

Counting : How many possible outcomes satisfy some event.

### Step Rule of Counting (Product Rule)

If an experiment has two steps, where  
The first step's outcome are from set A,  
where  $|A| = m$ , and the second step's  
outcome are from step B, where  $|B| = n$ ,  
and  $|B|$  is unaffected by outcome of first exp.  
Then the total no. of outcomes of the experiment  
is  $|A||B| = mn$ .

### Sum Rule of Counting

If the outcome of an experiment can be either  
from set A, where  $|A| = m$

or set B, where  $|B| = n$ .

Then the total no. of outcomes =  $|A| + |B|$   
 $= m + n$

The sets should be mutually ~~excl~~ exclusive.

$$\# = |A| + |B| - |A \text{ and } B|$$

↳ total no. of outcomes

## Permutations

A permutation is an ordered arrangement of objects.

The # of unique orderings of  $n$ -distinct objects is

$$n! = n \times (n-1) \times (n-2) \times \dots \times 2 \times 1$$

How do we find the # of permutations considering some objects are indistinct?

By the product rule, permutation of distinct object is a two step process:

permutation of  $n$  distinct objects = Permutation considering some objects are indistinct  $\times$  Permutation of just the indistinct objects

When there are  $n$  objects such that

$n_1$  are the same

$n_2$  are the same

$n_r$  are the same

$$|\text{Set A}| = |A| + |B| + \dots + |r| = \#$$

The # of unique orderings (permutations) is

$$\text{possible arrangements} = n!$$

$$n_1! n_2! \dots n_r!$$

Eg : MISSISSIPPI

# of letter orderings :

$$\frac{11!}{4! 4! 1! 2!}$$

## Combinations

A combination is an unordered selection of  $k$  objects from a set of  $n$  distinct objects.

The # of ways of =  $\frac{n!}{k!(n-k)!}$

$$\frac{n! \times 1}{k!} \times \frac{1}{(n-k)!}$$

↓ ↓ ↑ ↑

Order  $n$   
distinct  
objects

Taken first  
 $k$  as  
chosen

Overcounted  
any ordering of  
chosen group  
is same choice

Overcounted  
Any ordering of  
unchosen grp  
is same choice

Put  $n$  objects in  $r$  buckets:

(i) When the  $n$  objects are distinct

$$\# \text{ of ways} = r^n$$

(ii) When the  $n$  objects are indistinct

Divider Method:

To divide  $n$  objects in  $r$  buckets make  $r-1$  dividers (indistinct)

Step 0: Make objects & dividers distinct

Step 1: Order  $n$  distinct objects and  $r-1$  distinct dividers  $(n+r-1)!$

Step 2: Make  $n$  objects indistinct  $(1/n!)^n$

Step 3: Make  $r-1$  divider indistinct  $(1/(r-1))^{r-1}$

$$\text{Total # of ways} = \frac{(n+r-1)!}{n! (r-1)!}$$

## Sample Space :

Sample Space,  $S$ , is set of all possible outcomes of an experiment.

## Event Space :

Event,  $E$ , is some subset of  $S$ .

$$\{E \subseteq S\}$$

What is probability :

A number between 0 & 1 to which we ascribe meaning (our belief) that an event  $E$  occurs.

## Axioms of Probability :

- Axiom 1 :  $0 \leq P(E) \leq 1$
- Axiom 2 :  $P(S) = 1$
- Axiom 3 : If events  $E$  and  $F$  are mutually exclusive :

$$P(E \cup F) = P(E) + P(F)$$

$$P(E) = \frac{\# \text{ outcomes in } E}{\# \text{ outcomes in } S} = \frac{|E|}{|S|}$$

**?** Make indistinct items distinct to get equally likely sample space outcomes.

Equally Likely Outcomes :

Each possible result of an experiment has the same chance of occurring.

**?** When approaching an "Equally likely probability problem", start by defining sample space and event spaces.

## Conditional Probability

The conditional probability of E given F is the probability that E occurs given that F has already occurred. This is known as conditioning of F.

Written as :  $P(E|F)$

Means :  $P(E, \text{given } F \text{ already observed})$

$P(E|F) = \frac{\# \text{ of outcomes of } E \text{ consistent with } F}{\# \text{ of outcomes of } S \text{ consistent with } F}$

$$= \frac{|EF|}{|SF|} = \frac{|EF|}{|F|}$$

CHAIN RULE

$$P(E|F) = \frac{P(EF)}{P(F)} \Rightarrow P(EP) = P(F) P(E|F)$$

## Law of Total Probability

Let  $F$  be an event where  $P(F) > 0$ . For any event  $E$ ,

$$P(E) = P(E|F)P(F) + P(E|F^c)P(F^c)$$

complement

## Bayes' Theorem

Thm. For any event  $E$  and  $F$  where  $P(E) > 0$  &  $P(F) > 0$

$$P(F|E) = \frac{P(E|F)P(F)}{P(E)}$$

Expanded form :

$$P(F|E) = \frac{P(E|F)P(F)}{P(E|F)P(F) + P(E|F^c)P(F^c)}$$

## Inclusion | Exclusion with Three Events

$$P(E \cup F \cup G) = P(E) + P(F) + P(G) - P(E \cap F) - P(E \cap G) - P(F \cap G) + P(E \cap F \cap G)$$

## Independence

Two events A and B are called Independent if:

$P(A) = P(A|B)$  knowing that B happened  
doesn't change our belief.

Otherwise they are called dependent events.

$A \cap B$	$A$	$B$	$P(A B) = P(A)$
			$\frac{ A \cap B }{ B } = \frac{ A }{ S }$

## Conditional Independence

Two events E and F are called conditionally independent given  $G_1$ , if

$$P(EF|G_1) = P(E|G_1)P(F|G_1)$$

or, equivalently if

$$P(E|FG_1) = P(E|G_1)$$

$$(D = X)q$$