

## Lecture 6

### Discrete Random Variables and Probability Distributions

(DRV)

Recall that a discrete random variable,  $X$ , is a function that assigns real values to possible outcomes of a sample space  $S$ , and such that the range of  $X$  is finite or countably infinite.

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A probability distribution for a random variable  $X$  is a description of the probabilities associated with the possible values of  $X$ .

For a DRV, this is often just a list of the possible outcomes together with probabilities for each. (More complicated situations  $\rightarrow$  a formula).

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Ex: Consider the experiment where we toss three fair coins. Let  $H$  be the random variable "# of heads".

The range of  $H$  is

range  $\{0, 1, 2, 3\}$ ,

so  $H$  is a DVR.

The probability distribution is given by:

$$P(H=0) = P(t, t, t) = 1/8$$

$$P(H=1) = P(tth, thh, htt) = 3/8$$

$$P(H=2) = P(hht, hth, thh) = 3/8$$

$$P(H=3) = P(hhh) = 1/8.$$

When the math concept is taught in its most basic form in class and the homework requires extra steps



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Sometimes, a probability distribution can be conveniently described by a function, called the probability mass function

Definition: Let  $X$  be a discrete random variable with range  $\{x_1, x_2, \dots\}$ . Then a probability mass function is a function  $f$  such that:

1)  $f(x_i) \geq 0 \quad \forall i$

2)  $\sum_i f(x_i) = 1.$

3)  $f(x_i) = P(\text{X} = x_i)$

Ex: In the coin tossing experiment we can set  $f(x) = \frac{\binom{3}{x}}{8}$ .

Note:  $1/8 + 3/8 + 3/8 + 1/8 = 1$ .

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Cumulative Distribution functions.

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Another way to describe a DVR is via a cumulative distribution function:

Defn: Given a DVR  $X$  and probability mass function  $f(x)$ , the cumulative Distribution Function of  $X$  is defined to be

$$F(x) = P(X \leq x) = \sum_{x_i \leq x} f(x_i).$$

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Note that Given a cumulative distribution function  $F(x)$  for a DVR, we can recover the probability mass function as

$$\begin{aligned} f(x_n) &= F(x_n) - F(x_{n-1}) \\ &= \sum_{x_i \leq x_n} f(x_i) - \sum_{x_i \leq x_{n-1}} f(x_i) \end{aligned}$$

Cumulative distribution functions satisfy the following (for a DVR):

$$1) F(x) = P(X \leq x) = \sum_{x_i \leq x} f(x_i)$$

$$2) 0 \leq F(x) \leq 1.$$

$$3) x \leq y \Rightarrow F(x) \leq F(y).$$

Ex: we can have a DVR  $X$  with Range  $\{1, 2, 3, \dots\}$  and  $P(X=n) = \frac{1}{2^n}$ .

$$\text{Then } F(n) = \sum_{i=1}^n \frac{1}{2^i} = \frac{1(1 - \frac{1}{2}^n)}{2(1 - \frac{1}{2})}$$

In particular,  $\lim_{n \rightarrow \infty} F(n) = 1$ .

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### Mean and Variance of a PRV

- Two numbers used to summarize a probability distribution; do not determine it uniquely.

- Let  $X$  be a DVR with prob. mass function  $f(x)$ .

Mean (expected Value)

$$\mu := E(X) = \sum_x x f(x)$$

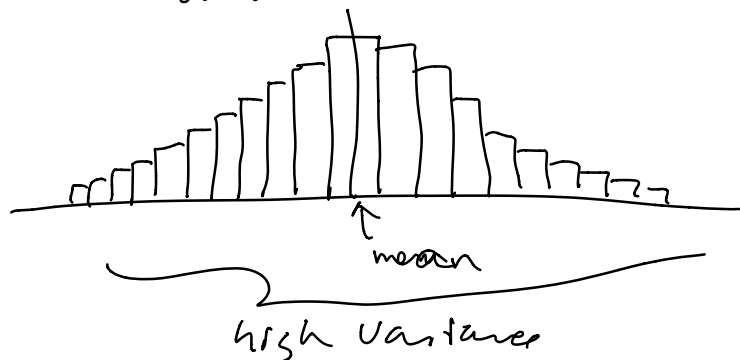
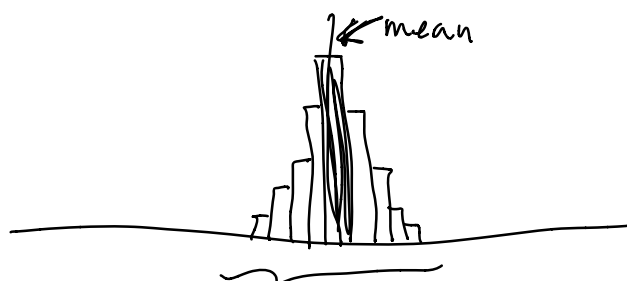
(F) (X)

- is the average of the outcomes, weighted by the probability.

## Variance

$$\begin{aligned}\sigma^2 = V(X) &= E(X - \mu)^2 = \sum_x (x - \mu)^2 f(x) \\ &= \sum_x x^2 f(x) - \mu^2\end{aligned}$$

- measures the "spread" of the possible outcomes / how much they vary away from the mean.



Ex. Consider the following game:

- a fair die is rolled
- if the roll  $n$  is 2, 3, 5, you win  $2 \cdot n$  dollars
- if the roll  $n = 1, 4, 6$ , you get nothing.

How much would you pay to play one round of this game?

Answer: We consider the expected value of the game! Let  $W$  be the discrete random variable "winnings", so  $\text{range}(W) = \{0, 2, 4, 6, 10\}$

$$P(W=0) = \frac{3}{6} \text{ (for a roll 1, 4, 6)}$$

$$P(W=4) = \frac{1}{6} \text{ (roll 2)}$$

$$P(W=6) = \frac{1}{6} \text{ (roll 3)}$$

$$P(W=10) = \frac{1}{6} \text{ (roll 5).}$$

$$\begin{aligned} \text{So } \mu = E(W) &= \frac{3}{6} \cdot 0 + \frac{1}{6} \cdot 4 + \frac{1}{6} \cdot 6 + \frac{1}{6} \cdot 10 \\ &= \frac{4 + 6 + 10}{6} = \frac{20}{6} \\ &= \frac{10}{3} \approx \$3.33. \end{aligned}$$

So on average, you should expect to win

\$3.33, so if you pay more than that per round, you should expect to lose money.

(A game where the cost equals the expected value)  
is called a fair game.

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We can also compute the variance of the game:

$$\begin{aligned} V(x) &= \sum_x (x - \mu)^2 f(x) \\ &= \left(0 - \frac{10}{3}\right)^2 \cdot \frac{3}{6} + \left(4 - \frac{10}{3}\right)^2 \frac{1}{6} + \left(6 - \frac{10}{3}\right)^2 \frac{1}{6} + \left(10 - \frac{10}{3}\right)^2 \frac{1}{6} \\ &= 14.22 \end{aligned}$$