

## CS 535: Notes II

### 1 Independence

Two events  $A$  and  $B$  are independent  $Pr(A \cap B) = Pr(A) \times Pr(B)$ . Observe that if  $A$  and  $B$  are independent, then  $\overline{A}$  and  $\overline{B}$  are also independent.

Consider the following experiment. Toss a fair coin, if the outcome is  $H$  then output 1, otherwise output 0. Toss another fair coin, if the outcome is  $H$  then output 1, otherwise output 0. Finally output the *xor* of the first two bits. In this experiment the sample space is  $\{000, 011, 101, 110\}$  and each sample point has probability  $1/4$ . Let  $E_i$ ,  $1 \leq i \leq 3$ , be the event that the  $i$ th bit is 1. We can see that  $E_1$  and  $E_3$  are independent, similarly  $E_2$  and  $E_3$  are also independent. Now consider the experiment where the second coin is a biased coin with probability of  $H$  being  $3/4$  (and the first coin is a fair coin). Are  $E_2$  and  $E_3$  independent? How about  $E_1$  and  $E_3$ ?

Let us return to the earlier experiment of tossing a fair coin  $n$  times. Now let us consider the probability that we see  $\log n/2$  consecutive heads. Let us denote this event with  $E$ . Divide the  $n$  coin tosses into  $m = 2n/\log n$  groups of size  $\log n/2$ . Let us call these groups  $G_1, \dots, G_m$ . Let  $E_i$  denote the probability that all coin tosses in group  $G_i$  are heads. Note that the events  $E_1, E_2, \dots, E_m$  are all mutually independent, and for every  $i$ ,  $1 \leq i \leq m$ ,  $P(E_i) = (1/2)^{\log n/2} = \frac{1}{\sqrt{n}}$ . Since  $\cup_i E_i \subseteq E$ ,

$$\begin{aligned} P(E) &\geq P(\cup_i E_i) \\ &= 1 - P(\overline{\cup_i E_i}) \\ &= 1 - P(\overline{E_1} \cap \overline{E_2} \cap \dots \cap \overline{E_m}) \end{aligned}$$

Since  $\overline{E_1}, \dots, \overline{E_m}$  are mutually independent and  $P(\overline{E_i}) = 1 - \frac{1}{\sqrt{n}}$ ,

$$\begin{aligned} P(\overline{E_1} \cap \overline{E_2} \cap \dots \cap \overline{E_m}) &= \prod_i P(\overline{E_i}) \\ &= \left(1 - \frac{1}{\sqrt{n}}\right)^{2n/\log n} \\ &= \left(\left(1 - \frac{1}{\sqrt{n}}\right)^{\sqrt{n}}\right)^{2\sqrt{n}/\log n} \\ &= (1/e)^{2\sqrt{n}/\log n} \end{aligned}$$

Thus the probability that we see at least  $\log n/2$  consecutive heads is very close to 1. Thus with very high probability that numbers of consecutive heads lie between  $\log n/2$  and  $2 \log n$ .

## 2 Random Variables

A *random variable*  $X$  is a function from sample space to the real numbers,  $X : \Omega \rightarrow \mathbb{R}$ . Given a random variable  $X$  and a value  $\alpha$ ,

$$\begin{aligned}\Pr[X = \alpha] &= \Pr[\{v \mid X(v) = \alpha\}] \\ \Pr[X \geq \alpha] &= \Pr[\{v \mid X(v) \geq \alpha\}] \\ \Pr[X < \alpha] &= \Pr[\{v \mid X(v) < \alpha\}]\end{aligned}$$

Examples of random variables:

- Toss a coin  $n$  times. Number of heads.
- Toss a coin  $n$  times. Maximum number of consecutive heads.
- Roll a dice twice. Sum of the two outcomes.
- Toss a coin. The value of  $X$  is 1 if the outcome is head; otherwise  $X$  is 0.

The *expectation* of a random variable is defined as

$$E(X) = \sum_{\alpha} \Pr(X = \alpha) \times \alpha.$$

$E(X)$  is *average* value of  $X$ .

Two random variable  $X$  and  $Y$  are independent if for every  $\alpha$  and  $\beta$  the events  $\{v \mid X(v) = \alpha\}$  and  $\{v \mid Y(v) = \beta\}$  are independent. This definition can be extended to more than two random variables.

Consider random variable from fourth experiment above. The expectation of  $X$  is calculated as follows:

$$1 \times \Pr[X = 1] + 0 \times \Pr[X = 0] = 1/2.$$

Throw a dice. Let  $Y$  be the random variable that denotes the outcome.

$$E[Y] = \sum_{i=1}^6 i \times \Pr[Y = i] = 1/6 + 2/6 + 3/6 + 4/6 + 5/6 + 6/6 = 3.5$$

Consider the experiment with  $n$  fair coin tosses. We can define a random variable  $X$  to be the number of heads. What is the expectation of  $X$ ? It can be calculated by using the definition of expectation.

$$\begin{aligned}E[X] &= \sum_{i=0}^n \Pr[X = i] \times i \\ &= \sum_{i=0}^n \binom{n}{i} \frac{1}{2^n}\end{aligned}$$

It is not that easy to compute the value of the expression  $\sum_{i=0}^n \binom{n}{i} \frac{1}{2^n}$ . We will compute the expectation using a different route. A very useful property of expectation is that it is linear, i.e., If  $X$  and  $Y$  are two random variable and  $a$  is a real number, then

$$E(aX + Y) = aE(X) + E(Y).$$

This property can be used to compute the expectation of some random variables easily. Let  $X$  be the number of heads in  $n$  fair coin tosses. For  $1 \leq i \leq n$ , define random variable  $X_i$  as follows:  $X_i = 1$ , if the  $i$ th coin toss is a head, else  $X_i$  is zero.

$$E(X_i) = \Pr(X_i = 1) \times 1 + \Pr(X_i = 0) \times 0 = 1/2.$$

It is clear that  $X = X_1 + X_2 + \dots + X_n$ .

$$\begin{aligned} E(X) &= E(X_1) + E(X_2) + \dots + E(X_n) \\ &= 1/2 + 1/2 + \dots + 1/2 \\ &= n/2. \end{aligned}$$

Let  $\phi(x_1, \dots, x_n)$  be a 3-CNF formula with  $n$  variables and  $m$  clauses. Randomly assign a value to  $x_i$ ,  $1 \leq i \leq n$ , from  $\{T, F\}$ . Let  $X$  denote the number of clauses satisfied. We can again calculate the expectation of  $X$  using linearity. Let  $X_i$  denote a random variable whose value is 1 if  $i$ th clause of  $\phi$  is satisfied; otherwise  $X_i$  is 0. Note that  $X = X_1 + \dots + X_m$ . Thus  $E[X] = E[X_1 + \dots + X_m] = E[X_1] + \dots + E[X_m]$ . Given  $X_i$  what is the expectation of  $X_i$ ?

$$E[X_i] = 1 \times \Pr[X_i = 1] + 0 \times \Pr[X_i = 0] = \Pr[X_i = 1]$$

Since a clause has three distinct variables, and the clause is a disjunction; the probability that a random assignment satisfies the clause is  $7/8$ . Thus  $\Pr[X_i = 1] = 7/8$ . Thus  $E[X_i] = 7/8$ , thus  $E[X] = 7m/8$ .

Let  $X$  be a random variable that takes values in  $\{0, 1, 2, \dots\}$ .

**Claim.**  $E(X) = \sum_i P(X > i)$ .

$$\begin{aligned} P(X > 0) &= P(X = 1) + P(X = 2) + P(X = 3) + \dots \\ P(X > 1) &= \phantom{P(X = 1)} + P(X = 2) + P(X = 3) + \dots \\ P(X > 2) &= \phantom{P(X = 1)} \phantom{+ P(X = 2)} + P(X = 3) + \dots \end{aligned}$$

Thus

$$\begin{aligned} \sum_i P(X > i) &= \sum_{i=1} i \times P(X = i) \\ &= E(X) \end{aligned}$$

Suppose we have biased coin with probability of head being  $p$ . Let us consider the following experiment: Toss the coin till head appears. Let  $X$  be a random variable that denotes the number of coin tosses made. What is the expectation of  $X$ ? Since  $X$  takes values in  $\{0, 1, \dots\}$ , by previous Claim,

$$E(X) = \sum_i P(X > i).$$

$P(X > i)$  is the probability that the first  $i$  tosses result in tails. Thus  $P(X > i) = (1 - p)^i$ . Thus

$$\begin{aligned} E(X) &= \sum_i P(X > i) \\ &= \sum_i (1 - p)^i \\ &= \frac{1}{p} \end{aligned}$$