

**MthStat 465, Spring 2005, Lecture Number 13**  
**Normal Approximation to the Binomial Distribution.**

We show how the normal distribution arises as an approximation to the binomial distribution. This is summarized by the DeMoivre-Laplace Theorem, which we will take up in the next lecture.

1. TWO BUILDING BLOCKS

We need two results. The first is Stirling's formula:

**Theorem 1** (Stirling's Formula).

$$(1) \quad \lim_{N \rightarrow \infty} \frac{N!}{N^N e^{-N} \sqrt{2\pi N}} = 1.$$

*Proof.* For a proof see Ash, *Basic Probability Theory*. □

We are concerned with the symmetric binomial probabilities. In particular, in  $2N$  tosses of a fair coin, what is the probability that the number of heads equals the number of tails? Stirling's Formula suggests the following algebraic manipulation:

$$\begin{aligned} \binom{2N}{N} \left(\frac{1}{2}\right)^{2N} &= \frac{(2N)!}{N!N!} \frac{1}{2^{2N}} \\ &= \frac{(2N)!}{(2N)^{2N} e^{-2N} \sqrt{2\pi 2N}} \left(\frac{N^N e^{-N} \sqrt{2\pi N}}{N!}\right)^2 \\ &\quad \times \frac{(2N)^{2N} e^{-2N} \sqrt{2\pi 2N}}{(N^N e^{-N} \sqrt{2\pi N})^2} \frac{1}{2^{2N}} \\ &= \frac{(2N)!}{(2N)^{2N} e^{-2N} \sqrt{2\pi 2N}} \left(\frac{N^N e^{-N} \sqrt{2\pi N}}{N!}\right)^2 \frac{1}{\sqrt{N\pi}} \end{aligned}$$

from which we see that

$$(2) \quad \lim_{N \rightarrow \infty} \sqrt{N\pi} \binom{2N}{N} \left(\frac{1}{2}\right)^{2N} = 1.$$

We can restate (2) as

$$(3) \quad \binom{2N}{N} \left(\frac{1}{2}\right)^{2N} \approx \frac{1}{\sqrt{\pi N}}.$$

The other tool we need is that for  $|x| \ll 1$  we have

$$(4) \quad \log(1+x) \approx x.$$

This should be familiar from calculus.

## 2. WHERE THE NORMAL DISTRIBUTION COMES IN

The idea is to measure all the other binomial probabilities as ratios with the left hand side of (3). We consider for each  $k \in \{-N, \dots, N\}$  the ratio

$$\begin{aligned}
\frac{\binom{2N}{N+K} \left(\frac{1}{2}\right)^{2N}}{\binom{2N}{N} \left(\frac{1}{2}\right)^{2N}} &= \frac{\binom{2N}{N+K}}{\binom{2N}{N}} \\
&= \frac{(2N)!}{(N+k)!(N-k)!} \frac{N!N!}{(2N)!} \\
&= \frac{N!}{(N-k)!} \frac{N!}{(N+k)!} \\
&= \frac{N(N-1)\cdots(N-(k-1))}{(N+k)(N+k-1)\cdots(N+1)} \\
&= \frac{N(N-1)\cdots(N-(k-1))N^{-k}}{(N+k)(N+k-1)\cdots(N+1)N^{-k}} \\
&= \frac{\left(1 - \frac{1}{N}\right)\cdots\left(1 - \frac{k-1}{N}\right)}{\left(1 + \frac{k}{N}\right)\left(1 + \frac{k-1}{N}\right)\cdots\left(1 + \frac{1}{N}\right)}
\end{aligned}$$

so

$$(5) \quad \frac{\binom{2N}{N+K} \left(\frac{1}{2}\right)^{2N}}{\binom{2N}{N} \left(\frac{1}{2}\right)^{2N}} = \frac{\left(1 - \frac{1}{N}\right)\cdots\left(1 - \frac{k-1}{N}\right)}{\left(1 + \frac{k}{N}\right)\left(1 + \frac{k-1}{N}\right)\cdots\left(1 + \frac{1}{N}\right)}.$$

We now estimate each of these products by first estimating their logarithms using (4). The key here is that we have to make sure that  $|k| \ll N$  so that the approximations are good enough.

$$\begin{aligned}
\log\left(\left(1 - \frac{1}{N}\right)\cdots\left(1 - \frac{k-1}{N}\right)\right) &= \log\left(1 - \frac{1}{N}\right) + \cdots + \log\left(1 - \frac{k-1}{N}\right) \\
&\approx -\left(\frac{1}{N} + \cdots + \frac{k-1}{N}\right) \\
&= -\frac{(k-1)k}{2N} \\
&= \frac{-k^2 + k}{2N}
\end{aligned}$$

so

$$(6) \quad \left(1 - \frac{1}{N}\right)\cdots\left(1 - \frac{k-1}{N}\right) \approx \exp\left(\frac{-k^2 + k}{2N}\right),$$

where  $\exp(x) = e^x$ . Similarly,

$$\begin{aligned} \log\left(\left(1 + \frac{k}{N}\right) \cdots \left(1 + \frac{1}{N}\right)\right) &= \log\left(1 + \frac{k}{N}\right) + \cdots + \log\left(1 + \frac{1}{N}\right) \\ &\approx \left(\frac{k}{N} + \cdots + \frac{1}{N}\right) \\ &= \frac{k(k+1)}{2N} \\ &= \frac{k^2 + k}{2N}, \end{aligned}$$

so

$$(7) \quad \left(1 + \frac{k}{N}\right) \cdots \left(1 + \frac{1}{N}\right) \approx \exp\left(\frac{k^2 + k}{2N}\right).$$

Combining (5), (6) and (7) we have

$$\frac{\binom{2N}{N+K} \left(\frac{1}{2}\right)^{2N}}{\binom{2N}{N} \left(\frac{1}{2}\right)^{2N}} \approx \exp\left(-\frac{k^2}{N}\right),$$

or, more usefully,

$$(8) \quad \binom{2N}{N+K} \left(\frac{1}{2}\right)^{2N} \approx \binom{2N}{N} \left(\frac{1}{2}\right)^{2N} \exp\left(-\frac{k^2}{N}\right).$$

Combining (3) with (8) and cleverly re-arranging, we arrive at

$$(9) \quad \binom{2N}{N+K} \left(\frac{1}{2}\right)^{2N} = \frac{1}{\sqrt{\pi}} \exp\left(-\left(\frac{k}{\sqrt{N}}\right)^2\right) \frac{1}{\sqrt{N}}.$$

So, if we are looking at an expression of the form

$$\sum_{k=A}^B B \binom{2N}{N+K} \left(\frac{1}{2}\right)^{2N}$$

we see it can be approximated by a sum

$$\sum_{k=A}^B \frac{1}{\sqrt{\pi}} \exp\left(-\left(\frac{k}{\sqrt{N}}\right)^2\right) \frac{1}{\sqrt{N}}.$$

which is a Riemann sum for an integral of the form

$$\int_a^b \frac{1}{\sqrt{\pi}} e^{-x^2} dx,$$

provided we can show that all the approximations are good enough. The relationship between  $a$  and  $A$  is that  $a = A/\sqrt{N}$  and, similarly,  $b = B/\sqrt{N}$ .