

Probability and Stochastic Modelling 1

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Dr Alexander Browning

TARANG JANAWALKAR

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1 Events and Probability

1.1 Events

Definition 1.1 (Event). An event is a set of outcomes in a random experiment commonly denoted by a capital letter. Events can be simple (a single event) or compound (two or more simple events).

Definition 1.2 (Sample space). The set of all possible outcomes of an experiment is known as the sample space for that experiment and is denoted Ω .

Definition 1.3 (Intersection). An intersection between two events A and B describes the set of outcomes that occur in both A and B . The intersection can be represented using the set AND operator (\cap) — $A \cap B$ (or AB).

Definition 1.4 (Disjoint). Disjoint (mutually exclusive) events are two events that cannot occur simultaneously, or have no common outcomes.

Theorem 1.1.1 (Intersection of disjoint events). *The intersection of disjoint events results in the null set (\emptyset).*

Lemma 1.1.1.1. *Disjoint events are **dependent** events as the occurrence of one means the other cannot occur.*

Definition 1.5 (Union). A union of two events A and B describes the set of outcomes in either A or B . The union is represented using the set OR operator (\cup) — $A \cup B$.

Definition 1.6 (Complement). The complement of an event E is the set of all other outcomes in Ω . The complement of E is denoted \bar{E} .

Theorem 1.1.2 (Intersection of complement set).

$$A\bar{A} = \emptyset$$

Theorem 1.1.3 (Union of complement set).

$$A \cup \bar{A} = \Omega$$

Definition 1.7 (Subset). A is a (non-strict) subset of B if all elements in A are also in B . This can be denoted as $A \subset B$.

Theorem 1.1.4. *All events E are subsets of Ω .*

Theorem 1.1.5. *Given $A \subset B$*

$$AB = A \quad \text{and} \quad A \cup B = B$$

Corollary 1.1.5.1. *Given $\emptyset \subset E$*

$$\emptyset E = \emptyset \quad \text{and} \quad \emptyset \cup E = E$$

Theorem 1.1.6 (Associative Identities).

$$A(BC) = (AB)C$$

$$A \cup (B \cup C) = (A \cup B) \cup C$$

Theorem 1.1.7 (Distributive Identities).

$$A(B \cup C) = AB \cup AC$$

$$A \cup BC = (A \cup B)(A \cup C)$$

1.2 Probability

Definition 1.8 (Probability). Probability is a measure of the likeliness of an event occurring. The probability of an event E is denoted $\Pr(E)$ (sometimes $P(E)$).

$$0 \leq \Pr(E) \leq 1$$

where a probability of 0 never happens, and 1 always happens.

Theorem 1.2.1 (Probability of Ω).

$$\Pr(\Omega) = 1$$

Theorem 1.2.2 (Complement rule). *The probability of the complement of E is given by*

$$\Pr(\bar{E}) = 1 - \Pr(E)$$

Theorem 1.2.3 (Multiplication rule for independent events). *The probability of the intersection between two independent events A and B is given by*

$$\Pr(AB) = \Pr(A) \Pr(B)$$

Theorem 1.2.4 (Addition rule for independent events). *The probability of the union between two independent events A and B is given by*

$$\Pr(A \cup B) = \Pr(A) + \Pr(B) - \Pr(AB).$$

If A and B are disjoint, then $\Pr(AB) = 0$, so that $\Pr(A \cup B) = \Pr(A) + \Pr(B)$.

Corollary 1.2.4.1 (Addition rule for 3 events). *The addition rule for 3 events is as follows*

$$\Pr(A \cup B \cup C) = \Pr(A) + \Pr(B) + \Pr(C) - \Pr(AB) - \Pr(AC) - \Pr(BC) + \Pr(ABC).$$

Proof. If we write $D = A \cup B$ and apply the addition rule twice, we have

$$\begin{aligned} \Pr(A \cup B \cup C) &= \Pr(D \cup C) \\ &= \Pr(D) + \Pr(C) - \Pr(DC) \\ &= \Pr(A \cup B) + \Pr(C) - \Pr((A \cup B)C) \\ &= \Pr(A) + \Pr(B) - \Pr(AB) + \Pr(C) - \Pr(AC \cup BC) \\ &= \Pr(A) + \Pr(B) - \Pr(AB) + \Pr(C) - (\Pr(AC) + \Pr(BC) - \Pr(ACBC)) \\ &= \Pr(A) + \Pr(B) + \Pr(C) - \Pr(AB) - \Pr(AC) - \Pr(BC) + \Pr(ABC) \end{aligned}$$

□

Theorem 1.2.5 (Law of total probability). *By writing the event A as $AB \cup A\bar{B}$, and noting that AB and $A\bar{B}$ are disjoint:*

$$\Pr(A) = \Pr(AB) + \Pr(A\bar{B})$$

Theorem 1.2.6 (De Morgan's laws). *Recall De Morgan's Laws:*

$$\begin{aligned} \overline{A \cup B} &= \bar{A} \bar{B} \\ \overline{AB} &= \bar{A} \cup \bar{B}. \end{aligned}$$

Taking the negation of both sides and applying the complement rule yields

$$\begin{aligned} \Pr(A \cup B) &= 1 - \Pr(\bar{A} \bar{B}) \\ \Pr(AB) &= 1 - \Pr(\bar{A} \cup \bar{B}) \end{aligned}$$

1.3 Circuits

A signal can pass through a circuit if there is a functional path from start to finish.

We can define a circuit where each component i functions with probability p , and is independent of other components.

Then W_i to be the event in which the associated component i functions, we can determine the event S in which the system functions, and probability $\Pr(S)$ that the system functions.

As the probability that any component functions is p , in other words

$$\Pr(W_i) = p,$$

$\Pr(S)$ will be a function of p defined $f : [0, 1] \rightarrow [0, 1]$.

2 Independence

Definition 2.1 (Conditional probability). When discussing multiple events, it is possible that the occurrence of one event changes the probability that another will occur. This can be denoted using a vertical bar, and is read as “the probability of event A given B ”:

$$\Pr(A | B) = \frac{\Pr(AB)}{\Pr(B)}.$$

Definition 2.2 (Multiplication rule). For events A and B , the general multiplication rule states that

$$\Pr(AB) = \Pr(A | B) \Pr(B)$$

Theorem 2.0.1 (Independent events). *If A and B are independent events then*

$$\Pr(A | B) = \Pr(A)$$

$$\Pr(B | A) = \Pr(B)$$

Theorem 2.0.2 (Complement of independent events). *If A and B are independent, all complement pairs are also independent. Given*

$$\Pr(A | B) = \Pr(A)$$

$$\Pr(B | A) = \Pr(B)$$

the following statements are also true

$$\Pr(A | \overline{B}) = \Pr(A)$$

$$\Pr(B | \overline{A}) = \Pr(B)$$

$$\Pr(\overline{A} | B) = \Pr(\overline{A})$$

$$\Pr(\overline{B} | A) = \Pr(\overline{B})$$

$$\Pr(\overline{A} | \overline{B}) = \Pr(\overline{A})$$

$$\Pr(\overline{B} | \overline{A}) = \Pr(\overline{B})$$

2.1 Probability Rules with Conditional

All probability rules hold when conditioning on some event C .

Theorem 2.1.1 (Complement rule with condition).

$$\Pr(\bar{A} \mid C) = 1 - \Pr(A \mid C)$$

Theorem 2.1.2 (Addition rule with condition).

$$\Pr(A \cup B \mid C) = \Pr(A \mid C) + \Pr(B \mid C) - \Pr(AB \mid C)$$

Theorem 2.1.3 (Multiplication rule with condition).

$$\Pr(AB \mid C) = \Pr(A \mid BC) \Pr(B \mid C)$$

In the above examples, all probabilities are conditional on the sample space, hence we are effectively changing the sample space.

2.2 Conditional Independence

Definition 2.3 (Conditional independence). Suppose events A and B are not independent, i.e.,

$$\Pr(A \mid B) \neq \Pr(A)$$

but they become independent when conditioned with another event C , i.e.,

$$\Pr(A \mid BC) = \Pr(A \mid C)$$

Here we say that A and B are **conditionally independent** given C . Furthermore

$$\Pr(AB \mid C) = \Pr(A \mid C) \Pr(B \mid C)$$

Conversely, events A and B may be conditionally dependent but unconditionally independent, i.e.,

$$\begin{aligned} \Pr(A \mid B) &= \Pr(A) \\ \Pr(A \mid BC) &\neq \Pr(A \mid C) \\ \Pr(AB \mid C) &= \Pr(A \mid BC) \Pr(B \mid C) \end{aligned}$$

Theorem 2.2.1. *Given events A , B , and C . Pairwise independence does not imply mutual independence. I.e.,*

$$\begin{cases} \Pr(AB) = \Pr(A) \Pr(B) \\ \Pr(AC) = \Pr(A) \Pr(C) \\ \Pr(BC) = \Pr(B) \Pr(C) \end{cases}$$

does not imply

$$\Pr(ABC) = \Pr(A) \Pr(B) \Pr(C).$$

In summary, independence should not be assumed unless explicitly stated.

2.3 Disjoint Events

Theorem 2.3.1 (Probability of disjoint events). *The probability of disjoint events A and B is given by*

$$\begin{aligned}\Pr(AB) &= 0 \\ \Pr(\emptyset) &= 0.\end{aligned}$$

Disjoint events are highly dependent events, since the occurrence of one means the other cannot occur. This implies

$$\Pr(A \mid B) = 0$$

2.4 Subsets

Theorem 2.4.1 (Probability of subsets). *If $A \subset B$ then $\Pr(A) \leq \Pr(B)$. We also know that $\Pr(AB) = \Pr(A)$ and $\Pr(A \cup B) = \Pr(B)$.*

Here, if A happens, then B definitely happens.

$$\Pr(B \mid A) = 1$$

Given $\Pr(AB) = \Pr(A)$

$$\Pr(A \mid B) = \frac{\Pr(A)}{\Pr(B)}$$

These events are also highly dependent.