## STAT241/251 Lecture Notes Chapter 6 Part 1

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## Discrete random variables

First, We recall some concepts from Ch. 4. An example of discrete r.v.:

| X      | 1   | 2   | 3   | 4   |
|--------|-----|-----|-----|-----|
| P(X=x) | 0.4 | 0.3 | 0.2 | 0.1 |

How to find E(X)?

$$E(X) = (1 \times 0.4) + (2 \times 0.3) + (3 \times 0.2) + (4 \times 0.1)$$

$$= 2$$

$$Var(X) = E(X^{2}) - [E(X)]^{2}$$

$$E(X) = (1^{2} \times 0.4) + (2^{2} \times 0.3) + (3^{2} \times 0.2) + (4^{2} \times 0.1)$$

$$= 5$$

$$Var(X) = E(X^{2}) - [E(X)]^{2}$$

$$= 5 - 2^{2}$$

$$= 5 - 4$$

$$= 1$$

So if you need  $E(X^3)$ , what do you do?

Answer: 
$$E(X^3) = (1^3 \times 0.4) + (2^3 \times 0.3) + (3^3 \times 0.2) + (4^3 \times 0.1)$$

Note: For continuous r.v., we have pdf (probability density function). For discrete r.v., we don't call it a pdf but instead, we call it a pmf (probability mass function).

A pmf can be in the form of an equation, such as

$$P(X = x) = {}^{n}C_{x}p^{x}(1-p)^{n-x}, x = 0, 1, 2, 3, \dots, n$$

or it can be in the form of a table:

For discrete r.v., the meaning of cdf is the same as for continuous r.v.

e.g. 
$$F(x) = P(X \le x)$$
 note the sign, it is  $\le$  not  $<$ 

However, you have to be very careful when dealing with  $P(a \le X \le b)$  because

$$P(a \le X \le b) = F(b) - F(a - b)$$

Why? because you want to <u>include</u> a. this means largest possible value of X less than a

Chapter 6 covers the following discrete random variables.

- (a) Bernoulli r.v.—We start with these two.
- (b) Binonial r.v.
- (c) Geometric r.v.
- (d) Poisson r.v.

## Bernoulli r.v.

If a random variable can take on only one of two possible values (e.g. head or tail, success or failure, etc) then we call that a Bernoulli random variable.

We usually assign the 2 possible values 0 and 1.

The pmf of a Bernoulli r.v. is

$$P(X = x) = \begin{cases} 1 - p, & x = 0\\ p, & x = 1\\ 0, & \text{otherwise} \end{cases}$$

Note: We can also express the above pmf in a single line:

$$P(X = x) = (1 - p)^{1-x}p^x$$
 x=0,1

To see that both ways of expressing are really the same, check: when x=0,

$$P(X = 0) = (1 - p)^{1-0}p^{0}$$
$$= 1 - p$$

when x = 1,

$$P(X = 1) = (1 - p)^{1 - 1} p^{1}$$
$$= p$$

You should be able to calculate E(X) and Var(X) for a Bernoulli random variable. You will probably recall how to calculate them if you draw a table as we did in class:

| X      | 0   | 1 |
|--------|-----|---|
| P(X=x) | 1-p | р |

$$E(X) = 0 \times (1 - p) + (1 \times p)$$

$$= p$$

$$Var(X) = E(X^{2}) - [E(X)]^{2}$$

$$E(X^{2}) = [0^{2} \cdot (1 - p)] + [1^{2} \cdot p]$$

$$= p$$

$$Var(X) = E(X^{2}) - [E(X)]^{2}$$

$$= p - p^{2}$$

$$= p(1 - p)$$