### 4.1.3 Bionomial distribution

Consider a random variable with two outcomes $A$ and $B$ (e.g. a coin toss) of relative probabilities $p_{A}$ and $p_{B}=1-p_{A}$. The probability that in $N$ trials the event $A$ occurs exactly $N_{A}$ times (e.g. 5 heads in 12 coin tosses), is given by the binomial distribution

$$
\begin{equation*}
p_{N}\left(N_{A}\right)=\left(\frac{N!}{N_{A}!\left(N-N_{A}\right)!}\right) p_{A}^{N_{A}} p_{B}^{N-N_{A}} . \tag{4.1.8}
\end{equation*}
$$

The prefactor is just the coefficient obtained in the binomial expansion of $\left(p_{A}+p_{B}\right)^{N}$, and gives the number of possible orderings of $N_{A}$ events $A$ and $N_{B}=N-N_{A}$ events $B$.

The generating function for the binomial distribution is

$$
\begin{equation*}
G_{N}(\lambda)=\left\langle e^{\lambda N_{A}}\right\rangle=\sum_{N_{A}=0}^{N} \frac{N!}{N_{A}!\left(N-N_{A}\right)!} p_{A}^{N_{A}} p_{B}^{N-N_{A}} e^{\lambda N_{A}}=\left(p_{A} e^{\lambda}+p_{B}\right)^{N} \tag{4.1.9}
\end{equation*}
$$

The resulting cumulant generating function is

$$
\begin{equation*}
\ln G_{N}(\lambda)=N \ln \left(p_{A} e^{-i k}+p_{B}\right)=N \ln G_{1}(\lambda) \tag{4.1.10}
\end{equation*}
$$

where $\ln G_{1}(\lambda)$ is the cumulant generating function for a single step. Hence, the cumulants after $N$ steps are simply $N$ times the cumulants in a single step. In each step, the allowed values of $N_{A}$ are 0 and 1 with respective probabilities $p_{B}$ and $p_{A}$, leading to $\left\langle N_{A}^{\ell}\right\rangle=p_{A}$, for all $\ell$. After $N$ trials the first two cumulants are

$$
\begin{equation*}
\left\langle N_{A}\right\rangle_{c}=N p_{A} \quad, \quad\left\langle N_{A}^{2}\right\rangle_{c}=N\left(p_{A}-p_{A}^{2}\right)=N p_{A} p_{B} \tag{4.1.11}
\end{equation*}
$$

A measure of fluctuations around the mean is provided by the standard deviation, which is the square root of the variance. While the mean of the binomial distribution scales as $N$, its standard deviation only grows as $\sqrt{N}$. Hence, the relative uncertainty becomes smaller for large $N$.

The binomial distribution is straightforwardly generalized to a multinomial distribution, when the several outcomes $\{A, B, \cdots, M\}$ occur with probabilities $\left\{p_{A}, p_{B}, \cdots, p_{M}\right\}$. The probability of finding outcomes $\left\{N_{A}, N_{B}, \cdots, N_{M}\right\}$ in a total of $N=N_{A}+N_{B} \cdots+N_{M}$ trials is

$$
\begin{equation*}
p_{N}\left(\left\{N_{A}, N_{B}, \cdots, N_{M}\right\}\right)=\frac{N!}{N_{A}!N_{B}!\cdots N_{M}!} p_{A}^{N_{A}} p_{B}^{N_{B}} \cdots p_{M}^{N_{M}} \tag{4.1.12}
\end{equation*}
$$

