Math 3338: Probability (Fall 2006)

Jiwen He

Section Number: 10853

http://math.uh.edu/~jiwenhe/math3338fall06.html



2.2 Axioms, Interpretations, and Properties of Probability



Definition and three axioms

• **Definition:** A *probability measure* on the sample space Ω is a function of *events*,

$$P: A \subset \Omega \mapsto P(A) \in R$$

satisfying three axioms

Axiom 1: For any event $A, P(A) \ge 0$.

Axiom 2: $P(\Omega) = 1$.

Axiom 3: If A_1, A_2, \ldots is an infinite collection of *disjoint events*, then

$$P(A_1 \cup A_2 \cup A_3 \cup \cdots) = \sum_{i=1}^{\infty} P(A_i).$$

Proposition:

- 1. $P(\emptyset) = 0$ where \emptyset is the null event.
- 2. Axiom 3 is valid for a *finite* collection of events.

Proof: 1. Consider $A_1 = \emptyset$, $A_2 = \emptyset$, Axiom 3 gives $P(\emptyset) = \sum P(\emptyset)$, which can happen only if $P(\emptyset) = 0$. 2. Consider k disjoint events A_1, A_2, \ldots, A_k and append to these the infinite collection $A_{k+1} = \emptyset$, $A_{k+2} = \emptyset$, Again invoking Axiom 3,

$$P(\bigcup_{i=1}^{k} A_i) = P(\bigcup_{i=1}^{\infty} A_i) = \sum_{i=1}^{\infty} P(A_i) = \sum_{i=1}^{k} P(A_i)$$



as desired.

Deductions from the axioms

• Proposition:

- 1. For any event A, $P(A) = 1 P(A^c)$. Then $0 \le P(A) \le 1$.
- 2. For any two events A and B,

$$P(A \cup B) = P(A) + P(B) - P(A \cap B).$$

3. For any three events A, B, and C,

 $P(A\cup B\cup C)=P(A)+P(B)+P(C)-P(A\cap B)-P(A\cap C)-P(B\cap C)+P(A\cap B\cap C).$

4. For any *n* events A_1, A_2, \ldots, A_n ,

$$P(\bigcup_{i=1}^{n} A_i) = P_1 - P_2 + P_3 - P_4 + \dots + (-1)^{n+1} P_n$$

where

$$P_1 = \sum_{i=1}^n P(A_i), P_2 = \sum_{1 \le i \le j \le n} P(A_i \cap A_j), P_3 = \sum_{1 \le i \le j \le k \le n} P(A_i \cap A_j \cap A_k), \dots$$

Proof: 1 - 3. by the partition, 4. by induction.





Jiwen He, University of Houston, jiwenhe@math.uh.edu Math 3338: Probability (Fall 2006), August 28- September 1, 2006

Examples

- 2.11: $\Omega = \{U, D\}$. For any $0 \le p \le 1$, P(U) = p, P(D) = 1 p is a probability assignment consistent with the axioms: $1 = P(\Omega) = P(U) + p(D)$.
- 2.11: Ω = {S, FS, FFS, FFFS, For any 0 ≤ p ≤ 1, suppose the probability of any particular battery being satisfactory is p. Then it can be shown that P(S) = p, P(FS) = (1 − p)p, , P(FFS) = (1 − p)²p, ..., is a probability assignment consistent with the axioms:

$$1 = P(\Omega) = P(S) + p(FS) + P(FFS) + \dots = p \sum_{i=0}^{\infty} (1-p)^{i} = p \frac{1}{1-(1-p)}.$$

2.13: Let A be the event that the system fails. Suppose the probability of any particular component being satisfactory is p, 0 ≤ p ≤ 1.

$$P(A) = 1 - P(A^{c}) = 1 - P(SSSSS) = 1 - p^{5}.$$

• **2.14**: P(subscribes to at least one of the two newspapers)

 $P(A \cup B) = P(A) + P(B) - P(A \cap B) = .6 + .8 - .5 = .9$





More properties

Proposition:

1. For any two events A and B such that $A \subset B$,

$$P(B - A) = P(B) - P(A).$$

Then

P(A) $\leq P(B)$ 2. (Boole's Inequality) For any *n* events A_1, A_2, \dots, A_n ,

$$P(\bigcup_{i=1}^{n} A_i) \le \sum_{i=1}^{n} P(A_i)$$

Proof:

- 1. By the partition: B = A + (B A), we have P(B) = P(A) + P(B A).
- 2. By induction. For n = 2, it follows that

$$P(A \cup B) = P(A) + P(B) - P(A \cap B) \le P(A) + P(B).$$

Suppose that it holds for any n-1 events. Then

$$P(\bigcup_{i=1}^{n} A_i) = P((\bigcup_{i=1}^{n-1} A_i) \cup A_n) \le P(\bigcup_{i=1}^{n-1} A_i) + P(A_n)$$
$$\le \sum_{i=1}^{n-1} P(A_i) + P(A_n) = \sum_{i=1}^{n} P(A_i)$$



Determining probability systematically

- A simple way to determine probability measure for any *countable* sample space $\Omega = \{E_1, E_2, \dots, E_i, \dots\}$ is to
 - 1. Determine probabilities $P(E_i)$ for all simple events such that

$$P(E_i) \ge 0, \quad \sum_i P(E_i) = 1.$$

2. For any event A, the probability P(A) is computed by adding together the $P(E_i)$'s for all E_i 's in A:

$$P(A) = \sum_{E_i \in A} P(E_i)$$

Thus, $P(\cdot)$ is a function defined for all events and it can be shown that $P(\cdot)$ is a probability measure satisfying the three axioms.

• Example 2.15 - tossing a die: $\Omega = \{1, 2, \dots, 6\} =: \{E_1, E_2, \dots, E_6\}$. An probability assignment to simple events is

$$P(E_1) = P(E_3) = P(E_5) = 1/9, \quad P(E_2) = P(E_4) = P(E_6) = 2/9.$$

Then for the event $A = \{$ outcome is even $\}$, we have

$$A = \{E_2, E_4, E_6\}, \quad P(A) = P(E_2) + P(E_4) + P(E_6) = 6/9.$$



Equally likely outcomes

In the very special case that Ω is finite and contains exactly N point, i.e.,
Ω = {E₁, E₂,..., E_i,..., E_N} and |Ω| = N, we may attach equal probability p to all N points, P(E_i) = p for every i,

$$1 = \sum_{i=1}^{N} P(E_i) = pN \quad \text{ so that } \quad p = \frac{1}{N}$$

That is, if there are N possible outcomes, then the probability assigned to each is 1/N.

• For any event A, with |A| denoting the number of points contained in A. Then

$$P(A) = \sum_{E_i \in A} P(E_i) = \frac{|A|}{|\Omega|}.$$

- Once we have *counted* the number N of outcomes in the sample spaces, to compute the probability of any event we must *count* the number of outcomes contained in that event and take the ratio. When outcomes are *equally likely*, computing probability reduces to counting.
- Example 2.16 Rolling two dice separately: N = 36. If both the dices are fair, all 36 outcomes are equally likely, so $P(E_i) = 1/36$. Then the event $A = \{\text{sum of two numbers} = 7\}$ consists of the six outcomes: (1, 6), (2, 5), (3, 4), (4, 3), (5, 2), and (6, 1), so



$$P(A) = \frac{|A|}{|\Omega|} = \frac{6}{36} = 1/6.$$