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**The Derivations of the Edgeworth Expansion:
A Tutorial**

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1 Background Material:

A general introduction to asymptotic approximations to distributions is given in Wallace [5] and the notation of his article will be used at least initially. The Central Limit Theorem has its limitations, especially when the sample size n , is small. The Edgeworth expansion is aimed at improving the approximation for such samples. The expansion is derived from a formal identity which relates the characteristic functions of two distributions. For simplicity we shall assume that both distributions are continuous and share the same domain on the real line. Let $F(x)$ be the CDF of the distribution to be approximated and let $\Psi(x)$ be the CDF of the distribution to be used in making the approximation. In general, $\Psi(x)$ need not be the normal distribution, but later it will be what we use to obtain the Edgeworth expansion. Let $\psi(t)$ be the characteristic function of $\Psi(x)$ and let $\gamma_1, \gamma_2, \dots$ be its cumulants. Then we know that $\psi(t)$ can be written as

$$\psi(t) = \exp\left(\sum_{r=1}^{\infty} \gamma_r \frac{(it)^r}{r!}\right) \quad (1.1)$$

where i is the complex number defined as $i = \sqrt{-1}$. It follows then that it is the case that

$$\psi(t) \exp\left(\sum_{r=1}^{\infty} -\gamma_r \frac{(it)^r}{r!}\right) = 1 \quad (1.2)$$

If we denote the characteristic function for the distribution $F(x)$ by $f(t)$ and its associated cumulants by $\kappa_1, \kappa_2, \dots$ then

$$f(t) = \exp\left(\sum_{r=1}^{\infty} \kappa_r \frac{(it)^r}{r!}\right) \quad (1.3)$$

Combining equation (1.3) and equation (1.2) the characteristic functions satisfy the formal identity

$$f(t) = \exp\left(\sum_{r=1}^{\infty} [\kappa_r - \gamma_r] \frac{(it)^r}{r!}\right) \psi(t) \quad (1.4)$$

Next quoting from Wallace [5]

If now, Ψ and all its derivatives vanish at the extreme range of x and exist for all x in that range, then by integration by parts, $(it)^r \psi(t)$ is the characteristic function of $(-1)^r \Psi^{(r)}(x)$. Introducing the differential operator D to represent differentiation with respect to x , the formal identity corresponds term-wise in any formal expansion to the formal identity

$$F(x) = \exp\left(\sum_{r=1}^{\infty} [\kappa_r - \gamma_r] \frac{(-D)^r}{r!}\right) \Psi(x) \quad (1.5)$$

One can formally and apparently construct a distribution with prescribed cumulants by choosing Ψ and formally expanding.

The most important developing function $\Psi(x)$ is a normal distribution and with that choice, the formal expansion has been given earlier by Chebyshev,[2] Edgeworth [3] and Charlier [1].

2 Developing the Expansion

Let X_1, X_2, \dots, X_n be an i.i.d. sample from a distribution with $E(X_i) = \mu = \kappa_1$ and $Var(X_i) = \sigma^2 = \kappa_2$ and higher cumulants denoted by κ_r and with $-\infty < X < \infty$. Our objective will be to develop an approximation for the distribution of the statistic

$$Y_n = \frac{\sqrt{n}(\bar{X}_n - \mu)}{\sigma} = \frac{1}{\sqrt{n}} \sum_{i=1}^n \left(\frac{X_i - \mu}{\sigma}\right) \quad (2.1)$$

where $Y_n \xrightarrow{L} N(0, 1)$ as $n \rightarrow \infty$ which is better than $N(0, 1)$ when n is not too large. In our development we shall use the standard normal distribution with CDF $\Phi(x)$ as the approximating distribution $\Psi(x)$ of equation (1.5). For the standard normal it is well known that the cumulants are $\gamma_1 = 0, \gamma_2 = 1$ and $\gamma_r = 0, r \geq 3$. Next we must find the characteristic function for Y_n . Let $g(t)$ be the characteristic function for the distribution of the X_i , Then it is well known by properties of the characteristic function that for any X_i the characteristic function of

$$\frac{1}{\sqrt{n}} \left(\frac{X_i - \mu}{\sigma} \right)$$

is

$$\begin{aligned} h(t) &= \exp \left(-t \frac{\mu}{\sqrt{n}\sigma} \right) g \left(\frac{1}{\sqrt{n}\sigma} t \right) \\ &= \exp \left(-\left[\frac{\mu}{\sqrt{n}\sigma} \right](it) + \left[\frac{\kappa_1}{\sqrt{n}\sigma} \right](it) + \frac{\kappa_2}{n\sigma^2} \frac{(it)^2}{2!} + \sum_{r=3}^{\infty} \frac{\kappa_r}{(\sqrt{n}\sigma)^r} \frac{(it)^r}{r!} \right) \end{aligned} \quad (2.2)$$

Since $\kappa_1 = \mu$ and $\kappa_2 = \sigma^2$ we see that the $h(t)$ reduces to

$$h(t) = \exp \left(\frac{1}{2n} (it)^2 + \sum_{r=3}^{\infty} \frac{\kappa_r}{(\sqrt{n}\sigma)^r} \frac{(it)^r}{r!} \right)$$

Finally, the characteristic function of Y_n is just equal to $h(t)^n$ so we have

$$w(t) = h(t)^n = \exp \left(\frac{1}{2} (it)^2 + \sum_{r=3}^{\infty} \frac{1}{n^{r/2-1}} \frac{\kappa_r}{\sigma^r} \frac{(it)^r}{r!} \right)$$

Next we note that for the standard normal distribution, $\gamma_1 = 0, \gamma_2 = 1$ and $\gamma_r = 0, r \geq 3$, so that the characteristic function of the standard normal is $\exp[-(1/2)(it)^2]$. Plugging into equation (1.5) leads to the expression

$$F_{Y_n}(x) = \exp \left(\sum_{r=3}^{\infty} \frac{1}{n^{r/2-1}} \frac{\kappa_r}{\sigma^r} \frac{(-D)^r}{r!} \right) \Phi(x) \quad (2.3)$$

where as has been previously stated D is the differential operator, $D^r = d^r/dx^r$. The next step is to expand the exponential function in its MacLaurin

expansion; that is

$$\begin{aligned}
\exp\left(\sum_{r=3}^{\infty} \frac{1}{n^{r/2-1}} \frac{\kappa_r}{\sigma^r} \frac{(-D)^r}{r!}\right) &= 1 + \left(\sum_{r=3}^{\infty} \frac{1}{n^{r/2-1}} \frac{\kappa_r}{\sigma^r} \frac{(-D)^r}{r!}\right) \\
&+ \frac{1}{2!} \left(\sum_{r=3}^{\infty} \frac{1}{n^{r/2-1}} \frac{\kappa_r}{\sigma^r} \frac{(-D)^r}{r!}\right)^2 \\
&+ \frac{1}{3!} \left(\sum_{r=3}^{\infty} \frac{1}{n^{r/2-1}} \frac{\kappa_r}{\sigma^r} \frac{(-D)^r}{r!}\right)^3 \\
&+ \frac{1}{4!} \left(\sum_{r=3}^{\infty} \frac{1}{n^{r/2-1}} \frac{\kappa_r}{\sigma^r} \frac{(-D)^r}{r!}\right)^4 \\
&+ \dots
\end{aligned} \tag{2.4}$$

At this point each term in the formal relationship is expanded and the terms gathered in powers of $1/\sqrt{n}$. This is a daunting task even with today's modern computer tools for doing the algebra involved. This can be accomplished using a program like Maple or Mathematica. Written in terms of the cumulants and powers of D the first few terms of this expansion are,

$$\begin{aligned}
&1 - \frac{1}{\sqrt{n}} \left[\frac{\kappa_3}{6\sigma^3} D^3 \right] + \frac{1}{n} \left[\frac{1}{72} \left(\frac{\kappa_3}{\sigma^3}\right)^2 D^6 + \frac{1}{24} \left(\frac{\kappa_4}{\sigma^4}\right) D^4 \right] \\
&- \frac{1}{n^{3/2}} \left[\frac{1}{144} \left(\frac{\kappa_3}{\sigma^3}\right) \left(\frac{\kappa_4}{\sigma^4}\right) D^7 + \frac{1}{120} \left(\frac{\kappa_5}{\sigma^5}\right) D^5 + \frac{1}{1296} \left(\frac{\kappa_3}{\sigma^3}\right)^3 D^9 \right] \\
&+ \frac{1}{n^2} \left[\frac{1}{1152} \left(\frac{\kappa_4}{\sigma^4}\right)^2 D^8 + \frac{1}{720} \left(\frac{\kappa_6}{\sigma^6}\right) D^6 + \frac{1}{1728} \left(\frac{\kappa_3}{\sigma^3}\right)^2 \left(\frac{\kappa_4}{\sigma^4}\right) D^{10} \right. \\
&\left. + \frac{1}{31104} \left(\frac{\kappa_3}{\sigma^3}\right)^4 D^{12} + \frac{1}{720} \left(\frac{\kappa_3}{\sigma^3}\right) \left(\frac{\kappa_5}{\sigma^5}\right) D^8 \right] + \mathcal{O}\left(\frac{1}{n^{5/2}}\right)
\end{aligned} \tag{2.5}$$

When this lengthy expression is applied as an operator on $\Phi(x)$ the result is the Edgeworth expansion. This can be written in a number of ways, where the effects of the differentiation can be expressed most simply in terms of the Hermite polynomials $He_n(x)$ defined by the relationship

$$He_n(x) = (-1)^n \frac{\phi^{(n)}(x)}{\phi(x)} \quad \text{or} \quad \phi^{(n)}(x) = (-1)^n \phi(x) He_n(x)$$

where $\phi(x) = d\Phi(x)/dx$. The polynomial functions, $He_n(x)$ can be found recursively as

$$He_{n+1}(x) = xHe_n(x) - nHe_{n-1}(x) \text{ for } n \geq 1 \quad (2.6)$$

given that $He_0 = 1$ and $He_1 = x$. Application of equation(2.5) to the function $\Phi(x)$ leads to the Expansion for the distribution of the random variable Y_n ,

$$\begin{aligned} F_{Y_n}(x) = & \Phi(x) - \frac{1}{\sqrt{n}} \left[\frac{\kappa_3}{6\sigma^3} He_2(x) \right] \phi(x) + \frac{1}{n} \left[\frac{1}{72} \left(\frac{\kappa_3}{\sigma^3} \right)^2 (-He_5(x)) \right. \\ & + \left. \frac{1}{24} \left(\frac{\kappa_4}{\sigma^4} \right) (-He_3(x)) \right] \phi(x) - \frac{1}{n^{3/2}} \left[\frac{1}{144} \left(\frac{\kappa_3}{\sigma^3} \right) \left(\frac{\kappa_4}{\sigma^4} \right) He_6(x) \right. \\ & + \left. \frac{1}{120} \left(\frac{\kappa_5}{\sigma^5} \right) He_4(x) + \frac{1}{1296} \left(\frac{\kappa_3}{\sigma^3} \right)^3 He_8(x) \right] \phi(x) \\ & + \frac{1}{n^2} \left[\frac{1}{1152} \left(\frac{\kappa_4}{\sigma^4} \right)^2 (-He_7(x)) + \frac{1}{720} \left(\frac{\kappa_6}{\sigma^6} \right) (-He_5(x)) \right. \\ & + \frac{1}{1728} \left(\frac{\kappa_3}{\sigma^3} \right)^2 \left(\frac{\kappa_4}{\sigma^4} \right) (-He_9(x)) + \frac{1}{31104} \left(\frac{\kappa_3}{\sigma^3} \right)^4 (-He_{11}(x)) \\ & + \left. \frac{1}{720} \left(\frac{\kappa_3}{\sigma^3} \right) \left(\frac{\kappa_5}{\sigma^5} \right) (-He_7(x)) \right] \phi(x) \\ & + \mathcal{O}\left(\frac{1}{n^{5/2}}\right) \end{aligned} \quad (2.7)$$

Or after adjusting for the minus signs on the odd ordered Hermite polynomials, the expansion becomes

$$\begin{aligned}
F_{Y_n}(x) &= \Phi(x) - \frac{1}{\sqrt{n}} \left[\frac{\kappa_3}{6\sigma^3} He_2(x) \right] \phi(x) - \frac{1}{n} \left[\frac{1}{72} \left(\frac{\kappa_3}{\sigma^3} \right)^2 He_5(x) \right. \\
&\quad \left. + \frac{1}{24} \left(\frac{\kappa_4}{\sigma^4} \right) He_3(x) \right] \phi(x) - \frac{1}{n^{3/2}} \left[\frac{1}{144} \left(\frac{\kappa_3}{\sigma^3} \right) \left(\frac{\kappa_4}{\sigma^4} \right) He_6(x) \right. \\
&\quad \left. + \frac{1}{120} \left(\frac{\kappa_5}{\sigma^5} \right) He_4(x) + \frac{1}{1296} \left(\frac{\kappa_3}{\sigma^3} \right)^3 He_8(x) \right] \phi(x) \\
&\quad - \frac{1}{n^2} \left[\frac{1}{1152} \left(\frac{\kappa_4}{\sigma^4} \right)^2 He_7(x) + \frac{1}{720} \left(\frac{\kappa_6}{\sigma^6} \right) He_5(x) \right. \\
&\quad \left. + \frac{1}{1728} \left(\frac{\kappa_3}{\sigma^3} \right)^2 \left(\frac{\kappa_4}{\sigma^4} \right) He_9(x) + \frac{1}{31104} \left(\frac{\kappa_3}{\sigma^3} \right)^4 He_{11}(x) \right. \\
&\quad \left. + \frac{1}{720} \left(\frac{\kappa_3}{\sigma^3} \right) \left(\frac{\kappa_5}{\sigma^5} \right) He_7(x) \right] \phi(x) \\
&\quad + \mathcal{O}\left(\frac{1}{n^{5/2}}\right)
\end{aligned} \tag{2.8}$$

Some sources express the expansion in terms of the central moments of the the distribution of X and these first few of the relationships between the cumulants and the central moments are

$$\begin{aligned}
\kappa_1 &= \mu \\
\kappa_2 &= \mu_2 = \sigma^2 \\
\kappa_3 &= \mu_3 \\
\kappa_4 &= \mu_4 - 3\mu_2^2 \\
\kappa_5 &= \mu_5 - 10\mu_2\mu_3
\end{aligned} \tag{2.9}$$

Given that $\kappa_3 = \mu_3$ and that $He_2(x) = x^2 - 1$ the term of order $1/\sqrt{n}$ in equation (2.7) can be written as

$$-\frac{1}{\sqrt{n}} \left[\frac{\mu_3}{6\sigma^3} (x^2 - 1) \right] \phi(x) = \frac{1}{\sqrt{n}} \left[\frac{\mu_3}{6\sigma^3} (1 - x^2) \right] \phi(x) \tag{2.10}$$

Similarly, given that $He_3 = x^3 - 3x$, $He_4 = x^4 - 6x^2 + 3$ and $He_5 = x^5 - 10x^3 + 15x$ the second term in the expansion (term of order $1/n$) becomes

$$-\frac{1}{n} \left[\frac{1}{72} \left(\frac{\mu_3}{\sigma^3} \right)^2 (x^5 - 10x^3 + 15x) + \frac{1}{24} \left(\frac{\mu_4 - 3\mu_2^2}{\sigma^4} \right) (x^3 - 3x) \right] \phi(x) \tag{2.11}$$

The terms given in equations (2.10) and (2.11) correspond to those given by Lehmann [4] on pages 81 and 83.

3 Remarks

In this report we have developed the Edgeworth expansion for the case of a random variable Y_n which has the property that

$$Y_n = \frac{\sqrt{n}(\bar{X}_n - \mu)}{\sigma} \xrightarrow{L} N(0, 1) \text{ as } n \rightarrow \infty$$

where \bar{X}_n is the sample mean of an i.i.d. sample from a distribution with cumulants $\kappa_1, \kappa_2, \dots$. If the moment generating function, $M(t)$ of the distribution of a random variable is known then the cumulant generating function is

$$K(t) = \ln(M(t)) = \kappa_1 t + \frac{\kappa_2}{2!} t^2 + \frac{\kappa_3}{3!} t^3 + \dots$$

The cumulants are then found by differentiating repeatedly by t and evaluating the respective derivatives at $t = 0$. From a computational point of view, the recursion given in equation (2.6) is very handy, since for any particular value of x , the numerical values of the polynomials $He_r(x)$ can be found without actually finding the polynomial form.

Appendix

In this appendix we give the expressions for the terms of order $n^{-5/2}$ and n^{-3} . These are long and involve high order cumulants and high order Hermite polynomials.

$$\begin{aligned} & -\frac{\phi(x)}{n^{5/2}} \left[\frac{1}{933120} \left(\frac{\kappa_3}{\sigma^3}\right)^5 He_{14}(x) + \frac{1}{31104} \left(\frac{\kappa_3}{\sigma^3}\right)^3 \left(\frac{\kappa_4}{\sigma^4}\right) He_{12}(x) \right. \\ & + \frac{1}{8640} \left(\frac{\kappa_3}{\sigma^3}\right)^2 \left(\frac{\kappa_5}{\sigma^5}\right) He_{10}(x) + \frac{1}{6912} \left(\frac{\kappa_3}{\sigma^3}\right) \left(\frac{\kappa_4}{\sigma^4}\right)^2 He_{10}(x) \\ & + \frac{1}{4320} \left(\frac{\kappa_3}{\sigma^3}\right) \left(\frac{\kappa_6}{\sigma^6}\right) He_8(x) + \frac{1}{2880} \left(\frac{\kappa_4}{\sigma^4}\right) \left(\frac{\kappa_5}{\sigma^5}\right) He_8(x) \\ & \left. - \frac{1}{5040} \left(\frac{\kappa_7}{\sigma^7}\right) He_6(x) \right] \end{aligned}$$

and

$$\begin{aligned}
& -\frac{\phi(x)}{n^3} \left[\frac{1}{33592320} \left(\frac{\kappa_3}{\sigma^3}\right)^6 He_{17}(x) + \frac{1}{746496} \left(\frac{\kappa_3}{\sigma^3}\right)^4 \left(\frac{\kappa_4}{\sigma^4}\right) He_{15}(x) \right. \\
& + \frac{1}{155520} \left(\frac{\kappa_3}{\sigma^3}\right)^3 \left(\frac{\kappa_5}{\sigma^5}\right) He_{13}(x) + \frac{1}{82944} \left(\frac{\kappa_3}{\sigma^3}\right)^2 \left(\frac{\kappa_4}{\sigma^4}\right)^2 He_{13} \\
& + \frac{1}{51840} \left(\frac{\kappa_3}{\sigma^3}\right)^2 \left(\frac{\kappa_6}{\sigma^6}\right) He_{11}(x) + \frac{1}{17280} \left(\frac{\kappa_3}{\sigma^3}\right)^4 \left(\frac{\kappa_6}{\sigma^6}\right) He_{11}(x) \\
& + \frac{1}{17280} \left(\frac{\kappa_3}{\sigma^3}\right) \left(\frac{\kappa_4}{\sigma^4}\right) \left(\frac{\kappa_6}{\sigma^6}\right) He_{11}(x) + \frac{1}{82944} \left(\frac{\kappa_4}{\sigma^4}\right)^3 He_{11}(x) \\
& - \frac{1}{30240} \left(\frac{\kappa_3}{\sigma^3}\right) \left(\frac{\kappa_7}{\sigma^7}\right) He_9(x) + \frac{1}{17280} \left(\frac{\kappa_4}{\sigma^4}\right) \left(\frac{\kappa_6}{\sigma^6}\right) He_9(x) \\
& \left. + \frac{1}{28800} \left(\frac{\kappa_5}{\sigma^5}\right)^2 He_9(x) + \frac{1}{40320} \left(\frac{\kappa_8}{\sigma^8}\right) He_7(x) \right]
\end{aligned}$$

References

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