

(1)

Some of my white board writing from week 11

1.) I focused on the relationships between moments and cumulants.

The moment generating function (MGF) $M_X(t) = E(\exp(tX))$ has the obvious property $M_X^{(r)}(t)|_{t=0} = E(X^r) = \mu_r'$ with μ_r' being a short-hand notation for the raw moment $E(X^r)$ (as opposed to $\mu_r = E(X - \mu_1')^r$ used to denote the central moments).

We have by simple Taylor expansion then

$$M_X(t) = 1 + \mu_1' t + \mu_2' \frac{t^2}{2!} + \dots + \mu_r' \frac{t^r}{r!} + O(t^{r+1})$$

as $t \rightarrow 0$. The cumulant generating function is $K_X(t) = \log M_X(t)$ (i.e. $e^{K_X(t)} = M_X(t)$ holds).

Since $K_X(0) = 0$ we get the Taylor expansion

$$K_X(t) = \kappa_1 t + \kappa_2 \frac{t^2}{2!} + \dots + \kappa_r \frac{t^r}{r!} + O(t^{r+1}).$$

Substituting in the relation $\frac{\uparrow}{\text{we get:}}$

$$e^{\kappa_1 t} e^{\kappa_2 \frac{t^2}{2!}} e^{\kappa_3 \frac{t^3}{3!}} \dots = 1 + \mu_1' t + \mu_2' \frac{t^2}{2!} + \dots \quad (1)$$

and expanding the exponents on the LHS we get.

$$(1 + \kappa_1 t + \kappa_1^2 \frac{t^2}{2!} + \dots) (1 + \kappa_2 \frac{t^2}{2!} + \frac{1}{2!} (\kappa_2 \frac{t^2}{2!})^2 + \dots) (1 + \kappa_3 \frac{t^3}{3!} + \frac{1}{2!} (\kappa_3 \frac{t^3}{3!})^2 + \dots)$$

Equating the coefficients in front of the powers of t in the LHS and RHS of (1) we get:

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$$R_1 = \mu_1'$$

$$R_2 + R_1^2 = \mu_2' \rightarrow R_2 = \mu_2' - (\mu_1')^2 = E(X)^2 - (EX)^2 = \text{Var} X = \sigma^2$$

$$R_3 = 2\mu_1'^3 - 3\mu_1'\mu_2' + \mu_3'$$

$$R_4 = -6(\mu_1')^4 + 12(\mu_1')^2\mu_2' - 3(\mu_2')^2 - 4\mu_1'\mu_3' + \mu_4'$$

In particular for $N(0,1)$ we get $R_4 = 0$ since $\mu_4' = 3(\mu_2')^2$ holds for $N(0,1)$

To summarize: - the first cumulant is the first moment, the second cumulant is the variance, the third cumulant is the skewness, the fourth cumulant is the kurtosis.

We also introduce:

- STANDARDIZED SKEWNESS: $\beta_3 = R_3 / (R_2^{3/2})$

- Standardized kurtosis $\beta_4 = R_4 / (R_2^2)$

These are useful in the forthcoming Edgeworth expansions.

2) Then I spoke about Cramer's condition and its importance and discussed the details in the formulation of Theorem 8.1. I specifically stressed that

the constants $G_1(F), G_2(F), G_3(F)$ can be expressed by using β_3 and β_4 . Namely: $G_1(F) = \frac{1}{6} \frac{R_3}{\sigma^3} = \frac{1}{6} \beta_3$; $G_2(F) = \frac{1}{24} \beta_4$

$$G_3(F) = \frac{G_2^2(F)}{2} = \frac{\beta_3^2}{72}$$

In section 8.3.2, when discussing formula (19), I also wrote all the first 6 Hermite polynomials explicitly.

$$H_1(z) = z; H_2(z) = z^2 - 1; H_3(z) = z^3 - 3z; H_4(z) = z^4 - 6z^2 + 3; \\ H_5(z) = z^5 - 10z^3 + 15z; H_6(z) = z^6 - 15z^4 + 45z^2 - 15$$

3) Next, I discussed the Cornish-Fisher expansion. I gave heuristic justification to the formula given in Theorem 8.2. It goes as follows:

Since $Z_n = \sqrt{n}(\bar{X} - \mu) \approx$ standard normal, the quantile Z_α , which is theoretically defined as the solution of $F_{Z_n}(Z_\alpha) = 1 - \alpha$, should be in a vicinity of the u_α quantile defined as the solution of $\Phi(u_\alpha) = 1 - \alpha$. I also discussed some "famous" u_α quantiles such as

$$u_{0.01} = 2.326, \quad u_{0.025} = 1.96, \quad u_{0.05} = 1.645, \quad u_{0.1} = 1.28$$

Then the argument, by using Taylor expansion, goes as follows:

$$1 - \alpha = F_{Z_n}(Z_\alpha) \stackrel{\text{using Theorem 8.1}}{\approx} \Phi(Z_\alpha) - \frac{G_1(F)p_1(Z_\alpha)\phi(Z_\alpha)}{\sqrt{n}} - \frac{G_2(F)p_2(Z_\alpha) + G_3(F)p_3(Z_\alpha)}{n}$$

$\approx \Phi(u_\alpha) + \phi(u_\alpha)(Z_\alpha - u_\alpha) + \phi(u_\alpha)[\text{polynomials containing } Z_\alpha, u_\alpha]$
 apply Taylor by expanding "everywhere" around u_α

Since $\Phi(u_\alpha) = 1 - \alpha$, we cancel with $(1 - \alpha)$ on the LHS

and get $\phi(u_\alpha) \underbrace{[\text{some polynomials of } Z_\alpha, u_\alpha]}_{\neq 0} = 0$

Set this $= 0$ and express Z_α by using u_α from the resulting relation. In this way we finally obtain the expression given in Theorem 8.2

$$Z_\alpha = u_\alpha + \frac{(u_\alpha^2 - 1)p_3}{6\sqrt{n}} + \frac{(u_\alpha^3 - 3u_\alpha)p_4}{24n} - \frac{(2u_\alpha^3 - 5u_\alpha)p_3^2}{36n} + o(n^{-1})$$

4) I then applied this Theorem to illustrate the power and accuracy of the Cornish-Fisher expansion on an example given on p. 69:

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Example from p.69 in detail:

$\frac{W_n}{n}$ is an average of n i.i.d. squared standard normals.

Hence the CLT will give us: (Note: for χ_1^2 r.v. $E\chi_1^2 = 1, \text{Var}\chi_1^2 = 2$)

$\frac{\sqrt{n} \left(\frac{W_n}{n} - 1 \right)}{\sqrt{2}} = \frac{W_n - n}{\sqrt{2n}}$ is about standard normal (this is the first order asymptotics). Then since

$P(W_n < z_\alpha) = P\left(\frac{W_n - n}{\sqrt{2n}} < \frac{z_\alpha - n}{\sqrt{2n}}\right)$ we know that

$\frac{z_\alpha - n}{\sqrt{2n}}$ should be "close" to u_α so then $\underline{z}_\alpha = n + \sqrt{2n} u_\alpha$ is the first order approximation of the $(1-\alpha)100\%$ quantile of W_n .

To improve it by using higher order Cornish-Fisher expansion we proceed as follows:

The mgf of the χ_n^2 random variable (denoted generically as X here) is known to be $M_X(t) = (1-2t)^{-\frac{n}{2}}$

Hence $K_X(t) = -\frac{n}{2} \log(1-2t)$ and we get:

$$K'_X(t) = \frac{n}{1-2t}, \quad K''_X(t) = \frac{2n}{(1-2t)^2}, \quad K'''_X(t) = \frac{8n}{(1-2t)^3}, \quad K^{(IV)}_X(t) = \frac{48n}{(1-2t)^4}$$

Hence $K'_X(0) = n, K''_X(0) = 2n, K'''_X(0) = 8n, K^{(IV)}_X(0) = 48n$

In our case we only need to specialise this for χ_1^2 random variable so we get $K'_X(0) = 1, K''_X(0) = 2, K'''_X(0) = 8, K^{(IV)}_X(0) = 48$

$$\text{Then } \gamma_3 = \frac{8}{2^{3/2}} = 2\sqrt{2}, \quad \gamma_4 = \frac{48}{2^2} = 12$$

$$\text{This leads to } \eta \approx n + \sqrt{2n} \left[u_\alpha + \frac{(u_\alpha^2 - 1)2\sqrt{2}}{6\sqrt{n}} + \frac{(u_\alpha^3 - 3u_\alpha)12}{24n} - \frac{(u_\alpha^3 - 5u_\alpha)8}{36n} \right]$$

We apply these approximations e.g., for $n = 5$ to get:

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$$5 + \sqrt{2 \times 5} \times 2.326 = 12.36$$

$$5 + \sqrt{2 \times 5} \left(2.326 + \frac{2.326^2 - 1}{6\sqrt{5}} \cdot 2\sqrt{2} \right) = 15.296$$

$$5 + \sqrt{2 \times 5} \left(2.326 + \frac{2.326^2 - 1}{6\sqrt{5}} 2\sqrt{2} + \frac{(2.326^3 - 3 \times 2.326) 12}{24 \times 5} - \frac{(2 \times 2.326^3 - 5 \times 2.326^2)}{36 \times 5} \right) = 15.16$$

The ~~true~~ value of the quantile is 15.09 and we see the increasing precision popping up when we increase the order of the approximation.

5.) I then started discussing the idea of the saddlepoint method. I believe that the derivations, as presented in the lecture notes, are detailed enough and I did not write anything specific on the white board. I still need to go through pages 72-73 of this lecture at the beginning of lecture in week 12.