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The experiment consists of drawing some predetermined number n of chips from the urn, **without replacement**, and noting the number of red and black chips.

The random variable associated with the experiment is usually defined to be the number of red chips drawn.

Had we replaced the chip after each draw, a bit of thought should convince you that the number of red chips drawn should have a binomial distribution, because we conduct a fixed number n of Bernoulli trials each with probability of success

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The slight modification in the experiment that produces the hypergeometric distribution instead of the binomial is that we **do not** replace each chip after it is drawn, but conduct the next draw from whatever chips are left in the urn.

Because we do not replace chips as we draw them, the probability of drawing a red chip does not remain the same on successive draws.

As it turns out, the probability of a red chip on the second draw depends on the outcome of the first draw.

If the first draw results in a red chip, the probability of a red chip on the second draw is:

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If the first draw results in a red chip, the probability of a red chip on the second draw is:

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However, if the first chip drawn is black, the probability of a red chip on the second draw is:

$$p = \frac{r}{r+b-1}$$

Because we violate both the assumption of independence of successive trials, and the assumption of constant probability of success, this is not a binomial experiment.

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The distribution that results from this experiment is called the **hypergeometric** distribution.

The author's notation for the hypergeometric distribution is:

$$P(X = x) = h(x; n, M, N)$$

where, in the parlance of the urn experiment,

- \checkmark N represents the number of chips in the urn at the start
- \blacksquare M represents the number of red chips
- *n* represents the number of chips drawn (and not replaced) during the experiment
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As we see, the hypergeometric is a bit more complicated than the others we have studied.

The probability mass function for the hypergeometric distribution is:

$$P(X = x) = h(x; n, M, N) = \frac{\binom{M}{x}\binom{N-M}{n-x}}{\binom{N}{n}}$$

with some restrictions on the values that x and n can take:

- \bullet x cannot be smaller than zero
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With a bit of tedious algebra, one can show that the expected value and variance of a hypergeometric random variable are:

$$E(X) = n \cdot \frac{M}{N}$$
 and $V(X) = \left(\frac{N-n}{N-1}\right) \cdot n \cdot \frac{M}{N} \cdot \left(1 - \frac{M}{N}\right)$

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Note that if we let p = M/N and we let N become large while p remains fixed, we end up with the mean and variance of a binomial random variable.

Once again, there is not universal agreement on the h(x; n, M, N) notation.

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In terms of the author's notation,

$$P(X=x)=h(x;n,M,N)=\operatorname{dhyper}\left(\mathbf{x}\,\text{,}\,\mathrm{M}\,\text{,}\,\mathrm{N-M}\,\text{,}\,\mathrm{n}\,\right)$$

Example: Keno

In the lottery game called Keno, 20 of the 80 numbers from 1 to 80 are randomly selected. Prior to the selection, players fill out a card specifying up to 10 numbers they think will be chosen.

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The experiment is then hypergeometric with 20 numbers chosen without replacement.

The probablility of getting 5 hits out of 10 numbers chosen is:

$$P(X=5)=h(5;20,10,80)={\tt dhyper(5,10,80-10,20)}=.05$$

The hypergeometric can also be visualized by a tree diagram.