

Probability and statistics

August - 24

Lecture #4



R: Random expt.

Ω : Sample space, \mathcal{F} : σ -algebra, P: probability measure.

In case, additional information is available about R, ($B \in \mathcal{F}$ occurs) conditional probability.

$$P(A|B) = \frac{P(A \cap B)}{P(B)} \quad P(B) > 0$$

$$P(A) = P(A|\Omega) = \frac{P(A \cap \Omega)}{P(\Omega)} = \frac{P(A)}{1} = P(A)$$

disjoint cover: $\{B_1, \dots, B_n\}$ disjoint cover of Ω

$$P(A) = P(A|B_1)P(B_1) + \dots + P(A|B_n)P(B_n)$$

Bayes's rule: $P(A \cap B) = P(A|B)P(B) = \underbrace{P(B|A)P(A)}$

Independence of events.

Product rule: $P(A \cap B) = P(A|B) P(B)$ —(1)
 $= P(B|A) P(A)$ —(2)

Definition: $A, B \in \mathcal{F}$ are said to be independent (statistically independent) if

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

$$\left\{ \begin{array}{l} \text{From (1), } P(A|B) = P(A) \\ \text{From (2), } P(B|A) = P(B) \end{array} \right\}$$

Ex:

R : Random exp²

$$\Omega = \{1, 2, 3, 4\}$$

Assumption: All outcomes are equally likely.

$$A = \{1, 2\}, \quad B = \{1, 3\}, \quad C = \{1, 4\}$$

$$P(A) = \frac{2}{4} = \frac{1}{2}$$

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

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

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

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

A & B are independent events.

□

Pairwise independence of events

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

In the previous example A, B are pairwise independent and so are B, C and A, C .

$$P(A \cap B \cap C) = \frac{1}{4} \neq P(A) \cdot P(B) \cdot P(C) = \frac{1}{8}$$

Definition: Let A, B, C be three events from (Ω, \mathcal{F}, P) . We call A, B, C mutually independent if

- 1) A, B, C are pairwise independent.
 $P(A \cap B) = P(A) P(B)$, $P(B \cap C) = P(B) P(C)$, $P(A \cap C) = P(A) P(C)$
- 2) $P(A \cap B \cap C) = P(A) P(B) P(C)$

We can generalize the definition of mutual independence to the events A_1, A_2, \dots, A_n in an inductive way.

- Any subcollection of A_1, \dots, A_n containing at least 2 events and atmost $n-1$ events is mutually independent.
- $P(A_1 \cap A_2 \cap \dots \cap A_n) = P(A_1) P(A_2) \dots P(A_n)$

Application of independence in setting up probability space in case of repeated random expt. with only 2 possible outcomes.

Note: A random expt. with only 2 outcomes is called as Bernoulli trial.

$$\Omega = \{0, 1\}$$

$$\mathcal{F} = 2^\Omega = \{\emptyset, \{1\}, \{0\}, \{0, 1\}\}$$

Assignment

of probability : $P\{1\} = p$ \leftarrow probability of success.

$$\Rightarrow P\{0\} = 1 - p$$

Aim: Set up the probability space when the random expt. R/ Bernoulli trial is repeated 'n' number of times with the assumption that the trials / repetitions are independent of each other.

For the compound exp't / n-repetition of Bernoulli trial.

$$\Omega_C = \{y_1, y_2, \dots, y_n \mid y_1, y_2, \dots, y_n \in \{0, 1\}^n\}$$

0100...011, 11...1, 00...0

$$\#\Omega_C = 2^n \text{ (finite set)}$$

$$f_C = \text{power set } \Omega_C = 2^{\Omega_C}$$

Probability assignment:

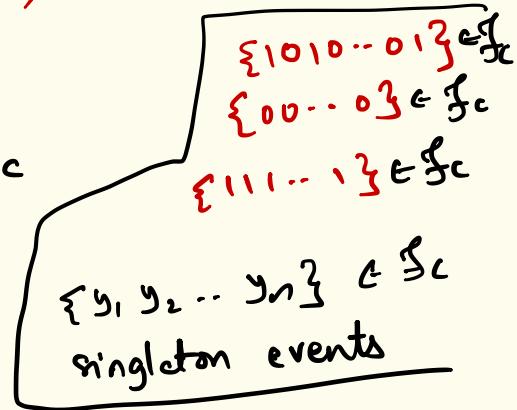
It is enough to assign

probability to singleton events

$\{y_1, y_2, \dots, y_n\} \in \mathcal{F}_C$

Consider an element $w = y_1 y_2 \dots y_n \in \Omega_C$ such

that there are exactly k number of 1's at specified location.



Let us assume that all the 1's are in the first k position.

$$y_1 = 1, y_2 = 1, \dots, y_k = 1; y_{k+1} = 0 = y_{k+2} = \dots = y_n$$

$$\omega = \left(\underbrace{11\cdots 1}_k \underbrace{00\cdots 0}_{n-k} \right)$$

(There are exactly k number of successes in n independent Bernoulli trials at specified locations)

A_i = Success at i^{th} trial $i=1, 2, \dots, n$.

$$\begin{aligned} p(\omega) &= P\left(\underbrace{11\cdots 1}_k \underbrace{00\cdots 0}_{n-k}\right) = P(A_1 \cap A_2 \cap \dots \cap A_k \cap \overline{A_{k+1}} \cap \dots \cap \overline{A_n}) \\ &= P(A_1) P(A_2) \dots P(A_k) P(\overline{A_{k+1}}) \cap \dots \cap P(\overline{A_n}) \\ &= p^k (1-p)^{n-k} \end{aligned}$$

Assumption of
mutually independent
Bernoulli trials

Consider event A_k = All outcomes with exactly k number of successes.

$$P(A_k) = \binom{n}{k} p^k (1-p)^{n-k} \quad k=0,1,2,\dots,n$$

$$A_k = \bigcup \{y_1 \dots y_n\}$$



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

<u>Example:</u>	A box contains	$n=10$	balls
6 - red	(60%)		
4 - blue	(40%)		<div style="display: flex; align-items: center;"> Any ball is equally likely </div> <div style="display: flex; align-items: center; margin-top: 10px;"> to be picked </div>

case(1) R: Pick a ball and replace it
 in the urn and a pick a ball again. } sampling
 A₁ = red ball is picked in the with
 first trial. replacement.

A₂ = blue ball is picked in the
 second trial.

$$P(A_2 | A_1) = \frac{4}{10} ; P(A_2) = \frac{4}{10}$$

$$P(A_2) = P(A_2 | A_1) P(A_1) + P(A_2 | A_1^c) P(A_1^c)$$

case(2): R: Pick a ball ; do not replace in the box and pick a ball again.

} Sampling without replacement.

A_1 = red ball in the first trial.

A_2 = blue ball in the second trial.

$$P(A_2 | A_1) = \frac{4}{9}$$

$$\begin{aligned} P(A_2) &= P(A_2 | A_1) P(A_1) + P(A_2 | A_1^c) P(A_1^c) \\ &= \frac{4}{9} \cdot \frac{6}{10} + \frac{3}{9} \cdot \frac{4}{10} \end{aligned}$$

$$P(A_2) = \frac{4}{10}$$

$$P(A_2 | A_1) \neq P(A_2)$$

Simple exercise: Sampling without replacement.

$$n = 10 \quad (0.044) \quad P(A_2 | A_1) = \frac{4}{9} \quad ; \quad P(A_2) = \frac{4}{10}$$

$$n = 100 \quad (0.004) \quad P(A_2 | A_1) = \frac{40}{99} \quad ; \quad P(A_2) = \frac{40}{100}$$

$$n = 1000 \quad P(A_2 | A_1) = \frac{400}{999} \quad ; \quad P(A_2) = \frac{400}{1000}$$

$$(0.0004) \quad n = 10000 \quad P(A_2 | A_1) = \frac{4000}{9999} \quad ; \quad P(A_2) = \frac{4000}{10000}$$

$$(0.00004)$$

$$P(A_2 | A_1) \rightarrow P(A_2) \quad \text{as} \quad n \rightarrow \infty$$

60% red balls & 40% blue balls.

- Recap:
- Ω
 - \mathcal{F} , \mathbb{P}
 - axiomatic definition of probability.
 - several properties
 - Uniform probability spaces
 - conditional probability
 - total probability law
 - Baye's rule
 - independence of events
 - setting new probability space from repeating indep. Bernoulli trials.
 - sampling with /without replacement.

Tutorial
sheet #1.

Problem: A machine contains 4 components in parallel with 0.1, 0.2, 0.3, 0.4 as their probabilities of failures resp. The machine fails if all the components fail simultaneously. The failure of components is independent. Then what is the probability that the machine once started will not fail??

Let A_i : Event that component i fails.

$A_1 \cap A_2 \cap A_3 \cap A_4$: machine fails

$$\begin{aligned}
 P(A_1 \cap A_2 \cap A_3 \cap A_4)^c &= 1 - P(A_1 \cap A_2 \cap A_3 \cap A_4) \\
 &= 1 - P(A_1) P(A_2) P(A_3) P(A_4) \quad \text{dependence} \\
 &\approx 1 - 0.1 \times 0.2 \times 0.3 \times 0.4
 \end{aligned}$$

Problem: A course on Probability is a very famous course and students are allowed to register after the teacher consents. It is observed that 20% times after obtaining the consent, student fails to register for the course. There are only 100 seats available in the class. If teacher consents 102 students, what is the probability that all the students (who try to register) are accommodated in the class??

Ans: $1 - (0.8)^{102} - 102 \times (0.8)^{101} \times 0.2$

Problem : 3 printers

P_1, P_2, P_3

Printer P_1 gets 50% jobs

P_2 30%

P_3 20%

P_1 has 0.15 as its probability of failure,

P_2 0.1

P_3 0.2

If a randomly selected printing job is a failure,
what is the probability that it was printed
by printer 1 ??

Problem: If $A_1, A_2 \in \mathcal{F}$ are indep., can we say $\underline{\underline{A_1^C \text{ & } A_2^C}}$ are also indep. ??

Ans: Yes

$$(A_1 \cup A_2)^C = A_1^C \cap A_2^C$$

$$P(A_1 \cup A_2)^C = 1 - P(A_1 \cup A_2)$$

$$\begin{aligned} P(A_1^C \cap A_2^C) &= 1 - P(A_1) - P(A_2) + P(A_1 \cap A_2) \\ &= 1 - P(A_1) - P(A_2) + P(A_1) P(A_2) \\ &= (1 - P(A_1))(1 - P(A_2)) \\ &= P(A_1^C) P(A_2^C) \end{aligned}$$

$\Rightarrow A_1^C$ and A_2^C are indep.

Problem: $A_1, A_2 \in \mathcal{F}$ are indep.

Are A_1, A_2^c also indep??

$$P \Rightarrow q$$

|||

$$\neg q \Leftrightarrow \neg b$$

Solu:

$$A_1 = (A_1 \cap A_2^c) \cup (A_1 \cap A_2)$$

$$\begin{aligned} P(A_1) &= P(A_1 \cap A_2^c) + P(A_1 \cap A_2) \\ &= P(A_1 \cap A_2^c) + P(A_1)P(A_2) \end{aligned}$$

$$\begin{aligned} P(A_1 \cap A_2^c) &= P(A_1) - P(A_1)P(A_2) \\ &= P(A_1)(1 - P(A_2)) \end{aligned}$$

$$\boxed{P(A_1 \cap A_2^c) = P(A_1)P(A_2^c)}$$

$\Rightarrow A_1 \& A_2^c$ are indep.

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