

Fuzzy Rayleigh Distribution Model for the Expected Salivary Excretion of Oxytocin in Humans

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Abstract: The study is set to investigate the oxytocin was developed by preset on-line solid-phase micro extraction coupled with liquid chromatography-tandem mass spectrometry in saliva samples using the fuzzy Rayleigh distribution. Parameter of Rayleigh distribution find out by the Maximum Likelihood Estimator. The fuzzy mean values and variance values are calculated for different alpha values. The result shows that the mean and variance values are increasing for lower alpha values and decreasing for upper alpha values.

Keywords: Oxytocin, Rayleigh distribution, Maximum Likelihood Estimator.

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I. Introduction

The Rayleigh distribution is named after the British physicist Lord Rayleigh (1842–1919), also known as Baron John William Strutt Rayleigh and Nobel Prize winner in physics 1904. Consequent to the exponential law, the Rayleigh distribution is the mainly far and wide renowned particular case of the Weibull distribution. It comes up through the Weibull density when the shape parameter is set equivalent to two. Similarly the square root of a chi-squared χ_v^2 random variable with $v = 2$, that is of an exponential random variable, follows the Rayleigh distribution [8]. The Rayleigh distribution was firstly derived in association with an obstacle in acoustics, and has been used in modelling certain features of electronic waves and as the distance distribution between individuals in a spatial Poisson process. Most frequently however it appears as a suitable model in life testing and reliability theory. Heading for additional particulars on the Rayleigh distribution the reader is referred to Johnson et al. (1994) [6]. Approximate Maximum Likelihood Estimator (MLE) of the Scale Parameter of the Rayleigh Distribution with Censoring sample was discussed by Balakrishnan. N [1].

Oxytocin is mammalian neurohypophysial nonapeptide hormone secreted by the posterior pituitary gland revealed to perform vital roles in numerous perceiving tasks. For example oxytocin behaves as a neuromodulator, and has been shown to be involved in stress, anxiety, trust, empathy, social recognition, orgasm, parturition, lactation, maternal behaviors, and mother-child and pair bonding [2-4], [7] and [10-12]. Oxytocin in biotic fluids has been measured by radioimmunoassay, enzyme immunoassay, high performance liquid chromatography (HPLC), and liquid chromatography (LC) plus tandem mass spectrometry (MS/MS). In-tube solid-phase micro extraction (SPME) using an open tubular fused-silica capillary with an inner surface coating as the SPME device, is a simple method that can be easily coupled with LC [9].

The objective of this work is to analyze the on-line in-tube solid-phase microextraction coupled with liquid chromatography-tandem mass spectrometry (online in tube SPME LC-MS/MS) method in a fuzzy environment via estimate the fuzzy expected values and fuzzy variance for salivary excretion of oxytocin through fuzzy Rayleigh distribution by finding the parameter of the Rayleigh distribution in the method of MLE.

II. Notations

| | |
|-----------------------|---|
| β | – Scale Parameter of Rayleigh distribution. |
| $\bar{\beta}(\alpha)$ | – Alpha cut of the scale parameter |
| $E(X)$ | – Mean of X |
| $V(X)$ | – Variance of X |
| $\bar{E}(X)$ | – Fuzzy Mean of X |
| $\bar{V}(X)$ | – Fuzzy Variance of X |

III. Materials And Methods

Rayleigh Distribution:

The cumulative function of random variable X follows Rayleigh distribution is denoted by $X \sim R(x, \beta)$ and it is defined by

$$F(x) = P(X \leq x) = 1 - e^{-\frac{1}{2}\left(\frac{x}{\beta}\right)^2}, \text{ where } 0 \leq x \leq \infty, \beta > 0.$$

The probability density function (pdf) is given by

$$\begin{aligned} f(x) &= \frac{d}{dx} F(x) \\ &= \frac{x}{\beta^2} e^{-\frac{1}{2}\left(\frac{x}{\beta}\right)^2} \end{aligned}$$

The mean value of Rayleigh distribution is $E(X) = \int_0^{\infty} x f(x) dx = \beta \sqrt{\frac{\pi}{2}}$

The variance value of Rayleigh distribution is $V(X) = E(X^2) - [E(X)]^2 = \beta^2 \left(2 - \frac{\pi}{2}\right)$.

Parameter Estimation by MLE

Here we extant the method of Maximum Likelihood Estimation as this technique gives simpler estimation as compared to the Method of moments and the Local frequency ratio method of estimation. Now we are estimate the parameter of the Rayleigh distribution from which the sample comes. Let x_1, x_2, \dots, x_n be a random sample of n observations from the Rayleigh population with pdf

$$f(x) = \frac{x}{\beta^2} e^{-\frac{1}{2}\left(\frac{x}{\beta}\right)^2}, x > 0.$$

The Likelihood function for this sample is

$$L = \prod_{i=1}^n f(x) = \prod_{i=1}^n \frac{x}{\beta^2} e^{-\frac{1}{2}\left(\frac{x}{\beta}\right)^2} = \frac{\prod_{i=1}^n x_i}{(\beta^2)^n} \cdot e^{-\frac{1}{2}\sum_{i=1}^n \left(\frac{x_i}{\beta}\right)^2}$$

$$\log L = \log \left(\frac{\prod_{i=1}^n x_i}{(\beta^2)^n} \right) - \frac{1}{2} \sum_{i=1}^n \left(\frac{x_i}{\beta} \right)^2$$

$$\log L = \log \left(\prod_{i=1}^n x_i \right) - \log \left((\beta^2)^n \right) - \frac{1}{2} \sum_{i=1}^n \left(\frac{x_i}{\beta} \right)^2$$

$$\log L = \log \left(\prod_{i=1}^n x_i \right) - 2n \log \beta - \frac{1}{2} \sum_{i=1}^n \left(\frac{x_i}{\beta} \right)^2$$

The likelihood equation is $\frac{\partial \log L}{\partial \beta} = 0$.

$$\Rightarrow -\frac{2n}{\beta} + \frac{\sum_{i=1}^n x_i^2}{\beta^3} = 0.$$

$$\hat{\beta} = \sqrt{\frac{\sum_{i=1}^n x_i^2}{2n}} \quad \text{----- (1)}$$

Fuzzy Rayleigh Distribution

In life time applications, fickleness is not the loneattribute of vagueness. In many fields of application, owing tothe fuzziness of environment and the negligence of observers, it is sometimes impossible to obtain exact annotationsoflifetime [9]. The acquired lifetime data may be “contaminated” and wooly most of the time. In addition, constrained byhuman being and other wherewithal in experiment, mainly for novequipment’s, unusually long-life equipment’s, andnon-mass-production products, for which there is no comparative dependability information available, more often thannot, the lifetime is based upon subjective evaluation or rough estimate. That leads to the fuzziness of lifetime data. In the circumstance Rayleigh distribution consider with fuzzy rules.

Now consider the Rayleigh distribution with fuzzy parameter $\bar{\beta}$ that is swapped with β . The probability of a random variable X follows Fuzzy Rayleigh distribution is denoted by $X \sim \text{FR}(x; \bar{\beta})$ the fuzzy probability density function of a random variable $X \sim \text{FR}(x; \bar{\beta})$ is defined by

$$f(x, \bar{\beta}) = \{f(x)[\alpha], \mu_{f(x)} \mid f(x)[\alpha] = [f_{\min}(x)[\alpha], f_{\max}(x)[\alpha]], \mu_{f(x)} = \alpha\}$$

$$f_{\min}(x)[\alpha] = \inf \{f(x, \beta)(\alpha) \mid \beta \in \bar{\beta}(\alpha)\},$$

$$f_{\max}(x)[\alpha] = \sup \{f(x, \beta)(\alpha) \mid \beta \in \bar{\beta}(\alpha)\}.$$

$$f(x, \bar{\beta}) = \frac{x}{\bar{\beta}^2} e^{-\frac{1}{2}(\frac{x}{\bar{\beta}})^2}, x > 0, \bar{\beta} \in \bar{\beta}(\alpha)$$

The Mean value of FR distribution is given by

$$\bar{E}(X) = \{E(X)[\alpha], \mu_{E(X)} \mid E(X)[\alpha] = [E_{\min}(X)[\alpha], E_{\max}(X)[\alpha]], \mu_{E(X)} = \alpha\}$$

$$E_{\min}(X)[\alpha] = \inf \{E(X) \mid \beta \in \bar{\beta}(\alpha)\}$$

$$E_{\max}(X)[\alpha] = \sup \{E(X) \mid \beta \in \bar{\beta}(\alpha)\}$$

$$\bar{E}(X) = \bar{\beta} \sqrt{\frac{\pi}{2}}, \bar{\beta} \in \bar{\beta}(\alpha).$$

Let the life time random variable is a fuzzy random variable with p.d.f. $f(x, \bar{\beta})$, then the fuzzy variance is defined as follows

$$\bar{V}(X) = \{V(X)[\alpha], \mu_{V(X)} \mid V(X)[\alpha] = [V_{\min}(X)[\alpha], V_{\max}(X)[\alpha]], \mu_{V(X)} = \alpha\}$$

$$V_{\min}(X)[\alpha] = \inf \{V(X) \mid \beta \in \bar{\beta}(\alpha)\}$$

$$V_{\max}(X)[\alpha] = \sup \{V(X) \mid \beta \in \bar{\beta}(\alpha)\}$$

$$\bar{V}(X) = (\bar{\beta})^2 \left(2 - \frac{\pi}{2}\right)$$

IV. Result And Discussion

Let us consider the trial in Shujitsu University, School of Pharmacy, in Japan [8]. Tocalculate the salivary secretion ofoxytocin, 2mgmL⁻¹oxytocin solution was directed byfour bouquets (containing ca. 1.47 mg of oxytocin) into the adenoidal caves of 59 male volunteers.Salivawas collected by rinsing the mouth of each subject with water, followed by thecollectionofsalivasamplesinSalisofttubes containing polypropylene-

polyethylenesponge (Assist,Tokyo,Japan).After saliva samples were collected into Salisoft tubes containing polypropylene-polyethylene sponges, followed by ultracentrifugation with Amicon Ultra to eliminate the proteins. To eradicate salivary interfering substances such as mucin, the filtrate was extracted with MonoTip C18, a monolithic silica adsorbent packed into a micro-tip. The saliva samples were successfully analyzed without interference peaks using the established in-tube SPME LC–MS/MS method with MRM mode detection.Fig 4.1 shows the salivary excretion of oxytocin after intranasal oxytocin administration.

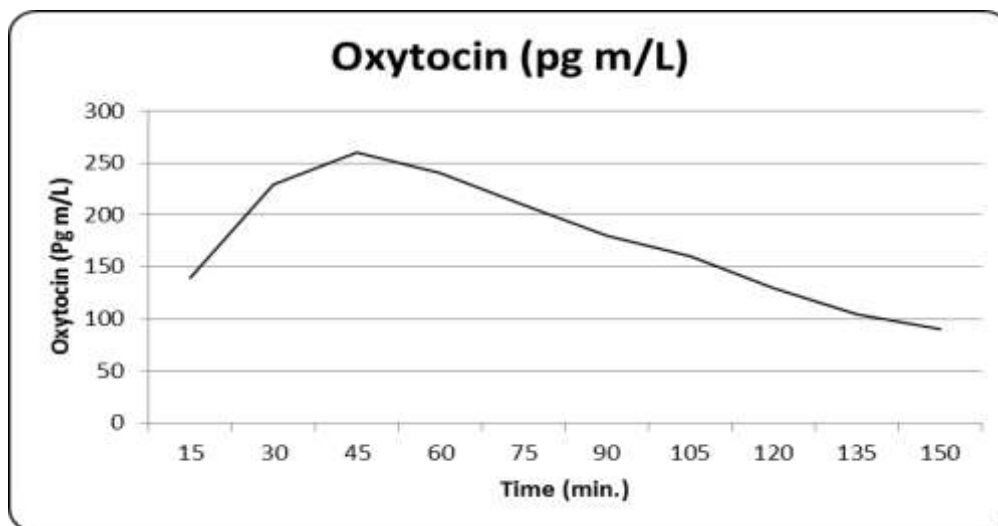


Fig. 4.1. Salivary excretion of Oxytocin

Based on the above observation sample the parameter of Rayleigh distribution by MLE (1) is $\beta = 65.812$. The corresponding fuzzy triangular numbers are [63.917, 65.812, 67.480]. The corresponding α -cut is

$\bar{\beta}(\alpha) = [63.967 + 1.845\alpha, 65.812, 67.480 - 1.668\alpha]$. The fuzzy mean values and fuzzy variance values for α -cuts are presented in Table 4.1.

Table 4.1. Mean and Variance values for alpha cuts

| Alpha Values | $E_{\min} [X](\alpha)$ | $E_{\max} [X](\alpha)$ | $V_{\min} [X](\alpha)$ | $V_{\max} [X](\alpha)$ |
|--------------|------------------------|------------------------|------------------------|------------------------|
| 0 | 80.171 | 84.574 | 1756.206 | 1954.401 |
| 0.05 | 80.286 | 84.469 | 1761.275 | 1949.573 |
| 0.1 | 80.402 | 84.365 | 1766.351 | 1944.751 |
| 0.15 | 80.518 | 84.260 | 1771.435 | 1939.935 |
| 0.2 | 80.633 | 84.156 | 1776.526 | 1935.124 |
| 0.25 | 80.749 | 84.051 | 1781.624 | 1930.320 |
| 0.3 | 80.864 | 83.946 | 1786.730 | 1925.522 |
| 0.35 | 80.980 | 83.842 | 1791.843 | 1920.730 |
| 0.4 | 81.096 | 83.737 | 1796.963 | 1915.944 |
| 0.45 | 81.211 | 83.633 | 1802.090 | 1911.164 |
| 0.5 | 81.327 | 83.528 | 1807.225 | 1906.389 |
| 0.55 | 81.443 | 83.424 | 1812.367 | 1901.621 |
| 0.6 | 81.558 | 83.319 | 1817.517 | 1896.859 |
| 0.65 | 81.674 | 83.215 | 1822.674 | 1892.102 |
| 0.7 | 81.789 | 83.110 | 1827.838 | 1887.352 |
| 0.75 | 81.905 | 83.006 | 1833.009 | 1882.608 |
| 0.8 | 82.021 | 82.901 | 1838.188 | 1877.869 |
| 0.85 | 82.136 | 82.797 | 1843.374 | 1873.137 |
| 0.9 | 82.252 | 82.692 | 1848.567 | 1868.410 |
| 0.95 | 82.367 | 82.588 | 1853.767 | 1863.690 |
| 1 | 82.483 | 82.483 | 1858.975 | 1858.975 |

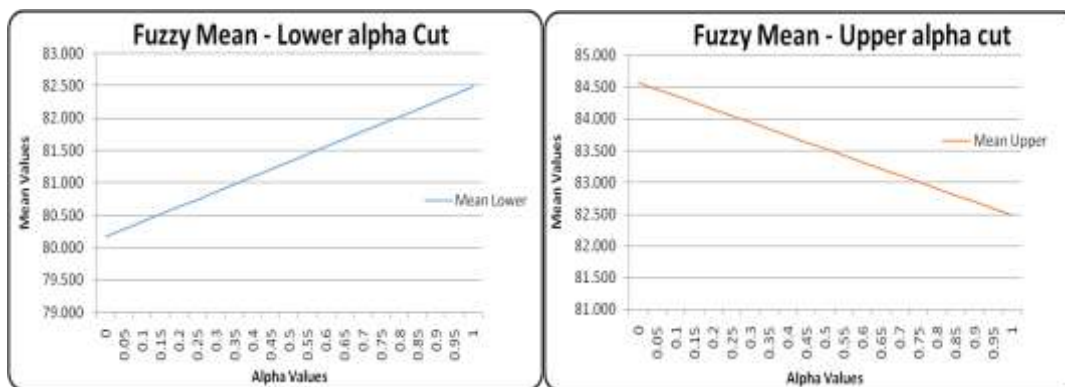


Fig. 4.2. Fuzzy Mean values of salivary excretion of Oxytocin for Lower & Upper alpha cuts

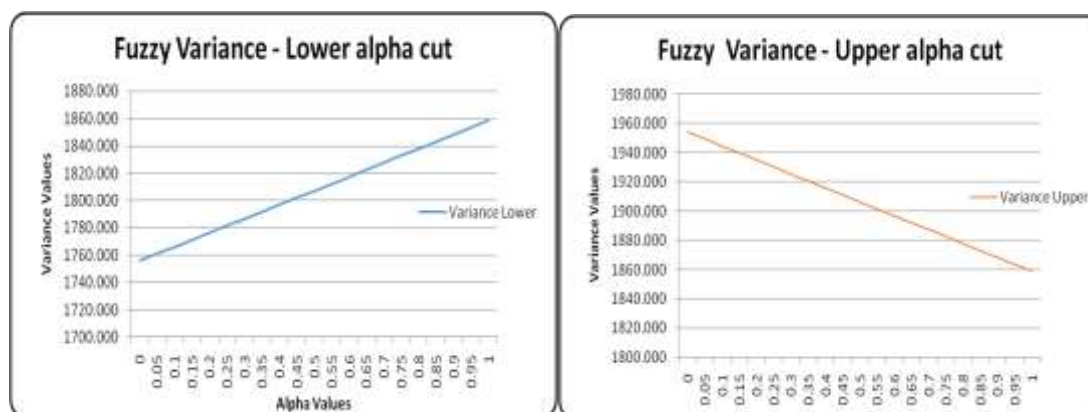


Fig. 4.3. Fuzzy Variance values of salivary excretion of Oxytocin for Lower & Upper alpha cuts

V. Conclusion

The parameter for Rayleigh distribution was calculated successfully by using MLE. The mean and variance values are estimated for the unremitting drawing out and concentration of oxytocin in saliva samples analysis using fuzzy Rayleigh distribution. Analyzing of fuzzy mean and variance shows that for lower alpha cuts has increasing expected salivary excretion than the upper alpha cuts. The fuzzy Rayleigh distribution model for investigation of oxytocin analyzed by online in tube SPME LC-MS/MS methods is very handy for drug examples and for impartial assessment of the biological belongings of oxytocin.

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