

# APPLICATIONS OF H – FUNCTION IN STATISTICAL DISTRIBUTION THEORY

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**Abstract**—The H- function is a special function which has wide applications in the field of statistical distribution . By using the definitions and its properties one can easily express most of the statistical distributions in terms of H- function particularly for a continuous random variable. In this paper we attempt to prove the same results of mean , variance and moment generating functions of Gamma distributions by using Mellin- BARNES Integral , properties of Gamma function and H- function

**Keywords**— H- function , Gamma function , MELLIN- BARNES Integral , Gamma Distribution , Mean , Variance

**Introduction**

H function is denoted by  $H(x) = H_{p,q}^{m,n}(z)$  [1] , [2] and which can be defines in two different forms

$$H(z) = H_{p,q}^{m,n} \left[ z / (a_p, A_p) : (b_q, B_q) \right] = H_{p,q}^{m,n} \left[ \begin{matrix} z / (a_p, A_p) \\ (b_q, B_q) \end{matrix} \right]$$

$$H_{p,q}^{m,n} \left[ \begin{matrix} / (a_1, A_1), \dots (a_n, A_n), (a_{n+1}, A_{n+1}) \dots (a_p, A_p) \\ z / (b_1, B_1), \dots (b_m, B_m), (b_{m+1}, B_{m+1}) \dots (b_q, B_q) \end{matrix} \right] = \frac{1}{2\pi i} \int_{L_1} \theta(s) z^{-s} ds$$

Where  $\theta(s) = \frac{\prod_{j=1}^m \Gamma(b_j + B_j s) \prod_{j=1}^n \Gamma(1 - a_j - A_j s)}{\prod_{j=m+1}^q \Gamma(1 - b_j - B_j s) \prod_{j=n+1}^p \Gamma(a_j + A_j s)}$ .

here  $A_j$  and  $B_j$  are positive real numbers and  $z$  and all  $a_j$  and  $b_j$  may be real or complex numbers . Also the parameters  $m, n, p, q \in \mathbb{Z}$  such that  $0 \leq m \leq q$  and  $0 \leq n \leq p$  . Note that empty product forms unity .  $L_1$  is a contour in the complex plane which runs from  $c - i\infty$  to

$c + i\infty$  such that all the poles  $\Gamma(b_j + B_j s)$  ,  $j = 1, 2 \dots m$  lies to the left of  $L_1$  and the poles

$\Gamma(1 - a_j - A_j s)$  ,  $j = 1, 2 \dots n$  lies to the right of  $L_1$

Alternatively H function can be defined as follows

$$H(z) = H_{p,q}^{m,n} \left[ z / (a_p, A_p) : (b_q, B_q) \right] = H_{p,q}^{m,n} \left[ \begin{matrix} z / (a_p, A_p) \\ (b_q, B_q) \end{matrix} \right] = H_{p,q}^{m,n} \left[ \begin{matrix} z / (a_1, A_1), \dots (a_n, A_n), (a_{n+1}, A_{n+1}) \dots (a_p, A_p) \\ z / (b_1, B_1), \dots (b_m, B_m), (b_{m+1}, B_{m+1}) \dots (b_q, B_q) \end{matrix} \right] = \frac{1}{2\pi i} \int_{L_2} \theta^*(s) z^{+s} ds$$

Where  $\theta^*(s) = \frac{\prod_{j=1}^m \Gamma(b_j - B_j s) \prod_{j=1}^n \Gamma(1 - a_j + A_j s)}{\prod_{j=m+1}^q \Gamma(1 - b_j + B_j s) \prod_{j=n+1}^p \Gamma(a_j - A_j s)}$

Here also  $L_2$  is a contour in the complex plane which runs from  $c - i\infty$  to  $c + i\infty$  such that all the poles  $\Gamma(b_j - B_j s)$  ,  $j = 1, 2 \dots m$  lies to the right of  $L_2$  and the poles  $\Gamma(1 - a_j + A_j s)$  ,  $j = 1, 2 \dots n$  lies to the left of  $L_2$  .

**A . Preliminary Results in the Literature**

In this section some important results were presented in the literature given without proof

**1. Pochhammer Symbol** :- It is denoted by  $(\alpha)_r$

$$\therefore (\alpha)_r = \alpha (\alpha + 1)(\alpha + 2) \dots (\alpha + r - 1) ,$$

$$(\alpha)_0 = 1, \alpha \neq 0$$

**2. Gamma Function** :- It is denoted by  $\Gamma(\alpha)$

- (a)  $\Gamma(\alpha + 1) = \alpha \Gamma(\alpha)$
- (b)  $\Gamma(\alpha) = \int_0^\infty x^{\alpha-1} e^{-x} dx$
- (c)  $\int_0^\infty x^{\alpha-1} e^{-ax} dx = \frac{\Gamma(\alpha)}{a^\alpha}$

**3. Mellin - Barnes Integral for an Exponential function [5]**

- (i)  $f(z) = \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} \Gamma(s) z^{-s} ds = e^{-z}$   
 $\Rightarrow e^{-z} = \sum_{v=0}^\infty \frac{(-1)^v}{v!} z^v = H_{1,1}^{1,0} \left[ z / (0,1) \right]$
- (ii)  $H_{1,1}^{1,0} \left[ z / (b,1) \right] = z^b e^{-z}$
- (iii)  $f(z) = \frac{1}{2\pi i} \frac{1}{\Gamma(a)} \int_{c-i\infty}^{c+i\infty} \Gamma(-s) \Gamma(s+a) (-z)^s ds$   
 $= (1-z)^{-a}$   
 $\Rightarrow (1-z)^{-a} = \sum_{v=0}^\infty \frac{(a)_v}{v!} z^v = H_{1,1}^{1,1} \left[ -z / (1-a, 1) \right]$

**4 .Properties of H function :-**

(a) Reciprocal argument :-

$$H_{p,q}^{m,n} \left[ z / (a_p, A_p) \right] = H_{p,q}^{m,n} \left[ \frac{1}{z} / (1-b_q, B_q) \right]$$

(b) Argument to a real power :-

$$H_{p,q}^{m,n} \left[ z^k / (a_p, A_p) \right] = (-k)^{-1} H_{q,p}^{n,m} \left[ z / (1-b_q, \frac{-B_q}{k}) \right] \text{ when } k > 0$$

$$H_{p,q}^{m,n} \left[ z / (a_p, A_p) \right] = -k H_{q,p}^{n,m} \left[ z^k / (1-b_q, -kB_q) \right] \text{ when } k < 0$$

(c) Multiplication by the Argument to a power k :-

$$z^k H_{p,q}^{m,n} \left[ z^k / (a_p, A_p) \right] = H_{p,q}^{m,n} \left[ z^k / (a_p + kA_p, A_p) \right]$$

**3. Derivatives of H function :-**

The  $r^{th}$  derivative of a H function can be defined using SKIBINSKI's formula [5]

$$z^r \frac{d^r}{dx^r} H_{p,q}^{m,n} \left[ z^k / (a_p, A_p) \right] =$$

$$H_{p+1,q+1}^{m,n+1} \left[ z^k / (0,k), (a_p, A_p) \right]$$

**B. Main Results :-**

In this section we are expressing Gamma function and its distribution in terms of H functions and H distributions hence we find the Mean, Variance, and Moment generating function using definition of H function and hence arrive the same result as given in the literature.

**C. Gamma Distribution :-**

A Continuous Random Variable X is called Gamma distribution if its probability density function is given by

$$f(x) = \begin{cases} \frac{1}{\Gamma(m)} a^m x^{m-1} e^{-ax}, & x \geq 0 \\ 0, & \text{otherwise} \end{cases}$$

The above function can also be expressed in the form of H function as follows

$$f(x) = \begin{cases} \frac{a}{\Gamma(m)} H_{0,1}^{1,0} \left[ ax / (m-1, 1) \right], & x \geq 0 \\ 0, & \text{otherwise} \end{cases}$$

By using gamma function and definition of H function one can derive the above result

$$\begin{aligned} \text{Now consider } f(x) &= \frac{a}{\Gamma(m)} H_{0,1}^{1,0} \left[ ax / (m-1, 1) \right] \\ &= \frac{a}{\Gamma(m)} \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} \Gamma(m-1+s) (ax)^{-s} ds \\ &= \frac{a}{\Gamma(m)} \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} \Gamma(s+m-1) (ax)^{-(s+m-1-m+1)} ds \\ &= \frac{a}{\Gamma(m)} \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} \Gamma(s+m-1) (ax)^{-(s+m-1)} (ax)^{m-1} ds \\ &= \frac{a}{\Gamma(m)} \frac{1}{2\pi i} (a)^{m-1} (x)^{m-1} \int_{c-i\infty}^{c+i\infty} \Gamma(s+m-1) (ax)^{-(s+m-1)} ds \\ &= \frac{a^m}{\Gamma(m)} (x)^{m-1} e^{-ax} \text{ which completes the proof} \end{aligned}$$

**D. Mean and variance of Gamma distribution in terms H function :-**

H function is a generalized form of any special functions which can be reduced in to any continuous distribution .

In this section we deal with mean and variance of gamma distribution using the definition of H function as special function In the statistical literature the pdf of gamma function is

$$f(x) = \frac{1}{\Gamma(m)} a^m x^{m-1} e^{-ax} \text{ then its mean =}$$

$$E(X) = \int_0^\infty x f(x) dx = \frac{m}{a} \text{ and}$$

$$\text{Variance} = V(X) = E(X^2) - [E(X)]^2 = \frac{m}{a^2} \quad [4]$$

The gamma function can be expressed in the form of H -function

$$f(x) = \frac{a}{\Gamma(m)} H_{0,1}^{1,0} \left[ \frac{ax}{(m-1,1)} \right] =$$

$$H(x) = \frac{a}{\Gamma(m)} (ax)^{m-1} e^{-ax}$$

$$\text{Mean} = E(X) = \int_0^\infty x f(x) dx$$

$$= \int_0^\infty x H(x) dx \quad \text{integrating by parts}$$

$$= \left\{ H(x) \frac{x^2}{2} - \int_0^\infty \frac{x^2}{2} H'(x) dx \right\}$$

$$= \left\{ \int_0^\infty \frac{x^2}{2} \frac{d}{dx} \left( \frac{a}{\Gamma(m)} (ax)^{m-1} e^{-ax} \right) dx \right\}$$

$$= \frac{a}{\Gamma(m)} \cdot \frac{-1}{2} \cdot (a)^{m-1} \left\{ \int_0^\infty x^2 [x^{m-1} e^{-ax} \cdot -a + e^{-ax} (m-1)x^{m-2}] dx \right\}$$

$$= -\frac{a^m}{\Gamma(m)} \left( \frac{1}{2} \right) \left\{ \int_0^\infty -a x^{(m+2)-1} e^{-ax} dx + \int_0^\infty x^{(m+1)-1} e^{-ax} dx \right\}$$

$$= -\frac{a^m}{\Gamma(m)} \left( \frac{1}{2} \right) \left\{ -a \frac{\Gamma(m+2)}{a^{m+2}} + (m-1) \frac{\Gamma(m+1)}{a^{m+1}} \right\}$$

$$= -\frac{a^m}{\Gamma(m)} \left( \frac{1}{2} \right) \left\{ -\frac{\Gamma(m+2) + (m-1)\Gamma(m+1)}{a^{m+1}} \right\}$$

$$= -\frac{1}{\Gamma(m)} \left( \frac{1}{2} \right) \left\{ -\frac{(m+1)(m)\Gamma(m) + (m-1)m\Gamma(m)}{a} \right\}$$

$$= -\frac{1}{\Gamma(m)} \left( \frac{1}{2} \right) (m)\Gamma(m) \left\{ \frac{-m-1+m-1}{a} \right\}$$

$$= \frac{m}{a}$$

By similar calculations we can obtain the following result

$$E(X^2) = \int_0^\infty x^2 f(x) dx = \int_0^\infty x^2 H(x) dx$$

$$= -\frac{a^m}{\Gamma(m)} \left( \frac{1}{3} \right) \left\{ -a \frac{\Gamma(m+3)}{a^{m+3}} + (m-1) \frac{\Gamma(m+2)}{a^{m+2}} \right\}$$

$$=$$

$$-\frac{a^m}{\Gamma(m)} \left( \frac{1}{3} \right) \left\{ -\frac{(m+2)(m+1)(m)\Gamma(m) + (m-1)(m+1)m\Gamma(m)}{a^{m+2}} \right\}$$

$$= \frac{m(m+1)}{a^2}$$

Hence variance =  $\frac{m}{a^2}$

**7. Moment Generating function (MGF) of Gamma function in the form of H function :-**

MGF about the origin is  $M_X(e^{tx}) = E(e^{tx})$

$$= \left( 1 - \frac{t}{a} \right)^{-m}$$

Now  $f(z) = \frac{1}{2\pi i} \frac{1}{\Gamma(a)} \int_{c-i\infty}^{c+i\infty} \Gamma(-s) \Gamma(s+a) (-z)^s ds$

$$= (1-z)^{-a} = \sum_{v=0}^\infty \frac{(a)_v}{v!} z^v$$

$$= \frac{1}{\Gamma(a)} H_{1,1}^1 \left[ \frac{-z}{(0,1)} \right]$$

Consider  $\frac{1}{\Gamma(a)} \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} \theta(s) (-z)^{-s} ds$

$$= \frac{1}{\Gamma(a)} \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} \Gamma(0+s)\Gamma(1-(1-m)-s) (-z)^{-s} ds$$

$$= \frac{1}{\Gamma(a)} \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} \Gamma(s)\Gamma(-s+m) (-z)^{-s} ds$$

By putting  $-s = s \quad z = \frac{t}{a}$  one can reduces in to

$$f(z) = \frac{1}{2\pi i} \frac{1}{\Gamma(a)} \int_{c-i\infty}^{c+i\infty} \Gamma(-s) \Gamma(s+a) \left( -\frac{t}{a} \right)^s ds =$$

$$= \left( 1 - \frac{t}{a} \right)^{-a}$$

$$= \sum_{v=0}^\infty \frac{(m)_v}{v!} \left( \frac{t}{a} \right)^v$$

**E. Moment about the origin**

In this section , Using the above relation one can find moment about the origin , by putting  $v = 1, 2, \dots$

When  $v = 1$  , we get  $\frac{(m)_1}{1!} \left( \frac{t}{a} \right)^1$

$$= \frac{m}{1!} \cdot \frac{t}{a}$$

Then  $\mu_1' =$  co efficient of  $\frac{t}{1!} = \frac{m}{a} =$  mean

When  $v = 2$  , we get  $\frac{(m)_2}{2!} \left( \frac{t}{a} \right)^2$

$$= \frac{m(m+1)}{2!} \cdot \left( \frac{t}{a} \right)^2$$

Then  $\mu_2' =$  co efficient of  $\frac{t^2}{2!} = \frac{\Gamma(m+2)}{a^2 \Gamma(m)}$

$\mu_3' =$  co efficient of  $\frac{t^3}{3!} = \frac{\Gamma(m+3)}{a^3 \Gamma(m)}$

$\mu_4' =$  co efficient of  $\frac{t^4}{4!} = \frac{\Gamma(m+4)}{a^4 \Gamma(m)}$

#### **F. Conclusion :-**

Similar idea and methodology can be applied in Beta distributions of type – I and type- II , exponential and uniform distribution also.

#### **References**

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