

CS7GV2: Mathematics of Light and Sound, M.Sc. in Computer Science.

Lecture#8: Intensity sums

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Descriptive statistics of a sum of two terms

The means of random intensities I_X and I_Y are,

$$\overline{I_X} = \mathbf{E}[I_X] \quad \text{and} \quad \overline{I_Y} = \mathbf{E}[I_Y].$$

Linearity of expectation allows the mean of their sum to be expressed as,

$$\mathbf{E}[I_X + I_Y] = \mathbf{E}[I_X] + \mathbf{E}[I_Y]$$

$$\overline{I_Z} = \overline{I_X + I_Y} = \overline{I_X} + \overline{I_Y}.$$

When I_X and I_Y are independent,

$$\mathbf{E}[I_X I_Y] = \mathbf{E}[I_X] \mathbf{E}[I_Y]$$

$$\overline{I_X I_Y} = \overline{I_X} \overline{I_Y}.$$

Hence,

$$\begin{aligned} \overline{I_Z^2} &= \overline{(I_X + I_Y)^2} \\ &= \overline{I_X^2 + 2 I_X I_Y + I_Y^2} \\ &= \overline{I_X^2} + 2 \overline{I_X} \overline{I_Y} + \overline{I_Y^2}, \end{aligned}$$

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For an exponential PDF,

$$\overline{I_X^2} = 2 \overline{I_X}^2 \quad \text{and} \quad \overline{I_Y^2} = 2 \overline{I_Y}^2.$$

So the variance can be expressed as,

$$\begin{aligned} \sigma_Z^2 &= \mathbf{E}[I_Z^2] - \mathbf{E}[I_Z]^2 = \overline{I_Z^2} - \overline{I_Z}^2 \\ &= 2 \overline{I_X}^2 + 2 \overline{I_X} \overline{I_Y} + 2 \overline{I_Y}^2 \\ &\quad - \overline{I_X}^2 - 2 \overline{I_X} \overline{I_Y} - \overline{I_Y}^2 \\ &= \overline{I_X}^2 + \overline{I_Y}^2. \end{aligned}$$

$$C = \frac{\sigma_Z}{\overline{I_Z}} = \frac{\sqrt{\overline{I_X}^2 + \overline{I_Y}^2}}{\overline{I_X} + \overline{I_Y}} = \frac{\sqrt{1 + r^2}}{1 + r}$$

where $r = \overline{I_Y} / \overline{I_X}$ is the ratio of one mean to the other.

Hence when $r = 1$ the contrast is minimized,

$$C = \sqrt{2}/2 = (\sqrt{2}\sqrt{2})/(2\sqrt{2}) = 2/(2\sqrt{2}) = 1/\sqrt{2}.$$

Note, for the denominator,

$$\frac{\sqrt{a^2 + b^2}}{a} = \frac{\sqrt{a^2 + b^2}}{\sqrt{a^2}} = \sqrt{\frac{a^2}{a^2} + \frac{b^2}{a^2}} = \sqrt{1 + \left(\frac{b}{a}\right)^2}.$$

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Descriptive statistics of a sum of N terms

Note that $(a + b + c)^2$

$$\begin{aligned} &= (a + b + c)(a + b + c) \\ &= a^2 + b^2 + c^2 \\ &\quad + 2(ab + ac + bc) \\ &= \sum_{n=\{a,b,c\}} n^2 + \sum_{n=\{a,b,c\}} \sum_{\substack{m=\{a,b,c\}, \\ m \neq n}} nm. \end{aligned}$$

For a sum of random intensities,

$$\begin{aligned} I_Z &= I_1 + I_2 + \dots + I_N, \\ \text{the mean } \overline{I_Z} &= \sum_{n=1}^N \overline{I_n} \text{ by LOE.} \\ \overline{I_Z^2} &= \sum_{n=1}^N \sum_{m=1}^N \overline{I_n I_m} \\ &= \sum_{n=1}^N \overline{I_n^2} + \sum_{n=1}^N \sum_{\substack{m=1, \\ m \neq n}}^N \overline{I_n I_m} \end{aligned}$$

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Using $\overline{I_n^2} = 2\overline{I_n}^2$,

$$\begin{aligned} \overline{I_Z^2} &= 2 \sum_{n=1}^N \overline{I_n}^2 + \sum_{n=1}^N \sum_{\substack{m=1, \\ m \neq n}}^N \overline{I_n I_m} \\ &= \sum_{n=1}^N \overline{I_n}^2 + \left(\sum_{n=1}^N \overline{I_n} \right)^2 \\ &= \sum_{n=1}^N \overline{I_n}^2 + \overline{I_Z}^2. \end{aligned}$$

$$\sigma_Z^2 = \overline{I_Z^2} - \overline{I_Z}^2 = \sum_{n=1}^N \overline{I_n}^2$$

$$C = \frac{\sigma_Z}{\overline{I_Z}} = \frac{\sqrt{\sum_{n=1}^N \overline{I_n}^2}}{\sum_{n=1}^N \overline{I_n}}$$

When mean intensities $\overline{I_n}$ are equal,

$$C = \frac{\sqrt{N\overline{I}^2}}{N\overline{I}} = \frac{\sqrt{N}\sqrt{\overline{I}^2}}{N\overline{I}} = \frac{\sqrt{N}}{N} = \frac{1}{\sqrt{N}}.$$

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PDF of a sum of two terms

For random intensities I_X and I_Y described as exponential PDFs,

$$\begin{aligned} f_{I_X}(i_X) &= \frac{1}{I_X} \exp\left\{-\frac{i_X}{I_X}\right\} \\ f_{I_Y}(i_Y) &= \frac{1}{I_Y} \exp\left\{-\frac{i_Y}{I_Y}\right\}, \end{aligned}$$

their sum $I_Z = I_X + I_Y$ is described by another PDF, $f_{I_Z}(i_Z)$.

For *discrete* random quantities $X, Y, Z \in \mathbb{N}_0$, let $Z = X + Y$. A sum of 3 could result from $0 + 3$ or $1 + 2$ or $2 + 1$ or $3 + 0$ so,

$$\begin{aligned} \mathbf{P}(Z = 3) &= \mathbf{P}(X = 0) \mathbf{P}(Y = 3) \\ &\quad + \mathbf{P}(X = 1) \mathbf{P}(Y = 2) \\ &\quad + \mathbf{P}(X = 2) \mathbf{P}(Y = 1) \\ &\quad + \mathbf{P}(X = 3) \mathbf{P}(Y = 0) \end{aligned}$$

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$$= \sum_{x=0}^3 \mathbf{P}(X = x) \mathbf{P}(Y = 3 - x).$$

Letting $p_Z(z)$ etc. be discrete equivalents of a PDF to denote probabilities,

$$p_Z(z) = \sum_{x=0}^z p_X(x) p_Y(z - x).$$

Such a sum (or integral) of products of functions is called a *convolution*.

For *continuous* random quantities $X, Y, Z \in \mathbb{R}_{\geq 0}$, let $Z = X + Y$. The PDF for Z has an integral instead of a sum,

$$f_Z(z) = \int_0^z f_X(x) f_Y(z - x) dx.$$

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Let $f_X(x) = \lambda_X e^{-\lambda_X x}$ and $f_Y(y) = \lambda_Y e^{-\lambda_Y y}$ be exponential PDFs with rate parameters λ_X and λ_Y . The PDF of their sum is,

$$f_Z(z) = \int_0^z \lambda_X e^{-\lambda_X x} \lambda_Y e^{-\lambda_Y (z-x)} dx$$

$$= \lambda_X \lambda_Y e^{-\lambda_Y z} \int_0^z e^{(\lambda_Y - \lambda_X)x} dx.$$

By FTC the definite integral is the difference between the antiderivative evaluated at the limits,

$$\frac{1}{\lambda_Y - \lambda_X} e^{(\lambda_Y - \lambda_X)x} \Big|_0^z$$

$$= \frac{1}{\lambda_Y - \lambda_X} e^{(\lambda_Y - \lambda_X)z} - \frac{1}{\lambda_Y - \lambda_X}.$$

Or $\int_0^z 1 dx = z$ when $\lambda_X = \lambda_Y$.

$$f_Z(z) = \begin{cases} \frac{\lambda_X \lambda_Y}{\lambda_Y - \lambda_X} (e^{-\lambda_X z} - e^{-\lambda_Y z}) & \text{when } \lambda_Y \neq \lambda_X. \\ z \lambda^2 e^{-\lambda z} & \text{when } \lambda = \lambda_X = \lambda_Y. \end{cases}$$

$\lambda = 1/\bar{t}$ for $\bar{T} = \overline{I_X} = \overline{I_Y}$, so the PDF of the sum is a scaled* version of the originals,

$$f_{I_Z}(iz) = \frac{iz}{\bar{T}} \frac{1}{\bar{T}} \exp\left\{\frac{-iz}{\bar{T}}\right\}.$$

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PDF of a sum of N terms

A bit complicated.

If interested, see Goodman,

Speckle Phenomena in Optics,

page 42.

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