Theorem If $X_i \sim \text{binomial}(n_i, p)$ for i = 1, 2, ..., m, then $\sum_{i=1}^n X_i \sim \text{binomial}(\sum_{i=1}^m n_i, p)$. **Proof** The probability mass function of X_i is

$$f_{X_i}(x_i) = \binom{n_i}{x_i} p^{x_i} (1-p)^{1-x_i}$$
 $x = 0, 1, 2, \dots, n_i,$

for i = 1, 2, ..., m and $-\infty < t < \infty$. The moment generating function for a binomial random variable is

$$M_{X_i}(t) = (pe^t + (1-p))^{n_i}$$

for i = 1, 2, ..., m. Let the random variable $Y = \sum_{i=1}^{m} X_i$. The moment generating function of Y is

$$E\left[e^{tY}\right] = E\left[e^{t\left(\sum_{i=1}^{m} X_{i}\right)}\right]$$

$$= E\left[e^{tX_{1}}e^{tX_{2}}\dots e^{tX_{n}}\right]$$

$$= E\left[e^{tX_{1}}\right]E\left[e^{tX_{2}}\right]\dots E\left[e^{tX_{n}}\right]$$

$$= \left(pe^{t} + (1-p)\right)^{n_{1}}\left(pe^{t} + (1-p)\right)^{n_{2}}\dots\left(pe^{t} + (1-p)\right)^{n_{m}}$$

$$= \left(pe^{t} + (1-p)\right)^{\sum_{i=1}^{m} n_{i}}$$

for $-\infty < t < \infty$, which is the moment generating function of a binomial random variable with parameters $\sum_{i=1}^{m} n_i$ and p.

APPL illustration: The APPL statements

X1 := BinomialRV(n1, p);
X2 := BinomialRV(n2, p);
Y := Convolution(X1, X2);
MGF(Y);

fail to provide the appropriate moment generating function.