

Principles of Counting

Discrete Uniform Law

- **Uniform** sample space: all sample points are equally likely
- **Discrete** sample space: has finite or countably infinite number of possible outcomes.
- If sample space Ω is discrete and uniform, then:

$$P(A) = \frac{\text{number of elements of } A}{\text{total number of sample points}} = \frac{|A|}{|\Omega|}$$

- Therefore, you need to count number of elements of A , and total number of sample points.
- For example, if $|\Omega| = N$ and $|A| = M$, then each sample point will happen with probability $\frac{1}{N}$. Therefore, A will happen with probability $\frac{M}{N}$.

Basic Counting Principle

- There are r stages.
- There are n_i choices at stage i .
- Then, total number of choices will be $n_1 n_2 \cdots n_r$.

Example:

- Number of license plates with 3 letters and 4 digits:

$$26 \cdot 26 \cdot 26 \cdot 10 \cdot 10 \cdot 10 \cdot 10$$

- Number of license plates with 3 letters and 4 digits, with no repetition:

$$26 \cdot 25 \cdot 24 \cdot 10 \cdot 9 \cdot 8 \cdot 7$$

Basic Counting Principle

- **Permutations:** all different ways of ordering n elements
- Number of permutations for n elements:

$$n \cdot (n - 1) \cdots 2 \cdot 1 = n!$$

Example:

- Permutations of ABC :

$ABC, ACB, BAC, BCA, CAB, CBA$

- $n! = 3! = 6$

Basic Counting Principle

- Number of subsets of $\{1, 2, \dots, n\}$:
- Each element i can be either in the subset or not, therefore we have two options for each element.

- Total number of subsets:

$$2 \cdot 2 \cdots 2 = 2^n$$

- How many subsets include n ? (only one option for n : include it)

$$2 \cdot 2 \cdots 2 \cdot 1 = 2^{n-1}$$

- How many subsets do not include n ? (only one option for n : exclude it)

$$2 \cdot 2 \cdots 2 \cdot 1 = 2^{n-1}$$

Example:

- All subsets of $\{1, 2, 3\}$: $\emptyset, \{1\}, \{2\}, \{3\}, \{1, 2\}, \{1, 3\}, \{2, 3\}, \{1, 2, 3\}$
- $2^3 = 8$ subsets.

Basic Counting Principle

- Probability that six rolls (independent rolls) of a six-sided dice (fair dice) all give different numbers?
 - Number of outcomes that make the event happen: $6!$
 - Number of elements in the sample space: 6^6
 - Answer: $\frac{6!}{6^6}$
- What is the probability of seeing a specific sequence of rolls?
 - $P((2, 3, 3, 1, 6, 5)) = ?$
 - $P((2, 3, 3, 1, 6, 5)) = P(2)P(3)P(3)P(1)P(6)P(5) = \frac{1}{6^6}$

Combinations

- How many ways do we have to choose k elements out of n elements?
- Number of k -element subsets of a given n -element set?
- The answer is $\binom{n}{k} = \frac{n!}{k!(n-k)!}$
- Two ways of constructing an ordered sequence of k distinct items:
 - 1) Choose the k items one at a time:

$$n(n-1)\cdots(n-k+1) = \frac{n!}{(n-k)!}$$

- 2) Choose k items, then order them ($k!$ possible orders):

$$\binom{n}{k} k!$$

- Therefore,

$$\binom{n}{k} k! = \frac{n!}{(n-k)!}$$

Combinations

- $\binom{n}{k}$ is also called “binomial coefficient”
- $\sum_{k=0}^n \binom{n}{k} = \text{Total number of subsets} = 2^n$
- $\binom{n}{0} = \binom{n}{n} = \frac{n!}{n!0!} = 1 \Rightarrow 0! = 1$
- Note that $\binom{n}{k} = \binom{n}{n-k} = \frac{n!}{k!(n-k)!}$
- Binomial formula:

$$(p + q)^n = \sum_{k=0}^n \binom{n}{k} p^k q^{n-k}$$

- If $p = q = 1$:

$$(2)^n = \sum_{k=0}^n \binom{n}{k}$$

Binomial Probabilities

- We have n independent coin tosses.
- $P(\text{Head}) = P(H) = p$ and $P(\text{Tail}) = P(T) = 1 - p$.
- $P(\text{sequence}) = p^{\# \text{ of heads}}(1 - p)^{\# \text{ of tails}}$

$$P(k \text{ heads}) = \sum_{k\text{-head seq}} P(\text{seq}) = (\# \text{ of } k\text{-head seqs}) p^k (1 - p)^{n-k}$$

$$= \binom{n}{k} p^k (1 - p)^{n-k}$$

- $\# \text{ of } k\text{-head seqs} = \binom{n}{k}$: because out of n tosses k of them are heads.

Binomial Probabilities—Example 1

- $P(HTT) = P(H)P(T)P(T) = p(1-p)(1-p) = p(1-p)^2$
- $P(THT) = P(T)P(H)P(T) = (1-p)p(1-p) = p(1-p)^2$
- $P(TTH) = P(T)P(T)P(H) = (1-p)(1-p)p = p(1-p)^2$

$$P(1 \text{ heads}) = \binom{3}{1} p(1-p)^2 = 3p(1-p)^2$$

$$P(1 \text{ heads}) = P(HTT) + P(THT) + P(TTH) = 3p(1-p)^2$$

Binomial Probabilities—Example 2

- $P(HHTT) = P(H)P(H)P(T)P(T) = pp(1-p)(1-p) = p^2(1-p)^2$
- $P(HTHT) = P(H)P(T)P(H)P(T) = p(1-p)p(1-p) = p^2(1-p)^2$
- $P(HTTH) = P(H)P(T)P(T)P(H) = p(1-p)(1-p)p = p^2(1-p)^2$
- $P(THHT) = P(T)P(H)P(H)P(T) = (1-p)pp(1-p) = p^2(1-p)^2$
- $P(THTH) = P(T)P(H)P(T)P(H) = (1-p)p(1-p)p = p^2(1-p)^2$
- $P(TTTH) = P(T)P(T)P(H)P(H) = (1-p)(1-p)pp = p^2(1-p)^2$

$$P(2 \text{ heads out of 4 tosses}) = \binom{4}{2} p^2(1-p)^2 = 6p^2(1-p)^2$$

$$\begin{aligned} &= P(HHTT) + P(HTHT) + P(HTTH) + P(THHT) + P(THTH) + P(TTTH) \\ &= 6p^2(1-p)^2 \end{aligned}$$

Binomial Probabilities

- Claim:

$$\sum_{k=0}^n \binom{n}{k} p^k (1-p)^{n-k} = 1$$

- Proof:

- $\binom{n}{k} p^k (1-p)^{n-k}$ is $P(\text{seeing } k \text{ heads out of } n \text{ tosses})$

- $\sum_{k=0}^n \binom{n}{k} p^k (1-p)^{n-k} =$

$$P(0 \text{ heads}) + P(1 \text{ heads}) + P(2 \text{ heads}) + \cdots + P(n \text{ heads}) = 1$$

- Second proof:

- Use Binomial formula for p and $q = 1 - p$:

$$1 = (p + (1-p))^n = \sum_{k=0}^n \binom{n}{k} p^k (1-p)^{n-k}$$

Coin Tossing Problem

- We have 10 independent coin tosses.
- $P(\text{Head}) = P(H) = p$ and $P(\text{Tail}) = P(T) = 1 - p$
- Event $B = 3$ out of 10 tosses are “heads”
- Event $A =$ first 2 tosses are “heads”, what is $P(A|B)$?
- Number of outcomes in B : $\binom{n}{k} = \binom{10}{3}$
- Number of outcomes in $A \cap B$:
 - first 2 tosses are “heads” and 3 out of 10 tosses are “heads”
 - HH _ _ _ _ _
 - There are 8 positions for the third “head”
 - Therefore, number of outcomes in $A \cap B$ is 8
- All outcomes in set B are equally likely with probability $p^3(1 - p)^7$
- $$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{8 p^3(1-p)^7}{\binom{10}{3} p^3(1-p)^7} = \frac{8}{\binom{10}{3}}$$

Conditional Probability—Discrete Uniform Law

- B is a subset of the sample space Ω and number of outcomes in B is N
- All outcomes in set B are equally likely with probability q
- A is also a subset of the sample space Ω , number of outcomes in A is M and number of outcomes in $A \cap B$ is K
- Then $P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{K \cdot q}{N \cdot q} = \frac{K}{N}$
- **Another Interpretation:**
 - In universe B , there are N outcomes that are equally likely
 - Therefore, probability of seeing one single element in B is $\frac{1}{N}$
 - $A \cap B$ is an event in universe B and has K outcomes.
 - In discrete uniform universe B :

$$P(\text{seeing } K \text{ elements}) = K \cdot P(\text{seeing one element}) = K \cdot \frac{1}{N}$$