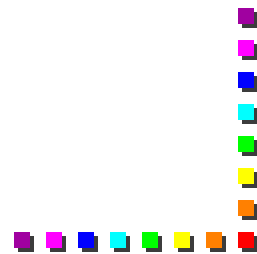


# Noise

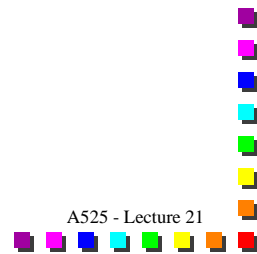
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# Outline

- Spectral density
- Correlation function
- Weiner-Khintchine Theorem
- Campbell's Theorem
- Carson's Theorem
  - Spectral Density of Photon Noise
  - Systems with time constants
- Non-ideal Photon Detection
  - Johnson Noise



## Noise: What is it?

- The rms deviation from the time averaged signal is a measure of the “noise” in the detection system.

$$\Delta v_{\text{rms}} = \sqrt{\overline{(v(t) - \bar{v})^2}}$$

time-averaged signal

signal

where

$$\overline{(\Delta v)^2} = (\Delta v_{\text{rms}})^2 = \text{variance}$$

- This is not the whole story. It only gives us limited information on the signal properties.

Discussion follows Boyd

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## Noise: What is it?

$v_1(t)$

$t$

$v_2(t)$

$t$

- Consider two signals that have the same average and variance, as shown in the plots above. We have

$$\bar{v}_1 = \bar{v}_2 \quad \& \quad (\Delta v_1)_{\text{rms}} = (\Delta v_2)_{\text{rms}}$$

- But, the character (time scales) of the fluctuations are different because of the different frequency components in the noise.
- How do we quantitatively make a distinction?

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## Fourier Transform

- We can look at the harmonic content of the signal via the Fourier Transform:

$$V(f) = \int_{-\infty}^{\infty} v(t) e^{-i2\pi ft} dt$$

The inverse transform gives back the signal

$$v(t) = \int_{-\infty}^{\infty} V(f) e^{i2\pi ft} dt$$

- However, if  $v(t)$  is a stationary, stochastic function of time for  $-\infty < t < \infty$ , then  $v(t)$  is not square-integrable and  $V(f)$  does not exist.

**stochastic:** non-deterministic behavior (current state does not fully determine the next)  
**stationary:** probability distribution is the same for all times (mean, variance, etc. are constant)

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## Spectral Density

- We can use a truncated signal  $v_T(t)$

$$v_T(t) = \begin{cases} v(t) & -T/2 < t < T/2 \\ 0 & \text{otherwise} \end{cases}$$

Then

$$V_T(f) = \int_{-T/2}^{T/2} v(t) e^{-i2\pi ft} dt$$

- The spectral density,  $S(f)$ , is defined by (here  $\langle \rangle$  indicates an ensemble average)

$$S(f) = \lim_{T \rightarrow \infty} \left\langle \frac{1}{T} |V_T(f)|^2 \right\rangle$$

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## Spectral Density

- For most signals, we avoid performing the ensemble average and calculate  $S(f)$  over some (perhaps very small) frequency interval

$$S(f) = \lim_{T \rightarrow \infty} \frac{1}{T} |V_T(f)|^2$$

This is the definition Boyd adopts for the spectral density (ergodicity)

- $S(f)$  = power per unit frequency interval that a voltage signal  $V(t)$  could deliver to a  $1 \Omega$  resistor.

**ergodic hypothesis:** assume that the average of a process parameter over time and the average over the statistical ensemble are the same.

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## Variance from the Spectral Density

- Now

$$\overline{v^2} = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} v^2(t) dt = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-\infty}^{\infty} v_T^2(t) dt$$

- By Parseval's theorem for the transform pair,  $x$  and  $X$

$$\int_{-\infty}^{\infty} |x(t)|^2 dt = \int_{-\infty}^{\infty} |X(f)|^2 df$$

Then using the fact that since  $v_T(t)$  is real, the average power can be expressed as

$$\begin{aligned} \overline{v^2} &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-\infty}^{\infty} |V_T(f)|^2 df \\ &= \int_{-\infty}^{\infty} S(f) df \end{aligned}$$

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## The Correlation Function

- The limiting process required to compute  $S(f)$

$$S(f) = \lim_{T \rightarrow \infty} \frac{1}{T} |V_T(f)|^2$$

is awkward to use in practice. As it turns out the correlation function can be related to the spectral density

- In general the correlation function looks like

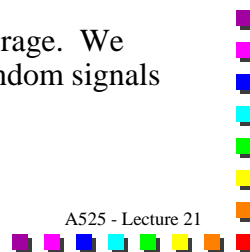
$$C(t_1, t_2) = \langle v(t_1)v(t_2) \rangle$$

- Where the brackets indicate an ensemble average. We usually deal with signals that are ergodic (random signals with time average = an ensemble average)

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## Stationary Signals

- For stationary signals  $C(t_1, t_2)$  must only depend on the difference  $\tau = t_1 - t_2$ . Thus

$$C(\tau) = \lim_{T \rightarrow \infty} C_T(\tau)$$

where

$$C_T(\tau) = \frac{1}{T} \int_{-T/2}^{T/2} v(t)v(t+\tau)dt$$

- The correlation function is usually represented as

$$C(\tau) = \overline{v(t)v(t+\tau)} \quad \leftarrow \text{time average}$$

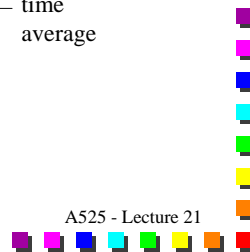
At  $\tau = 0$ ,

$$C(0) = \overline{v^2}$$

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## Properties of Correlation Function

- The correlation function is an even function of  $\tau$

$$C(-\tau) = \overline{v(t)v(t-\tau)} = \overline{v(t+\tau)v(t)} = C(\tau)$$

We also have

$$[v(t) \pm v(t+\tau)]^2 \geq 0$$

$$\Rightarrow v^2(t) + v^2(t+\tau) \geq \pm 2v(t)v(t+\tau)$$

- and taking the time average of both sides give

$$\overline{v^2} = \pm \overline{v(t)v(t+\tau)}$$

$$\Rightarrow C(0) \geq |C(\tau)|$$

Perfect correlation  
with zero delay –  
can't beat that.

## Wiener-Khinchine Theorem

- The W-K theorem states that  $S(f)$  and  $C(\tau)$  are a Fourier Transform pair. Consider

$$\begin{aligned} \int_{-\infty}^{\infty} C(\tau) e^{-i2\pi f\tau} d\tau &= \frac{1}{T} \int_{-\infty}^{\infty} d\tau \int_{-\infty}^{\infty} dt v_T(t) v_T(t+\tau) e^{-i2\pi f\tau} \\ &= \frac{1}{T} \int_{-\infty}^{\infty} dt v_T(t) e^{i2\pi ft} \int_{-\infty}^{\infty} d\tau v_T(t+\tau) e^{-i2\pi f(t+\tau)} \\ &= \frac{1}{T} V_T(-f) V_T(f) \end{aligned}$$

$$\Rightarrow \int_{-\infty}^{\infty} C(\tau) e^{-i2\pi f\tau} d\tau = \frac{1}{T} |V_T(f)|^2$$

- This is a special case of the convolution theorem, which states that the Fourier transform of the convolution of two functions equals the product of the Fourier transforms.

## Wiener-Khintchine Relations

- Letting  $T \rightarrow \infty \Rightarrow S(f) = \int_{-\infty}^{\infty} C(\tau) e^{-i2\pi f\tau} d\tau$

- Taking the inverse transform gives

$$C(\tau) = \int_{-\infty}^{\infty} S(f) e^{i2\pi f\tau} df$$

- This and the previous equation are the Wiener-Khintchine relations. If  $C(\tau)$  is an even function of  $\tau$  then

$$S(f) = 2 \int_0^{\infty} C(\tau) \cos 2\pi f\tau d\tau$$

- Which means  $S(f)$  is an even function of  $f$ , hence

$$C(\tau) = 2 \int_0^{\infty} S(f) \cos 2\pi f\tau df$$

- These two equations are another form of the W-K relations

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## Wiener-Khintchine Relations

- In practice we have  $0 < f < \infty$ , rather than  $-\infty < f < \infty$ . The spectral density of  $v(t)$  is then often expressed as

$$\overline{v^2}(f) = 2S(f)$$

- We then have for the W-K relations

$$\overline{v^2}(f) = 4 \int_0^{\infty} C(\tau) \cos 2\pi f\tau d\tau$$

and

$$C(\tau) = \int_0^{\infty} \overline{v^2}(f) \cos 2\pi f\tau df$$

- And the mean square of  $v$  becomes

$$\overline{v^2} = \int_{-\infty}^{\infty} S(f) df = 2 \int_0^{\infty} S(f) df = \int_0^{\infty} \overline{v^2}(f) df$$

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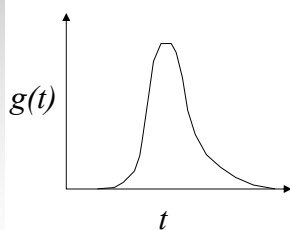
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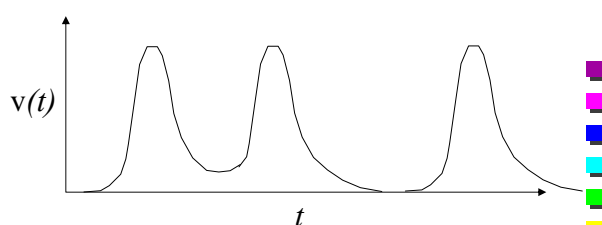
## Example: Identical Events

- Consider identical events occurring at random times (see Boyd, pg. 125-127)

$$v_T = \begin{cases} \sum_{i=1}^N g(t-t_i) & -\frac{T}{2} \leq t \leq \frac{T}{2} \\ 0 & \text{otherwise} \end{cases}$$



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## Average Signal and Event Rate

- The time average of \$v(t)\$ is then

$$\bar{v} = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} \sum_{i=1}^N g(t-t_i) dt$$

Which can be expressed as

$$\bar{v} = r \int_{-\infty}^{\infty} g(t) dt \quad \text{where} \quad r = \left\langle \frac{N}{T} \right\rangle = \text{average rate of events}$$

$$\text{or} \quad r = \lim_{T \rightarrow \infty} \frac{N}{T} \quad (\text{ergodic})$$

- The average signal is just the event rate times the integrated amount in one event

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## Statistical Properties

- To look at the statistical properties of these events we look at the correlation function

$$\begin{aligned}
 C_T(\tau) &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} v(t)v(t+\tau)dt \\
 &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-\infty}^{\infty} \sum_{i=1}^N g(t-t_i) \sum_{i=1}^N g(t+\tau-t_i) dt \\
 &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-\infty}^{\infty} dt \left\{ \sum_{i=1}^N g(t-t_i)g(t+\tau-t_i) + \sum_{i=1}^N \sum_{\substack{j=1 \\ i \neq j}}^N g(t-t_i)g(t+\tau-t_j) \right\}
 \end{aligned}$$

- The first term in  $\{ \}$  is just the correlation function of  $g(t)$  with itself. Since no correlation exists between individual pulses we have

$$C_v(\tau) = r \int_{-\infty}^{\infty} g(t)g(t+\tau)dt + \bar{v}^2$$

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## Campbell's Theorem

- Using the variance in the signal as a measure of the noise

$$\overline{(\Delta v)^2} = \overline{[v(t) - \bar{v}]^2} = \overline{v^2(t)} - \bar{v}^2$$

$$\text{Since } \overline{v^2(t)} = C_v(0) \quad \overline{(\Delta v)^2} = r \int_{-\infty}^{\infty} g^2(t)dt$$

- This expression and the earlier one for the average signal are known as **Campbell's theorem**, that is,

$$\bar{v} = r \int_{-\infty}^{\infty} g(t)dt \quad \& \quad \overline{(\Delta v)^2} = r \int_{-\infty}^{\infty} g^2(t)dt$$

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## Carson's Theorem

- The spectral density can be computed from the correlation function  $S(f) = r \int_{-\infty}^{\infty} \left( \int_{-\infty}^{\infty} g(t)g(t+\tau)dt + \bar{v}^2 \right) e^{-i2\pi f\tau} d\tau$

Proceeding as we did with the W-K theorem and using the fact that

$$\delta(f) = \int_{-\infty}^{\infty} e^{-i2\pi f\tau} d\tau \quad \text{where} \quad \int_{-\infty}^{\infty} \delta(f) e^{i2\pi f\tau} df = 1$$

gives

$$S(f) = r |G(f)|^2 + \bar{v}^2 \delta(f) \quad \text{where} \quad G(f) = \int_{-\infty}^{\infty} g(t) e^{-i2\pi ft} dt$$

- This is called **Carson's theorem**.
- The  $\delta(f)$  term is the average power in the dc component of  $G(t)$ , i.e.  $\delta(f) \neq 0$  only for  $f = 0$ .
- The  $G(f)$  term represents the noise in  $v(t)$  which is related to the spectral content of each random event.

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## Spectral Density of Photon Noise

- Recalling earlier we wrote  $\bar{v}^2(f) = 2S(f)$
- Or we could write for the noise current due to photons

$$\overline{i_N^2}(f) = 2S(f)$$

- And using our derived expression for the spectral density for random uncorrelated events we have

$$\overline{i_N^2}(f) = 2r |G(f)|^2$$

- Where the dc term [  $\delta(f)$  ] has been dropped since this is just the average current.

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## Spectral Density of Photon Noise

- If the pulses are unresolved

$$g(t) = e\delta(t) \Rightarrow G(f) = \int_{-\infty}^{\infty} g(t)e^{-i2\pi ft} dt = e$$

- And the current noise spectral density is

$$\overline{i_N^2}(f) = 2re^2 = 2e\bar{i}$$

- The total noise power is obtained by multiplying by the electrical bandwidth,  $\Delta f$

$$(\overline{\Delta i})^2 = \overline{i_N^2}(f)\Delta f = 2e\bar{i}\Delta f$$

- Note that this is “white noise”

## Real System: Time constant

- In a real system we might expect a time constant, e.g. an exponential decay when a low pass filter is used

$$g(t) = \begin{cases} 0 & t < 0 \\ \frac{e}{\tau} e^{-t/\tau} & t \geq 0 \end{cases} \Rightarrow G(f) = \frac{e}{1 + i2\pi f\tau}$$

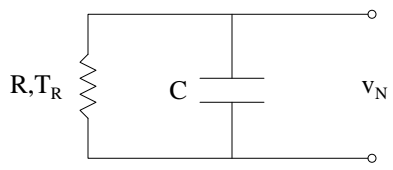
- The spectral density of the current noise is then

$$\overline{i_N^2}(f) = \frac{2e\bar{i}}{1 + (2\pi f\tau)^2}$$

- For  $f \ll (2\pi\tau)^{-1}$  this approximates white noise again

## Johnson Noise

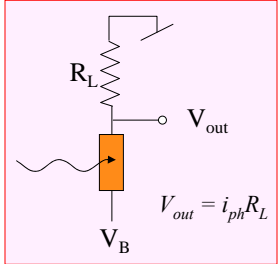
- Thermal agitation of electrons in a resistor cause noise



$$\text{stored energy} = \frac{1}{2} C \overline{v_N^2}$$

- Equipartition  $\Rightarrow$  average energy stored in our one dimensional classical system is equal to  $kT_R/2$

$$\Rightarrow \overline{v_N^2} = \frac{kT_R}{C}$$



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## Correlation Function/Spectral Density

- The correlation function of the voltage fluctuations gives the transient behavior of the circuit

$$C(\tau) = \overline{v(t)v(t+\tau)} = \overline{v(t)v(t)} e^{-|\tau|/RC} = \overline{v_N^2} e^{-|\tau|/RC}$$

- The spectral density of the voltage noise is then (slide 14)

$$\overline{v_N^2}(f) = 4 \int_0^\infty C(\tau) \cos 2\pi f \tau d\tau$$

- So that

$$\overline{v_N^2}(f) = \frac{4 \overline{v_N^2} RC}{1 + (2\pi f RC)^2}$$

$$= 4 \overline{v_N^2} RC \quad \text{For } f \ll 1/RC$$

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## Johnson Noise: Result

- From the equipartition result

$$\overline{v_N^2}(f) = 4kT_R R$$

Johnson  
Noise

- Note: there is no  $C$  in the final result. Thus we can take  $C = 0$ , i.e. valid for all frequencies. This is another example of white noise. (This is only true for  $hf \ll kT$ )
- For **Johnson noise limited detection**

$$NEP = \frac{h\nu}{\eta e} \sqrt{\frac{4kT_R}{R}}$$

←

$$v_{\text{signal}} = i_{ph} R = \frac{\eta e P_s}{h\nu} R$$

$$\frac{S}{N} = \frac{v_{\text{signal}}}{(\overline{v_N^2})^{1/2}} = \frac{\eta e P_s}{h\nu} R \frac{1}{\sqrt{4kT_R R}}$$

$$NEP = P_s \text{ for } S/N = 1$$

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