

# Assignment 13

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# Outline

- 1 Problem
- 2 Properties of Normal Random Variables
- 3 Laplace Transform
- 4 Solution

# Problem Statement

**(Papoulis/Pillai, Exercise 5-48)** The random variable  $X$  is  $N(0; \sigma^2)$ .

- Using Laplace transforms, show that if  $g(x)$  is a function such that  $g(x) \exp(-\frac{x^2}{2\sigma^2}) \rightarrow 0$  as  $|x| \rightarrow \infty$ , then (Price's Theorem)

$$\frac{dE\{g(X)\}}{dv} = \frac{1}{2} E \left\{ \frac{d^2 g(X)}{dX^2} \right\} \quad (1)$$

- The moments  $\mu_n$  of  $X$  are functions of  $v$ . Using (1), show that

$$\mu_n(v) = \frac{n(n-1)}{2} \int_0^v \mu_{n-2}(\beta) d\beta \quad (2)$$

Here,  $v = \sigma^2$ .

# PDF and Partial Derivatives of $N(\mu; \sigma^2)$

Since  $X \sim N(\mu; \sigma^2)$ , the PDF of  $X$  is given by

$$f(x) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right) \quad (3)$$

$$f(x, v) = \frac{1}{\sqrt{2\pi v}} \exp\left(-\frac{(x-\mu)^2}{2v}\right) \quad (4)$$

Notice that

$$\frac{\partial f}{\partial v} = \frac{\exp\left(\frac{(x-\mu)^2}{2v}\right)((x-\mu)^2 - v)}{2\sqrt{2\pi v^5}} \quad (5)$$

$$\frac{\partial^2 f}{\partial x^2} = \frac{\exp\left(\frac{(x-\mu)^2}{2v}\right)((x-\mu)^2 - v)}{\sqrt{2\pi v^5}} \quad (6)$$

$$\Rightarrow \frac{\partial f}{\partial v} = \frac{1}{2} \frac{\partial^2 f}{\partial x^2} \quad (7)$$

# Definition and Properties of Laplace Transform

The Laplace Transform of a real function  $f(t)$  is given by

$$\mathcal{L}\{f\}(s) = \int_0^{\infty} e^{-st} f(t) dt \quad (8)$$

where  $s \in \mathbb{C}$ . Here,  $k(s, t) = e^{-st}$  is called the kernel function. A useful property of Laplace transforms is the following, where the  $n^{\text{th}}$  derivative of  $f$  is assumed to be of exponential type.

$$\mathcal{L}\{f^{(n)}\} = s^n \mathcal{L}\{f\} - \sum_{k=1}^n s^{n-k} f^{(k-1)}(0^+) \quad (9)$$

For  $n = 1$ ,

$$\mathcal{L}\{f'\} = s \mathcal{L}\{f\} - f(0^+) \quad (10)$$

# Solution

In this solution, we assume that  $g^{(k)}(x) \exp(-\frac{x^2}{2\sigma^2}) \rightarrow 0$  as  $|x| \rightarrow \infty$  for  $k = 0, 1, 2$ . Applying a Laplace transform, and noting that at  $f(x, 0^+) = 0$ ,

$$\frac{dE\{g(X)\}}{dv} \xleftrightarrow{\mathcal{L}} s\mathcal{L}\{f\} \quad (11)$$

$$= s \int_0^\infty e^{-sv} \left( \int_{-\infty}^\infty g(x) f(x, v) dx \right) dv \quad (12)$$

$$= \int_{-\infty}^\infty g(x) \left( \int_0^\infty s e^{-sv} f(x, v) dv \right) dx \quad (13)$$

$$= - \int_{-\infty}^\infty g \left( \frac{\partial}{\partial v} f \right) dx \quad (14)$$

$$= \int_{-\infty}^\infty g \left( e^{-sv} \frac{\partial f}{\partial v} \right) dx \quad (15)$$

Using (7), and integrating repeatedly by parts using our assumptions

$$\frac{dE\{g(X)\}}{dv} \xleftrightarrow{\mathcal{L}} \frac{1}{2} \int_0^\infty e^{-sv} \left( \int_{-\infty}^\infty g \frac{\partial^2 f}{\partial x^2} dx \right) dv \quad (16)$$

$$= \frac{1}{2} \int_0^\infty e^{-sv} \left( \int_{-\infty}^\infty f \frac{\partial^2 g}{\partial x^2} dx \right) dv \quad (17)$$

$$= \mathcal{L}\left\{E\left\{\frac{d^2 g(X)}{dX^2}\right\}\right\}(s) \xleftrightarrow{\mathcal{L}} \frac{1}{2} E\left\{\frac{d^2 g(X)}{dX^2}\right\} \quad (18)$$

For the second part, observe that  $\mu_n = E[X^n]$  and hence it is a function of  $v$ . Further, using the exponential power series, note that for any positive integer  $n$ ,

$$\exp\left(\frac{x^2}{2\sigma^2}\right) > \frac{x^{2n}}{(2\sigma^2)^n n!} \quad (19)$$

$$\Rightarrow 0 < \frac{x^n}{\exp\left(\frac{x^2}{2\sigma^2}\right)} < \frac{(2\sigma^2)^n n!}{x^n} \quad (20)$$

and using the Sandwich Theorem, we can choose  $g(x) = x^n$  to use in (1).

$$\mu'_n(v) = \frac{1}{2} E \{ n(n-1)x^{n-2} \} = \frac{n(n-1)}{2} \mu_{n-2}(v) \quad (21)$$

However, note that if  $v = 0$ , then from (4),  $f(x) = 0$  and consequently  $\mu_n(0) = 0$ . Integrating (21) and changing variables, we get

$$\mu_n(v) = \frac{n(n-1)}{2} \int_0^v \mu_{n-2}(\beta) d\beta \quad (22)$$