

Conditional probability example:

(1)

Suppose that 10 out of 1000 people are affected by a certain disease.

Suppose that, if a person has the disease, the test results will be positive 97% of the time.  
(3% false negatives)

Also suppose that, if the person doesn't have the disease, the test results will be negative 95% of the time.  
(5% false positives)

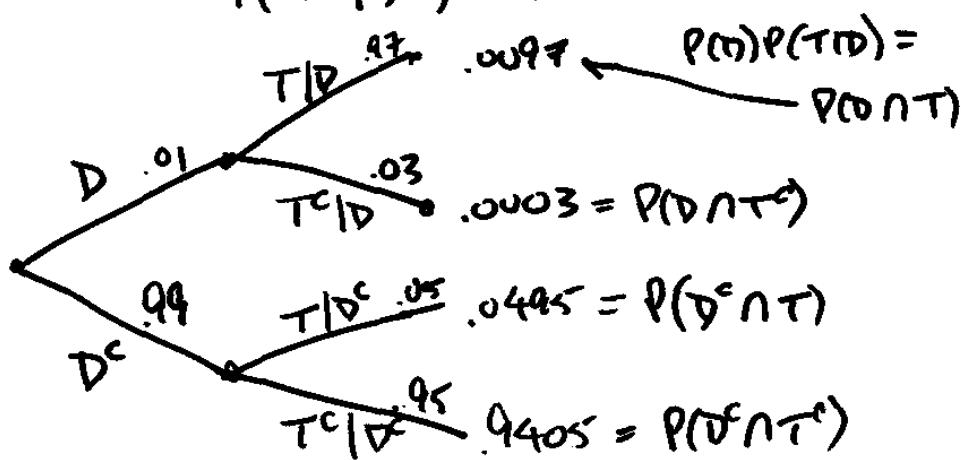
(2)

$$D: \text{disease} \quad P(D) = \frac{10}{1000} = .01$$

T: test results are positive

$$P(T|D) = .97$$

$$P(T^c|D^c) = .95$$



(3)

		$T$	$T^c$	true positives
		.0017	.0003	false negatives
		.0495	.9405	.99
		.0592	.9408	true negatives
<u>false positives</u>				

Question: If a person tests positive, what is the probability that they have the disease?

$$P(D | T) = \frac{P(D \cap T)}{P(T)} = \frac{.0017}{.0592} = .164$$

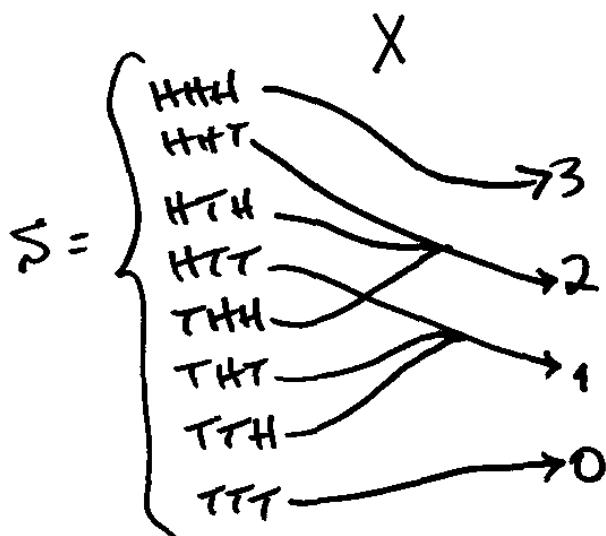
### Random variables

(4)

Defn.: A random variable is a variable whose value is determined by the outcome of an experiment.

Actually, a random variable is a function whose domain is  $S$  and whose range is a subset of  $\mathbb{R}$

Example: Flip 3 coins. Let  $X = \#$  heads (5)



If the range of  $X$  is finite or countably infinite, then  $X$  is called a discrete random variable.

If the range of  $X$  is an interval, then  $X$  is called a continuous random variable.

## Discrete Random variables (6)

Define a probability (mass) function

$$p(x) = P(X=x)$$

Properties:  $0 \leq p(x) \leq 1$

$$\sum_{\text{all } x} p(x) = 1$$

3 coin example:

$x$	$p(x)$
0	1/8
1	3/8
2	3/8
3	1/8 / 1

$p(0) = P(X=0) = P(\text{TTT})$   
 $p(1) = P(X=1) = P(\text{HTT}, \text{THT}, \text{TTH})$

Example: Roll 2 dice

(7)

Let  $Y = \text{smallest of the two}$

$y$	$p(y)$
1	$\frac{1}{36}$
2	$\frac{2}{36}$
3	$\frac{3}{36}$
4	$\frac{4}{36}$
5	$\frac{5}{36}$
6	$\frac{6}{36}$
	1

$$S = \left\{ \begin{array}{cccc} 11 & 12 & \cdots & 16 \\ 21 & 22 & \cdots & 26 \\ \vdots & \vdots & & \vdots \\ 61 & 62 & \cdots & 66 \end{array} \right\}$$

44, 45, 52, 46, 64  
 55, 56, 65

Note:  
 $p(y) = \frac{2(6-y)+1}{36} \quad y=1,2,3,\dots,6$

Defn: For a discrete random variable  $X$ ,

the expected value of  $X$  is

$$\mu = E(X) = \sum_{\text{all } X} x p(x) \quad (\text{1st moment})$$

"mu"

3-Coin example:  $\mu = \sum x p(x)$

$$= 0 \cdot \frac{1}{8} + 1 \cdot \frac{3}{8} + 2 \cdot \frac{3}{8} + 3 \cdot \frac{1}{8}$$

$$= \frac{0+3+6+3}{8} = \frac{12}{8} = 1.5$$

(8)

(9)

Defn:  $E(X^2)$  is called the  $2^{\text{nd}}$  moment

$$= \sum_{\text{all } x} x^2 p(x)$$

3 coin example:

$$\begin{aligned} E(X^2) &= 0^2 \cdot \frac{1}{8} + 1^2 \cdot \frac{3}{8} + 2^2 \cdot \frac{3}{8} + 3^2 \cdot \frac{1}{8} \\ &= \frac{0+3+12+9}{8} = \frac{24}{8} = 3 \end{aligned}$$

(10)

Defn:  $E[(X-\mu)^2]$  is the variance of  $X$

$$\sigma^2 = \text{Var}(X) = \sum_{\text{all } x} (x-\mu)^2 p(x)$$

"Sigma squared"

$$\text{Note: } \sigma^2 = \sum (x-\mu)^2 p(x) = \sum (x^2 - 2\mu x + \mu^2) p(x)$$

$$= \sum x^2 p(x) - 2\mu \underbrace{\sum x p(x)}_{\mu} + \mu^2 \underbrace{\sum p(x)}_1$$

$$= E(X^2) - \mu^2$$

(11)

3 coin example:

$$\sigma^2 = E(X^2) - \mu^2 = 3 - (1.5)^2 \\ = .75$$

Defn: The standard deviation of  $X$  is  $\sigma = \sqrt{\sigma^2}$

HW #2 due 10/13

P.72 # 7, 8, 9, 13

P.78 # 4, 6