

## Expected values - 1/11/06

### Binomial distribution (in book)

If random variable  $X$  is the number of heads in  $n$  tosses of a coin with bias  $0 \leq \alpha \leq 1$ , it has binomial distribution

$$p_X(k) = \binom{n}{k} \alpha^k (1 - \alpha)^{n-k} =: B_n(k)$$

Expected value

$$\begin{aligned} E(X) &:= \sum_{k=0}^n k p_X(k) \\ &= \sum_{k=1}^n k p_X(k) \quad \text{since the } k=0 \text{ term contributes zero to sum} \\ &= \sum_{k=1}^n k \frac{n!}{k!(n-k)!} \alpha^k (1-\alpha)^{n-k} \\ &= \sum_{k=1}^n \frac{n!}{(k-1)!(n-k)!} \alpha^k (1-\alpha)^{n-k} \quad \text{cancelling with } k! \\ &= n\alpha \sum_{k=1}^n \frac{(n-1)!}{(k-1)!(n-k)!} \alpha^{k-1} (1-\alpha)^{n-k} \quad \text{bring out factors of } n \text{ and } \alpha \\ &= n\alpha \sum_{j=0}^{n-1} \frac{(n-1)!}{j!(n-1-j)!} \alpha^j (1-\alpha)^{n-1-j} \quad \text{substitute } j = k-1 \text{ or } k = j+1 \\ &= n\alpha \sum_{j=0}^{n-1} B_{n-1}(j) \\ &= n\alpha \end{aligned}$$

Note how the limits of the sum changed by 1 when we changes  $k$  to  $j$ . The last step we recognized the sum as the sum of probabilities of the binomial with  $j$  heads out of  $n-1$  tosses. This is a standard mathematical 'trick' (it works to get the variance too, although this is a bit more messy).

So, a coin of bias  $\alpha = 0.7$  tossed  $n = 100$  times has an expected number of heads of  $0.7 \times 100 = 70$ , not surprisingly.

### Hypergeometric distribution (not in book...you're gonna like it!)

Random variable  $X$  is the number of red chips in a draw of  $n$  without replacement from urn of  $N = r + w$  red and white chips.

We start with an identity which can be verified by direct differentiation,

$$\left[ \frac{d}{d\mu} (1 + \mu)^r \right] \cdot (1 + \mu)^w = r(1 + \mu)^{N-1} \quad (1)$$

Now the coefficient of  $\mu^{k-1}$  appearing in the first LHS factor is  $k \binom{r}{k}$  (here you must binomial expand *before* taking the derivative). The coeff of  $\mu^{n-k}$  in the second factor is  $\binom{w}{n-k}$ . The coeff of  $\mu^{n-1}$  in the LHS (the product of the two factors) is then the following sum over  $k$  (just like your proof of LM3 3.3.10). But this must equal the coeff of  $\mu^{n-1}$  in the RHS, thus

$$\sum_{k=0}^n k \binom{r}{k} \binom{w}{n-k} = r \binom{N-1}{n-1}. \quad (2)$$

Dividing both sides by  $\binom{N}{n}$  turns the LHS into what we want,

$$\begin{aligned} E(X) &= \sum_{k=0}^n k \frac{\binom{r}{k} \binom{w}{n-k}}{\binom{N}{n}} = r \frac{\binom{N-1}{n-1}}{\binom{N}{n}} \\ &= r \frac{n}{N} = n \frac{r}{N}. \end{aligned} \quad (3)$$

Since  $r/N$  is the fraction of red chips in the urn, which is analogous to  $\alpha$  for the binomial, the expectation has exactly the same form as  $n\alpha$  for the binomial. How cool is that!

For example, the expected number of red chips when 2 are drawn from an urn of  $r = 3$ ,  $w = 2$ , is  $4/5$ .

How does this change if the chips are drawn *with replacement*?