Chapter 11

Laplace Transforms

There are different types of transforms

Back in Chapter 6 we covered a type of generating function called the z-transform.

The z-transform is particularly well suited to discrete, integer-valued random variables.

In this chapter we introduce a new generating function called the **Laplace transform**, which is well suited to common continuous random variables.

The structure of this chapter will closely mimic that of Chapter 6.

Motivation

Let $X \sim Exp(\lambda)$

What is $E[X^3]$?

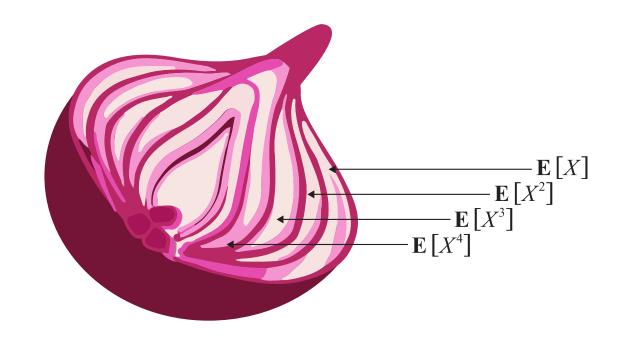
$$\mathbf{E}[X^3] = \int_0^\infty t^3 \cdot \lambda e^{-\lambda t} dt$$

Seems complicated to evaluate!

The Laplace transform will make this very easy!

The Laplace transform as an onion

Onion represents Laplace transform of r.v. X



Lower moments are in the outer layers \rightarrow less effort/tears Higher moments are deeper inside \rightarrow more effort/tears

Laplace transform of continuous r.v.

<u>Defn</u>: Let X be a non-negative continuous r.v. with p.d.f. $f_X(t)$. Then the **Laplace transform** of X is

$$\tilde{X}(s) = \mathbf{E}[e^{-sX}] = \int_0^\infty e^{-st} f_X(t) dt$$

Assume s is a constant where $s \geq 0$.

Note: The Laplace transform can be defined for any r.v., or even for just a function f(t), where $t \ge 0$. However convergence is only guaranteed when X is a non-negative r.v. and $s \ge 0$.

Pop Quiz

<u>Defn</u>: Let X be a non-negative continuous r.v. with p.d.f. $f_X(t)$. Then the **Laplace transform** of X is

$$\tilde{X}(s) = \mathbf{E}[e^{-sX}] = \int_0^\infty e^{-st} f_X(t) dt$$

Assume s is a constant where $s \geq 0$.

Q: What is $\widetilde{X}(0)$?

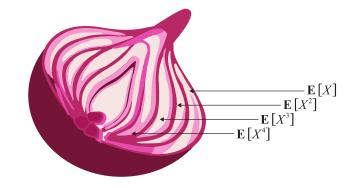
A:
$$\widetilde{X}(0) = E[e^{-0 \cdot X}] = 1$$

Example of Onion Building

$$X \sim Exp(\lambda)$$

Create the onion!

$$\tilde{X}(s) = \mathbf{E}[e^{-sX}] = \int_0^\infty e^{-st} \lambda e^{-\lambda t} dt$$
$$= \lambda \int_0^\infty e^{-(s+\lambda)t} dt$$
$$= \frac{\lambda}{s+\lambda}$$



$$\tilde{X}(s) = \mathbf{E}[e^{-sX}]$$

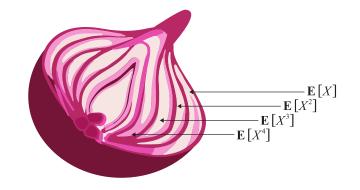
$$= \int_0^\infty e^{-st} f_X(t) dt$$

Example of Onion Building

$$X = 3$$

Create the onion!

$$\tilde{X}(s) = \mathbf{E}[e^{-sX}]$$
$$= \mathbf{E}[e^{-3s}]$$
$$= e^{-3s}$$

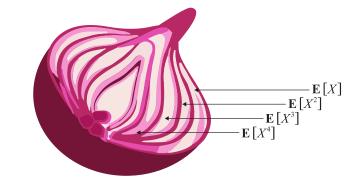


$$\tilde{X}(s) = \mathbf{E}[e^{-sX}]$$
$$= \int_0^\infty e^{-st} f_X(t) dt$$

Example of Onion Building

 $X \sim Uniform(a, b)$, where $a, b \ge 0$ Create the onion!

$$\tilde{X}(s) = \mathbf{E}[e^{-sX}] = \int_{a}^{b} e^{-st} \cdot \frac{1}{b-a} dt$$
$$= \frac{1}{b-a} \cdot \frac{1}{s} \cdot (e^{-sa} - e^{-sb})$$



$$\tilde{X}(s) = \mathbf{E}[e^{-sX}]$$

$$= \int_0^\infty e^{-st} f_X(t) dt$$

Convergence of Laplace transform

Theorem 11.7: $\tilde{X}(s)$ is bounded for any non-negative continuous r.v. X, assuming $s \geq 0$.

Proof:
$$e^{-t} \le 1$$
, $\forall t \ge 0$

$$\Rightarrow (e^{-t})^s \le 1$$
, $\forall s \ge 0$

$$\Rightarrow e^{-st} \le 1$$
, $\forall t, s \ge 0$

$$\Rightarrow \tilde{X}(s) = \int_0^\infty e^{-st} f_X(t) dt \le \int_0^\infty 1 \cdot f_X(t) dt = 1$$

Getting moments: Onion peeling

Theorem 11.8: (Onion Peeling) Let X be a non-negative, continuous r.v. with p.d.f. $f_X(t)$, $t \ge 0$. Then,

$$\left. \tilde{X}'(s) \right|_{s=0} = -\mathbf{E}[X]$$

$$\left. \tilde{X}^{\prime\prime}(s) \right|_{s=0} = \mathbf{E}[X^2]$$

$$\left. \tilde{X}^{\prime\prime\prime}(s) \right|_{s=0} = -\mathbf{E}[\mathbf{X}^3]$$

$$\left. \tilde{X}^{\prime\prime\prime\prime\prime}(s) \right|_{s=0} = \mathbf{E}[X^4]$$

If can't evaluate at s=0, instead consider limit as $s\to 0$ (use L'Hospital's Rule).

Proof of onion peeling theorem

$$e^{-st} = 1 - (st) + \frac{(st)^2}{2!} - \frac{(st)^3}{3!} + \frac{(st)^4}{4!} - \cdots$$
 (Taylor Series Expansion)

$$e^{-st}f(t) = f(t) - (st)f(t) + \frac{(st)^2}{2!}f(t) - \frac{(st)^3}{3!}f(t) + \frac{(st)^4}{4!}f(t) - \cdots$$

$$\int_0^\infty e^{-st} f(t) dt = \int_0^\infty f(t) dt - \int_0^\infty (st) f(t) dt + \int_0^\infty \frac{(st)^2}{2!} f(t) dt - \int_0^\infty \frac{(st)^3}{3!} f(t) dt + \cdots$$

$$\tilde{X}(s) = 1 - s E[X] + \frac{s^2}{2!} E[X^2] - \frac{s^3}{3!} E[X^3] + \frac{s^4}{4!} E[X^4] - \frac{s^5}{5!} E[X^5] + \cdots$$

Proof of onion peeling theorem

$$\tilde{X}(s) = 1 - s E[X] + \frac{s^2}{2!} E[X^2] - \frac{s^3}{3!} E[X^3] + \frac{s^4}{4!} E[X^4] - \frac{s^5}{5!} E[X^5] + \frac{s^6}{6!} E[X^6] \cdots$$

$$\tilde{X}'(s) = -\mathbf{E}[X] + s\mathbf{E}[X^2] - \frac{s^2}{2!}\mathbf{E}[X^3] + \frac{s^3}{3!}\mathbf{E}[X^4] - \frac{s^4}{4!}\mathbf{E}[X^5] + \frac{s^5}{5!}\mathbf{E}[X^6] \cdots$$

$$\tilde{X}'(0) = -\mathbf{E}[X]$$

$$\tilde{X}''(s) = \mathbf{E}[X^2] - s\mathbf{E}[X^3] + \frac{s^2}{2!}\mathbf{E}[X^4] - \frac{s^3}{3!}\mathbf{E}[X^5] + \frac{s^4}{4!}\mathbf{E}[X^6] \cdots$$

$$\tilde{X}^{\prime\prime}(0) = \mathbf{E}[X^2] \quad \checkmark$$

$$\tilde{X}'''(s) = -\mathbf{E}[X^3] + s\mathbf{E}[X^4] - \frac{s^2}{2!}\mathbf{E}[X^5] + \frac{s^3}{3!}\mathbf{E}[X^6] \cdots$$

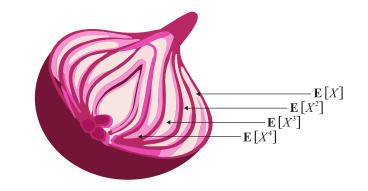
$$\tilde{X}'''(0) = -E[X^3]_{\text{"Invoduction to Probability for Computing", Harchol-Balter '24}$$

Example of onion peeling

$$X \sim Exp(\lambda)$$

$$\tilde{X}(s) = \frac{\lambda}{\lambda + s} = \lambda(\lambda + s)^{-1}$$

Q: Peel the onion to get E[X], $E[X^2]$, $E[X^3]$, $E[X^4]$,...



$$\tilde{X}'(s) = -\lambda(\lambda + s)^{-2}$$

$$\Rightarrow E[X] = \frac{1}{\lambda}$$

$$\tilde{X}''(s) = 2\lambda(\lambda + s)^{-3}$$

$$\Rightarrow \mathbf{E}[X^2] = \frac{2}{\lambda^2}$$

$$\tilde{X}'''(s) = -3! \lambda (\lambda + s)^{-4}$$

$$\Rightarrow \mathbf{E}[X^3] = \frac{3!}{\lambda^3}$$

$$\Longrightarrow \mathbf{E}[X^k] = \frac{k!}{\lambda^k}$$

Linearity of Transforms

Theorem 11.10: (Linearity) Let X and Y be independent continuous r.v.s. Let

$$Z = X + Y$$

Then the Laplace transform of Z is:

$$\tilde{Z}(s) = \tilde{X}(s) \cdot \tilde{Y}(s)$$

Proof:
$$ilde{Z}(s) = \mathbf{E}[e^{-sZ}] = \mathbf{E}[e^{-s(X+Y)}]$$

$$= \mathbf{E}[e^{-sX} \cdot e^{-sY}]$$

$$= \mathbf{E}[e^{-sX}] \cdot \mathbf{E}[e^{-sY}]$$

$$= \tilde{X}(s) \cdot \tilde{Y}(s)$$

Conditioning with Transforms

Theorem 11.11: Let X, A, and B be continuous r.v.s. where

$$X = \begin{cases} A & \text{w.p.} & p \\ B & \text{w.p.} & 1-p \end{cases}$$

Then,

$$\tilde{X}(s) = p \cdot \tilde{A}(s) + (1-p) \cdot \tilde{B}(s)$$

Proof:
$$\tilde{X}(s) = E[e^{-sX}]$$

$$= \mathbf{E}[e^{-sX}|X=A] \cdot p + \mathbf{E}[e^{-sX}|X=B] \cdot (1-p)$$

$$= \mathbf{E}[e^{-sA}] \cdot p + \mathbf{E}[e^{-sB}] \cdot (1-p)$$

$$= p \cdot \tilde{A}(s) + (1 - p) \cdot \tilde{B}(s)$$

Conditioning

Theorem 11.12:

Let Y be a continuous r.v. and let X_Y be a continuous r.v. that dependes on Y. Let $f_Y(y)$ denote the p.d.f. of Y.

Then:

$$\widetilde{X_Y}(s) = \int_{y=0}^{\infty} \widetilde{X_Y}(s) \cdot f_Y(y) dy$$

$$\widetilde{X_Y}(s) = \mathbf{E}[e^{-sX_Y}] = \int_{y=0}^{y=\infty} \mathbf{E}[e^{-sX_Y}|Y=y] \cdot f_Y(y) dy$$
$$= \int_{y=0}^{y=\infty} \mathbf{E}[e^{-sX_Y}] \cdot f_Y(y) dy$$
$$= \int_{v=0}^{\infty} \widetilde{X_Y}(s) \cdot f_Y(y) dy$$