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1 recursion formular - simple version

Consider generating function for Poisson-Binomial of you case. m Bernoulli trials with success p and n Bernoulli trials with success q .

$$m + n = N \quad (1.1)$$

$$q + p = 1 \quad (1.2)$$

$$G(x) = (q + px)^m \cdot (p + qx)^n \quad (1.3)$$

The derivative of $\ln(f(x))$ is given by:

$$[\ln(g(x))]' = \frac{g'(x)}{f(x)} = \frac{\left(\sum_{i=0}^N a_i x^i\right)'}{\sum_{i=0}^N a_i x^i} = \frac{\sum_{i=1}^N i \cdot a_i x^{i-1}}{\sum_{i=0}^N a_i x^i} \quad (1.4)$$

On the other hand

$$[\ln(g(x))]' = (m(q + px) + n(p + qx))' = \frac{mp}{q + px} + \frac{nq}{p + qx} = \frac{xpq(m + n) + mp^2 + nq^2}{x^2pq + x(p^2 + q^2) + pq} \quad (1.5)$$

Equating both expressions we get

$$\frac{xpqN + mp^2 + nq^2}{x^2pq + x(p^2 + q^2) + pq} = \frac{\sum_{i=1}^N i \cdot a_i x^{i-1}}{\sum_{i=0}^N a_i x^i} \quad (1.6)$$

$$(xpqN + mp^2 + nq^2)\left(\sum_{i=0}^N a_i x^i\right) = (x^2pq + x(p^2 + q^2) + pq)\left(\sum_{i=1}^N i \cdot a_i x^{i-1}\right) \quad (1.7)$$

Multiplying and equating terms with same power of x we get:

$$a_i(mp^2 + nq^2) + a_{i-1}pqN = a_{i+1}pq(i + 1) + a_i(p^2 + q^2)i + a_{i-1}pq(i - 1) \quad (1.8)$$

$$a_i(mp^2 + nq^2) + a_{i-1}pqN = a_{i+1}pqi + a_i(p^2 + q^2)i + a_{i-1}pqi + (a_{i+1} - a_{i-1}) \quad (1.9)$$

Ignore the difference of $a_{i+1} - a_{i-1}$, and denote the expectation of successes as μ . Then the expression simplifies to:

$$\frac{N - i}{\frac{a_{i+1}}{a_i}i - (\frac{\mu - Npq}{pq} - \frac{p^2 + q^2}{pq}i)} = \frac{a_i}{a_{i-1}} \quad (1.10)$$

$$\frac{N - i}{N - i + (\frac{a_{i+1}}{a_i} - 1)i - \frac{\mu - i}{pq}} = \frac{a_i}{a_{i-1}} \quad (1.11)$$

$$(1.12)$$

Denote the distance between i and μ as l . Then:

$$\mu - i = l \quad (1.13)$$

$$i = \mu - l \quad (1.14)$$

$$\frac{a_{i+1}}{a_i} = f_l \quad (1.15)$$

$$\frac{a_i}{a_{i-1}} = f_{l+1} \quad (1.16)$$

$$\frac{N - \mu + l}{N - \mu + l + (f_l - 1)(\mu - l) - \frac{l}{pq}} = f_{l+1} \quad (1.17)$$

$$f_{l+1} = \frac{1}{1 - \frac{1}{N - \mu + l}(\frac{l}{pq} - (f_l - 1)(\mu - l))} \quad (1.18)$$

Differentiating by μ , we have:

$$F'(\mu) = -\frac{pq((f_l - 1)Npq - l)}{((f_l - 2)pqx + (n + (2 - f_l)l)pq - l)^2} \quad (1.19)$$

Since the denominator is always positive, the derivative is always negative as long as for all l the inequality below holds:

$$(f_l - 1)Npq > l \quad (1.20)$$

This is a magic formula. No matter what I do, and how I twist the proof, it always comes down to that inequality. I tested it empirically and it indeed works, and in fact $(f_l - 1)Npq$ grows much much faster than l . But i had difficulties proving that analytically. A very interesting observation that Npq is the variance, which would suggest using Chebusheff inequality, but this, again, didn't get me anywhere. I know the inequality holds, but not sure how to prove it.

Note that when $l = 0$ and $f_0 = 1$, $f_1 = 1$ for all μ . When $l = 1$, f_1 is given by

$$f_1 = \frac{1}{1 - \frac{1}{N - \mu + 1} \frac{1}{pq}} \quad (1.21)$$

Clearly, f_1 is largest for the smallest μ which is reached when $m = 0$ and the smallest when $m = N$. And we also know that f_l when $m = 0$ is always greater than f_l for $m = N$, perhaps there's some way to utilize monotonicity of the f_{l+1} when f_l is fixed.