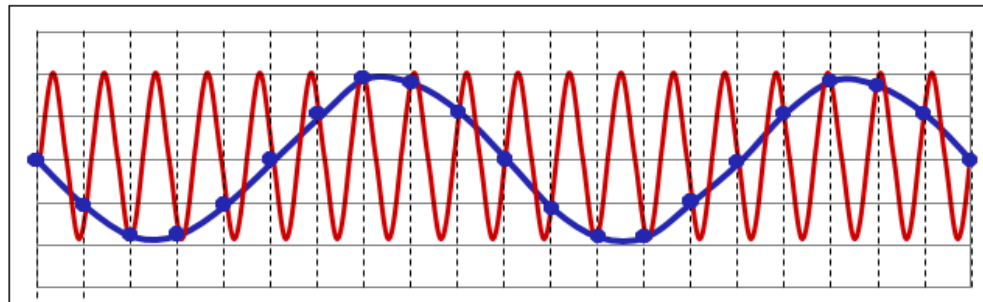
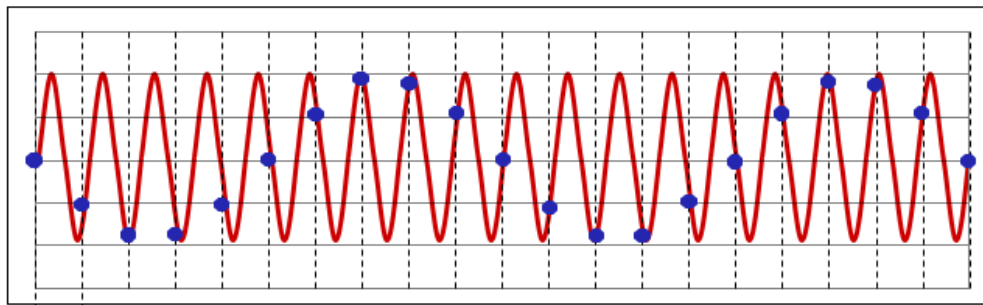
$$N \Rightarrow \text{Nyquist}$$

Aster and Borchers Figure 3.1: The Shah function and its Fourier Transform; Fourier Transform of a Sampled Function (slightly aliased)



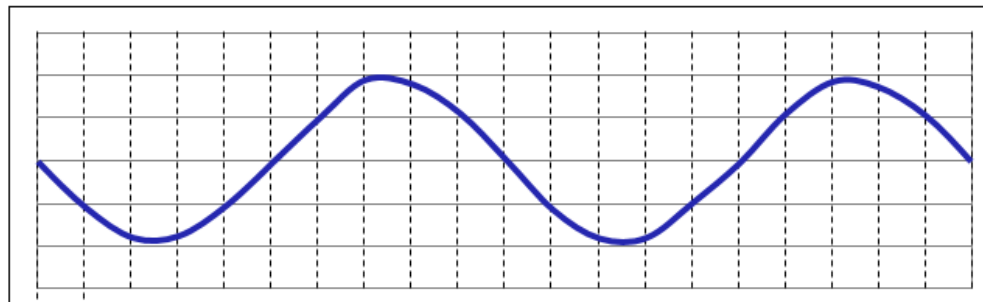
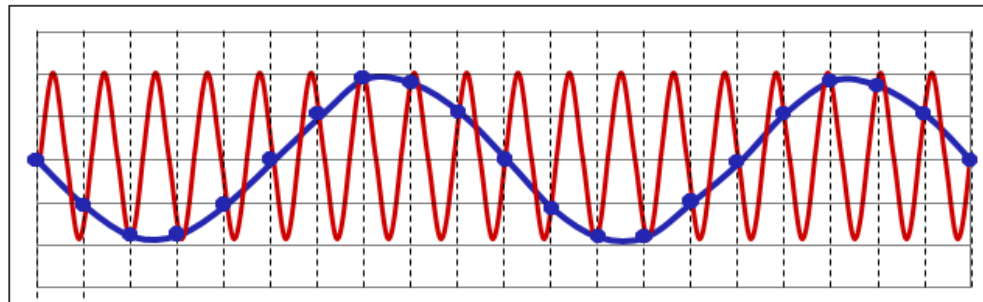
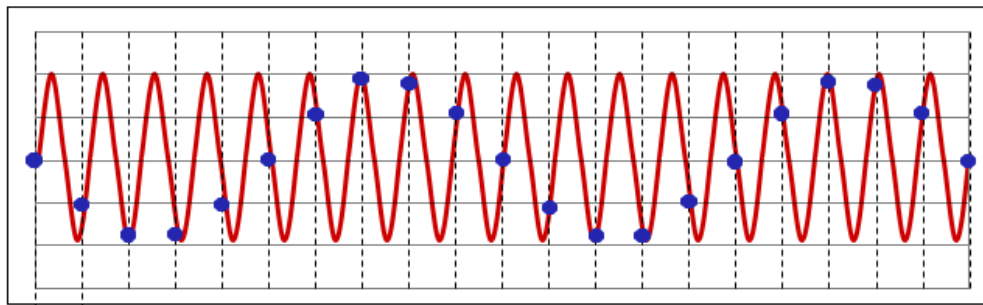


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Aliased signals map to  $(-f_{max}, f_{max})$  by  $f_a = \pm(r - f)$  where  $f_a$  is the aliased frequency on  $(-f_{max}, f_{max})$ ,  $r$  is the sample rate (or Nyquist frequency) and  $f$  is the true unaliased frequency.



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e.g.  $f = 60\text{Hz}$ ,  $r = 100\text{sps}$ ,  
 $f_a = \pm 40\text{Hz}$

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$$\begin{aligned} \text{Let } s_n &= e^{i2\pi f n} \\ g_n &= x_n * s_n = \sum_{k=-\infty}^{\infty} x_k e^{i2\pi(n-k)} = e^{i2\pi f n} \sum_{k=-\infty}^{\infty} x_k e^{-i2\pi f k} \\ &= e^{i2\pi f n} X(f) \end{aligned}$$

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$$\begin{aligned} \text{Much like} \quad g(t) &= \phi(t) * e^{i2\pi f t} = \int_{-\infty}^{\infty} \phi(\tau) e^{i2\pi f(t-\tau)} d\tau \\ &= e^{i2\pi f t} \int_{-\infty}^{\infty} \phi(\tau) e^{-i2\pi f \tau} d\tau = e^{i2\pi f t} \Phi(f) \end{aligned}$$

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The integral is 0 for  $rt - n \neq 0$  and 1 for  $t = n/r$ .

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Let  $r = 1$  for simplicity which implies  $f_{max} = 1/2$ .

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While this is a transform pair ( $x_n$  and  $X(f)$ ), it is not symmetric and  $n$  is discrete but infinite and  $f$  is continuous but finite.

We have this less than ideal transform pair,

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$$x_n = \frac{1}{N} \sum_{k=0}^{N-1} X_k e^{i2\pi kn/N} = IDFT[X_k] \quad X_k = \sum_{n=0}^{N-1} x_n e^{-i2\pi kn/N} = DFT[x_n]$$

This is the discrete Fourier Transform.

To verify the transform pair we apply the forward DFT both sides of the IDFT (though we must take care with our choice of summation indices since  $k$  is already used for the sequence  $X_k$ ).

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When  $(k - m)$  is a multiple of  $N$ ,  $(k-m)/N$  is an integer and,

$$e^{i2\pi(k-m)/N} = e^{i2\pi j} = 1 \quad \begin{array}{l} \sin(j2\pi) = 0 \\ \cos(j2\pi) = 1 \end{array} \quad \text{For integer } j$$

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When  $(k - m)$  is not a multiple of  $N$  it's a bit more complicated.

Consider the geometric series,

$$s = a + ar + ar^2 + ar^3 + \cdots + ar^{n-1}$$



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$$s = a \frac{1 - r^n}{1 - r}$$

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Remember we're working with values of  $(k - m)$  not integer multiples of  $N$ .

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So that

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We're only working with values of  $k$  and  $m$  on the interval  $(0, N - 1)$ ,

$$\sum_{n=0}^{N-1} (e^{i2\pi(k-m)/N})^n = N\delta_{k-m}$$

Now back to the DFT.

$$\sum_{n=0}^{N-1} x_n e^{-i2\pi nm/N} = \frac{1}{N} \sum_{k=0}^{N-1} X_k \sum_{n=0}^{N-1} \left( e^{i2\pi(k-m)/N} \right)^n$$

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So we successfully verified our DFT/IDFT pair.

$$\Phi_k = \sum_{n=0}^{N-1} \phi_n e^{-i2\pi kn/N} \qquad \phi_n = \frac{1}{N} \sum_{k=0}^{N-1} \Phi_k e^{i2\pi kn/N}$$

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Similarly to the continuous FT, we can decompose  $\phi_n$  into its constituent frequency components  $\Phi_k$ , and we can reconstruct  $\phi_n$  from those frequency components. Except here, both frequency and time are finite length, periodic, and discrete. (very handy for working with computers)

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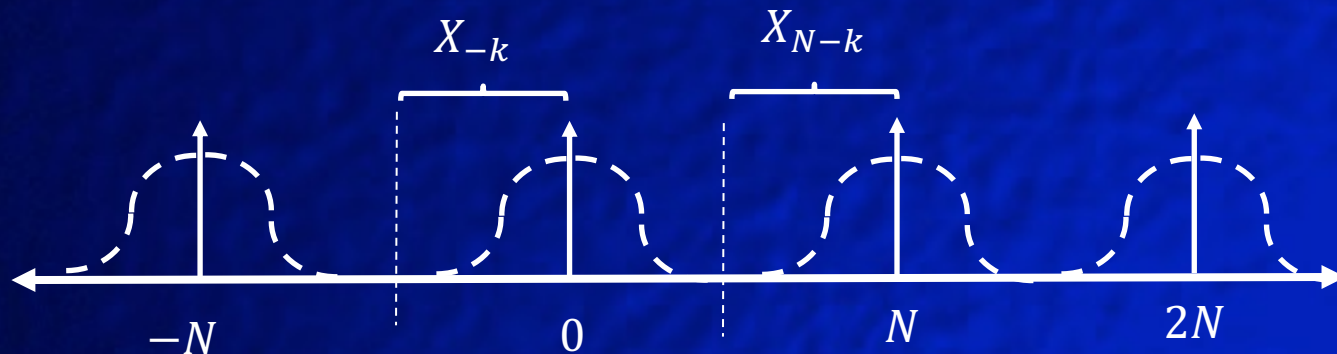
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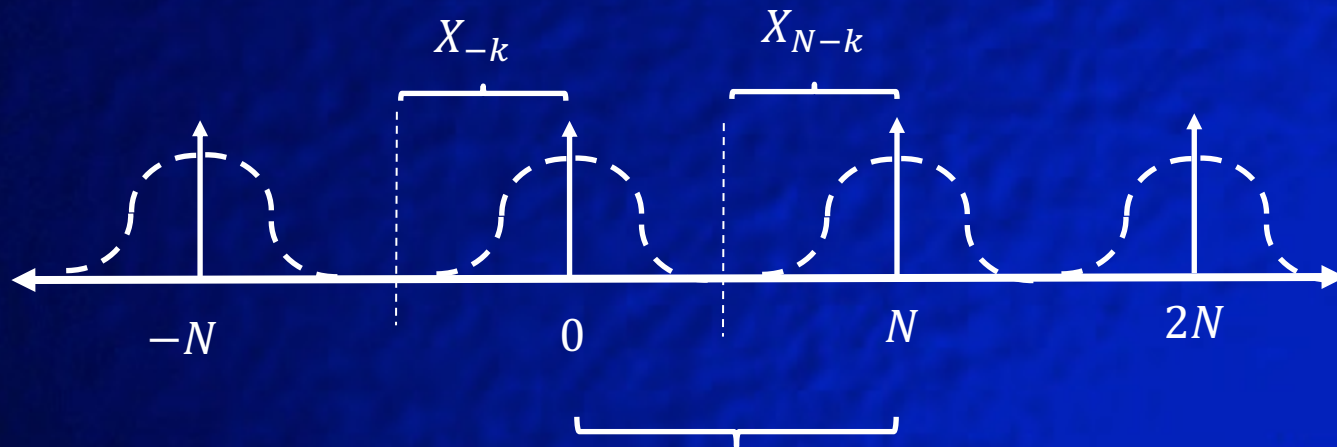


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Matlab FFT vector.



## Indexing

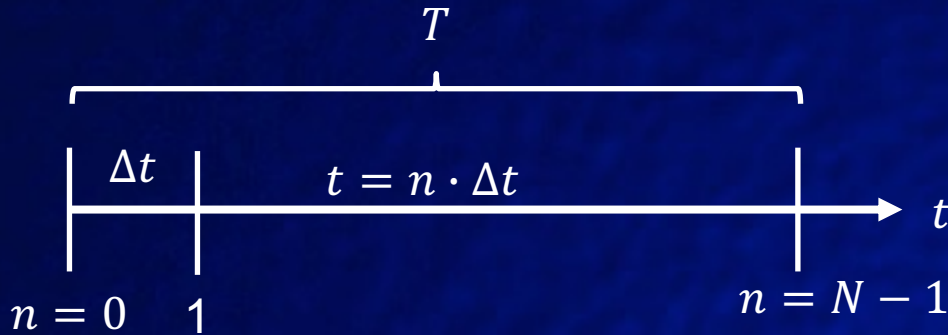
If  $N$  is even, plot  $-\frac{N}{2} \leq k \leq \left(\frac{N}{2} - 1\right)$

If  $N$  is odd, plot  $-\frac{(N-1)}{2} \leq k \leq \frac{(N-1)}{2}$

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Time resolution

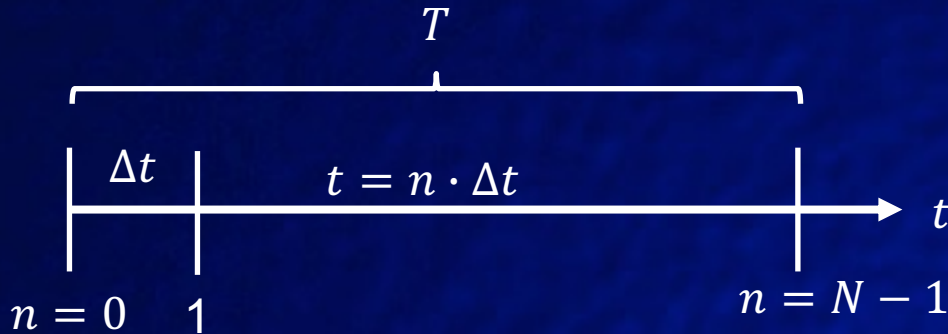
$$\Delta t = \frac{T}{N}, \quad N = T \cdot r$$

$r = \text{sample rate, samps/sec}$

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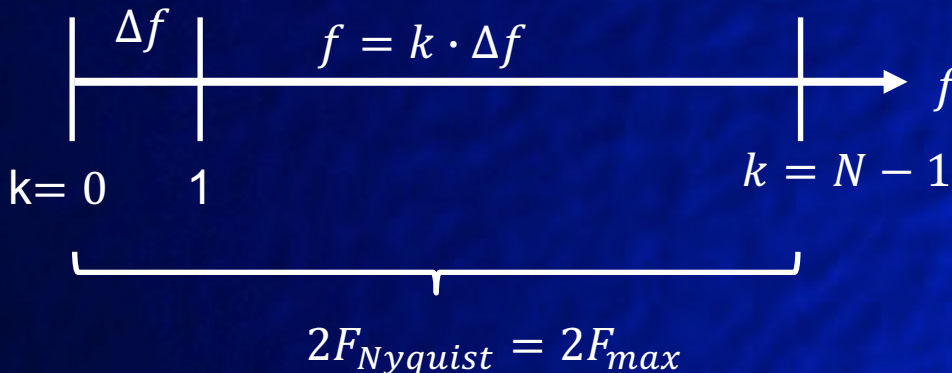
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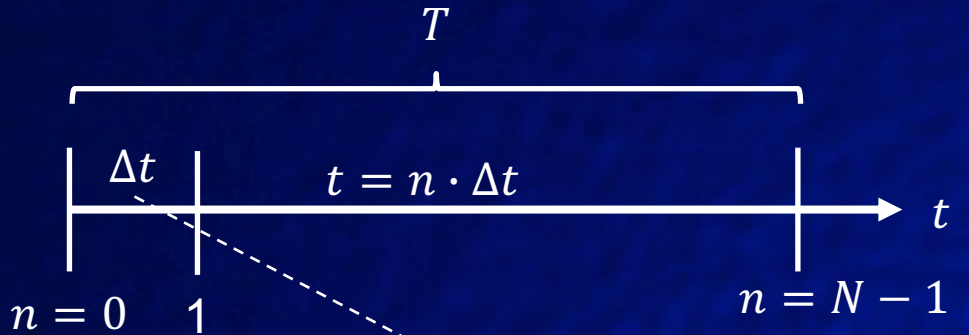
Frequency resolution

$$\Delta f = \frac{2F_{max}}{N} = \frac{r}{N}$$

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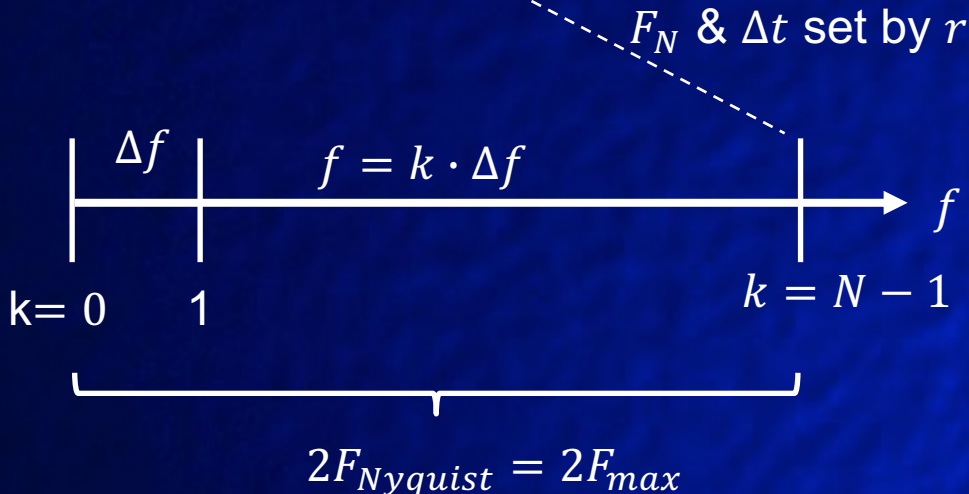
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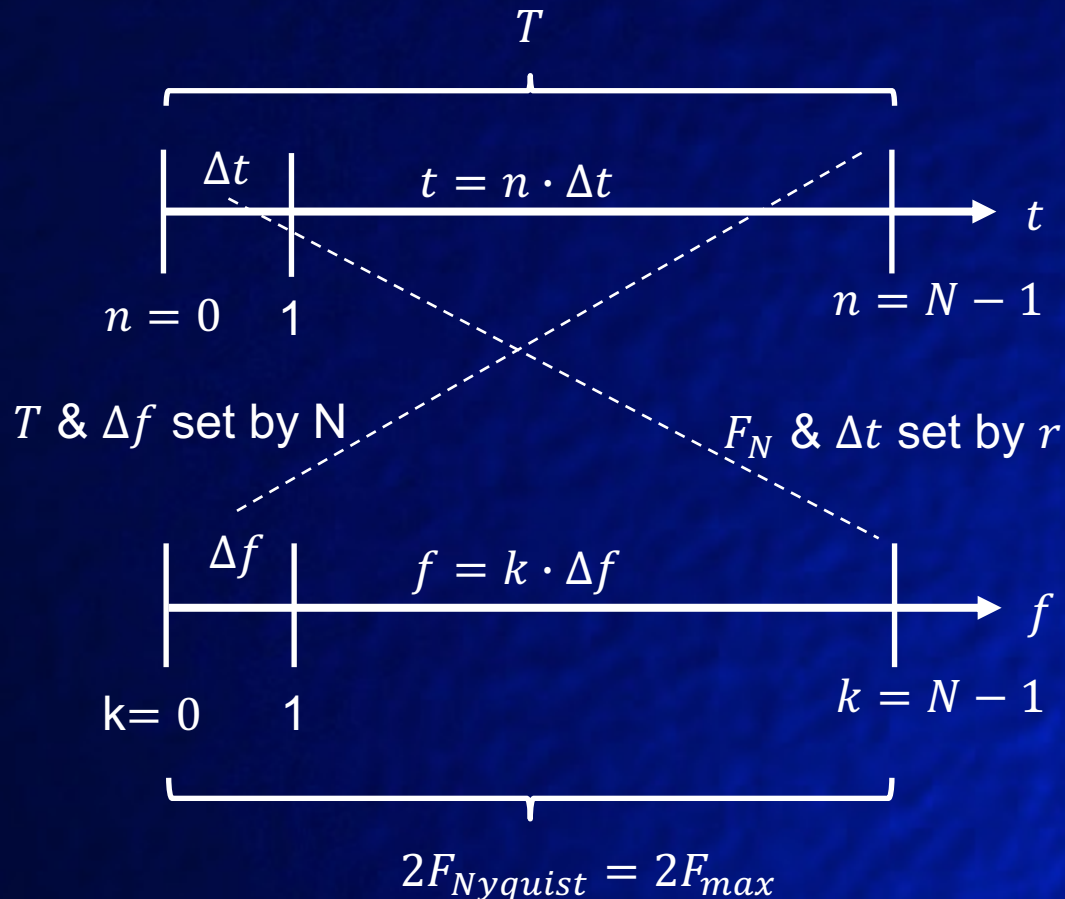
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$\Delta f$ , or frequency resolution, is limited by  $N$  or  $T$ .

## Discrete Convolution

Let  $z_n = x_n * y_n$

$$DFT[z_n] = X_k Y_k$$

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$$= \frac{1}{N} \sum_{l=0}^{N-1} x_l \sum_{m=0}^{N-1} y_m \sum_{k=0}^{N-1} e^{i2\pi k(n-m-l)/N}$$

From before, 
$$\sum_{k=0}^{N-1} e^{i2\pi(n-m-l)/N} = \begin{cases} N, & n - m - l = 0 \\ 0, & \text{otherwise} \end{cases}$$



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The discrete convolution.



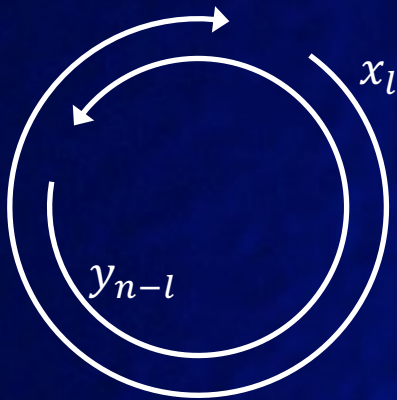
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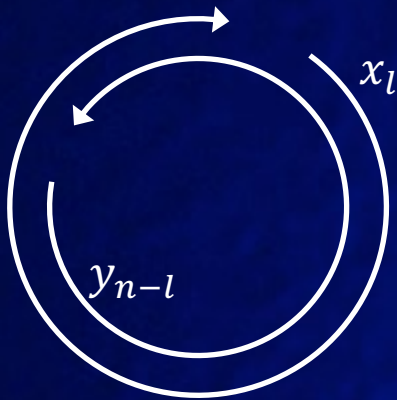
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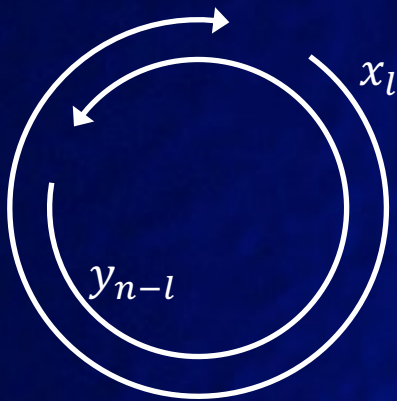
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Multiply and sum, then rotate one point. Repeat for  $N - 1$  points.



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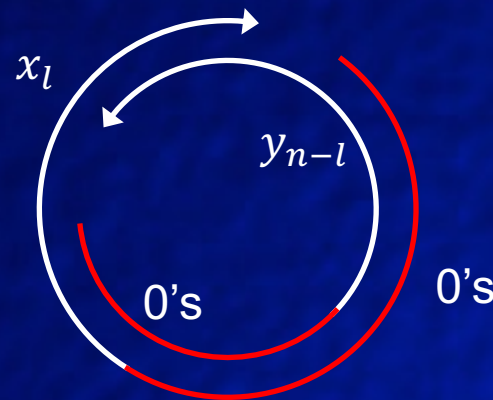
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We almost always want linear convolution but only have  $N - 1$  and summing over  $(-\infty, \infty)$  is not feasible for discrete finite length series. So we approximate linear convolution using circular convolution by padding with zeros.

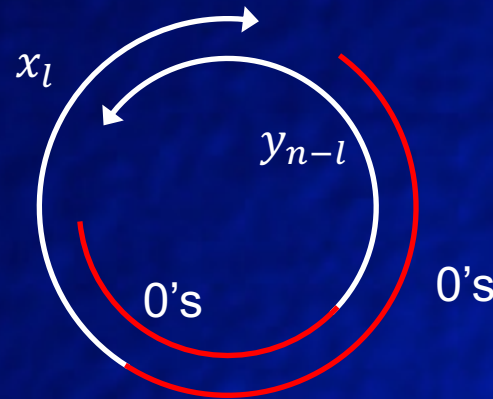
If  $x_n$  has  $N$  points and  $y_n$  has  $M$  points, we pad  $x$  with  $M$  points and  $y$  with  $N$  points then keep the first  $N + M - 1$  values in the convolution.

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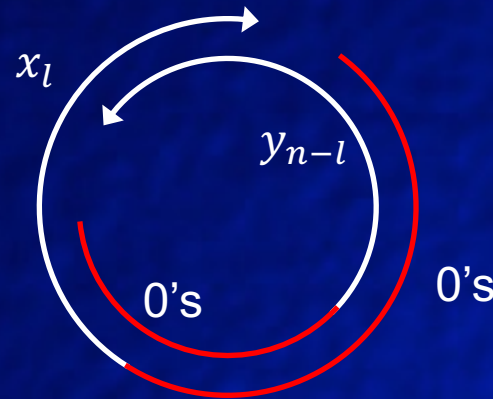


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Another way of looking at it is windowing  $x$  and  $y$  by a boxcar prior to convolution so that all points outside the window are 0.



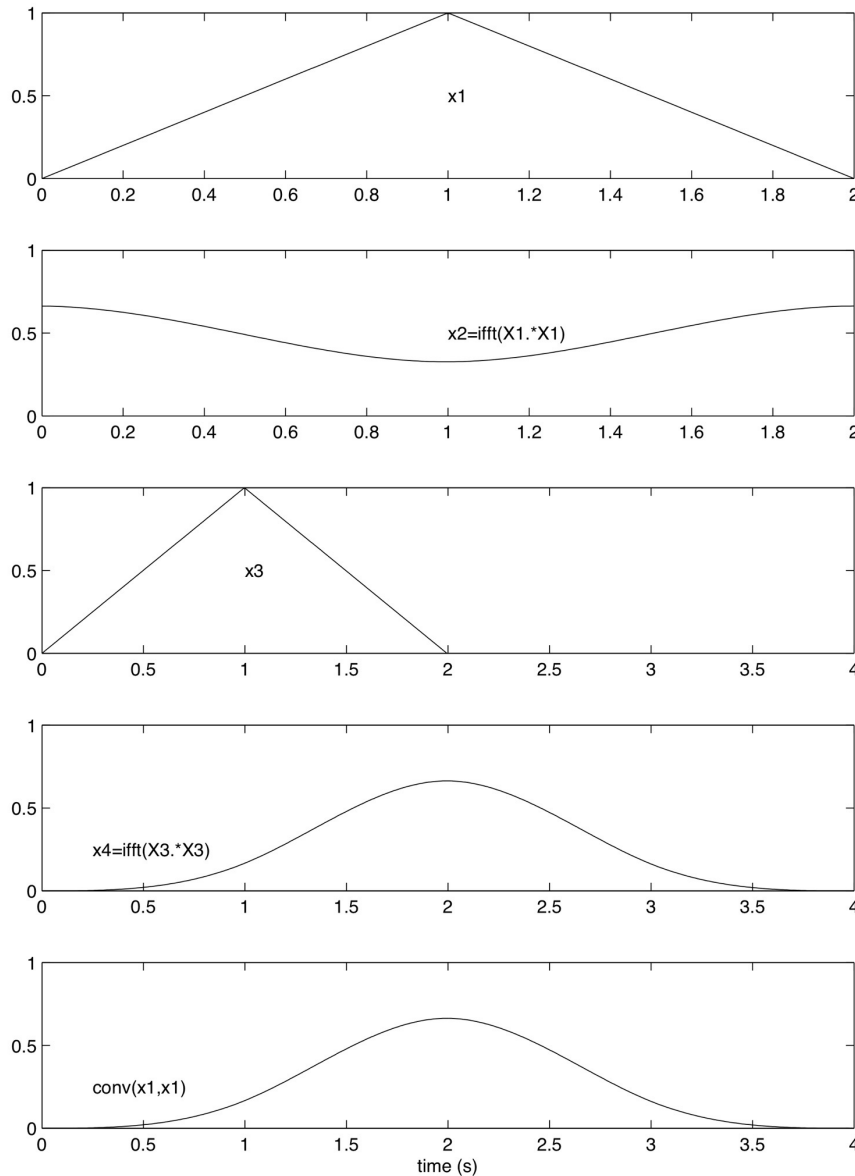
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Our data are finite length, so we're still performing circular convolution, but we approximate the linear convolution by windowing the data.



The matlab *conv* command takes care of the padding for you and returns  $N + M - 1$ , points.

Figure to the left is from *wrap.m* that we reviewed earlier.

More discrete analogs to the continuous FT.

$$DFT[x_{n-n_0}] = \sum_{n=0}^{N-1} x_{n-n_0} e^{-i2\pi kn/N}$$

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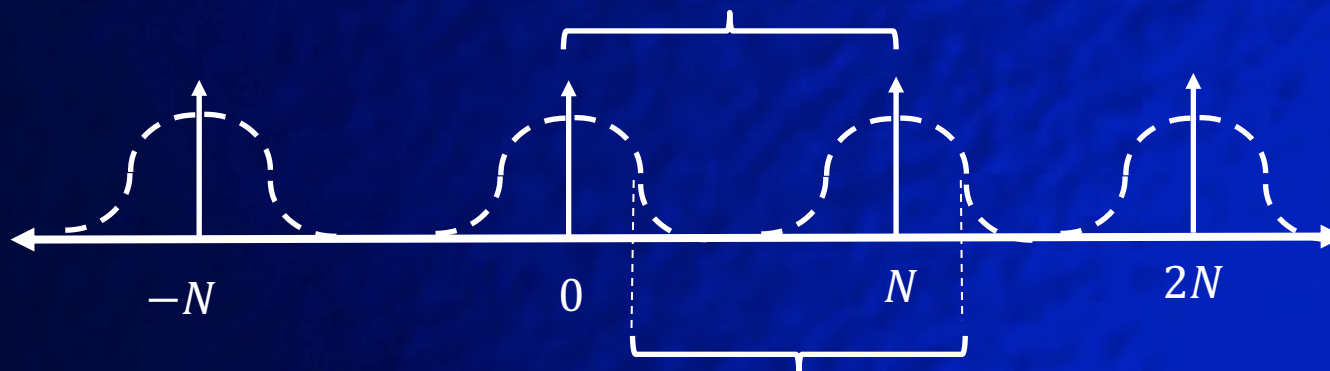
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Time shift



Phase shift

$$\sum_{n=0}^{N-1} |x_n|^2 = \sum_{n=0}^{N-1} x_n x_n^*$$

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$$\begin{aligned}
\sum_{n=0}^{N-1} |x_n|^2 &= \sum_{n=0}^{N-1} x_n x_n^* = \sum_{n=0}^{N-1} \left[ \frac{1}{N} \sum_{k=0}^{N-1} X_k e^{i2\pi kn/N} \right] \left[ \frac{1}{N} \sum_{l=0}^{N-1} X_l e^{i2\pi ln/N} \right]^* \\
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\end{aligned}$$

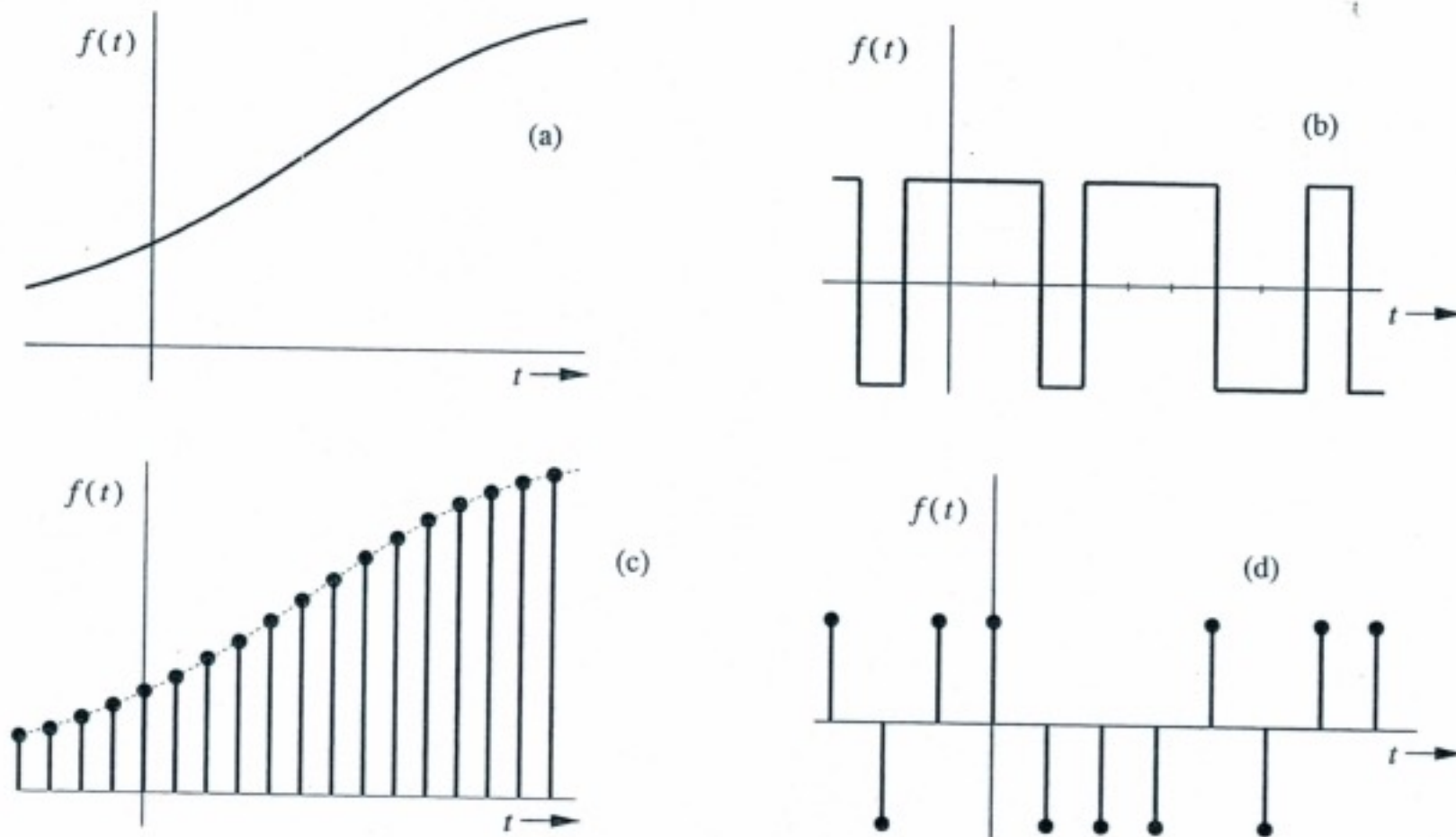
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\sum_{n=0}^{N-1} |x_n|^2 &= \sum_{n=0}^{N-1} x_n x_n^* = \sum_{n=0}^{N-1} \left[ \frac{1}{N} \sum_{k=0}^{N-1} X_k e^{i2\pi kn/N} \right] \left[ \frac{1}{N} \sum_{l=0}^{N-1} X_l e^{i2\pi ln/N} \right]^* \\
&= \sum_{n=0}^{N-1} \left[ \frac{1}{N} \sum_{k=0}^{N-1} X_k e^{i2\pi kn/N} \right] \left[ \frac{1}{N} \sum_{l=0}^{N-1} X_l^* e^{-i2\pi ln/N} \right] \\
&= \frac{1}{N^2} \sum_{n=0}^{N-1} \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} X_k X_l^* e^{i2\pi n(k-l)/N} = \frac{1}{N^2} \sum_{n=0}^{N-1} \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} X_k X_l^* (e^{i2\pi(k-l)/N})^n \\
&= \frac{1}{N^2} \sum_{n=0}^{N-1} \sum_{k=0}^{N-1} X_k \sum_{l=0}^{N-1} X_l^* \delta_{k-l}
\end{aligned}$$



$$\begin{aligned}
\sum_{n=0}^{N-1} |x_n|^2 &= \sum_{n=0}^{N-1} x_n x_n^* = \sum_{n=0}^{N-1} \left[ \frac{1}{N} \sum_{k=0}^{N-1} X_k e^{i2\pi kn/N} \right] \left[ \frac{1}{N} \sum_{l=0}^{N-1} X_l e^{i2\pi ln/N} \right]^* \\
&= \sum_{n=0}^{N-1} \left[ \frac{1}{N} \sum_{k=0}^{N-1} X_k e^{i2\pi kn/N} \right] \left[ \frac{1}{N} \sum_{l=0}^{N-1} X_l^* e^{-i2\pi ln/N} \right] \\
&= \frac{1}{N^2} \sum_{n=0}^{N-1} \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} X_k X_l^* e^{i2\pi n(k-l)/N} = \frac{1}{N^2} \sum_{n=0}^{N-1} \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} X_k X_l^* (e^{i2\pi(k-l)/N})^n \\
&= \frac{1}{N^2} \sum_{n=0}^{N-1} \sum_{k=0}^{N-1} X_k \sum_{l=0}^{N-1} X_l^* \delta_{k-l} = \frac{1}{N^2} \sum_{k=0}^{N-1} X_k X_k^* \sum_{n=0}^{N-1} 1
\end{aligned}$$

$$\begin{aligned}
\sum_{n=0}^{N-1} |x_n|^2 &= \sum_{n=0}^{N-1} x_n x_n^* = \sum_{n=0}^{N-1} \left[ \frac{1}{N} \sum_{k=0}^{N-1} X_k e^{i2\pi kn/N} \right] \left[ \frac{1}{N} \sum_{l=0}^{N-1} X_l e^{i2\pi ln/N} \right]^* \\
&= \sum_{n=0}^{N-1} \left[ \frac{1}{N} \sum_{k=0}^{N-1} X_k e^{i2\pi kn/N} \right] \left[ \frac{1}{N} \sum_{l=0}^{N-1} X_l^* e^{-i2\pi ln/N} \right] \\
&= \frac{1}{N^2} \sum_{n=0}^{N-1} \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} X_k X_l^* e^{i2\pi n(k-l)/N} = \frac{1}{N^2} \sum_{n=0}^{N-1} \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} X_k X_l^* (e^{i2\pi(k-l)/N})^n \\
&= \frac{1}{N^2} \sum_{n=0}^{N-1} \sum_{k=0}^{N-1} X_k \sum_{l=0}^{N-1} X_l^* \delta_{k-l} = \frac{1}{N^2} \sum_{k=0}^{N-1} X_k X_k^* \sum_{n=0}^{N-1} 1 \\
&= \frac{1}{N} \sum_{k=0}^{N-1} |X_k|^2
\end{aligned}$$

Discrete Analog to Parseval.

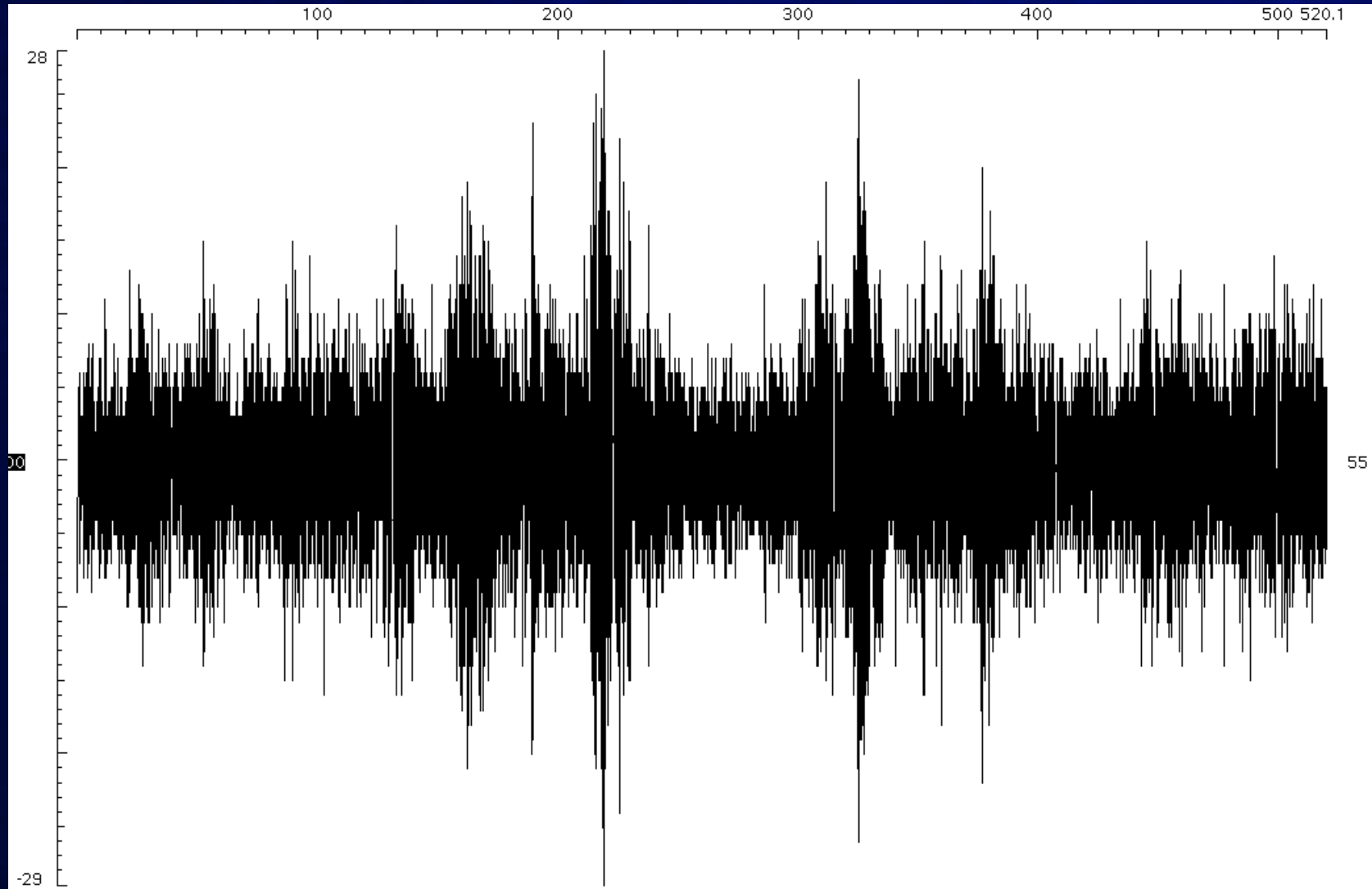


**Fig. 1.5** Examples of Signals: (a) analog, continuous-time (b) digital, continuous-time (c) analog, discrete-time (d) digital, discrete-time.

# Example of amplitude resolution CATM.EHZ.NM.00

$$2^{12} = \pm 2048$$

$$2^{24} = \pm 8,388,608$$





Examples of Discrete Processes from Signal Processing and Linear Systems by B.P. Lathi, 1998, Section 8.5, pp 562-564.

Discrete systems don't necessarily need to be digitized versions of continuous systems.

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Consider a person who makes bank deposits at regular intervals,  $T$  (e.g. once each month). The bank pays interest on the balance during  $T$ . We wish to find the output (the account balance) of the “system” to the input (the deposit).

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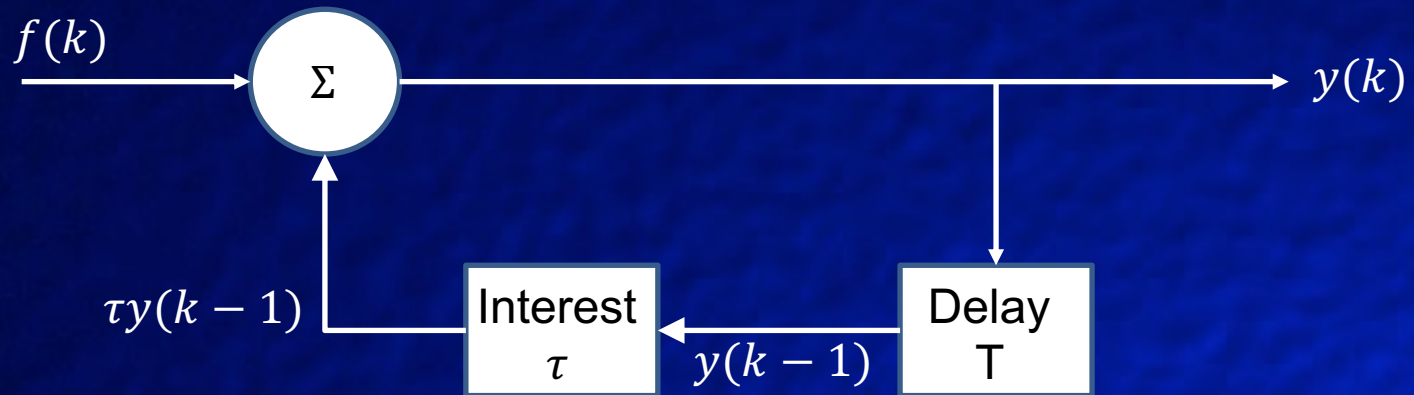
Consider a person who makes bank deposits at regular intervals,  $T$  (e.g. once each month). The bank pays interest on the balance during  $T$ . We wish to find the output (the account balance) of the “system” to the input (the deposit).

$f(k)$  = deposit made at  $k^{th}$  interval

$y(k)$  = account balance at  $k$  immediately after the deposit

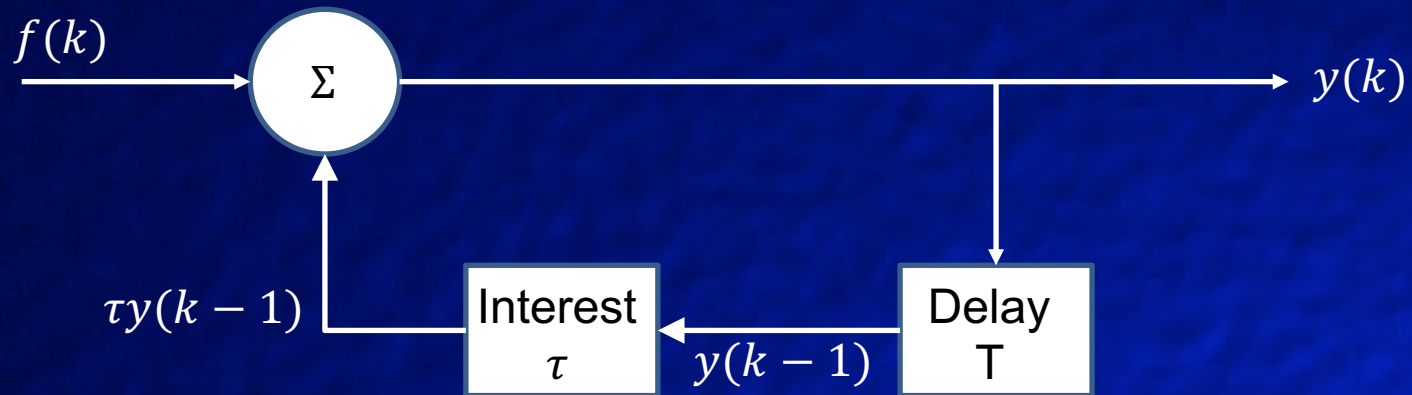
$\tau$  = interest rate per dollar per  $T$

The balance,  $y(k)$ , is the sum of the previous balance,  $y(k - 1)$ , the interest earned on  $y(k - 1)$  during  $T$ , and the deposit  $f(k)$ .





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$$y(k) = y(k - 1) + \tau y(k - 1) + f(k)$$

$$= (1 + \tau)y(k - 1) + f(k)$$

If we assume monthly deposits are a constant,  $D$ , then

$$y_0 = D$$

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$$y_0 = D$$

$$y_1 = \tau y_0 + D = \tau D + D$$

$$y_2 = \tau y_1 + D = \tau^2 D + \tau D + D$$

$$y_N = \sum_{n=0}^{N-1} \tau^n D$$

$f(k)$  students register for a class in the  $k^{th}$  semester and must buy a book. Books last three semesters.  $1/4$  of the books from the previous semester are sold as used.  $y(k)$  is the number of new books sold.

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$y(k)$  = number of new books sold in semester  $k$ .

$y(k - 1)$  = number of new books sold in semester  $k - 1$ .

$1/4 y(k - 1)$  of these new books are sold as used in semester  $k$ .



$f(k)$  students register for a class in the  $k^{th}$  semester and must buy a book. Books last three semesters.  $1/4$  of the books from the previous semester are sold as used.  $y(k)$  is the number of new books sold.

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$y(k - 1)$  = number of new books sold in semester  $k - 1$ .

$1/4 y(k - 1)$  of these new books are sold as used in semester  $k$ .

$y(k - 2)$  = number of new books sold in semester  $k - 2$ .

$1/4 y(k - 2)$  of these new books are sold as used in semester  $k - 1$  and  $1/4$  of those, or  $1/16 y(k - 2)$  are sold as used in semester  $k$ .

$f(k)$  students register for a class in the  $k^{th}$  semester and must buy a book. Books last three semesters.  $\frac{1}{4}$  of the books from the previous semester are sold as used.  $y(k)$  is the number of new books sold.

$y(k)$  = number of new books sold in semester  $k$ .

$y(k - 1)$  = number of new books sold in semester  $k - 1$ .

$\frac{1}{4}y(k - 1)$  of these new books are sold as used in semester  $k$ .

$y(k - 2)$  = number of new books sold in semester  $k - 2$ .

$\frac{1}{4}y(k - 2)$  of these new books are sold as used in semester  $k - 1$  and  $\frac{1}{4}$  of those, or  $\frac{1}{16}y(k - 2)$  are sold as used in semester  $k$ .

$$f(k) = y(k) + \frac{1}{4}y(k - 1) + \frac{1}{16}y(k - 2)$$

$$f(k) = y(k) + \frac{1}{4}y(k-1) + \frac{1}{16}y(k-2)$$

