## ÉCOLE POLYTECHNIQUE FÉDÉRALE DE LAUSANNE

School of Computer and Communication Sciences

Handout 3
Solutions to homework 1

Information Theory and Coding Sep. 24, 2013

PROBLEM 1. Note that  $E_0 = E_1 \cup E_2 \cup E_3$ .

- (a) (1) For disjoint events,  $P(E_0) = P(E_1) + P(E_2) + P(E_3)$ , so  $P(E_0) = 3/4$ .
  - (2) For independent events,  $1 P(E_0)$  is the probability that none of the events occur, which is the product of the probabilities that each one doesn't occur. Thus  $1 P(E_0) = (3/4)^3$  and  $P(E_0) = 37/64$ .
  - (3) If  $E_1 = E_2 = E_3$ , then  $E_0 = E_1$  and  $P(E_0) = 1/4$ .
- (b) (1) From the Venn diagram in Fig. 1,  $P(E_0)$  is clearly maximized when the events are disjoint, so max  $P(E_0) = 3/4$ .

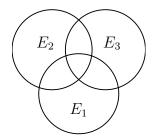


Figure 1: Venn Diagram for problem 1 (b)(1)

(2) The intersection of each pair of sets has probability 1/16. As seen in Fig. 2,  $P(E_0)$  is maximized if all these pairwise intersections are identical, in which case  $P(E_0) = 3(1/4 - 1/16) + 1/16 = 5/8$ . One can also use the formula  $P(E_0) = P(E_1) + P(E_2) + P(E_3) - P(E_1 \cap E_2) - P(E_1 \cap E_3) - P(E_2 \cap E_3) + P(E_1 \cap E_2 \cap E_3)$ , and notice that all the terms except the last is fixed by the problem, and the last term cannot be made more than  $\min_{i,j} P(E_i \cap E_j) = 1/16$ .

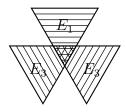


Figure 2: Venn Diagram for problem 1 (b)(2)

(c) Same considerations as in (b)(2) yields the upper bound  $P(E_0) \leq 3p-2p^2$  As  $P(E_0) = 1$ , we find that  $p \geq 1/2$ .

PROBLEM 2. Let L be the event that the loaded die is picked and H the event that the honest die is picked. Let  $A_i$  be the event that i is turned up on the first roll, and  $B_i$  be the event that i is turned up on the second roll. We are given that P(L) = 1/3, P(H) = 2/3;  $P(A_i | L) = 2/3$ ;  $P(A_i | L) = 1/15$   $2 \le i \le 6$ ;  $P(A_i | H) = 1/6$   $1 \le i \le 6$ . Then

$$P(L \mid A_{1}) = \frac{P(L, A_{1})}{P(A_{1})} = \frac{P(A_{1} \mid L) P(L)}{P(A_{1} \mid L) P(L) + P(A_{1} \mid H) P(H)} = \frac{2}{3}.$$

This is the probability that the loaded die was picked conditional on the first roll showing a 1. For two rolls we make the assumption from the physical mechanism involved in rolling a die that the outcome on the two successive rolls of a given die are independent. Thus  $P(A_1B_1 \mid L) = (2/3)^2$  and  $P(A_1B_1 \mid H) = (1/6)^2$ . It follows as before that

$$P\left(L\mid A_1B_1\right) = \frac{8}{9}.$$

PROBLEM 3. Since A, B, C, D form a Markov chain their probability distribution is given by

$$p(a)p(b|a)p(c|b)p(d|c) (1)$$

- (a) Yes: Summing (1) over d shows that A, B, C have the probability distribution p(a)p(b|a)p(c|b).
- (b) Yes: The reverse of a Markov chain is also a Markov chain. Applying this to A, B, C, D and using part (a) we get that D, C, B is a Markov chain. Reversing again we get the desired result.
- (c) Yes: Since A, B, C, D is a Markov chain, given C, D is independent of B, and thus p(d|c) = p(d|(b,c)). So (1) can be written as

$$p(a, (b, c), d) = p(a)p((b, c)|a)p(d|(b, c)).$$

(d) Yes, by a similar (in fact easier) reasoning as (c).

PROBLEM 4. No. Take for example A = D and let A be independent of the pair (B, C). Then both A, B, C and B, C, A (same as B, C, D) are Markov chains. But A, B, C, D is not: A is not independent of D when B and C are given.

Problem 5.

(a)

$$E[X + Y] = \sum_{x,y} (x + y) P_{XY}(x,y)$$

$$= \sum_{x,y} x P_{XY}(x,y) + \sum_{x,y} y P_{XY}(x,y)$$

$$= \sum_{x} x P_{X}(x) + \sum_{y} y P_{Y}(y)$$

$$= E[X] + E[Y].$$

Note that independence is not necessary here and that the argument extends to nondiscrete variables if the expectation exists. (b)

$$E[XY] = \sum_{x,y} xy P_{XY}(x,y)$$

$$= \sum_{x,y} xy P_X(x) P_Y(y)$$

$$= \sum_x x P_X(x) \sum_y y P_Y(y)$$

$$= E[X] E[Y].$$

Note that the statistical independence was used on the second line. Let X and Y take on only the values  $\pm 1$  and 0. An example of uncorrelated but dependent variables is

$$P_{XY}(1,0) = P_{XY}(0,1) = P_{XY}(-1,0) = P_{XY}(0,-1) = \frac{1}{4}.$$

An example of correlated and dependent variables is

$$P_{XY}(1,1) = P_{XY}(-1,-1) = \frac{1}{2}.$$

(c) Using (a), we have

$$\sigma_{X+Y}^2 = E[(X - E[X] + Y - E[Y])^2]$$
  
=  $E[(X - E[X])^2] + 2E[(X - E[X])(Y - E[Y])] + E[(Y - E[Y])^2].$ 

The middle term, from (a), is 2(E[XY] - E[X]E[Y]). For uncorrelated variables that is zero, leaving us with  $\sigma_{X+Y}^2 = \sigma_X^2 + \sigma_Y^2$ .

PROBLEM 6. We solve the problem for a general vehicle with n wheels.

- (a) Out of n! possible orderings (n-1)! has the tyre 1 in its original place. Thus tyre 1 is installed in its original position with probability 1/n.
- (b) All tyres end up in their original position in only 1 of the n! orders. Thus the probability of this event is 1/n!.
- (c) Let  $X_i$  be the indicator random variable that tyre i is installed in its original position, so that the number of tyres installed in their original positions is  $N = \sum_{i=1}^{n} X_i$ . By (a),  $E[X_i] = 1/n$ . By the linearity of expectation, E[N] = n(1/n) = 1. Note that the linearity of the expectation holds even if the  $X_i$ 's are not independent (as it is in this case).
- (e) Let  $A_i$  be the event that the *i*th tyre remains in its original position. Then, the event we are interested in is the complement of the event  $\bigcup_i A_i$  and thus has probability  $1 \Pr(\bigcup_i A_i)$ . Furthermore, by the inclusion/exclusion formula,

$$\Pr(\bigcup_{i} A_{i}) = \sum_{i} \Pr(A_{i}) - \sum_{i_{1} < i_{2}} \Pr(A_{i_{1}} \cap A_{i_{2}}) + \sum_{i_{1} < i_{2} < i_{3}} P(A_{i_{1}} \cap A_{i_{2}} \cap A_{i_{3}}) - \dots$$

The jth sum above consists of  $\binom{n}{j}$  terms, each term having the value  $P(A_1 \cap \cdots \cap A_j)$ . Note that this is the probability of the event that tyres 1 through j have remained in their original positions, and equals (n-j)!/n!. Consequently,

$$\Pr\left(\bigcup_{i} A_{i}\right) = \sum_{j=1}^{n} (-1)^{j-1} \binom{n}{j} \frac{(n-j)!}{n!} = \sum_{j=1}^{n} (-1)^{j-1} 1/j!,$$

and the event that no tyre remains in its original position has probability

$$1 - \Pr\left(\bigcup_{i} A_{i}\right) = \sum_{j=0}^{n} \frac{(-1)^{j}}{j!}.$$

(For the case n = 4, the value is 3/8.)

## Problem 7.

(a) Let  $A_i$  denote the event that  $X_i \neq X$ . The event that X does not appear in the inventory is thus

$$A = A_1 \cap \dots A_n$$
.

Note that the events  $A_1, \ldots, A_n$  are not independent—because they involve the common random variable X. However, they become independent when conditioned on the value of X, with  $P(A_i|X=x)=1-p(x)$ . Thus,

$$P(A|X = x) = (1 - p(x))^n.$$

Consequently  $P(A) = \sum_{x} p(x)(1 - p(x))^{n}$ ...

- (b) With p the uniform distribution on n items, the above value for P(A) equals  $(1 1/n)^n$ .
- (c) For n large,  $(1 1/n)^n$  approaches  $1/e \approx 37\%$ .