EEL 5544 Lecture 9

IMPORTANT RANDOM VARIABLES-CONT.

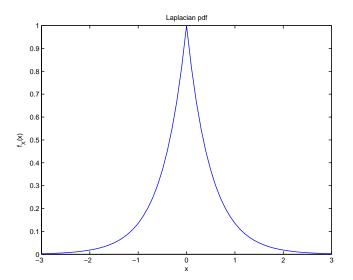
4. Laplacian RV

- The Laplacian random variable is often used to model the difference between correlated data sources
- For example, the differences between adjacent samples of speech, images, or video are often modeled using Laplacian random variables
- The density function for a Laplacian random variable is

$$f_X(x) = \frac{c}{2} \exp\{-c|x|\}, -\infty < x < \infty$$

where c > 0

• The pdf for a Laplacian RV with c = 2 is shown below



5. Chi-square RV

- Let X be a Gaussian random variable
- Then $Y = X^2$ is a chi-square random variable

• But in fact, chi-square random variables are much more general. For instance, if X_i , i = 1, 2, ..., n are s.i. *Gaussian* random variables with identical variances σ^2 , then

$$Y = \sum_{i=1}^{n} X_i^2 \tag{1}$$

is a chi-square random variable with n degrees of freedom

- The chi-square random variable is usually classified as either *central* or *non-central*
- If in (1), all of the Gaussian random variables have mean $\mu = 0$, then Y is a central chi-square random variable
- The density of a central chi-square random variable is given by

$$f_Y(y) = \begin{cases} \frac{1}{\sigma^n 2^{n/2} \Gamma(n/2)} y^{(n/2)-1} e^{-y/(2\sigma^2)}, & y \ge 0, \\ 0, & y < 0 \end{cases}$$

where $\Gamma(p)$ is the gamma function defined as

$$\begin{cases} \Gamma(p) = \int_0^\infty t^{p-1} e^{-t} dt, & p > 0\\ \Gamma(p) = (p-1)!, & p \text{ an integer } > 0\\ \Gamma(1/2) = \sqrt{\pi}\\ \Gamma(3/2) = \frac{1}{2}\sqrt{\pi} \end{cases}$$

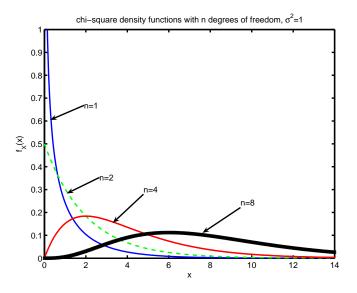
• Note that for n = 2,

$$f_Y(y) = \begin{cases} \frac{1}{2\sigma^2} e^{-y/(2\sigma^2)}, & y \ge 0, \\ 0, & y < 0 \end{cases}$$

which is an exponential density

- Thus, the sum of the squares of two s.i. zero-mean Gaussian RVs with equal variance is exponential!
- Note also that although the Gaussian random variables that make up the central chi-square RV have zero mean, the chi-square RV has mean > 0

• The pdf for chi-square RVs with 1,2, 4, or 8 degrees of freedom and $\sigma^2 = 1$ are shown below



• The CDF for a central chi-square RV can be found through repeated integration by parts to be

$$F_Y(y) = \begin{cases} 1 - e^{-y/(2\sigma^2)} \sum_{k=0}^{m-1} \frac{1}{k!} \left(\frac{y}{2\sigma^2}\right)^k, & y \ge 0\\ 0, & y < 0 \end{cases}$$

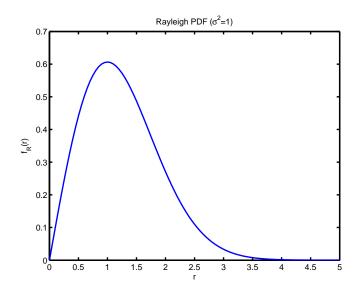
• The non-central chi-square random variable has more complicated pdfs and CDFs, and the CDF cannot be written in closed form

6. Rayleigh RV

- Consider an exponential RV Y, which is equivalent to a central chi-square RV with 2 degrees of freedom, which is equivalent to the sum of the squares of independent, zero-mean Gaussian RVs with common variance
- Let $R = \sqrt{Y}$
- Then R is a Rayleigh RV with pdf

$$f_R(r) = \begin{cases} \frac{r}{\sigma^2} e^{-r^2/(2\sigma^2)}, & r \ge 0\\ 0, & r < 0 \end{cases}$$

• The pdf for Rayleigh RV with $\sigma^2 = 1$ is shown below



• The CDF for a Rayleigh RV is

$$F_R(r) = \begin{cases} 1 - e^{-r^2/(2\sigma^2)}, & r \ge 0\\ 0, & r < 0 \end{cases}$$

• The amplitude of a radio signal in a dense multipath environment is often modeled as a complex Gaussian random variable with independent real and imaginary parts. Thus, the amplitude of the waveform is a Rayleigh random variable.

7. Lognormal RV

- Consider a random variable L such that $X = \ln L$ is a Gaussian RV
- For instance in communications, we find that shadowing of communication signals by buildings and foliage has a Gaussian distribution when the signal loss is expressed in decibels
- Then L is a lognormal random variable with pdf given by

$$f_L(\ell) = \begin{cases} \frac{1}{\ell\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{(\ln\ell-\mu)^2}{2\sigma^2}\right\}, & \ell \ge 0\\ 0, & \ell < 0 \end{cases}$$

CONDITIONAL DISTRIBUTIONS AND DENSITIES

• Given (S, \mathcal{A}, P) and $X : S \to \mathbf{R}$ (a RV on S):

DEFN For an event $A \in \mathcal{A}$ with $P(A) \neq 0$, the conditional distribution of X given A is

$$f_X(x|A) =$$

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$$f_X(x|A) =$$

• Then $F_X(x|A)$ is a distribution function, and $f_X(x|A)$ is a density function.

Example Suppose that the waiting time in minutes at Wendy's in the Reitz Union is an exponential random variable with parameter $\lambda = 1$. Given that you have already been waiting for two minutes, then what is

- (a) the conditional distribution for your total wait time?
- (b) the conditional distribution for your remaining wait time?

TOTAL PROBABILITY AND BAYE'S THEOREM

• If A_1, A_2, \ldots, A_n form a partition of S, then

$$F_X(x) = F_X(x|A_1)P(A_1) + F_X(x|A_2)P(A_2) + \dots + F_X(x|A_n)P(A_n)$$

and

$$f_X(x) = \sum_{i=1}^n f_X(x|A_i) P(A_i).$$

Point Conditioning:

- Suppose we want to evaluate the probability of an event given that X = x, where X is a continuous random variable.
- Clearly, P(X=x) = 0, so the previous definition of conditional prob. will not work.

$$P(A|X = x) = \lim_{\Delta x \to 0} P(A|x < X \le x + \Delta x)$$

$$= \lim_{\Delta x \to 0} \frac{F_X(x + \Delta x|A) - F_X(x|A)}{F_X(x + \Delta x) - F_X(x)} P(A)$$

$$= \lim_{\Delta x \to 0} \frac{\frac{F_X(x + \Delta x|A) - F_X(x|A)}{\Delta x}}{\frac{F_X(x + \Delta x) - F_X(x)}{\Delta x}} P(A)$$

$$= \frac{f_X(x|A)}{f_X(x)} P(A),$$

if $f_X(x|A)$ and $f_X(x)$ exist.

Implication of Point Conditioning

• Note that

$$P(A|X = x) = \frac{f_X(x|A)}{f_X(x)}P(A)$$

$$\Rightarrow P(A|X = x)f_X(x) = f_X(x|A)P(A)$$

$$\Rightarrow \int_{-\infty}^{\infty} P(A|X = x)f_X(x)dx = \int_{-\infty}^{\infty} f_X(x|A)dxP(A)$$

$$\Rightarrow P(A) = \int_{-\infty}^{\infty} P(A|X = x)f_X(x)dx$$

(Continuous Version of Law of Total Probability)

• Once we have the Law of Total Probability, it is easy to derive the **Continuous Version** of **Baye's Theorem**:

$$f_X(x|A) = \frac{P(A|X=x)}{P(A)} f_X(x)$$
$$= \frac{P(A|X=x) f_X(x)}{\int_{-\infty}^{\infty} P(A|X=t) f_X(t) dt}.$$

Example on board.