



On the moments of a distribution defined by the Gaussian polynomials

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Abstract

An alternative method is presented to compute the moments of the probability distribution defined by the Gaussian polynomials. It computes the cumulants first.

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Di Bucchianico (1999) has considered the probability distribution defined by the (probability) generating function

$$F_{m,n}(q) = \frac{\begin{bmatrix} m+n \\ m \end{bmatrix}_q}{\begin{bmatrix} m+n \\ m \end{bmatrix}},$$

with the Gaussian polynomials

$$\begin{bmatrix} m+n \\ m \end{bmatrix}_q = \frac{(1-q)(1-q^2)\cdots(1-q^{m+n})}{(1-q)(1-q^2)\cdots(1-q^m)(1-q)(1-q^2)\cdots(1-q^n)}.$$

He discussed several methods to compute the moments of this distribution.

In this short note, I want to draw the attention of the reader to another (potentially superior) method that is due to Panny (1986). This method first computes the *cumulants* and translates them into the *moments* by a standard formula. This yields *explicit* formulæ for all the moments.

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Note that

$$F_{m,n}(q) = \frac{g_1(q) \cdots g_{m+n}(q)}{g_1(q) \cdots g_m(q)g_1(q) \cdots g_n(q)}$$

with

$$g_k(q) = \frac{1 - q^k}{k(1 - q)},$$

where $g_k(q)$ is a probability generating function of a random variable X_k . The function $\log g_k(e^{it})$ is the generating function of the cumulants.¹ Panny has computed that as

$$\log g_k(e^{it}) = \frac{(k - 1)it}{2} + \sum_{j \geq 1} \frac{B_{2j}}{2j} (k^{2j} - 1) \frac{(it)^{2j}}{(2j)!}, \quad |t| < \frac{2\pi}{k},$$

where B_i denotes the i th Bernoulli number.

Reading off coefficients we find the cumulants of X_k as

$$\kappa_1 = \frac{k - 1}{2}, \quad \kappa_{2r} = \frac{(k^{2r} - 1)B_{2r}}{2r}, \quad \kappa_{2r+1} = 0, \quad r = 1, 2, 3, \dots$$

Since

$$\log F_{m,n}(e^{it}) = \sum_{k=1}^{m+n} \log g_k(e^{it}) - \sum_{k=1}^m \log g_k(e^{it}) - \sum_{k=1}^n \log g_k(e^{it}),$$

the cumulants $\kappa_r^{m,n}$ are obtained by summing the κ_r 's.

So we get

$$\begin{aligned} \kappa_1^{m,n} &= \frac{\binom{m+n}{2}}{2} - \frac{\binom{m}{2}}{2} - \frac{\binom{n}{2}}{2} = \frac{mn}{2}, \\ \kappa_{2r}^{m,n} &= \frac{B_{2r}}{2r(2r + 1)} (B_{2r+1}(m + n + 1) - B_{2r+1}(m + 1) - B_{2r+1}(n + 1)), \\ \kappa_{2r+1}^{m,n} &= 0. \end{aligned}$$

Here is a little list:

$$\begin{aligned} \kappa_1^{m,n} &= \frac{mn}{2}, \\ \kappa_2^{m,n} &= \frac{mn(m + n + 1)}{12}, \\ \kappa_4^{m,n} &= -\frac{mn(m + n + 1)(m(m + 1) + mn + n(n + 1))}{120}. \end{aligned}$$

¹ We write $\log x$ for the (natural) logarithm with base e .

The formula to compute the moments μ_r from this is

$$\mu_r = \sum_{\pi_1+2\pi_2+\dots+r\pi_r=r} \left(\frac{\kappa_1}{1!}\right)^{\pi_1} \left(\frac{\kappa_2}{2!}\right)^{\pi_2} \dots \left(\frac{\kappa_r}{r!}\right)^{\pi_r} \frac{r!}{\pi_1!\pi_2!\dots\pi_r!},$$

which is a finite sum since $\pi_k \in \{0, 1, \dots, r\}$. Hence

$$\mu_1 = \frac{mn}{2},$$

$$\mu_2 = \kappa_1^2 + \kappa_2 = \frac{mn(m+n+3mn+1)}{12},$$

$$\mu_3 = \kappa_1^3 + 3\kappa_1\kappa_2 + \kappa_3 = \frac{m^2(m+1)n^2(n+1)}{8},$$

⋮

We finish by mentioning that, as in (Panny, 1986), one could also get asymptotic formulæ.

References

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