

The Hypergeometric Distribution

Klaus Pommerening
Fachbereich Physik, Mathematik, Informatik
der Johannes-Gutenberg-Universität
Saarstraße 21
D-55099 Mainz

February 2, 2003—English version February 14, 2015
last change February 15, 2015

The urn problem underlying the hypergeometric distribution is “drawing without replacement”. Assume the urn contains n balls s of which are black, and $t = n - s$ are white. Let

$$p := \frac{s}{n}$$

be the proportion of black balls, and assume without loss of generality that $p > \frac{1}{2}$. (The case $p = \frac{1}{2}$ is not interesting, the case $p < \frac{1}{2}$ is symmetric to the considered case.)

Draw r balls ($r \leq n$) by random. The probability that exactly ν of the balls are white is

$$q_r^{(s)}(\nu) = \frac{\binom{s}{r-\nu} \binom{t}{\nu}}{\binom{n}{r}}.$$

The function

$$q_r^{(s)}: \mathbb{Z} \longrightarrow \mathbb{R}$$

is called the **hypergeometric distribution** (with parameters n , s , and r). We have $q_r^{(s)}(\nu) = 0$ for $\nu < 0$ as well as for $\nu > r$. The probability of drawing more black balls than white ones is

$$p_r^{(s)} = \begin{cases} \sum_{\nu=0}^{\frac{r-1}{2}} q_r^{(s)}(\nu) & \text{if } r \text{ is odd,} \\ \sum_{\nu=0}^{\frac{r}{2}-1} q_r^{(s)}(\nu) + \frac{1}{2} q_r^{(s)}\left(\frac{r}{2}\right) & \text{if } r \text{ is even,} \end{cases}$$

in case of a tie we randomly decide between black and white with probability $\frac{1}{2}$.

In the uninteresting case $p = \frac{1}{2}$ obviously all $p_r^{(s)} = \frac{1}{2}$.

Lemma 1

- (i) $p_1^{(s)} = p$.
- (ii) $p_2^{(s)} = p_1^{(s)}$ (if $t \geq 1$).
- (iii) $p_3^{(s)} = \frac{s(s-1)}{n(n-1)} \cdot \left[3 - 2 \cdot \frac{s-2}{n-2} \right]$ (if $t \geq 2$).
- (iv) $p_4^{(s)} = p_3^{(s)}$ (if $t \geq 2$).
- (v) $p_r^{(s)} = 1$ for $r > 2t$.

Proof. (i) Trivial.

(ii) We draw two balls, and break the tie (in the case where we draw one ball of each type) by a random decision. Therefore the numerator is

$$\binom{s}{2} + \frac{1}{2} \binom{s}{1} \binom{t}{1} = \frac{s(s-1)}{2} + \frac{s(n-s)}{2} = \frac{s(n-1)}{2}.$$

The denominator is $\frac{n(n-1)}{2}$, and the quotient is

$$p_2^{(s)} = \frac{s(n-1)}{n(n-1)} = p.$$

(iii) Here the numerator is

$$\begin{aligned} \binom{s}{3} + \binom{s}{2} \cdot (n-s) &= \frac{s(s-1)(s-2) + 3s(s-1)(n-s)}{6} \\ &= \frac{s(s-1)}{6} \cdot [s-2 + 3 \cdot (n-s)] \\ &= \frac{s(s-1)}{6} \cdot [3 \cdot (n-2) - 2 \cdot (s-2)]. \end{aligned}$$

The denominator is $\frac{1}{6} \cdot n(n-1)(n-2)$, hence the asserted value of $p_3^{(s)}$.

(iv) We omit the calculation since the next lemma contains a more general statement.

(v) In this case we necessarily draw a majority of black balls. \diamond

Lemma 2 *If r is even and $2 \leq r \leq 2t$, then*

$$p_{r+1}^{(s)} > p_r^{(s)} = p_{r-1}^{(s)}.$$

Proof. Let $A_r^{(s)}(\nu) = \binom{n}{r} \cdot q_r^{(s)}(\nu)$ be the numerator of $q_r^{(s)}(\nu)$, and $B_r^{(s)} = \binom{n}{r} \cdot p_r^{(s)}$, the numerator of $p_r^{(s)}$.

After $r+1$ drawings we have a black majority in $B_{r+1}^{(s)}$ cases. Considering the change from r to $r+1$ we have:

- $\sum_{\nu=0}^{\frac{r}{2}-1} A_r^{(s)}(\nu)$ cases where the number of black balls is at least $\frac{r}{2} + 1$ after r drawings. We have $n - r$ possibilities for the $(r + 1)$ -th ball, but all of these cannot change the majority. So we get

$$X_1 = (n - r) \cdot \sum_{\nu=0}^{\frac{r}{2}-1} A_r^{(s)}(\nu)$$

cases with a black majority.

- $A_r^{(s)}(\frac{r}{2})$ cases where after r drawings we have exactly $\frac{r}{2}$ black balls. From the $n - r$ possibilities for the $(r + 1)$ -th ball
 - $s - \frac{r}{2}$ are black and give a black majority,
 - $t - \frac{r}{2}$ are white and give a white majority.

Thus we get another

$$X_2 = (s - \frac{r}{2}) \cdot A_r^{(s)}(\frac{r}{2})$$

cases with a black majority.

- In the remaining cases after r drawings we have at most $\frac{r}{2} - 1$ black balls. Therefore the $(r + 1)$ -th ball cannot change the white majority.

This count contains each resulting set exactly $r + 1$ times. Therefore

$$B_{r+1}^{(s)} = \frac{1}{r + 1} \cdot (X_1 + X_2) = \frac{n - r}{r + 1} \cdot \left[\sum_{\nu=0}^{\frac{r}{2}-1} A_r^{(s)}(\nu) + \frac{s - \frac{r}{2}}{n - r} \cdot A_r^{(s)}(\frac{r}{2}) \right].$$

For the coefficient of the last term we have

$$\frac{s - \frac{r}{2}}{n - r} > \frac{1}{2} \iff 2s - r > n - r \iff s > \frac{n}{2}.$$

(Since $r \leq 2t$ also $r < n$.) Therefore

$$B_{r+1}^{(s)} > \frac{n - r}{r + 1} \cdot B_r^{(s)},$$

and the first part of the assertion, $p_{r+1}^{(s)} > p_r^{(s)}$, follows.

Analyzing the change from $r - 1$ to r is somewhat more complicated. After r drawings we have a black majority in $B_r^{(s)}$ cases. Among these are:

- $\sum_{\nu=0}^{\frac{r}{2}-2} A_{r-1}^{(s)}$ cases where after $r - 1$ drawings we have at least $\frac{r}{2} + 1$ black balls. The $n - r + 1$ possibilities for the r -th ball can't change the decision. Hence we get

$$Y_1 = (n - r + 1) \cdot \sum_{\nu=0}^{\frac{r}{2}-2} A_{r-1}^{(s)}$$

cases with black majority.

- $A_{r-1}^{(s)}\left(\frac{r}{2} - 1\right)$ cases where after $r - 1$ drawings we have exactly $\frac{r}{2}$ black balls. The $n - r + 1$ possibilities for the r -th ball dissociate into

– $s - \frac{r}{2}$ black ones that result in a black majority. This makes

$$Y_2 = \left(s - \frac{r}{2}\right) \cdot A_{r-1}^{(s)}\left(\frac{r}{2} - 1\right)$$

additional cases.

– $t + 1 - \frac{r}{2}$ white ones where we randomly decide with probability $\frac{1}{2}$. This adds another

$$Y_3 = \frac{1}{2} \cdot \left(t + 1 - \frac{r}{2}\right) \cdot A_{r-1}^{(s)}\left(\frac{r}{2} - 1\right)$$

cases to our collection.

- $A_{r-1}^{(s)}\left(\frac{r}{2}\right)$ cases where after $r - 1$ drawings we have exactly $\frac{r}{2} - 1$ black balls. The $n - r + 1$ possibilities for the r -th ball dissociate into

– $s + 1 - \frac{r}{2}$ black ones where we randomly decide with probability $\frac{1}{2}$. This gives another

$$Y_4 = \frac{1}{2} \cdot \left(s + 1 - \frac{r}{2}\right) \cdot A_{r-1}^{(s)}\left(\frac{r}{2}\right)$$

cases.

– $t - \frac{r}{2}$ white ones that don't disturb the white majority.

- In the remaining cases after $r - 1$ drawings we have at most $\frac{r}{2} - 2$ black balls. The white majority is unchanged.

Each set of drawn balls is counted exactly r times. Therefore

$$\begin{aligned} B_r^{(s)} &= \frac{1}{r} \cdot (Y_1 + Y_2 + Y_3 + Y_4) \\ &= \frac{n - r + 1}{r} \cdot \sum_{\nu=0}^{\frac{r}{2}-2} A_{r-1}^{(s)} + \frac{1}{r} \cdot \left(s - \frac{r}{2} + \frac{t}{2} + \frac{1}{2} - \frac{r}{4}\right) \cdot A_{r-1}^{(s)}\left(\frac{r}{2} - 1\right) \\ &\quad + \frac{1}{2r} \cdot \left(s - \frac{r}{2} + 1\right) \cdot A_{r-1}^{(s)}\left(\frac{r}{2}\right) \end{aligned}$$

Since $s + \frac{t}{2} = n - \frac{t}{2}$ the coefficient of the middle term equals

$$s - \frac{r}{2} + \frac{t}{2} - \frac{r}{4} + \frac{1}{2} = n - \frac{t}{2} - r + \frac{r}{4} + 1 - \frac{1}{2} = (n - r + 1) - \frac{1}{2} \cdot \left(t - \frac{r}{2} + 1\right).$$

Hence

$$\begin{aligned} B_r^{(s)} &= \frac{n - r + 1}{r} \cdot \sum_{\nu=0}^{\frac{r}{2}-1} A_{r-1}^{(s)} \\ &\quad - \frac{1}{2r} \left(t - \frac{r}{2} + 1\right) \binom{s}{\frac{r}{2}} \binom{t}{\frac{r}{2} - 1} + \frac{1}{2r} \left(s - \frac{r}{2} + 1\right) \binom{s}{\frac{r}{2} - 1} \binom{t}{\frac{r}{2}}. \end{aligned}$$

The two last terms cancel. What remains is

$$B_r^{(s)} = \frac{n-r+1}{r} \cdot B_{r-1}^{(s)}.$$

This proves the second part of the assertion. \diamond

We conclude:

Proposition 1 *The probability $p_r^{(s)}$ grows monotonically with r from $p_1^{(s)} = p$ to $p_{2t+1}^{(s)} = 1$.*

If the quotients

$$\frac{rs}{n}, \frac{rt}{n}, \frac{(n-r)s}{n}, \frac{(n-r)t}{n}$$

are sufficiently large (by FISHER's rule of thumb: ≥ 5), the normal distribution approximates the hypergeometric distribution well. In particular

$$\sum_{\nu=0}^x q_r^{(s)}(\nu) \approx \Phi\left(\frac{x-\mu}{\sigma}\right) = \frac{1}{\sqrt{2\pi}} \cdot \int_{-\infty}^{\frac{x-\mu}{\sigma}} e^{-t^2/2} dt \quad (1)$$

where μ is the mean value and σ^2 is the variance of the hypergeometric distribution (with parameters n , s , and r), and Φ is the distribution function of the normal distribution. For mean value and variance we have:

Lemma 3

$$\begin{aligned} \mu &= \frac{rt}{n}, \\ \sigma^2 &= \frac{r(n-r) \cdot t(n-t)}{n^2(n-1)}. \end{aligned}$$

Proof. Take a random sample of r balls. Let $X_k : \Omega \rightarrow \mathbb{R}$ be a random variable that assumes the value 0 if the k -th ball is black, and 1 if it is white. Then $S = X_1 + \dots + X_r : \Omega \rightarrow \mathbb{R}$ is a random variable that counts the number of white balls in our sample. Then $\mu = E(S)$ is the expectation and $\sigma^2 = \text{Var}(S)$ is the variance of this random variable.

Since $E(X_k) = \frac{t}{n}$ we have $E(S) = r \cdot \frac{t}{n}$.

We note that $X_k^2 = X_k$ and derive

$$\text{Var}(X_k) = E(X_k^2) - E(X_k)^2 = \frac{t}{n} - \frac{t^2}{n^2} = \frac{t(n-t)}{n^2}.$$

Since $X_j X_k(\omega) = 1 \iff X_j(\omega) = 1 \text{ and } X_k(\omega) = 1$ the probability of this event is $\frac{t(t-1)}{n(n-1)}$. This gives the expectation $E(X_j X_k) = \frac{t(t-1)}{n(n-1)}$. Thus the covariance is

$$\begin{aligned} \text{Cov}(X_j, X_k) &= E(X_j X_k) - E(X_j)E(X_k) = \frac{t(t-1)}{n(n-1)} - \frac{t^2}{n^2} \\ &= \frac{t(n(t-1) - t(n-1))}{n^2(n-1)} = \frac{t(t-n)}{n^2(n-1)}. \end{aligned}$$

We deduce the variance of S :

$$\begin{aligned} \text{Var}(S) &= \sum_{k=1}^r \text{Var}(X_k) + 2 \cdot \sum_{1 \leq j < k \leq r} \text{Cov}(X_j, X_k) \\ &= \frac{rt(n-t)}{n^2} + r(r-1) \cdot \frac{t(t-n)}{n^2(n-1)} = \frac{rt(n-t)}{n^2} \cdot \left[1 - \frac{r-1}{n-1}\right] \\ &= \frac{rt(n-t)}{n^2(n-1)} \cdot [n-r], \end{aligned}$$

as claimed. \diamond

Proposition 2 (Asymptotic distribution) *The probability of a majority of black balls is*

$$p_r^{(s)} \approx \frac{1}{\sqrt{2\pi}} \cdot \int_{-\infty}^{\sqrt{r\lambda}} e^{-t^2/2} dt$$

with $\lambda = (2p-1)^2$, under the assumption that $p \approx \frac{1}{2}$, $r \ll n$, and r not too small.

[By FISHER's rule of thumb $10 \leq r \leq n-10$ suffices if $p \approx \frac{1}{2}$.

Note that this "proposition" lacks mathematical precision.]

Proof. We look at the upper boundary of the integral (1) for $x = \frac{r}{2}$:

$$\begin{aligned} \frac{x - \mu}{\sigma} &= \frac{(\frac{r}{2} - \frac{rt}{n}) \cdot n \cdot \sqrt{n-1}}{\sqrt{r(n-r)t(n-t)}} = \frac{(rn - 2rt)\sqrt{n-1}}{2 \cdot \sqrt{r(n-r)t(n-t)}} \\ &= \frac{\sqrt{r}\sqrt{n-1}}{\sqrt{n-r}} \cdot \frac{s-t}{2\sqrt{st}} = \frac{\sqrt{n-1}}{\sqrt{n-r}} \cdot \sqrt{r} \cdot \frac{2p-1}{2\sqrt{p(1-p)}} \\ &\approx 1 \cdot \sqrt{r} \cdot \frac{2p-1}{2 \cdot \sqrt{\frac{1}{4}}} = \sqrt{r\lambda}, \end{aligned}$$

as claimed. \diamond