

Laplace - Stieltjes Transform - The Laplace - Stieltjes transform (LST) of a continuous function  $f(t)$  is defined as

$$\tilde{f}(s) \equiv \int_0^{\infty} e^{-st} df(t)$$

We can relate  $\tilde{f}(s)$  to  $f^*(s)$  [when  $f^*(s)$  exists] by integrating by parts as follows -

$$\tilde{f}(s) = \int_0^{\infty} e^{-st} df(t) = e^{-st} f(t) \Big|_0^{\infty} + \int_0^{\infty} s e^{-st} f(t) dt$$

Assuming  $f(t)$  is such that as  $t \rightarrow \infty$ , it does not go to  $\infty$  as fast as  $e^{-t}$ , <sup>approaches to zero</sup> i.e.  $\lim_{t \rightarrow \infty} e^{-st} f(t) = 0$

So, we have

$$\tilde{f}(s) = s f^*(s) - f(0) = L[f'(t)]$$

If we consider the LST of a continuous C.d.f.,  $F(\cdot)$  where the random variable is non-negative, then  $F(\infty) = 1$  and  $F(0) = 0$ , so that

$$\tilde{F}(s) = s F^*(s)$$

and

$$\tilde{F}(s) = \int_0^{\infty} e^{-st} dF(t) = \int_0^{\infty} e^{-st} f(t) dt = \underline{f^*(s)}$$

i.e. LST of a Cdf = LT of the corresponding p.d.f.

Laplace - Stieltjes Transform of Stieltjes Convolution -

The integral

$$\int_0^t df(u) G(t-u) = F(t) \otimes G(t)$$

is called the Stieltjes convolution of two continuous functions  $F(t)$  and  $G(t)$ .

The LST of Stieltjes convolution is

$$\begin{aligned}
 \text{LS}[F(t) \circledast G(t)] &= \int_0^{\infty} e^{-st} d \int_0^t dF(u) G(t-u) \\
 &= e^{-st} \int_0^t dF(u) G(t-u) \Big|_0^{\infty} + s \int_0^{\infty} e^{-st} dt \left( \int_0^t dF(u) G(t-u) \right) \\
 &= 0 + s \int_0^{\infty} e^{-st} dt \int_0^t F'(u) G(t-u) du \\
 &= s \int_0^{\infty} e^{-st} dt [F'(t) \circledast G(t)] \\
 &= s \text{LT}[F'(t) \circledast G(t)] \\
 &= s \text{LT}[F'(t)] \cdot \text{LT}[G(t)] \\
 &= \text{LST}[F(t)] \cdot \text{LST}[G(t)] \\
 &= \tilde{F}(s) \cdot \tilde{G}(s), \text{ if } G(0) = 0
 \end{aligned}$$

Mean and Variance in terms of L.T. -

Let  $f(t)$  be a p.d.f of a non-negative continuous r.v.  $T$ . Then

$$\begin{aligned}
 \frac{d}{ds} f^*(s) &= \frac{d}{ds} \int_0^{\infty} e^{-st} f(t) dt \\
 &= - \int_0^{\infty} t e^{-st} f(t) dt
 \end{aligned}$$

$$\frac{d^2}{ds^2} f^*(s) = (-1)^2 \int_0^{\infty} t^2 e^{-st} f(t) dt$$

In general

$$\frac{d^n}{ds^n} f^*(s) = (-1)^n \int_0^{\infty} t^n e^{-st} f(t) dt$$

If we put  $s=0$ , we get

$$\frac{d^n}{ds^n} f^*(s) \Big|_{s=0} = (-1)^n \int_0^{\infty} t^n f(t) dt = (-1)^n E(T^n)$$

$$\text{or } E(T^n) = (-1)^n \frac{d^n}{ds^n} f^*(s) \Big|_{s=0}$$

In particular

$$E(T) = -\frac{d}{ds} f^*(s) \Big|_{s=0}, \quad E(T^2) = \frac{d^2}{ds^2} f^*(s) \Big|_{s=0}$$

so that  $V(T) = E(T^2) - [E(T)]^2$

Example-1 Let  $X$  follows negative exponential distribution with parameter  $\lambda$ . Its density function is

$$f(x) = \begin{cases} \lambda e^{-\lambda x}, & \lambda > 0; 0 \leq x < \infty \\ 0 & ; x < 0 \end{cases}$$

The L.T. of  $f(x)$  is

$$\begin{aligned} f^*(s) &= \int_0^{\infty} e^{-sx} f(x) dx \\ &= \lambda \int_0^{\infty} e^{-sx} e^{-\lambda x} dx = \frac{\lambda}{s+\lambda} \end{aligned}$$

Now, we have

$$\frac{d}{ds} f^*(s) = -\frac{\lambda}{(s+\lambda)^2}, \quad \frac{d^2}{ds^2} f^*(s) = \frac{2\lambda}{(s+\lambda)^3}$$

Therefore,

$$E(X) = -\frac{d}{ds} f^*(s) \Big|_{s=0} = \frac{1}{\lambda} \quad \text{and} \quad E(X^2) = \frac{d^2}{ds^2} f^*(s) \Big|_{s=0} = \frac{2}{\lambda^2}$$

so that

$$V(X) = \frac{2}{\lambda^2} - \frac{1}{\lambda^2} = \frac{1}{\lambda^2}$$

Example-2 Let  $X$  follows negative exponential distribution with parameter  $\lambda, \mu$ . Its density function is

$$f(x) = \begin{cases} \lambda e^{-\lambda(x-\mu)}, & \lambda, \mu > 0; \mu \leq x < \infty \\ 0 & ; x < \mu \end{cases}$$

The LT. of  $f(x)$  is

$$\begin{aligned}
f^*(s) &= \int_0^{\infty} e^{-sx} f(x) dx \\
&= \int_0^{\mu} e^{-sx} f(x) dx + \int_{\mu}^{\infty} e^{-sx} f(x) dx \\
&= \lambda \int_0^{\infty} e^{-sx} e^{-\lambda(x-\mu)} dx \\
&= \lambda e^{\lambda\mu} \int_{\mu}^{\infty} e^{-(s+\lambda)x} dx \\
&= \lambda e^{\lambda\mu} \left( \frac{e^{-(s+\lambda)\mu}}{s+\lambda} \right) = \frac{\lambda}{s+\lambda} e^{-s\mu}
\end{aligned}$$

Now,  $\frac{d}{ds} f^*(s) = \frac{\lambda}{s+\lambda} (-\mu) e^{-s\mu} + e^{-s\mu} (-1) \frac{\lambda}{(s+\lambda)^2}$

and  $\frac{d^2}{ds^2} f^*(s) = (-\mu) \left[ \frac{\lambda}{s+\lambda} (-\mu) e^{-s\mu} + e^{-s\mu} (-1) \frac{\lambda}{(s+\lambda)^2} \right] - \lambda \left[ e^{-s\mu} (-2) \frac{1}{(s+\lambda)^3} + \frac{1}{(s+\lambda)^2} (-\mu) e^{-s\mu} \right]$

So that

$$E(X) = -\frac{d}{ds} f^*(s) \Big|_{s=0} = \frac{1}{\lambda} + \mu$$

$$\begin{aligned}
\text{and } E(X^2) &= \frac{d}{ds^2} f^*(s) \Big|_{s=0} = \mu^2 + \frac{\mu}{\lambda} + \frac{2}{\lambda^2} + \frac{\mu}{\lambda} \\
&= \mu^2 + \frac{2}{\lambda^2} + \frac{2\mu}{\lambda}
\end{aligned}$$

Therefore,  $V(X) = \mu^2 + \frac{2}{\lambda^2} + \frac{2\mu}{\lambda} - \left( \frac{1}{\lambda} + \mu \right)^2 = \frac{1}{\lambda^2}$  (indep. of  $\mu$ )

Example-3 Let  $X$  follows two parameter gamma distribution with parameters  $\lambda, k$  ( $\lambda > 0$  is the scale parameter and  $k > 0$  is the shape parameter).

The density function of the r.v.  $X$  is

$$f(x) = \frac{\lambda^k e^{-\lambda x} x^{k-1}}{\Gamma(k)}, \quad x \geq 0$$

$$= 0, \quad x < 0$$

The L.T.  $f^*(s)$  is given by

$$f^*(s) = \int_0^{\infty} e^{-sx} f(x) dx$$

$$= \frac{\lambda^k}{\Gamma(k)} \int_0^{\infty} e^{-sx} e^{-\lambda x} x^{k-1} dx$$

$$= \frac{\lambda^k}{\Gamma(k)} \frac{\Gamma(k)}{(s+\lambda)^k}, \quad (\text{by gamma integral})$$

$$= \left( \frac{\lambda}{s+\lambda} \right)^k$$

Now, we have

$$\frac{d}{ds} f^*(s) = -\frac{k \lambda^k}{(s+\lambda)^{k+1}} \quad \text{and} \quad \frac{d^2}{ds^2} f^*(s) = \frac{k(k+1) \lambda^k}{(s+\lambda)^{k+2}}$$

Therefore,

$$E(X) = -\frac{d}{ds} f^*(s) \Big|_{s=0} = \frac{k}{\lambda}, \quad E(X^2) = \frac{d^2}{ds^2} f^*(s) \Big|_{s=0} = \frac{k(k+1)}{\lambda^2}$$

so that,  $V(X) = \frac{k}{\lambda^2}$

If we replace  $\lambda$  by  $\lambda k$  in the above gamma density, we get the density of Erlang distribution. i.e.

$$f(x) = \frac{(\lambda k)^k e^{-\lambda k x} x^{k-1}}{\Gamma(k)}, \quad x \geq 0$$

In this case  $f^*(s) = \left( \frac{\lambda k}{s+\lambda k} \right)^k$

and  $E(X) = \frac{1}{\lambda}, \quad V(X) = \frac{1}{k\lambda}$

Example-4 Let  $X$  be a random variable whose whole mass is concentrated at one single point say 'a' i.e. the r.v.  $X$  follows one point distribution as follows -

$$P(X=a) = 1 \text{ and } P(X \neq a) = 0$$

The L.T. of this distribution is given by

$$\begin{aligned} f^*(s) &= E[e^{-sX}] \\ &= e^{-sa} P(X=a) = e^{-sa} \end{aligned}$$

Now

$$\frac{d}{ds} f^*(s) = (-a)e^{-sa} \text{ and } \frac{d^2}{ds^2} f^*(s) = a^2 e^{-sa}$$

so that,

$$E(X) = -\frac{d}{ds} f^*(s) \Big|_{s=0} = a$$

$$\text{and } E(X^2) = \frac{d^2}{ds^2} f^*(s) \Big|_{s=0} = a^2$$

$$\text{Therefore, } V(X) = a^2 - a^2 = 0$$

Example-5 Let  $X$  follows Poisson distribution with parameter  $\lambda$  i.e.

$$P(X=x) = \frac{e^{-\lambda} \lambda^x}{x!}; \quad x=0, 1, 2, \dots$$

The L.T. of this probability function is given by

$$\begin{aligned} f^*(s) &= E[e^{-sX}] \\ &= \sum_{x=0}^{\infty} e^{-sx} P(X=x) \end{aligned}$$

$$\begin{aligned}
&= \sum_x e^{-\lambda x} \frac{e^{-\lambda} \lambda^x}{x!} \\
&= e^{-\lambda} \sum_{x=0}^{\infty} \frac{(\lambda e^{-\lambda})^x}{x!} = e^{-\lambda} e^{\lambda e^{-\lambda}} \\
&= \exp[\lambda(e^{-\lambda} - 1)]
\end{aligned}$$

Now

$$\begin{aligned}
E(X) &= -\frac{d}{d\lambda} \log f^*(\lambda) \Big|_{\lambda=0} = -1 \left[ e^{\lambda(e^{-\lambda} - 1)} \lambda e^{-\lambda} (-1) \right] \Big|_{\lambda=0} \\
&= \lambda
\end{aligned}$$

$$\begin{aligned}
E(X^2) &= \frac{d^2}{d\lambda^2} \log f^*(\lambda) \Big|_{\lambda=0} = (-\lambda) \left[ e^{\lambda(e^{-\lambda} - 1)} \lambda e^{-\lambda} (-1) \right. \\
&\quad \left. + e^{\lambda(e^{-\lambda} - 1)} e^{-\lambda} (-1) \right] \Big|_{\lambda=0} \\
&= (-\lambda) [-\lambda - 1] = \lambda^2 + \lambda
\end{aligned}$$

Therefore,

$$V(X) = \lambda^2 + \lambda - \lambda^2 = \lambda$$



Theorem — If  $L[f(t)] = f^*(s)$ , then

$$L\left[\frac{f(t)}{t}\right] = \int_s^{\infty} f^*(u) du$$

provided  $\lim_{t \rightarrow 0} \left\{ \frac{f(t)}{t} \right\}$  exists.

Proof — let  $\frac{f(t)}{t} = g(t)$

$$\therefore f(t) = t g(t)$$

$$\text{Now } L[f(t)] = L[t g(t)] = -\frac{d}{d\lambda} L[g(t)]$$

$$\text{or } f^*(\lambda) = -\frac{d}{d\lambda} g^*(\lambda) \tag{1}$$

Integrating both sides of (1) w.r.t.  $s$  from  $s$  to  $\infty$ , we have

$$\int_s^{\infty} f^*(u) du = - [g^*(u)]_s^{\infty}$$

$$= g^*(s) - g^*(\infty)$$

$$\text{or } g^*(s) = \int_s^{\infty} f^*(u) du \quad \because g^*(\infty) = 0$$

Proved

Ex. 1 Find L.T of the following functions -

(i)  $f(t) = t^n \quad (n > -1)$

(ii)  $f(t) = e^{at}$

(iii)  $f(t) = t^n e^{at} \quad (n > -1)$

(iv)  $f(t) = \sin kt$

Sol<sup>n</sup> (i) The L.T of  $f(t) = t^n$  is given by

$$L[t^n] = \int_0^{\infty} e^{-st} t^n dt$$

$$= \frac{\Gamma(n+1)}{s^{n+1}}$$

(ii) The L.T of  $f(t) = e^{at}$  is

$$L[e^{at}] = \int_0^{\infty} e^{-st} e^{at} dt$$

$$= \int_0^{\infty} e^{-(s-a)t} t^{1-1} dt$$

$$= \frac{1}{(s-a)}$$

(iii) The L.T of  $f(t) = t^n e^{at}$  is

$$L[t^n e^{at}] = \int_0^{\infty} e^{-st} t^n e^{at} dt$$



## Table of Laplace Transforms of some functions

S.N.	$f(t)$	$f^*(s)$	S.N.	$f(t)$	$f^*(s)$
1	1	$1/s$	11	$\cos at$	$s/(s^2 + a^2)$
2	$a f(t)$	$a f^*(s)$	12	$\sin at$	$a/(s^2 - a^2)$
3	$f_1(t) + f_2(t)$	$f_1^*(s) + f_2^*(s)$	13	$\cosh at$	$s/(s^2 - a^2)$
4	$f(t-a), a > 0$	$e^{-as} f^*(s)$	14	$e^{-at} \sin at$	$\frac{a}{(s+a)^2 + a^2}, s > -a$
5	$t f(t)$	$-\frac{d}{ds} f^*(s)$	15	$e^{at} \cos at$	$\frac{s-a}{(s-a)^2 + a^2}, s > -a$
6	$e^{at} f(t)$	$f^*(s+a)$	16	<del><math>t^k e^{-at} f(t)</math></del>	<del><math>\Gamma(k+1)/f^*(s+a)^{k+1}</math></del>
7	$e^{-at}$	$1/(s+a)$	17	$e^{-at} t^k / k!$	$1/(s+a)^{k+1}, s > -a$
8	$t^k (k > -1)$	$\Gamma(k+1)/s^{k+1}$	18	$e^{at} - e^{bt} (a > b)$	$\frac{a-b}{(s-a)(s-b)}, s > a$
9	$t^k e^{-at}$	$\Gamma(k+1)/(s+a)^{k+1}$	19	$f(t/a)$	$a f^*(as)$
10	$\ln at$	$f^*(s) - f^*(0)$	20	$f(t)$	$s f^*(s) - f(0)$