## CMSC 27400-1/37200-1 Combinatorics and Probability

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Lecture 19: May , 2005

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NOTE: Change in Monday's TA schedule; no change Tuesday and Thursday. TA SCHEDULE: TA sessions are held in Ryerson-255, Monday 7:30-8:30,

Tuesday and Thursday 5:30-6:30pm.

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## Martingales

Definition: If X, Y are random variables in the same probability space then the conditional expectation of X conditioned on Y, is a random variable  $Z := E(X|Y) : \Omega \to \mathbb{R}$  defined as Z(a) = E(X|Y = Y(a)).

Some Observations:

- (i) If c is a contant random variable, then E(X|c) = E(X).
- (ii) The number of different values taken by  $E(X|Y) \leq$  the number of different values taken by Y.
- (iii) If X and Y are independent then E(X|Y) = E(X).
- (iv) E(X|X) = X.
- (v)  $E(X^2|X) = X^2$ .

 $E(X|X^2)$  has no simple formula. It depends on the proportion of the positive and negative values of X in each part  $X^2 = a$ .

Definition: The stochastic process is a sequence of random variables,  $X_0, X_1, X_2, \ldots$ 

Example of a stochastic process: Gambler's ruin

Suppose a gambler starts with amount  $X_0$ . At every gamble, he loses with probability  $\frac{1}{2}$  and wins with probability  $\frac{1}{2}$ . The amount decreases by 1 when he loses and increases by 1 when he wins.  $X_i$  is the random variable denoting the amount that the gambler has after i gambles.  $X_0, X_1, \ldots$  is a stochastic process.

$$E(X_1) = X_0.$$

 $E(X_2) = X_0$ . (Why?)

 $E(X_2|X_1) = X_1.$ 

 $E(X_{i+1}|X_i) = X_i.$ 

Definition: The martingale is a stochastic process in which  $(\forall i)(E(X_{i+1}|X_i)=X_i)$ .

Theorem 19.1 (Hoeffding-Azuma: Concentration Inequality) Suppose  $X_0 = c$ , a constant.  $X_0, X_1, X_2, \ldots, X_n$  is a martingale and  $(\forall i)(|X_{i+1} - X_i| \leq 1)$  then

$$P(|X_n - X_0| \ge a) \le 2e^{\frac{-a^2}{2n}}$$
.

The Chernoff's bound is a special case of the above theorem. (Why?)

**Exercise 19.2** For i = 0, 1, ..., let  $Y_i$  be independent random variables such that  $|Y_i| \le 1$  and  $E(Y_i) = 0$ . Let  $X_i = Y_1 + \cdots + Y_i$ . Prove that  $X_0, X_1, dots$  is a martingale.

## Doob-martingale

Let  $Y_0 = c$ , a constant. Let  $Y_0, Y_1, \ldots, Y_n$  be a stochastic process. Let  $X_i = E(Y_n|Y_i)$ . So  $X_0 = E(Y_n)$ .  $X_n = E(Y_n|Y_n) = Y_n$ .  $X_0, X_1, \ldots, X_n$  is called Doob-martingale.

Exercise 19.3 
$$E(X_i) = E(Y_n) = X_0$$
.  $E(X_{i+1}|X_i) = X_i$ .

Chromatic number of random graphs Recall that for any graph G,  $\chi(G) \geq \frac{n}{\alpha(G)}$ .

**Theorem 19.4** For a random graph G on n vertices, almost always  $\alpha(G) = \Theta(\log n)$ , i.e.,  $P(\alpha(G) = \Theta(\log n)) \to 1$  as  $n \to \infty$ .

Exercise 19.5  $\chi(G) = \Theta(\frac{n}{\log n})$  almost always.

**Theorem 19.6**  $\exists k(n)$ )( $\forall g(n) \rightarrow infty$ )(almost always  $|\chi(G) - k(n)| < g(n)\sqrt{n}$ ).

Exercise 19.7 Modify the proof of Chernoff to prove this.

Exercise 19.8 Prove the theorem.