STATISTICS CENTRAL LIMIT THEOREM Proof for Bernoulli trials

THM: Given *n* Bernoulli trials with probability of success for each trial being *p*, the probability, P(m of n), of exactly *m* successes in *n* trials approaches the probability density of x = m for a normal (i.e., gaussian) distribution with $\mu = np$ and $\sigma^2 = npq$:

As
$$n \to \infty$$
, $P(m \text{ of } n) \to f(x = m) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(m-\mu)^2/2\sigma^2}$.

PROOF: We follow the general method of proof given in [1].

For Bernoulli trials we have the following value for P(m of n):

$$P(m \text{ of } n) = {}_{n}C_{m} \cdot p^{m}q^{n-m}$$

where ${}_{n}C_{m} = \frac{n!}{(n-m)!m!}$ is the combinatoric coefficient.

For the proof, we consider different values of n, and we will consider m to be a fixed number, k, of standard deviations from the mean as n increases.

$$m = \mu + k\sigma$$

NOTE: Although m is an integer, the method of proof allows k to have any real value.

We use Stirling's formula, [2], to approximate the factorials in ${}_{n}C_{m}$:

$$n! = \sqrt{2\pi} n^{n+\frac{1}{2}} e^{-n+\frac{\theta}{12n}}$$

where n > 0 and $0 < \theta < 1$.

NOTE: Stirling's formula is related to the Stirling series expansion of the gamma function in powers of 1/n, (see [3]). The Stirling series has the curious property that it produces very accurate approximations of the gamma functions with only a few terms—and actually diverges if all the terms are used.

Using Stirling's formula for the terms of ${}_{n}C_{m}$, yields the following expression:

$${}_{n}C_{m} = \frac{\sqrt{2\pi} n^{n+\frac{1}{2}} e^{-n+\frac{\theta_{1}}{12n}}}{\sqrt{2\pi} (n-m)^{n-m+\frac{1}{2}} e^{-n-m+\frac{\theta_{2}}{12(n-m)}} \sqrt{2\pi} m^{m+\frac{1}{2}} e^{-m+\frac{\theta_{3}}{12m}}}$$

As *n* becomes large, so do n - m and *m*, and the residual terms involving θ_1 , θ_2 , and θ_3 become vanishingly small. Thus, we may eliminate the θ terms and, after also canceling common factors of $\sqrt{2\pi}$ and the exponentials of *e*, write the following expression:

$$_{n}C_{m} \rightarrow \frac{n^{n+\frac{1}{2}}}{\sqrt{2\pi} (n-m)^{n-m+\frac{1}{2}} m^{m+\frac{1}{2}}} \text{ as } n \rightarrow \infty$$

If we split the n^n term into two pieces in the numerator, we can match up the exponents in the numerator and denominator:

$$_{n}C_{m} \rightarrow \frac{n^{n-m}n^{m}\sqrt{n}}{\sqrt{2\pi} (n-m)^{n-m}\sqrt{n-m} \cdot m^{m}\sqrt{m}} \text{ as } n \rightarrow \infty$$

or

$$_{n}C_{m} \rightarrow \frac{\sqrt{n}}{\sqrt{2\pi(n-m)m}} \left(\frac{n}{n-m}\right)^{n-m} \left(\frac{n}{m}\right)^{m} \text{ as } n \rightarrow \infty$$

Now we invert the terms being exponentiated and use the following formulas:

$$\frac{m}{n} = \frac{\mu + k\sigma}{n} = \frac{np + k\sigma}{n} = p\left(1 + \frac{k\sigma}{np}\right)$$

and

$$\frac{n-m}{n} = 1 - \frac{\mu + k\sigma}{n} = 1 - \frac{np + k\sigma}{n} = q \left(1 - \frac{k\sigma}{nq}\right)$$

Substituting these expressions yields the following equation:

$$_{n}C_{m} \rightarrow \frac{\sqrt{n}}{\sqrt{2\pi(n-m)m}} \left(q\left(1-\frac{k\sigma}{nq}\right)\right)^{-n+m} \left(p\left(1+\frac{k\sigma}{np}\right)\right)^{-m} \text{ as } n \rightarrow \infty$$

The terms having n in their denominators will become small as n becomes large. Thus, we use an approximation that exploits this behavior:

$$\ln(1+x) \approx x - \frac{x^2}{2}$$
 for x small

or

$$(1+x)^r = e^{r\ln(1+x)} \approx e^{r\left(x-\frac{x^2}{2}\right)}$$
 for x small (from Taylor series for ln)

Applying this identity to our formula for the combinatoric coefficient, we have the following expression:

$${}_{n}C_{m} \rightarrow \frac{\sqrt{n}}{\sqrt{2\pi(n-m)m}} q^{-n+m} p^{-m} e^{\left(-n+m\right)\left(-k\frac{\sigma}{nq}-\frac{k^{2}\sigma^{2}}{2n^{2}q^{2}}\right)} e^{-m\left(k\frac{\sigma}{np}-\frac{k^{2}\sigma^{2}}{2n^{2}p^{2}}\right)}$$

Using $m = np + k\sigma$ and $m - n = -nq + k\sigma$ we have

$${}_{n}C_{m} \rightarrow \frac{\sqrt{n}}{\sqrt{2\pi(n-m)m}}q^{-n+m}p^{-m}e^{\left(k\sigma-nq\right)\left(-k\frac{\sigma}{nq}-\frac{k^{2}\sigma^{2}}{2n^{2}q^{2}}\right)-\left(k\sigma+np\right)\left(k\frac{\sigma}{np}-\frac{k^{2}\sigma^{2}}{2n^{2}p^{2}}\right)}$$

If we consider just the exponent, we have the following calculation:

$$-\frac{k^{2}\sigma^{2}}{nq} - \frac{k^{2}\sigma^{2}}{np} + k\sigma \left(\frac{k^{2}\sigma^{2}}{2n^{2}p^{2}} - \frac{k^{2}\sigma^{2}}{2n^{2}q^{2}}\right) + nq\frac{k^{2}\sigma^{2}}{2n^{2}q^{2}} + np\frac{k^{2}\sigma^{2}}{2n^{2}p^{2}}$$
$$= -\frac{k^{2}\sigma^{2}p}{npq} - \frac{k^{2}\sigma^{2}q}{npq} + k\sigma \left(\frac{k^{2}\sigma^{2}q^{2}}{2n^{2}p^{2}q^{2}} - \frac{k^{2}\sigma^{2}p^{2}}{2n^{2}p^{2}q^{2}}\right) + (np + nq)\frac{k^{2}\sigma^{2}}{2n^{2}pq}$$

Using $\sigma^2 = npq$ the simplification of the exponent continues:

$$= -k^{2}p - k^{2}q + k\sigma\left(\frac{k^{2}q^{2}}{2\sigma^{2}} - \frac{k^{2}p^{2}}{2\sigma^{2}}\right) + \frac{k^{2}}{2}$$
$$= -k^{2} + k\left(\frac{k^{2}q^{2}}{2\sigma} - \frac{k^{2}p^{2}}{2\sigma}\right) + \frac{k^{2}}{2}$$
$$= -\frac{k^{2}}{2} + k\left(\frac{k^{2}q^{2}}{2\sigma} - \frac{k^{2}p^{2}}{2\sigma}\right)$$

We observe that the second term is proportional to $1/\sqrt{n}$ and vanishes as *n* becomes large. Dropping this term yields the following expression:

STATISTICS CENTRAL LIMIT THEOREM Proof for Bernoulli trials (cont.)

$$_{n}C_{m} \rightarrow \frac{\sqrt{n}}{\sqrt{2\pi(n-m)m}}q^{-n+m}p^{-m}e^{-\frac{k^{2}}{2}}$$
 as $n \rightarrow \infty$

If we now multiply by the probability, $p^m q^{n-m}$ of one particular pattern of *m* successes occurring, we obtain the following expression:

$$P(m \text{ of } n) \rightarrow \frac{\sqrt{n}}{\sqrt{2\pi(n-m)m}} e^{-\frac{k^2}{2}} \text{ as } n \rightarrow \infty$$

We have the following simplification for the factor in front:

$$\sqrt{\frac{n}{2\pi(n-m)m}} = \sqrt{\frac{1}{2\pi(1-\frac{m}{n})m}} = \sqrt{\frac{1}{2\pi(1-\frac{np+k\sigma}{n})(np+k\sigma)}}$$

For *n* large, $k\sigma$ is much smaller than *n*, leading to the following result:

$$\sqrt{\frac{n}{2\pi(n-m)m}} \approx \sqrt{\frac{1}{2\pi(1-\frac{np}{n})(np)}} = \sqrt{\frac{1}{2\pi qnp}} = \sqrt{\frac{1}{2\pi\sigma^2}}$$

With this substitution, and using $k^2 = \frac{(m-\mu)^2}{\sigma^2}$ we complete our proof:

$$P(m \text{ of } n) \rightarrow \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{k^2}{2}} = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(m-\mu)^2/2\sigma^2} \text{ as } n \rightarrow \infty$$

- **REF:** [1] Eugene Lukacs, *Probability and Mathematical Statistics, an Introduction*, New York, NY: Academic Press, 1972.
 - [2] Milton Abramowitz and Irene A. Stegun, Eds., Handbook of Mathematical Functions: National Bureau of Standards Applied Mathematics Series 55, Washington, D.C.: Government Printing Office, 1972.
 - [3] Carl M. Bender and Steven A. Orszag, *Advanced Mathematical Methods* for Scientists and Engineers, New York, NY: McGraw-Hill, 1978.