

Random Process Examples

Ex. #1: D-T White Noise

Let $x[k]$ be a sequence of RV's where...
each RV $x[k]$ in the sequence is
uncorrelated with all the others:

$$E\{ x[k] x[m] \} = 0 \quad \text{for } k \neq m$$

**This DEFINES a DT White Noise
Also called “Uncorrelated Process”**

Physically, uncorrelated means that knowing $x[k]$ provides no insight into what value $x[m]$ (for $m \neq k$) will be likely to take (roll a die; the value you get provides no insight into what you expect to get on any future roll)

Ex. #1: D-T White Noise

Recall: Gaussian processes are very common, so...

Also, let $x[k]$ be **Gaussian** with Zero mean and Variance σ^2

Define RP's PDF;
also called "Normal"

$X[k] \sim N(0, \sigma^2)$
Mean Variance

$$\text{PDF} : p(x; k) = \frac{1}{2\pi\sigma} e^{\left(\frac{-x^2}{2\sigma^2}\right)}$$

Note: Does not depend on k
→ at least the 1st order PDF is time invariant

Ex. #1: D-T White Noise

TASK : We have a model.... Find the mean, ACF, and check if WSS (also find variance of process)

MEAN of Process :

$$E \{ x[k] \} = 0 \quad \text{CONSTANT}$$

By definition !

ACF:

$$R_x(k_1, k_2) = E \{ x[k_1] \cdot x[k_2] \}$$

By our definition of white noise,
....this is 0 if $k_1 \neq k_2$

Ex. #1: D-T White Noise

Now for $k_1=k_2=k$:

$$R_x(k_1, k_2) \Big|_{k=k_1=k_2} = E\{x^2[k]\} = \sigma^2$$

Thus,

$$R_x(k_1, k_2) = \begin{cases} \sigma^2, & k_1 = k_2 \\ 0, & k_1 \neq k_2 \end{cases}$$

By definition of variance for zero-mean case

$$= \sigma^2 \delta[k_1 - k_2] \quad \begin{matrix} \nearrow \\ = m \\ \text{(like } \tau \text{ for cont-time ACF)} \end{matrix}$$

ACF for DT White RP

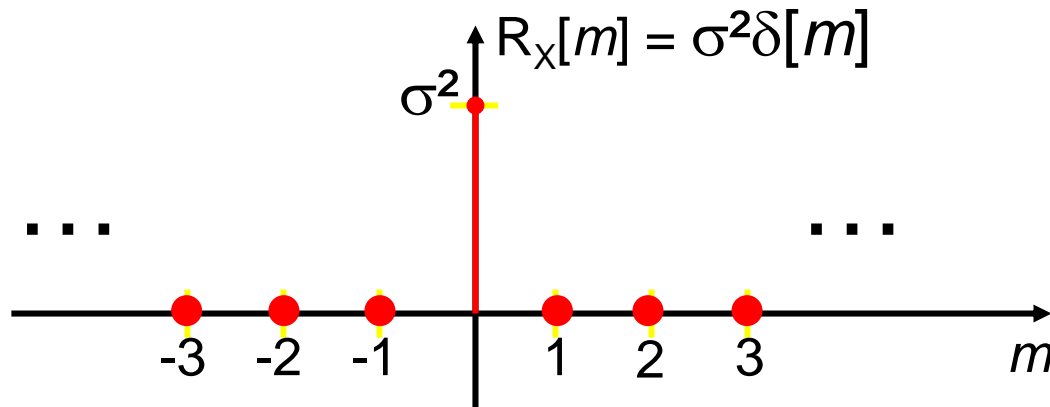


$$\Rightarrow R_x(m) = \sigma^2 \delta[m]$$



Ex. #1: D-T White Noise

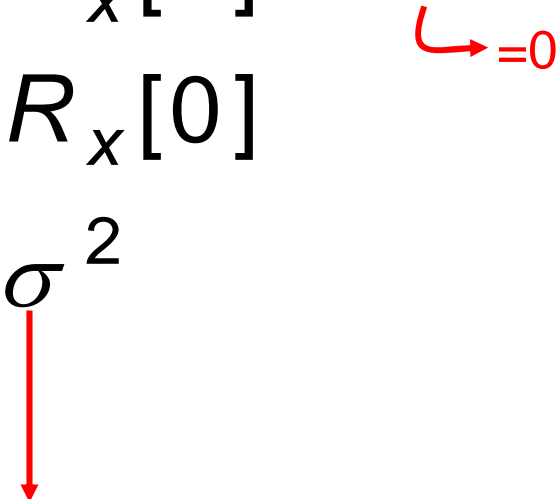
ACF displays lack of correlation between any pair of any time instants:



Now since we have constant mean and ACF depends only on $m = k_2 - k_1 \Rightarrow$ **WSS**

Ex. #1: D-T White Noise

Variance $\sigma_x^2 = R_x[0] - \bar{x}^2$
 $= R_x[0]$
 $= \sigma^2$



For this case:

Variance of the Process = Variance of the RV

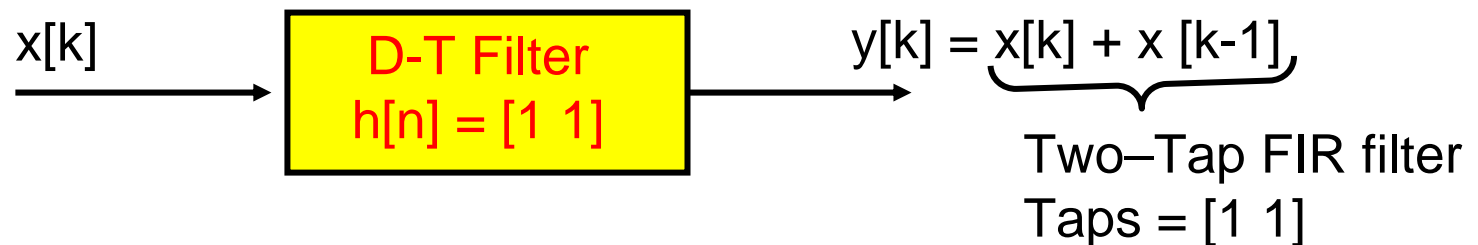
Ex. #2: Filtered D-T RP

Start with White RP $x[k]$ in previous example

Recall : Zero Mean Process

$$R_x [m] = \sigma^2 \delta[m]$$

\Rightarrow **WSS**



Ex. #2: Filtered D-T RP

TASK: Is $y[k]$ WSS?

\Rightarrow need to find mean & ACF

MEAN: Using filter output expressions gives

$$\begin{aligned} E\{y[k]\} &= E\{x[k] + x[k-1]\} \\ &= E\{x[k]\} + E\{x[k-1]\} \end{aligned}$$

$$\Rightarrow E\{y[k]\} = 0$$

Ex. #2: Filtered D-T RP

ACF :

Plug in Eq.
for output



$$\begin{aligned} R_y(k_1, k_2) &= E\{y[k_1]y[k_2]\} \\ &= E\{(x[k_1] + x[k_1 - 1])(x[k_2] + x[k_2 - 1])\} \\ &= \underbrace{E\{x[k_1]x[k_2]\}}_{R_x(k_2 - k_1)} + \underbrace{E\{x[k_1]x[k_2 - 1]\}}_{R_x(k_2 - k_1 - 1)} \\ &\quad + \underbrace{E\{x[k_1 - 1]x[k_2]\}}_{R_x(k_2 - k_1 + 1)} + \underbrace{E\{x[k_1 - 1]x[k_2 - 1]\}}_{R_x((k_2 - 1) - (k_1 - 1))} \end{aligned}$$

Ex. #2: Filtered D-T RP

$$R_y(k_1, k_2) = \underbrace{R_x(k_2 - k_1)}_{\sigma^2 \delta[m]} + \underbrace{R_x(k_2 - k_1 - 1)}_{\sigma^2 \delta[m-1]} \\ + \underbrace{R_x(k_2 - k_1 + 1)}_{\sigma^2 \delta[m+1]} + \underbrace{R_x((k_2 - 1) - (k_1 - 1))}_{\sigma^2 \delta[m]}$$

where $m = k_2 - k_1$

y[k] is WSS

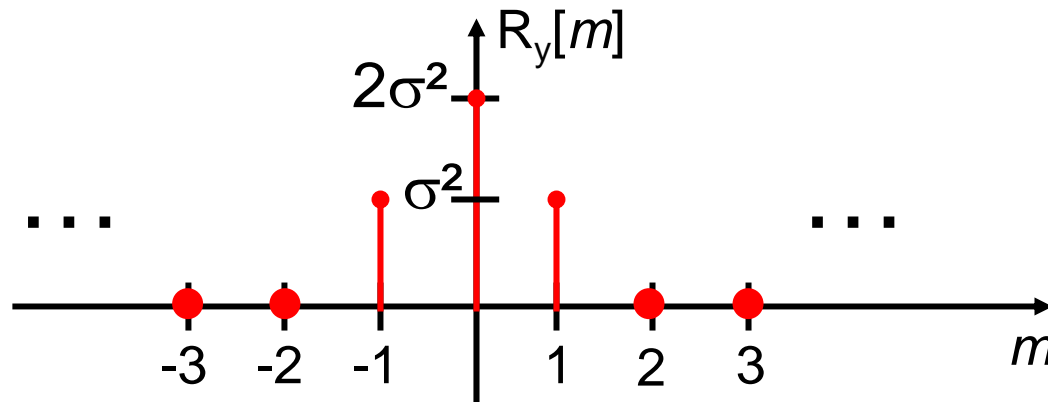
ACF for 2-Tap Filtered White RP



$$\Rightarrow R_Y(m) = \sigma^2[2\delta[m] + \delta[m-1] + \delta[m+1]]$$



Ex. #2: Filtered D-T RP

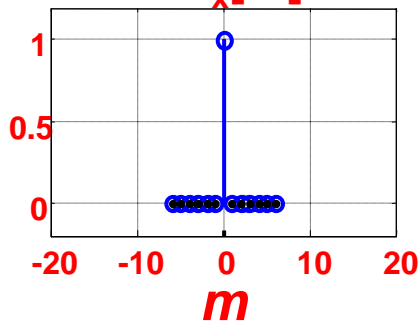


Note : Filter introduces correlation between adjacent samples - but still no correlation for samples 2 or more samples apart (for this filter)

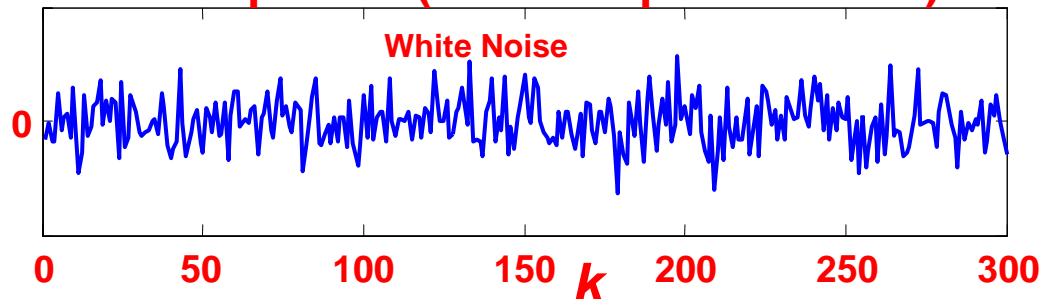
Big Picture: Filtered RP

Filters can be used to change the correlation structure of a RP:
RP:

ACF $R_x[m]$ of Input



Input RP (One Sample Function)



$x[k]$

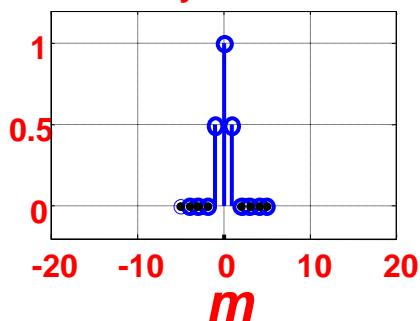


D-T Filter
 $h[n] = [1 \ 1]$

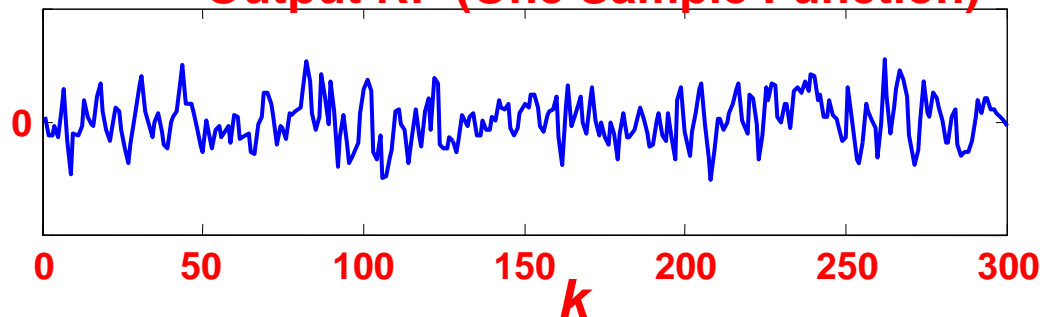
$y[k] = x[k] + x[k-1]$



ACF $R_y[m]$ of Output

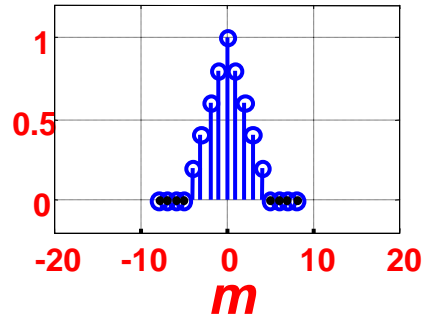


Output RP (One Sample Function)

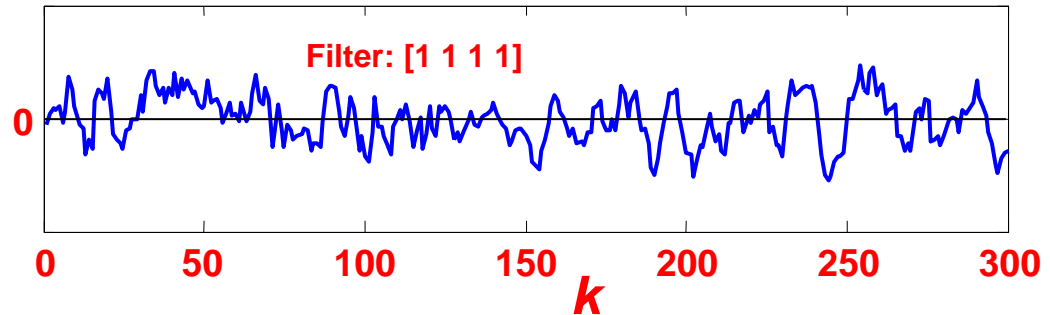


Big Picture: Filtered RP (cont)

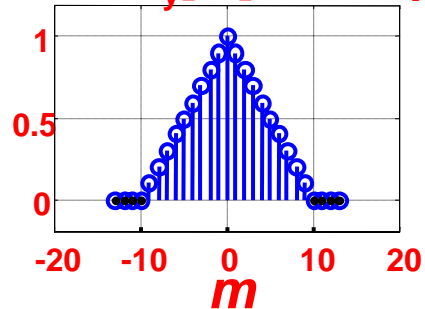
ACF $R_y[m]$ of Output



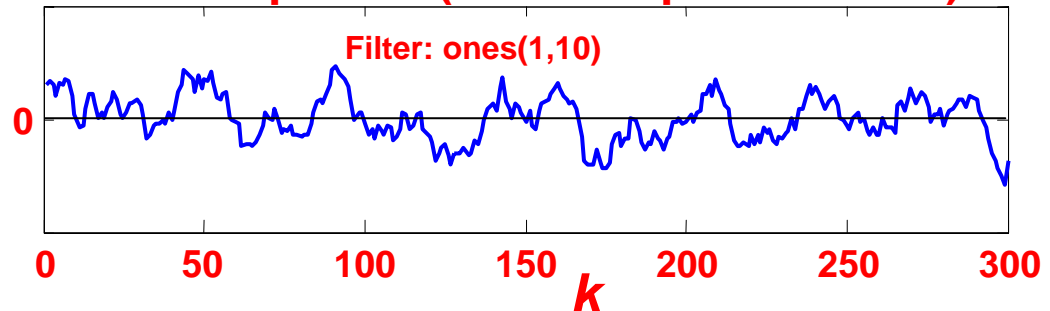
Output RP (One Sample Function)



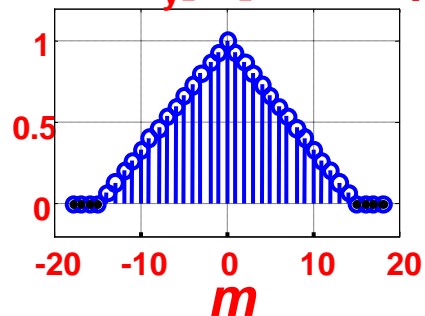
ACF $R_y[m]$ of Output



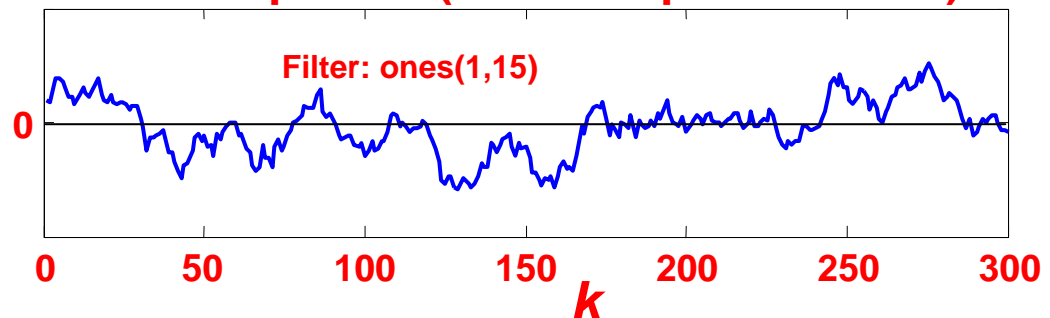
Output RP (One Sample Function)



ACF $R_y[m]$ of Output



Output RP (One Sample Function)



Filtered RPs: Insight

Our study of the ACFs of filtered random processes and the degree of “smoothness” of the sample functions shows the following general result:

Narrow ACF ↔ Rapid Fluctuations
Broad ACF ↔ Slow Fluctuations

Ex. #3: Exp. Sig. + White Noise

Define a RP as:

$$y[k] = a^k x + w[k]$$

Deterministic
Function

$a = \text{deter. \#}$

Random Variable
(picked once randomly
and then fixed)

White Noise
(get a new random value at
each k – uncorrelated
from sample-to-sample)

Ex. #3: Exp. Sig. + WN

Further Definition of this RP:

- $w[k]$ is **white noise**
- Each $w[k]$ is a Gaussian RV with $w[k] \sim N(0, \sigma_w^2)$
- X is a Gaussian RV with $X \sim N(0, \sigma_X^2)$
- RV X is independent of each RV $w[k]$
 $\Rightarrow E\{x w[k]\} = E\{x\} E\{w[k]\} = 0$
- The # a is a deterministic number

Ex. #3: Exp. Sig. + WN

TASK: Is this WSS?

= 0

= 0

MEAN: $E\{y[k]\} = a^k E\{X\} + E\{w[k]\}$
 $= 0$ (Constant)

ACF: $R_y(k, k+m) = E\{y[k] y[k+m]\}$
 $= E\{(a^k X + w[k]) (a^{k+m} X + w[k+m])\}$

$$= a^{2k+m} \underbrace{E\{X^2\}}_{\sigma_X^2} + a^k E\{X w[k+m]\} + a^{k+m} E\{X w[k]\} + E\{w[k] w[k+m]\}$$

$= \sigma_w^2 \delta[m]$

= 0
Due to
Indep.

Ex. #3: Exp. Sig. + WN

➔ $R_y(k, k+m) = \sigma_x^2 a^{2k+m} + \sigma_w^2 \delta[m]$

In General: Depends on k not just m

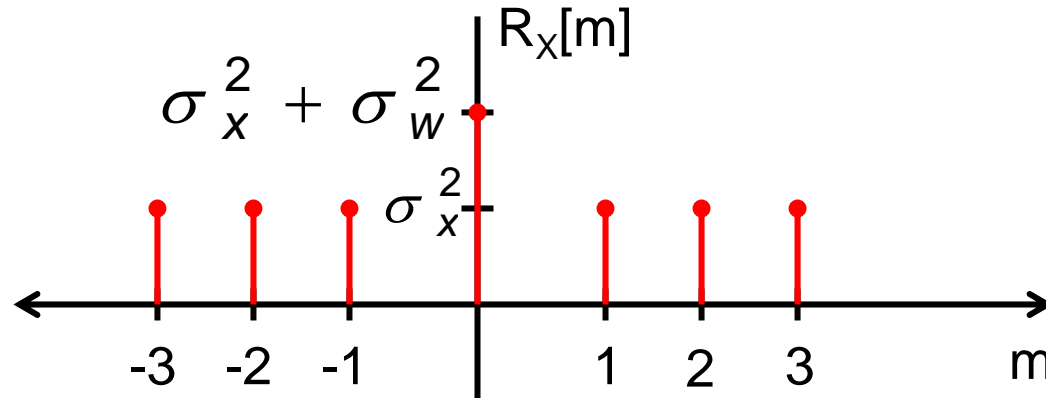
➔ Not WSS !!!

➔ But... If $a = \pm 1$ then $a^{2k+m} = a^{2k} a^m = a^m$

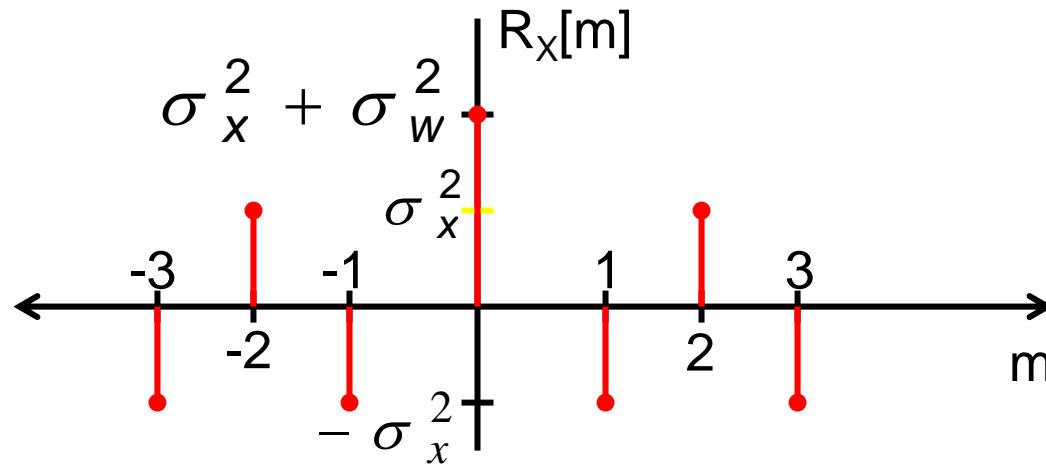
Then it is WSS

Ex. #3: Exp. Sig. + WN

If $a = +1$



If $a = -1$



Ex. #4: Weird Func of WN

Let $z[k]$ be white noise with zero mean
& variance of

$$\text{int}(D_3 D_2 D_1 \cdot d_1 d_2 d_3 d_4) = D_3 D_2 D_1$$

Define $x[n] = Z[\text{int}(n/2)]$

Then

$x[0]$	$= z[0]$	}	Always appear in pairs
$x[1]$	$= z[0]$		
$x[2]$	$= z[1]$	}	
$x[3]$	$= z[1]$		
\vdots	\vdots		

Ex. #4: Weird Func of WN

TASK: Is this RP WSS ?

MEAN: $E \{ x[n] \} = E \{ z[k] \} = 0$

$k = n/2$ if n is even
 $= (n-1)/2$ if n is odd

$R_x(n, n+m)$ = $E \{ x[n] x[n+m] \}$

If $|m| \geq 2$ then $x[n]$ & $x[n+m]$ are two different $z[.]$ values \Rightarrow uncorrelated

$\Rightarrow R_x(n, n+m) = 0$ for $|m| \geq 2$

Ex. #4: Weird Func of WN

If $m = 0$ then $R_X(n, n) = E \{ z^2[k] \} = \sigma_z^2$

If $m = 1$

... and n is even: $R_X(n, n+1) = \sigma_z^2$

... and n is odd: $R_X(n, n+1) = 0$

n dependence causes it not to be WSS

