

**Problem Set 3G: Solutions**  
**Due: February 23, 2005**

G1. We know that the PMF for a binomial random variable is:

$$\mathbf{P}(X = k) = \binom{n}{k} p^k (1-p)^{n-k}$$

To find the value of  $p$  that maximizes  $\mathbf{P}(X = k)$ , we use elementary calculus to determine which value of  $p$  maximizes the a priori probability for any experimental value of  $k$ .

$$\begin{aligned} \frac{d}{dp}[\mathbf{P}(X = k)] &= \binom{n}{k} [kp^{k-1}(1-p)^{n-k} - p^k(n-k)(1-p)^{n-k-1}] = 0 \\ k(1-p) - p(n-k) &= 0 * \\ k - pk - pn + pk &= 0 \\ p &= \boxed{k/n} \end{aligned}$$

For instance, if we observed 20 heads in a series of 45 coin tosses, then according to our maximum likelihood estimator, the best estimate for  $p$  (the probability of a head on any toss) would be  $p = \frac{20}{45}$

\*note, we divided through by the terms  $p^{k-1}$  and  $(1-p)^{n-k-1}$  in step 2

G2. Notice that only the relative distance between the fly and the spider matters here, and not the absolute positions of the fly and the spider.

Denote:

- $A_d$  the event that initially the spider and the fly are  $d$  units apart.
- $B_d$  the event that after one second the spider and the fly are  $d$  units apart.

Our approach will be to first apply the (conditional version of the) total expectation theorem to compute  $\mathbf{E}(T | A_1)$ , then use the result to compute  $\mathbf{E}(T | A_2)$ , and similarly compute sequentially  $\mathbf{E}(T | A_d)$  for all relevant values of  $d$ . We will then apply the (unconditional version of the) total expectation theorem to compute  $\mathbf{E}(T)$ .

We have

$$A_d = A_d B_d + A_d B_{d-1} + A_d B_{d-2}, \quad \text{if } d > 1.$$

This is because if the spider and the fly are at a distance  $d > 1$  apart, then one second later their distance will be  $d$  (if the fly moved away from the spider) or  $d - 1$  (if the fly did not move) or  $d - 2$  (if the fly moved towards the spider). We also have, for the case where the spider and the fly start one unit apart,

$$A_1 = A_1 B_1 + A_1 B_0.$$

Using the total expectation theorem, we obtain

$$\begin{aligned}\mathbf{E}(T | A_d) &= \mathbf{P}(A_d B_d) \mathbf{E}(T | A_d B_d) \\ &\quad + \mathbf{P}(A_d B_{d-1}) \mathbf{E}(T | A_d B_{d-1}) \quad \text{if } d > 1, \\ &\quad + \mathbf{P}(A_d B_{d-2}) \mathbf{E}(T | A_d B_{d-2})\end{aligned}$$

while for the case  $d = 1$ ,

$$\mathbf{E}(T | A_1) = \mathbf{P}(A_1 B_1) \mathbf{E}(T | A_1 B_1) + \mathbf{P}(A_1 B_0) \mathbf{E}(T | A_1 B_0).$$

It can be seen based on the problem data that

$$\begin{aligned}\mathbf{P}(A_1 B_1) &= 2p, & \mathbf{P}(A_1 B_0) &= 1 - 2p, \\ \mathbf{E}(T | A_1 B_1) &= 1 + \mathbf{E}(T | A_1), & \mathbf{E}(T | A_1 B_0) &= 1,\end{aligned}$$

so by applying the theorem with  $d = 1$ , we obtain

$$\mathbf{E}(T | A_1) = 2p(1 + \mathbf{E}(T | A_1)) + (1 - 2p),$$

or

$$\mathbf{E}(T | A_1) = \frac{1}{1 - 2p}.$$

By applying the theorem with  $d = 2$ , we obtain

$$\mathbf{E}(T | A_2) = p \mathbf{E}(T | A_2 B_2) + (1 - 2p) \mathbf{E}(T | A_2 B_1) + p \mathbf{E}(T | A_2 B_0).$$

We have

$$\begin{aligned}\mathbf{E}(T | A_2 B_0) &= 1, \\ \mathbf{E}(T | A_2 B_1) &= 1 + \mathbf{E}(T | A_1), \\ \mathbf{E}(T | A_2 B_2) &= 1 + \mathbf{E}(T | A_2),\end{aligned}$$

so by substituting these relations in the expression for  $\mathbf{E}(T | A_2)$ , we obtain

$$\begin{aligned}\mathbf{E}(T | A_2) &= p(1 + \mathbf{E}(T | A_2)) + (1 - 2p)(1 + \mathbf{E}(T | A_1)) + p \\ &= p(1 + \mathbf{E}(T | A_2)) + (1 - 2p) \cdot \left(1 + \frac{1}{1 - 2p}\right) + p.\end{aligned}$$

This equation yields after some calculation

$$\mathbf{E}(T | A_2) = \frac{2}{1 - p}.$$

Similarly, we obtain for  $d > 2$ ,

$$\begin{aligned}\mathbf{E}(T | A_d) &= p(1 + \mathbf{E}(T | A_d)) \\ &\quad + (1 - 2p)(1 + \mathbf{E}(T | A_{d-1})) \\ &\quad + p(1 + \mathbf{E}(T | A_{d-2})),\end{aligned}$$

so  $\mathbf{E}(T | A_d)$  can be generated recursively for any initial distance  $d$ , using as initial conditions the values of  $\mathbf{E}(T | A_1)$  and  $\mathbf{E}(T | A_2)$  obtained earlier.

Finally, the expected value of  $T$  can then be obtained using the given PMF for the initial distance  $d$  and the total expectation theorem:

$$\mathbf{E}(T) = \sum_{d_0} p_d(d_0) \mathbf{E}(T | A_{d_0}).$$