

$$\begin{aligned}
 \textcircled{2} \text{ a) } V(f_n) &= \sum_{k=0}^{N-1} v[k] \exp(-j 2\pi f_n k) \\
 &= \sum_{k=0}^{N-1} v[k] \cos(2\pi f_n k) - j \sum_{k=0}^{N-1} v[k] \sin(2\pi f_n k)
 \end{aligned}$$

$$\text{hence, } a_n = \sum_{k=0}^{N-1} v[k] \cos(2\pi f_n k)$$

$$\text{and } b_n = - \sum_{k=0}^{N-1} v[k] \sin(2\pi f_n k)$$

Notice that,  $a_n$  and  $b_n$  are sum of i.i.d Gaussian variables ( $\because v$  is  $\text{GWN}$ ) which are scaled by a constant factor ( $\cos(2\pi f_n k)$  for  $k$ th term of  $a_n$  and  $\sin(2\pi f_n k)$  for  $k$ th term of  $b_n$ )  
 Such a product remains Gaussian with a change in  $\sigma^2$ .  
 So  $a_n$  and  $b_n$  are just sum of independent

Gaussian Random Variables

$\Rightarrow$   $a_n$  and  $b_n$  themselves should be Gaussian distributed.

To prove that sum of Gaussian random variables is Gaussian, I will use the moment generating fn

i) For Gaussian  $X$ ,  $\text{MGF} = E(e^{sX})$

$$\text{Let } X \sim N(0,1) \Rightarrow \text{MGF}(X) = E(e^{sX}) = E(e^{sX + 1/2 X^2 - 1/2 X^2})$$

$$= \int \frac{1}{\sqrt{2\pi}} e^{-y^2/2} e^{s y} e^{-1/2 y^2} dy$$

$$= \frac{e^{\mu s}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \frac{e^{s\sigma y - \frac{1}{2}\sigma^2 y^2}}{\sqrt{2\pi}} dy$$

$$= e^{\mu s} \frac{e^{\frac{\sigma^2 s^2}{2}}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \frac{e^{-\frac{1}{2}(y^2 - 2\sigma s + \sigma^2 s^2)}}{\sqrt{2\pi}} dy$$

$$\Rightarrow M_X(s) = e^{\mu s + \frac{\sigma^2 s^2}{2}}$$

ii) Consider sum of i.i.d. RVs  $Y = \sum_{i=1}^N X_i$

$$M_Y(s) = E(e^{sY}) = E(e^{s\sum_{i=1}^N X_i}) = E\left(\prod_{i=1}^N e^{sX_i}\right)$$

$$\stackrel{\text{Since } \dots \text{ i.i.d.}}{\Rightarrow} M_Y(s) = \prod_{i=1}^N E(e^{sX_i})$$

$$\Rightarrow M_Y(s) = \prod_{i=1}^N E(e^{sX_i})$$

$$\text{From (i), } M_Y(s) = \prod_{i=1}^N e^{\mu_i s + \frac{\sigma_i^2 s^2}{2}}$$

$$= \exp\left(\left(\sum_{i=1}^N \mu_i\right)s + \frac{\sum_{i=1}^N \sigma_i^2}{2} s^2\right)$$

So  $Y$  is Gaussian distributed with

$$\mu = \sum_{i=1}^N \mu_i \quad \text{and} \quad \sigma^2 = \sum_{i=1}^N \frac{\sigma_i^2}{2}$$

$\therefore$  Sum of i.i.d. Gaussian RVs is a Gaussian RV.

$$b) \text{corr}(a_n, b_n) = \frac{\text{cov}(a_n, b_n)}{\sqrt{\sigma_{a_n}^2 \sigma_{b_n}^2}}$$

$$\Rightarrow \text{corr}(a_n, b_n) = \frac{E(a_n b_n) - E(a_n) E(b_n)}{\sqrt{\sigma_{a_n}^2 \sigma_{b_n}^2}} \quad \text{--- (1)}$$

Let's evaluate these moments one by one,

$$a_n = \sum_{k=0}^{N-1} v(k) \cos 2\pi f n k \quad \text{--- (2)}$$

$$E(a_n) = \sum E(v(k) \cos 2\pi f n k) = 0 \quad \text{--- (3)}$$

$$\text{var}(a_n) = \text{var}\left(\sum v(k) \cos 2\pi f n k\right)$$

$$= \sum \left(\text{var}(v(k)) (\cos(2\pi f n k))^2\right) \quad \text{(because } v_k \text{ is i.i.d.)}$$

$$\Rightarrow \text{var}(a_n) = \sum_{k=0}^{N-1} \cos^2(2\pi f n k) \quad \text{--- (4)} \quad (\sigma_v^2 = 1)$$

$$\text{Similarly, } b_n = -\sum v(k) \sin 2\pi f n k \quad \text{--- (5)}$$

$$\Rightarrow E(b_n) = 0 \quad \& \quad \text{var}(b_n) = \sum \sin^2 2\pi f n k \quad \text{--- (6)}$$

$$E(a_n b_n) = E\left(\left(\sum v(k) \cos(2\pi f n k)\right) \left(\sum v(k) \sin(2\pi f n k)\right)\right)$$

(from (2) & (5))

$$\text{Let } 2\pi f n = \omega_n$$

$$\Rightarrow E(a_n b_n) = E\left(\left(\sum v(k) \cos \omega_n k\right) \left(\sum v(k) \sin(\omega_n k)\right)\right)$$

Multiplying the summations,

$$= E\left(\left(\sum v^2(k) \cos \omega_n k \sin \omega_n k\right) + \sum_{k=0}^{N-1} \sum_{\substack{l=0 \\ l \neq k}}^{N-1} v(k) v(l) \cos \omega_n k \sin \omega_n l\right)$$

$$= E \left( \sum v^2(k) \cos \omega n k \sin \omega n k + \sum_{k=0}^{N-1} \sum_{\substack{l=0 \\ l \neq k}}^{N-1} v(k)v(l) \cos \omega n k \sin \omega n l \right)$$

$\because v[k]$  is white noise (GWN)  $\sim N(0,1)$

$$E(v^2(k)) = 1 \quad \text{or} \quad E(v(k)v(l)) = \sigma_{vv}[k-l] \\ = 0 \quad \text{if } l \neq k$$

$$\Rightarrow E(a_n b_n) = \sum E(v^2(k)) \cos \omega n k \sin \omega n k \\ + \sum_{k=0}^{N-1} \sum_{\substack{l=0 \\ l \neq k}}^{N-1} E(v(k)v(l)) \cos \omega n k \sin \omega n l \\ = \sum \cos \omega n k \sin \omega n k \quad \text{--- (8)}$$

From eqns ①, ③, ④, ⑥, ⑦, ⑧ we have

$$\text{corr}(a_n, b_n) = \frac{\sum_{k=0}^{N-1} \cos(2\pi f_n k) \sin(2\pi f_n k)}{\sqrt{\left( \sum_{k=0}^{N-1} \cos^2(2\pi f_n k) \right) \left( \sum_{k=0}^{N-1} \sin^2(2\pi f_n k) \right)}}$$

$$c) \quad \rho_n = \frac{2 P_{vv}(f_n)}{\gamma(f_n)} = \frac{2(a_n^2 + b_n^2)}{N \gamma(f_n)}$$

$$\gamma(f_n) = \sum_{l=-\infty}^{\infty} \sigma_{vv}[l] \exp(-j 2\pi f_n l)$$

$$\therefore \sigma_{vv}[l] = \begin{cases} 1 & l=0 \\ 0 & l \neq 0 \end{cases}$$

$$\Rightarrow \gamma(f_n) = 1$$



$$\therefore C_n = \frac{2(a_n^2 + b_n^2)}{N}$$

Substitute  $a_n$  &  $b_n$  (eqns 2 & 3)

$$= \frac{2}{N} \left[ \sum v^2(k) \cos^2(2\pi f_n k) + \sum \sum v(k)v(l) \cos(2\pi f_n k) \cos(2\pi f_n l) \right. \\ \left. + \sum v^2(k) \sin^2(2\pi f_n k) + \sum \sum v(k)v(l) \sin(2\pi f_n k) \sin(2\pi f_n l) \right]$$

group the  $\sin^2$  &  $\cos^2$  terms ( $\cos^2 \theta + \sin^2 \theta = 1$ )

$$\Rightarrow C_n = \frac{2}{N} \left[ \sum v^2(k) + \sum \sum v(k)v(l) [\cos(2\pi f_n k) \cos(2\pi f_n l) + \sin(2\pi f_n k) \sin(2\pi f_n l)] \right]$$

$$\Rightarrow C_n = \frac{2}{N} \left[ \sum_{k=0}^{N-1} v^2(k) + \sum_{k=0}^{N-1} \sum_{\substack{l=0 \\ l \neq k}}^{N-1} v(k)v(l) \cos(2\pi f_n (k-l)) \right]$$

Sum of the square of  $N$  independent Gaussian RV form the  $\chi^2$  distribution.

Also product of 2 independent Gaussian RVs

$XY$  can be represented as  $\frac{1}{4} ((x+y)^2 - (x-y)^2)$  sum

of 2  $\chi^2$  RVs

$\Rightarrow C_n$  is a weighted sum of  $\chi^2$  distributed R.V.s

$\Rightarrow C_n$  is a  $\chi^2$  distributed random variable

In limit  $N \rightarrow \infty$  when cross terms go to zero

$$C_n \xrightarrow{d} \chi^2(2)$$

i) Mean of  $e_n$

$$E(e_n) = \frac{2}{N} \left[ \sum E(v^2) + 2 \sum_{k=0}^{N-1} \sum_{\substack{l=0 \\ l \neq k}}^{N-1} E(v(k)v(l)) \cos(\omega_n(k-l)) \right]$$

$$E(v^2) = \sigma^2 = 1$$

$$E(v(k)v(l)) = \sigma_{vv}[l-k] = \begin{cases} \sigma^2 & l=k \\ 0 & l \neq k \end{cases}$$

$$= \frac{2}{N} \times N \Rightarrow \boxed{E(e_n) = 2}$$

ii) Variance of  $e_n$ .

$$\text{var}(e_n) = \frac{4}{N^2} \text{var} \left( \sum v_k^2 + \sum_{k=0}^{N-1} \sum_{\substack{l=0 \\ l \neq k}}^{N-1} v(k)v(l) \cos(\omega_n(k-l)) \right)$$

$$\text{var}(X+Y) = \text{var}(X) + 2\text{cov}(X,Y) + \text{var}(Y)$$

here first term is  $X$  and second term is  $Y$ .

$$\circ \text{var}(X) = \text{var} \left( \sum v_k^2 \right) ; v(k) \text{ is i.i.d}$$

$$\Rightarrow \text{var}(X) = \sum \sigma^4 = N \sigma^4 = N \times 1$$

$$\Rightarrow \text{var}(X) = N$$

$$(\because \text{var}(AB) = \text{var}(A) \text{var}(B) + \text{var}(A)(E(B))^2 + \text{var}(B)(E(A))^2$$

$$\Rightarrow \text{var}(v^2) = (1)^2 = \sigma^4$$

$$\circ \text{cov}(X,Y) = E(XY) - E(X)E(Y);$$

$$E(Y) = 0 \text{ because } E(v(k)v(l)) = 0 \text{ for } k \neq l.$$

$$= E \left( \sum (v(k))^2 \sum \sum v(k)v(l) \cos \omega_n(k-l) \right)$$

Notice that any term will be of the form  $(v(k))^2 v(l)$   
or  $v(j)v(k)v(l)$

$$E(v(i)v(k)v(l)) = E(v(i))E(v(k))E(v(l)) = 0$$

$$E((v(k))^2 v(l)) = E((v(k))^2) E(v(l)) = 0$$

(This is because all  $v_k$  are 0 mean &  
 $v(k), v(l)$  are uncorrelated for all  $k \neq l$ )

$$\Rightarrow \text{cov}(X, Y) = 0$$

$$\bullet \text{var}(Y) = \text{var}\left(\sum \sum v(k) v(l) \cos(\omega_n(k-l))\right)$$

We know that  $E(Y) = 0$

$$\Rightarrow \text{var}(Y) = E\left(\left(\sum \sum v(k) v(l) \cos(\omega_n(k-l))\right)^2\right)$$

$$= E\left(4 \sum_{k=0}^{N-1} \sum_{\substack{l=0 \\ k \neq l}}^{k-1} (v(k) v(l))^2 \cos^2(\omega_n(k-l)) + 8 \sum_{\substack{(i,j) \neq (k,l) \\ (i \neq j), (k \neq l)}} v(i) v(j) v(k) v(l) \cos(\omega_n(i-j)) \cos(\omega_n(k-l))\right)$$

The second term will never have all variables

having even powers, i.e.  $(v(k) v(l))^2$

So because of the i.i.d &  $E(v) = 0$ , we can say

that  $E(\text{second term})$  goes to zero

$$\Rightarrow \text{var}(Y) = 4 E\left(\sum_{k=0}^{N-1} \sum_{\substack{l=0 \\ k \neq l}}^{k-1} (v(k) v(l))^2 \cos^2(\omega_n(k-l))\right)$$

Once again we invoke  $\text{var}(A+B) = \text{var}(A)\text{var}(B) + \text{var}(A)(E(B))^2 + \text{var}(B)(E(A))^2$

$$\Rightarrow \text{var}(v(k)v(l)) = \text{var}(v(k))\text{var}(v(l)) = \sigma_v^4 = 1$$

Since  $E(v(k)v(l)) = 0$ , ( $\forall k \neq l$ ),

$$E((v(k)v(l))^2) = \text{var}(v(k)v(l))$$

$$\begin{aligned} \therefore \text{var}(Y) &= 4 \sum_{k=0}^{N-1} \sum_{l=0}^{k-1} E(v(k)v(l)) \cos^2(\omega_n(k-l)) \\ &= 4 \sum_{k=0}^{N-1} \sum_{\substack{l=0 \\ k \neq l}}^{k-1} \cos^2(\omega_n(k-l)) \\ &= 4 \sum_{k=0}^{N-1} \sum_{\substack{l=0 \\ k \neq l}}^{N-1} \frac{1 + \cos(2\omega_n(k-l))}{2} \end{aligned}$$

$$\begin{aligned} \sum_{k \neq l} \cos(2\omega_n(k-l)) &= \sum_{k \neq l} \cos(4\pi f_n(k-l)) \\ &= \sum_{k \neq l} \cos(4\pi \frac{n}{N}(k-l)) \end{aligned}$$

Using the fact that  $\sum_{k=0}^{N-1} \cos(2\pi \frac{n}{N} k) = 0$ ,

ie that  $\cos 0 = 1$  will be missing from the sum ( $\because k \neq l$ )

we get

$$\sum_{k=0}^{N-1} \sum_{\substack{l=0 \\ k \neq l}}^{k-1} \cos(4\pi \frac{n}{N}(k-l)) = \frac{-N}{2} \quad \text{(I also verified this in MATLAB)}$$

$$\Rightarrow \text{var}(Y) = 2 \left[ N \times \frac{(N-1)}{2} - \frac{N}{2} \right] = N(N-2)$$



$$\circ \circ \text{var}(C_n) = \frac{4}{N} + \frac{4}{N^2} \times N(N-2)$$

$$= \frac{4}{N} + \left( \frac{1}{N} + \frac{N-2}{N} \right)$$

$$\boxed{\text{var}(C_n) = 4 \frac{(N-1)}{N}}$$

As expected  $\lim_{N \rightarrow \infty} \text{var}(C_n) = 2 \times 2 = \lim_{N \rightarrow \infty} \left( \frac{1 - \frac{1}{N}}{1} \right) 4$   
 $= \boxed{4} \rightarrow \text{var}(X^2(2))$

d) ~~E~~ Mean squared convergence to be checked

$$\Rightarrow \lim_{N \rightarrow \infty} E \left( (P(f_n) - r_{vv}(f_n))^2 \right)$$

\* We already obtained that  $r_{vv}(f_n) = 1$ ,

$$\text{also } E(C_n) = 2 \Rightarrow E \left( \frac{2P(f_n)}{r(f_n)} \right) = 2$$

$$\Rightarrow E(P(f_n)) = r(f_n) = 1$$

So the RV  $E(P(f_n) - r_{vv}(f_n))$  has zero mean

$$\therefore \lim_{N \rightarrow \infty} E \left( (P(f_n) - r(f_n))^2 \right)$$

$$= E \text{var}((P(f_n) - 1)) + \cancel{E(P(f_n) - r(f_n))^2}$$

$$= \lim_{N \rightarrow \infty} \text{var}(P(f_n) - 1) = \lim_{N \rightarrow \infty} \text{var}(P(f_n))$$

from part (c)

$$\text{var}(\epsilon_n) = \frac{4(N-1)}{N}$$

$$\Rightarrow \text{var} \left( \frac{2P(f_n)}{Y(f_n)} \right) = \frac{4(N-1)}{N}$$

$$\Rightarrow \text{var}(P(f_n)) = \frac{(N-1)}{N}$$

$$\Rightarrow \lim_{N \rightarrow \infty} \text{var}(P(f_n)) = \frac{N-1}{N} \rightarrow 1$$

$$\therefore \lim_{N \rightarrow \infty} E \left( (P(f_n) - Y_{rv}(f_n))^2 \right) = 1 \neq 0$$

$\Rightarrow P(f_n)$  does not exhibit mean squared convergence to  $Y(f_n)$

however it is an unbiased estimator.