Math 3070, Applied Statistics

Section 1

August 30, 2019

Lecture Outline, 8/30

Section 2.5

- Independence
- Examples

Independence, Definition

Definition

Two events A and B are **independent** if P(A|B) = P(A) and are **dependent** otherwise.

A is independent of B implies that B is independent of A. Begin by assuming P(A|B) = P(A)

$$P(B|A) = \frac{P(B \cap A)}{P(A)}$$
(Multi. Rule for Cond. Prob.) =
$$\frac{P(A|B)P(B)}{P(A)}$$
((A is independent of B)) =
$$\frac{P(A)P(B)}{P(A)}$$
= $P(B)$

Independence, Definition

Definition

Two events A and B are **independent** if P(A|B) = P(A) and are **dependent** otherwise.

Intuition: The occurance of event B does not impact the probability of event A.

Multiplication Rule

Two events A and B are **independent** if and only if

$$P(A \cap B) = P(A) \cdot P(B)$$

Proof: Use the definition of multiplication rule of conditional probability

$$P(A \cap B) = P(A|B)P(B) = P(A)P(B)$$

This also means that P(A|B) = P(A) implies P(B|A) = P(B).

Independence, Coin Flip Example

Suppose two fair coins are flipped. Show that the first coin landing on heads independent of the second coin landing on heads.

The expected answer is yes. I will now show this in set notation.

$$S = \{(H, H), (H, T), (T, H), (T, T)\}$$

event
$$A =$$
 "first coin lands on heads" = $\{(H, H), (H, T)\}$
event $B =$ "second coin lands on heads" = $\{(H, H), (T, H)\}$

$$P(A)=\frac{1}{2}$$

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{1/4}{1/2} = \frac{1}{2} = P(A)$$

Independence, Comments

If A and B are independent so are A' and B'.

Proof:

$$P(A' \cap B') = 1 - P(A \cup B)$$

$$= 1 - (P(A) + P(B) - P(A \cap B))$$
(indep. of A and B) = 1 - (P(A) + P(B) - P(A)P(B))
$$P(A')P(B') = (1 - P(A))(1 - P(B))$$

$$= 1 - P(A) - P(B) + P(A)P(B)$$

$$= 1 - (P(A) + P(B) - P(A)P(B))$$

$$P(A' \cap B') = P(A')P(B')$$

Independence, Comments

If A and B are independent so are A and B'.

Proof:

$$P(A) = P(A \cap B) + P(A \cap B')$$

$$= P(A|B)P(B) + P(A|B')P(B')$$
(A and B are indep.) = $P(A)P(B) + P(A|B')P(B')$
rearrange
$$P(A|B')P(B') = P(A) - P(A)P(B)$$

$$= P(A)(1 - P(B))$$

$$= P(A)(P(B'))$$

$$P(A|B') = P(A)$$

Mutual Independence, Definition

Definition

Events A_1, \ldots, A_n are **mutually independent** if

$$P(A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}) = P(A_{i_1})P(A_{i_2}) \cdots P(A_{i_k})$$

for all choices of indices $1 \le i_1 < i_2 < \cdots < i_k \le n$

For example, saying three events A, B, C are independent means

$$P(A \cap B \cap C) = P(A)P(B)P(C)$$

$$P(A \cap B) = P(A)P(B)$$

$$P(A \cap C) = P(A)P(C)$$

$$P(B \cap C) = P(B)P(C)$$

If any two event are independent, then the events are **Pairwise independent**. Example, the last three statements show pairwise Independence of A, B and C.

Mutual Independence, Example

Suppose you flip 3 fair coins. Event A is that coin 1 lands on heads, event B is that coin 2 lands on heads and event C is that coin 3 lands on heads. Are the events mutually independent?

You can verify by applying the reasoning from the two coin example.

$$P(A \cap B) = P(A)P(B)$$

$$P(A \cap C) = P(A)P(C)$$

$$P(B \cap C) = P(B)P(C)$$

$$P(A \cap B \cap C) = P(\{(H, H, H)\}) = \frac{1}{8} = \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2}$$

Mutual Independence, Nonexample

Suppose you flip 2 fair coins. Event A is that coin 1 lands on heads, event B is that coin 2 lands on tails and event C is that the coins land on the same side. Are the events mutually independent?

$$A = \{(H, T), (H, H)\}, B = \{(H, T), (T, T)\}, C = \{(T, T), (H, H)\}$$
$$P(A \cap B \cap C) = P(\emptyset) = 0$$

nope, but they are pairwise independent.

$$P(A \cap B) = P(\{(H, T)\}) = \frac{1}{4} = P(A)P(B)$$

$$P(A \cap C) = P(\{(H, H)\}) = \frac{1}{4} = P(A)P(C)$$

$$P(B \cap C) = P(\{(T, T)\}) = \frac{1}{4} = P(B)P(C)$$

Independence, Summary

Independence of A and B:

$$P(A|B) = P(A), \quad P(A \cap B) = P(A)P(B)$$

- A is independent of B implies the converse.
- If A and B are independent then so are A' and B', and A and B'.
- Be careful not to assume events are independent.
- Pairwise Independence does not imply mutual Independence.
- The idea of Independence will come in handy later.

Multiplication Rule and Containment Example

A manufacturing process uses two machines, A and B, in serial. The probability that machine B fails is 1%. The probability that machine A fails is 5%. Assuming that the machines fail independently, what is the probability that machine A fails given that at least one machine fails?

event
$$A =$$
 "machine A fails" event $B =$ "machine B fails" , event $C =$ "at least one machine fails" ,

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

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

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

$$= 0.05 + 0.01 - 0.05(0.01)$$

$$= 0.0595$$

Multiplication Rule and Containment Example

A manufacturing process uses two machines, A and B. The probability that machine B fails is 1%. The probability that machine A fails is 5%. Assuming that the machines fail independently, what is the probability that machine A fails given that at least one machine fails?

$$P(A|C) = \frac{P(A \cap C)}{P(C)}$$
 (A is contained in C) = $\frac{P(A)}{P(C)} = \frac{0.05}{0.0595} \approx 0.840336134$

Containment shown on next slide.

Multiplication Rule and Containment Example

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(M_A, M_B) denotes the failure of each machine. M_A = F denotes that machine A fails. M_A = N denotes that machine A does not fails. M_B = N uses the same notation. A = \{(F, F), (F, N)\}
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 $C = \{(F, F), (F, N), (N, F)\}$

Serial Example

A company uses 17 machines for a manufacturing process. Each fail mutually independently of each other. Each has a 7% probability of failing. What is the probability that any are failing?

event
$$A_i$$
 = machine i fails

$$P(\text{"any are failing"}) = 1 - P(\text{"all are working"})$$

$$= 1 - P(A_1' \cap \ldots \cap A_{17}')$$

$$(\mathsf{indep}) = 1 - [P(A_1') \cdot \ldots \cdot P(A_{17}')]$$

$$P(A_i') = 1 - P(A_i) = 1 - 0.07 = 0.93$$

$$P("any are failing") = 1 - 0.93^{17} \approx 0.70878742441$$

Serial Example

If there are n events A_i , same probability and mutually independent, then

$$P(A_1 \cap \cdots \cap A_n) = P(A_1) \cdot \ldots \cdot P(A_n) = P(A_1)^n$$