

1.4 Conditional probability and independent events

Conditional probability(条件概率)

Example: A dice is thrown. The sample space is

$$\Omega = \{\omega_1, \omega_2, \dots, \omega_6\}, \quad \omega_i = \{i \text{ comes up}\}.$$

Consider the events

$$A = \{\text{an even point comes up}\} = \{\omega_2, \omega_4, \omega_6\},$$

$$B = \{\text{an odd point comes up}\} = \{\omega_1, \omega_3, \omega_5\}.$$

Then

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

Now, provided that a big point (≥ 4) comes up, the probability that the point is even is just the conditional probability of A given C , written as $P(A|C)$, where $C = \{ \text{a big point comes up} \}$.

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Provided that a big point comes up, the sample space becomes

$$\Omega_2 = C = \{\omega_4, \omega_5, \omega_6\}.$$

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$$P(A|C) = \frac{\#\{\omega_4, \omega_6\}}{\#\Omega_2} = \frac{2}{3}.$$

Similarly

$$P(B|C) = \frac{\#\{\omega_5\}}{\#\Omega_2} = \frac{1}{3}.$$

$$P(A|B)$$
$$= \frac{\text{Number of sample points contained by } A \text{ given } B}{\text{Number of sample points given } B}$$

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In general, given that the event B has occurred, the relevant sample space is no longer Ω but consists of outcomes in B ; A has occurred iff one of the outcomes in $A \cap B$ occurred, so that

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The proportionality constant $c = 1/P(B)$ is used to ensure that the probability $P(B|B)$ of the new sample space B equals 1.

Definition 1 If A , B are two events and $P(B) \neq 0$, then the conditional probability of A given B , written as $P(A|B)$, is defined to be

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

Properties of conditional probability

$P(\cdot|B) : \mathcal{F} \rightarrow [0, 1]$:

- ① (non-negativity) $P(A|B) \geq 0$ for all $A \in \mathcal{F}$;
- ② (normalization condition) $P(\Omega|B) = 1$;
- ③ (countable additivity) If A_1, \dots, A_n, \dots are mutually disjoint events ($A_i A_j = \emptyset, i \neq j$), then

$$P\left(\sum_{n=1}^{\infty} A_n | B\right) = \sum_{n=1}^{\infty} P(A_n | B).$$

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$$\begin{aligned}
 & P(A_1 A_2 \cdots A_n) \\
 &= P(A_n | A_1 A_2 \cdots A_{n-1}) P(A_1 A_2 \cdots A_{n-1}) \\
 &= P(A_n | A_1 A_2 \cdots A_{n-1}) P(A_{n-1} | A_1 A_2 \cdots A_{n-2}) \\
 &\quad \cdot P(A_1 A_2 \cdots A_{n-2}) \\
 &= \cdots .
 \end{aligned}$$

$$\begin{aligned}
 P(A_1 A_2 \cdots A_n) &= P(A_1) \cdot P(A_2 | A_1) \cdot P(A_3 | A_1 A_2) \\
 &\quad \cdots P(A_{n-1} | A_1 A_2 \cdots A_{n-2}) \\
 &\quad \cdot P(A_n | A_1 A_2 \cdots A_{n-1}).
 \end{aligned}$$

Example Two people A and B make an appointment to meet at a park between 7 o'clock and 8 o'clock and the person who first arrives at the park will keep waiting for another for 20 minutes. Find the probability that A arrives first if they meet each other.

Take 7 o'clock as the beginning time and assume that A arrives at x and B arrives at y . The sample space is

$$\Omega = \{(x, y) \mid 0 \leq x \leq 60, 0 \leq y \leq 60\}$$

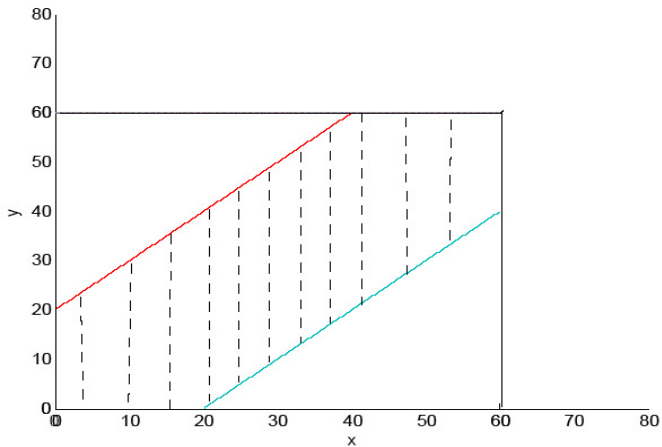
and

$$\begin{aligned} A &= \{\text{they meet each other}\} \\ &= \{(x, y) \mid |x - y| \leq 20, 0 \leq x, y \leq 60\}, \end{aligned}$$

$$P(A) = \frac{5}{9}.$$

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Conditional probability



On the other hand,

$$B = \{ A \text{ arrives first} \}$$

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So

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$$P(BA) = \frac{\frac{1}{2}60^2 - \frac{1}{2}(60 - 20)^2}{60^2} = \frac{5}{18}.$$

By definition, it follows that

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

Example 2. There is a prizewinning ticket in n lottery tickets. These n lottery tickets are supposed to be sold to n different persons randomly.

- (1) If the first $k - 1$ customers do not get the prizewinning ticket, find the probability that the k -th customer gets the prizewinning ticket;
- (2) Find the probability that the k -th customer gets the prizewinning ticket.

Solution (1). Let $A_i = \{\text{the } i\text{-th customer gets the prizewinning ticket}\}$. Then the event as condition in (1) is $\overline{A}_1 \overline{A}_2 \cdots \overline{A}_{k-1}$.

Solution (1). Let $A_i = \{\text{the } i\text{-th customer gets the prizewinning ticket}\}$. Then the event as condition in (1) is $\overline{A}_1 \overline{A}_2 \cdots \overline{A}_{k-1}$. If we consider the event A_k in the reduced sample space by $\Omega_2 = \overline{A}_1 \overline{A}_2 \cdots \overline{A}_{k-1}$, we can obtain by a direct application of classical probability model

$$P(A_k | \overline{A}_1 \overline{A}_2 \cdots \overline{A}_{k-1}) = \frac{1}{n - k + 1}.$$

As for (2), $A_k = \overline{A_1}\overline{A_2}\cdots\overline{A_{k-1}}A_k$ obviously holds.

So by the multiplication rule we have

$$P(A_k) = P(\overline{A_1}\overline{A_2}\cdots\overline{A_{k-1}}A_k)$$

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$$\begin{aligned} P(A_k) &= P(\overline{A_1}\overline{A_2}\cdots\overline{A_{k-1}}A_k) \\ &= P(\overline{A_1})P(\overline{A_2}|\overline{A_1})P(\overline{A_3}|\overline{A_1}\overline{A_2}) \\ &\quad \cdots P(A_k|\overline{A_1}\overline{A_2}\cdots\overline{A_{k-1}}) \end{aligned}$$

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So by the multiplication rule we have

$$\begin{aligned} P(A_k) &= P(\overline{A}_1 \overline{A}_2 \cdots \overline{A}_{k-1} A_k) \\ &= P(\overline{A}_1) P(\overline{A}_2 | \overline{A}_1) P(\overline{A}_3 | \overline{A}_1 \overline{A}_2) \\ &\quad \cdots P(A_k | \overline{A}_1 \overline{A}_2 \cdots \overline{A}_{k-1}) \\ &= \frac{n-1}{n} \cdot \frac{n-2}{n-1} \cdot \frac{n-3}{n-2} \cdots \frac{n-k+1}{n-k+2} \cdot \frac{1}{n-k+1} \\ &= \frac{1}{n}. \end{aligned}$$

Solution (2). Let $A_i = \{\text{the } i\text{-th customer gets the prizewinning ticket}\}$. The problem is equivalent to that a prizewinning ticket is assigned to one of n customers randomly. There are n assignment ways totally and only one way in A_k . So

$$P(A_k) = \frac{1}{n}.$$

Then the event $\overline{A_1}\overline{A_2}\cdots\overline{A_{k-1}}$ as condition in (1) is equivalent to that the prizewinning ticket is assigned to one of other $n - (k - 1)$ customers randomly, and there are $n - k + 1$ assignment ways. So

Then the event $\overline{A}_1\overline{A}_2\cdots\overline{A}_{k-1}$ as condition in (1) is equivalent to that the prizewinning ticket is assigned to one of other $n - (k - 1)$ customers randomly, and there are $n - k + 1$ assignment ways. So

$$P(\overline{A}_1\overline{A}_2\cdots\overline{A}_{k-1}) = \frac{n - k + 1}{n}.$$

Hence, by the definition of the condition probability,

$$\begin{aligned} & P(A_k | \overline{A}_1 \overline{A}_2 \cdots \overline{A}_{k-1}) \\ &= \frac{P(A_k \overline{A}_1 \overline{A}_2 \cdots \overline{A}_{k-1})}{P(\overline{A}_1 \overline{A}_2 \cdots \overline{A}_{k-1})} \end{aligned}$$

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Example

(Match problem (continued)) Suppose that a person types n letters, types the corresponding addresses on n envelopes, and then places the n letters in the n envelopes in random manner. We say that a match occurs if a letter is placed in the correct envelope. What is the probability of

- (a) no matches;
- (b) exactly k matches?

Solution. We denote the probability of exactly k matches by $P_k^{(n)}$.

(a) We have shown before that

$$P_0^{(n)} = \frac{1}{2!} - \frac{1}{3!} + \frac{1}{4!} - \cdots + \frac{(-1)^n}{n!}.$$

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(b) To obtain the probability of exactly k matches, we consider any fixed group of k letters, say the i_1, i_2, \dots, i_k -th letters. The probability that they, and only they, are placed in the correct envelopes is

$$\begin{aligned}
 & P(A_{i_1} \cdots A_{i_k} \bar{A}_{i_{k+1}} \cdots \bar{A}_{i_n}) \\
 &= P(A_{i_1})P(A_{i_2}|A_{i_1}) \cdots P(A_{i_k}|A_{i_1} \cdots A_{i_{k-1}}) \\
 &\quad \cdot P(\bar{A}_{i_{k+1}} \cdots \bar{A}_{i_n}|A_{i_1} \cdots A_{i_k}) \\
 &= \frac{1}{n} \frac{1}{n-1} \cdots \frac{1}{n-(k-1)} q_{n-k} = \frac{(n-k)!}{n!} q_{n-k},
 \end{aligned}$$

where q_{n-k} is the conditional probability that the other $n-k$ letters, being placed in their own envelopes, have no matches, and so

$$q_{n-k} = P_0^{(n-k)} = \frac{1}{2!} - \frac{1}{3!} + \frac{1}{4!} - \cdots + \frac{(-1)^{n-k}}{(n-k)!}.$$

As there are $\binom{n}{k}$ choices of a set of k letters, the desired probability of exactly k matches is

$$\begin{aligned} P_k^{(n)} &= \sum_{i_1 < \dots < i_k} P(A_{i_1} \cdots A_{i_k} \bar{A}_{i_{k+1}} \cdots \bar{A}_{i_n}) \\ &= \binom{n}{k} \cdot \frac{(n-k)!}{n!} q_{n-k} = \frac{P_0^{(n-k)}}{k!} \\ &= \frac{\frac{1}{2!} - \frac{1}{3!} + \frac{1}{4!} - \cdots + \frac{(-1)^{n-k}}{(n-k)!}}{k!}. \end{aligned}$$

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It is easily seen that

$$P_k^{(n)} \rightarrow e^{-1} \frac{1}{k!}.$$

Total probability formula and Bayes' rule

Definition 2 Suppose that $\{A_1, A_2, \dots, A_n, \dots\}$ is a set of events satisfying: (1) $A_i, i = 1, 2, \dots$, are mutually disjoint and $P(A_i) > 0$; (2) $\sum_{i=1}^{\infty} A_i = \Omega$. Then $\{A_1, A_2, \dots, A_n, \dots\}$ is called a set of *mutually exclusive and exhaustive* events in Ω , or a partition of Ω .

1.4 Conditional probability and independent events

Total probability formula and Bayes' rule

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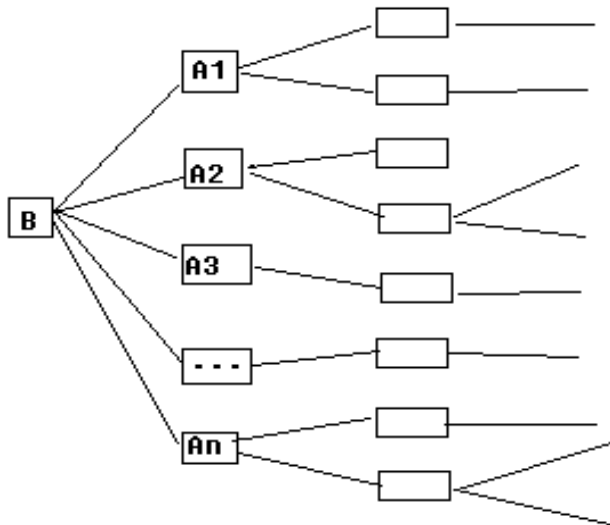
$$\begin{aligned} P(B) &= P\left(B \sum_{i=1}^{\infty} A_i\right) \\ &= P\left(\sum_{i=1}^{\infty} (BA_i)\right) \\ &= \sum_{i=1}^{\infty} P(BA_i) \\ &= \sum_{i=1}^{\infty} P(A_i)P(B|A_i). \end{aligned}$$

(Total probability formula) If $A_1, A_2, \dots, A_n, \dots$
are *mutually exclusive and exhaustive* events, then
for any event B

$$P(B) = \sum_{i=1}^{\infty} P(A_i)P(B|A_i).$$

1.4 Conditional probability and independent events

Total probability formula and Bayes' rule



Example 4 There are 3 new balls and 2 old balls in bag. If two balls are drawn in random and in succession without replacement, find the probability that the second is a new one.

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On the other hand

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So

$$P(B) = P(B|A)P(A) + P(B|\bar{A})P(\bar{A}) = \frac{3}{5}.$$

Example

(Match problem (continued)) Suppose that a person types n letters, types the corresponding addresses on n envelopes, and then places the n letters in the n envelopes in random manner. We say that a match occurs if a letter is placed in the correct envelope.

What is the probability of no matches?

Solution (2). We denote the probability of no matches by $P_0^{(n)}$. Let $E = \{ \text{no matches} \}$. Let A_i be the event that the i -th letter is placed its correct envelope. Then

$$P_0^{(n)} = P(E)$$

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$$P(E|A_1) = 0.$$

$P(E|\overline{A}_1)$ is the probability of no matches when $n - 1$ letters placed in a set of $n - 1$ envelopes that does not contain the envelope of one of these letters. (Say, the 1-st letter is placed to the i -th envelope)

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This can happen in either of two mutually exclusive ways.

$P(E|\overline{A}_1)$ is the probability of no matches when $n - 1$ letters placed in a set of $n - 1$ envelopes that does not contain the envelope of one of these letters. (Say, the 1-st letter is placed to the i -th envelope)

This can happen in either of two mutually exclusive ways.

- Either there are no matches and the extra letter is placed to the extra envelope (this being the i -th envelope that chose first letter, in the remained $(n - 2)$ letters and $(n - 2)$ envelopes there are no matches),
- or there are no matches and the extra letter is not placed in the extra envelope.

The probability of the first of these events is $\frac{1}{n-1}P_0^{(n-2)}$. The probability of the second event is just $P_0^{(n-1)}$, which is seen by regarding the extra envelope as "belonging" to the extra letter. So

$$P_0^{(n)} = P(E|\overline{A}_1) \frac{n-1}{n} = \left(P_0^{(n-1)} + \frac{1}{n-1} P_0^{(n-2)} \right) \frac{n-1}{n},$$

and thus,

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and thus,

$$P_0^{(n)} - P_0^{(n-1)} = -\frac{1}{n} \left(P_0^{(n-1)} - P_0^{(n-2)} \right).$$

Obviously, $P_0^{(1)} = 0$, $P_0^{(2)} = \frac{1}{2}$.

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Total probability formula and Bayes' rule

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Obviously, $P_0^{(1)} = 0$, $P_0^{(2)} = \frac{1}{2}$. Hence

$$P_0^{(n)} - P_0^{(n-1)} = -\frac{1}{n} \left(P_0^{(n-1)} - P_0^{(n-2)} \right) = \dots = \frac{(-1)^n}{n!}.$$

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It follows that

$$P_0^{(n)} = \frac{1}{2!} - \frac{1}{3!} + \frac{1}{4!} - \cdots + \frac{(-1)^n}{n!}.$$

Example. (The gambler's ruin problem) On each play of the game,

p ($0 < p < 1$)—gambler A will win one dollar from gambler B

$q = (1 - p)$ — B will win one dollar from A .

The initial fortune of A is i dollars, the initial fortune of B is $k - i$ dollars.

If one loses his all money, the game is over. Find the probability that B loses all his money.

Solution. Let p_i denote the probability that gambler A will win the game (the gambler B will ruin), given that his initial fortune is i dollars.

Obviously, $p_0 = 0$ and $p_k = 1$.

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Let $A_1(B_1)$ denote the event that gambler A wins (resp. losses) one dollar on the first play of the game; and let W denote the event that gambler A will win the game. Then

1.4 Conditional probability and independent events

Total probability formula and Bayes' rule

$$P(W) = P(A_1)P(W|A_1) + P(B_1)P(W|B_1)$$

1.4 Conditional probability and independent events

Total probability formula and Bayes' rule

$$\begin{aligned}P(W) &= P(A_1)P(W|A_1) + P(B_1)P(W|B_1) \\&= pP(W|A_1) + qP(W|B_1).\end{aligned}$$

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That is

$$p_i = pp_{i+1} + qp_{i-1}.$$

So

$$p_i - p_{i-1} = \frac{q}{p}(p_{i-1} - p_{i-2}) = \left(\frac{q}{p}\right)^{i-1} p_1, \quad i = 2, \dots, k.$$

Taking summation on both sides yields

1.4 Conditional probability and independent events

Total probability formula and Bayes' rule

$$\begin{aligned} 1 - p_1 &= p_1 \sum_{i=1}^{k-1} \left(\frac{q}{p}\right)^i \\ &= \begin{cases} p_1 \frac{(q/p)^k - q/p}{q/p - 1}, & \text{if } q \neq p, \\ (k-1)p_1, & \text{if } q = p. \end{cases} \end{aligned}$$

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Total probability formula and Bayes' rule

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Hence

$$p_1 = \begin{cases} \frac{q/p - 1}{(q/p)^k - 1}, & \text{if } p \neq 1/2, \\ 1/k, & \text{if } p = 1/2. \end{cases}$$

So, if $p \neq 1/2$, then

$$p_i = \frac{(q/p)^i - 1}{(q/p)^k - 1}, \quad \text{for } i = 1, \dots, k-1;$$

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if $p = 1/2$, then

$$p_i = \frac{i}{k}, \quad \text{for } i = 1, \dots, k-1.$$

(Bayes's rule) If $A_1, A_2, \dots, A_n, \dots$ are *mutually exclusive* and *exhaustive* events, then for any event B with $P(B) > 0$ we have

$$P(A_i|B) = \frac{P(BA_i)}{P(B)} = \frac{P(B|A_i)P(A_i)}{\sum_{j=1}^{\infty} P(B|A_j)P(A_j)}.$$

1.4 Conditional probability and independent events

Total probability formula and Bayes' rule

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$P(A_i)$ — priori probability(先验概率),

$P(A_i|B)$ — a posteriori probability(后验概率).

例：医生：如果患者患A病的可能性 $\geq 85\%$ ，则建议立即做手续；否则就建议做一些(昂贵)的检查；

Jonhson：开始时，得A病的可能性为60%，做了一项检查B呈阳性；

但同时得知他患有糖尿病，糖尿病导致检查B呈阳性的可能性为25%。

问医生是建议Jonhson立即做手术？还是做更多昂贵的检查？

1.4 Conditional probability and independent events

Total probability formula and Bayes' rule

解: 用 A 表示Jonhson患有A病, B 表示检查B 呈阳性, 已知 $P(A) = 0.6$. 如果患A病,则B呈阳性, 即 $P(B|A) = 1$.

解: 用 A 表示 Jonhson 患有 A 病, B 表示检查 B 呈阳性, 已知 $P(A) = 0.6$. 如果患 A 病, 则 B 呈阳性, 即 $P(B|A) = 1$. 在通常情况下, 如果病人没有患 A 病, 检查不会呈阳性, 即 $P(B|\bar{A}) = 0$,

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$$P(A|B) = \frac{P(AB)}{P(B)} = \frac{P(A)P(B|A)}{P(B|A)P(A) + P(B|\bar{A})P(\bar{A})}$$

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因此医生应该建议Jonhson立即做手术.

Example 7 A doctor uses to diagnose patients in order to see whether they suffer from liver cancer. Let C be the event that a patient suffers from liver cancer, A the event that a patient is diagnosed suffering from liver cancer (阳性). Suppose

$$P(A|C) = 0.95, P(A|\overline{C}) = 0.01(\text{假阳性}),$$

$P(C) = 0.0001$, find the probability that one patient diagnosed suffering from liver cancer suffers truly from liver cancer.

Solution: According to Bayes' formula, we have

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$$P(C|A) = \frac{P(C) \cdot P(A|C)}{P(C) \cdot P(A|C) + P(\overline{C}) \cdot P(A|\overline{C})}.$$

Solution: According to Bayes' formula, we have

$$P(C|A) = \frac{P(C) \cdot P(A|C)}{P(C) \cdot P(A|C) + P(\overline{C}) \cdot P(A|\overline{C})}.$$

In addition,

$$P(\overline{C}) = 1 - P(C) = 0.9999,$$

$$P(A|\overline{C}) = 0.01.$$

Solution: According to Bayes' formula, we have

$$P(C|A) = \frac{P(C) \cdot P(A|C)}{P(C) \cdot P(A|C) + P(\overline{C}) \cdot P(A|\overline{C})}.$$

In addition,

$$P(\overline{C}) = 1 - P(C) = 0.9999,$$

$$P(A|\overline{C}) = 0.01.$$

Substituting these numerical values into Bayes's formula

$$P(C|A) = 0.0094.$$

例8 某工厂有四条流水线生产同一种产品,其中每条流水线产量分别占总产量的12%,25%,25%和38%. 根据经验,每条流水线的不合格率分别为0.06, 0.05, 0.04, 0.03. 某客户够买该产品后,发现是不合格品,向厂家提出索赔10000元. 按规定,工厂要求四条流水线共同承担责任. 问每条流水线应该各赔付多少?

解: 用 B 表示“任取一件产品为不合格产品”,
 A_i 表示“任取一件产品是第 i 流水线生产的”,
 $i = 1, 2, 3, 4$.

解: 用 B 表示“任取一件产品为不合格产品”,
 A_i 表示“任取一件产品是第 i 流水线生产的”,
 $i = 1, 2, 3, 4$. 由题意得

$$P(B) = \sum_{i=1}^4 P(B|A_i)P(A_i)$$

$$= 0.12 \times 0.06 + 0.25 \times 0.05 + 0.25 \times 0.04 + 0.38 \times 0.03$$

$$= 0.0411.$$

上式表明该工厂产品不合格率为4.11%.

1.4 Conditional probability and independent events

Total probability formula and Bayes' rule

现在客户发现所购买产品为不合格品, 即 B 发生了, 我们要分析其发生的原因, 计算条件概率 $P(A_i|B)$, 并按其大小比例赔付客户.

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$$P(A_1|B) = \frac{P(B|A_1)P(A_1)}{P(B)} = \frac{0.12 \times 0.06}{0.0411} \simeq 0.175.$$

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类似地

$$P(A_2|B) \simeq 0.304, \quad P(A_3|B) \simeq 0.243, \quad P(A_4|B) \simeq 0.278.$$

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类似地

$$P(A_2|B) \simeq 0.304, \quad P(A_3|B) \simeq 0.243, \quad P(A_4|B) \simeq 0.278.$$

这样, 每条生产线应分别赔付1750元, 3040元, 2430元和2780元.

Independent events

1. Independence of two events:

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

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

Definition. Two events A and B are independent if

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Any event is independent of \emptyset .

Example 9. An urn contains a black balls and b white balls. If two balls are drawn in succession and we denote by A the event that the first ball drawn is black, B the event that the second ball drawn is black. Are A and B independent of each other? Consider two different situations: (1) with replacement, (2) without replacement.

Solution. For case (1),

$$P(B|A) = P(B|\bar{A}) = \frac{a}{a+b}.$$

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So

$$\begin{aligned} P(B) &= P(B|A)P(A) + P(B|\bar{A})P(\bar{A}) \\ &= \frac{a}{a+b}(P(A) + P(\bar{A})) = \frac{a}{a+b} \end{aligned}$$

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So

$$\begin{aligned} P(B) &= P(B|A)P(A) + P(B|\bar{A})P(\bar{A}) \\ &= \frac{a}{a+b}(P(A) + P(\bar{A})) = \frac{a}{a+b} \\ &= P(B|A), \end{aligned}$$

which shows that A and B are independent.

For case (2), we have

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$$\begin{aligned} P(B) &= \frac{a}{a+b} \cdot \frac{a-1}{a+b-1} + \frac{b}{a+b} \cdot \frac{a}{a+b-1} \\ &= \frac{a}{a+b} \neq P(B|A), \end{aligned}$$

which shows that A and B are not independent.

Example 11. Suppose A and B are two events independent of each other, show that so are A and \overline{B} , \overline{A} and B , \overline{A} and \overline{B} .

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Proof. Since A and B are indept.,

$P(AB) = P(A)P(B)$, then

$$P(A\overline{B}) = P(A - AB)$$

Example 11. Suppose A and B are two events independent of each other, show that so are A and \overline{B} , \overline{A} and B , \overline{A} and \overline{B} .

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$P(AB) = P(A)P(B)$, then

$$\begin{aligned} P(A\overline{B}) &= P(A - AB) = P(A) - P(AB) \\ &= P(A) - P(A)P(B) \end{aligned}$$

Example 11. Suppose A and B are two events independent of each other, show that so are A and \overline{B} , \overline{A} and B , \overline{A} and \overline{B} .

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$P(AB) = P(A)P(B)$, then

$$\begin{aligned} P(A\overline{B}) &= P(A - AB) = P(A) - P(AB) \\ &= P(A) - P(A)P(B) = P(A)(1 - P(B)) \end{aligned}$$

Example 11. Suppose A and B are two events independent of each other, show that so are A and \overline{B} , \overline{A} and B , \overline{A} and \overline{B} .

Proof. Since A and B are indept.,

$P(AB) = P(A)P(B)$, then

$$\begin{aligned} P(A\overline{B}) &= P(A - AB) = P(A) - P(AB) \\ &= P(A) - P(A)P(B) = P(A)(1 - P(B)) \\ &= P(A)P(\overline{B}). \end{aligned}$$

Definition Two σ -algebras \mathcal{F}_1 and \mathcal{F}_2 are said to be independent with regard to P , if

$$P(A_1 A_2) = P(A_1)P(A_2).$$

holds for arbitrary A_1, A_2 such that $A_1 \in \mathcal{F}_1$, and $A_2 \in \mathcal{F}_2$.

2. Independence of several events

Definition Events A , B and C are said to be independent if

$$\left. \begin{aligned} P(AB) &= P(A) \cdot P(B) \\ P(AC) &= P(A) \cdot P(C) \\ P(BC) &= P(B) \cdot P(C) \end{aligned} \right\} \quad (9)$$

and

$$P(ABC) = P(A) \cdot P(B) \cdot P(C).$$

Definition Suppose that A_1, A_2, \dots, A_n are n events. If for $1 \leq i < j < k < \dots \leq n$,

$$\left. \begin{aligned} P(A_i A_j) &= P(A_i)P(A_j), \\ P(A_i A_j A_k) &= P(A_i)P(A_j)P(A_k), \\ &\dots\dots\dots \\ P(A_1 A_2 \dots A_n) &= P(A_1)P(A_2) \dots P(A_n) \end{aligned} \right\} \quad (11)$$

hold, then A_1, A_2, \dots, A_n are said to be independent.

Example 13. Suppose that A_1, A_2, \dots, A_n are independent, and $P(A_i) = p_i, i = 1, 2, \dots, n$. Find the probabilities that

- (1) neither of them occurs;
- (2) at least one of them occurs;
- (3) only one of them occurs.

Solution:

(1) {neither of them occurs} = $\overline{A}_1 \overline{A}_2 \cdots \overline{A}_n$. We have

$$P(\overline{A}_1 \overline{A}_2 \cdots \overline{A}_n)$$

Solution:

(1) {neither of them occurs} = $\overline{A_1} \overline{A_2} \cdots \overline{A_n}$. We have

$$P(\overline{A_1} \overline{A_2} \cdots \overline{A_n}) = P(\overline{A_1})P(\overline{A_2}) \cdots P(\overline{A_n})$$

Solution:

(1) {neither of them occurs} = $\bar{A}_1 \bar{A}_2 \cdots \bar{A}_n$. We have

$$\begin{aligned} P(\bar{A}_1 \bar{A}_2 \cdots \bar{A}_n) &= P(\bar{A}_1)P(\bar{A}_2) \cdots P(\bar{A}_n) \\ &= \prod_{i=1}^n (1 - p_i). \end{aligned}$$

(2) {at least one of them occurs}=

$$A_1 \cup A_2 \cup \cdots \cup A_n = \overline{\overline{A_1} \overline{A_2} \cdots \overline{A_n}}. \text{ So}$$

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$$\begin{aligned} P(A_1 \cup A_2 \cup \cdots \cup A_n) &= 1 - P(\overline{A_1} \overline{A_2} \cdots \overline{A_n}) \\ &= 1 - \prod_{i=1}^n (1 - p_i). \end{aligned}$$

(3) {only one of them occurs}

$= \overline{A_1}\overline{A_2}\cdots\overline{A_{n-1}}A_n + \overline{A_1}\overline{A_2}\cdots A_{n-1}\overline{A_n} + \cdots +$
 $A_1\overline{A_2}\cdots\overline{A_n}$. Therefore, the desired probability is

$$P\left(\sum_{k=1}^n \overline{A_1}\overline{A_2}\cdots\overline{A_{k-1}}A_k\overline{A_{k+1}}\cdots\overline{A_n}\right)$$

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$= \overline{A}_1 \overline{A}_2 \cdots \overline{A}_{n-1} A_n + \overline{A}_1 \overline{A}_2 \cdots A_{n-1} \overline{A}_n + \cdots + A_1 \overline{A}_2 \cdots \overline{A}_n$. Therefore, the desired probability is

$$\begin{aligned} & P\left(\sum_{k=1}^n \overline{A}_1 \overline{A}_2 \cdots \overline{A}_{k-1} A_k \overline{A}_{k+1} \cdots \overline{A}_n\right) \\ &= \sum_{k=1}^n P(\overline{A}_1 \overline{A}_2 \cdots \overline{A}_{k-1} A_k \overline{A}_{k+1} \cdots \overline{A}_n) \end{aligned}$$

1.4 Conditional probability and independent events

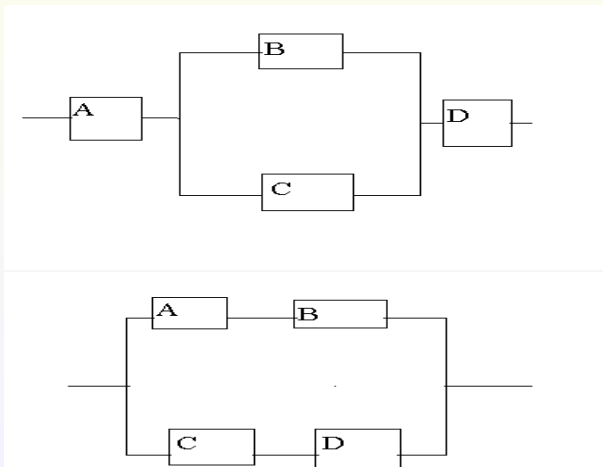
Independent events

(3) {only one of them occurs}

$= \bar{A}_1 \bar{A}_2 \cdots \bar{A}_{n-1} A_n + \bar{A}_1 \bar{A}_2 \cdots A_{n-1} \bar{A}_n + \cdots + A_1 \bar{A}_2 \cdots \bar{A}_n$. Therefore, the desired probability is

$$\begin{aligned} & P\left(\sum_{k=1}^n \bar{A}_1 \bar{A}_2 \cdots \bar{A}_{k-1} A_k \bar{A}_{k+1} \cdots \bar{A}_n\right) \\ &= \sum_{k=1}^n P(\bar{A}_1 \bar{A}_2 \cdots \bar{A}_{k-1} A_k \bar{A}_{k+1} \cdots \bar{A}_n) \\ &= \sum_{k=1}^n P(\bar{A}_1) P(\bar{A}_2) \cdots P(\bar{A}_{k-1}) P(A_k) P(\bar{A}_{k+1}) \cdots P(\bar{A}_n) \\ &= \sum_{k=1}^n p_k \prod_{i=1, i \neq k}^n (1 - p_i). \end{aligned}$$

Exmample 14. The reliability of each component is p , find the reliability of both systems.



Solution.

$$\begin{aligned} R_1 &= P(A \cap (B \cup C) \cap D) \\ &= P(ABD \cup ACD) \end{aligned}$$

Solution.

$$\begin{aligned} R_1 &= P(A \cap (B \cup C) \cap D) \\ &= P(ABD \cup ACD) \\ &= P(ABD) + P(ACD) - P(ABCD) \end{aligned}$$

Solution.

$$\begin{aligned}R_1 &= P(A \cap (B \cup C) \cap D) \\&= P(ABD \cup ACD) \\&= P(ABD) + P(ACD) - P(ABCD) \\&= P(A)P(B)P(D) + P(A)P(C)P(D) \\&\quad - P(A)P(B)P(C)P(D)\end{aligned}$$

Solution.

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$$R_2 = P(AB \cup CD)$$

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Example (分支过程) 设某种单性繁殖的生物群(如果是两性繁殖的生物, 只考虑男性及其男性的后代)中每个个体进行独立繁衍, 每个个体产生 k 个下一代个体的概率为 p_k , $k = 0, 1, 2, \dots$, 记 $m = \sum_{k=1}^{\infty} kp_k$. 设该生物群开始时(即第0代)只有一个个体. 证明: 如果 $m \leq 1$, $p_1 < 1$, 则这一生物群灭绝(即到某一代时个体数为0) 的概率为1.

证. 记 A 为该生物群灭绝这一事件, B_k 表示第一代有 k 个个体(即第0产生的 k 个子代), 由全概率公式知所求的概率为

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$$q = P(A) = \sum_{k=0}^{\infty} P(A|B_k)P(B_k) = \sum_{k=0}^{\infty} P(A|B_k)p_k.$$

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在事件 B_k 的条件下, 生物群有 k 个个体, 而以其中任意一个个体及其后代构成的生物子群灭绝的概率仍然为 q . 故 $P(A|B_k) = q^k$.

所以

$$q = \sum_{k=0}^{\infty} q^k p_k.$$

即 q 是方程 $g(s) = s$ 的解, 其中 $g(s) = \sum_{k=0}^{\infty} s^k p_k$
($0 \leq s \leq 1$). 显然, $g(1) = 1$.

所以

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($0 \leq s \leq 1$). 显然, $g(1) = 1$. 而当 $0 \leq s < 1$ 时, 函数 $g(s)$ 的导数为

$$g'(s) = \sum_{k=1}^{\infty} s^{k-1} k p_k = p_1 + \sum_{k=2}^{\infty} s^{k-1} k p_k.$$

如果 $p_0 + p_1 < 1$, 则必有一个 $p_k > 0, k \geq 2$, 这时

$$g'(s) = p_1 + \sum_{k=2}^{\infty} s^{k-1} k p_k < p_1 + \sum_{k=2}^{\infty} k p_k = m \leq 1;$$

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如果 $p_0 + p_1 = 1$, 这时

$$g'(s) = p_1 < 1.$$

所以总是有 $(g(s) - s)' < 0$, $0 \leq s < 1$. 从

而 $g(s) - s$ 在 $[0, 1]$ 上严格单调递减, 故 $q = 1$ 是方程 $g(s) = s$ 的唯一解. 结论得证.

The independence of experiments

Suppose E_1, E_2, \dots, E_n are n experiments, then each possible outcome of each experiment can be treated as an event. E_1, E_2, \dots, E_n are said to be independent if A_1, A_2, \dots, A_n are independent for any $A_1 \in E_1, A_2 \in E_2, \dots, A_n \in E_n$.

1.4 Conditional probability and independent events

The independence of experiments

$\Omega_i - E_i$. To describe these n experiments, we construct a compound experiment

$E = (E_1, E_2, \dots, E_n)$ with $\Omega = \Omega_1 \times \Omega_2 \times \dots \times \Omega_n$,

and let sample points $\omega = (\omega^1, \dots, \omega^n)$, where $\omega^i \in \Omega_i$.

In a compound sample space, event A^i can be represented as $\Omega_1 \times \dots \times A^i \times \dots \times \Omega_n$, which we still denote by A^i .

Then the independence of E_1, E_2, \dots, E_n can be expressed in terms of

$$P(A^1 A^2 \dots A^n) = P(A^1) P(A^2) \dots P(A^n),$$

for all A^i of E_i , $i = 1, 2, \dots, n$.

Repeated independent experiments.

4. The Bernoulli model

A trial is called Bernoulli trial if there are only two possible outcomes for each trial.

Let A denote "success" and \overline{A} "failure" in a Bernoulli trial, then

$$\Omega = \{\omega_1, \omega_2\}, \quad \omega_1 = A, \omega_2 = \overline{A},$$

$$\mathcal{F} = \{\emptyset, A, \overline{A}, \Omega\}.$$

Given $P(A) = p$, ($0 < p < 1$), $P(\overline{A}) = 1 - p$.

Repeated independent Bernoulli trials are widely studied. We call this probability model the Bernoulli model.

Its sample points are $\omega = (\omega^1, \dots, \omega^n)$, where ω^i is A or \overline{A} and the total number of its sample points is 2^n .

The Bernoulli model is **not a classical probability model** since the probabilities of its sample points is not necessarily equal.

Example 16. Consider a Bernoulli model of n repeated independent trials. Let $A_k = \{A \text{ occurs only in the first } k \text{ trials}\}$, $B_k = \{A \text{ occurs exactly } k \text{ times}\}$. Find (1) $P(A_k)$, (2) $P(B_k)$.

Solution. (1) It is easy to see

$$A_k = \underbrace{AA \cdots A}_k \underbrace{\overline{AA} \cdots \overline{A}}_{n-k}.$$

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Since A occurs independently with probability p each trial, then $P(A_k) = p^k q^{n-k}$.

Solution. (1) It is easy to see

$$A_k = \underbrace{AA \cdots A}_k \underbrace{\overline{AA} \cdots \overline{A}}_{n-k}.$$

Since A occurs independently with probability p each trial, then $P(A_k) = p^k q^{n-k}$.

(2) Note that

$$\begin{aligned} B_k = & \underbrace{AA \cdots A}_k \underbrace{\overline{AA} \cdots \overline{A}}_{n-k} + \underbrace{AA \overline{A}}_{k-1} \underbrace{A \cdots A}_{n-k-1} \underbrace{\overline{AA} \cdots \overline{A}}_{n-k-1} \\ & + \cdots + \underbrace{\overline{AA} \cdots \overline{A}}_{n-k} \underbrace{AA \cdots A}_k. \end{aligned}$$

So

$$\begin{aligned} P(B_k) &= b(k, n, p) \\ &\stackrel{\wedge}{=} \binom{n}{k} p^k q^{n-k} = \frac{n!}{k!(n-k)!} p^k q^{n-k}, \end{aligned}$$

$k = 0, 1, 2, \dots, n$, which appear in the expansion

$(p + q)^n = \sum_{k=0}^n \binom{n}{k} p^k q^{n-k}$ with the total sum 1.

So call $b(k, n, p)$ the binomial distribution.

例 考察由投掷两个均匀的骰子组成的独立重复试验, 问两个骰子点数之和为5的结果出现在它们的点数之和为7的结果之前的概率是多少?

解法1: 令 E_n 表示前 $n - 1$ 次试验5点和7点都没有出现而在第 n 次试验出现了5点这一事件,

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$$P\left(\bigcup_{n=1}^{\infty} E_n\right) = \sum_{n=1}^{\infty} P(E_n).$$

每次试验中5点的出现的概率为 $P(F) = \frac{4}{36}$, 而7点的出现的概率为 $P(S) = \frac{6}{36}$.

解法1: 令 E_n 表示前 $n - 1$ 次试验5点和7点都没有出现而在第 n 次试验出现了5点这一事件, 那么所求的概率为

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$$P(E_n) = (1 - P(F) - P(S))^{n-1} P(F).$$

从而有

$$\begin{aligned} P\left(\bigcup_{n=1}^{\infty} E_n\right) &= \sum_{n=1}^{\infty} (1 - P(F) - P(S))^{n-1} P(F) \\ &= \frac{P(F)}{P(F) + P(S)} = \frac{2}{5}. \end{aligned}$$

解法2: 令 E 表示5点出现在7点之前这一事件,
 F 表示第一次试验结果为5点, S 表示第一次试验
结果为7点, O 表示第一次试验结果为其它的点.

解法2: 令 E 表示5点出现在7点之前这一事件,
 F 表示第一次试验结果为5点, S 表示第一次试验
结果为7点, O 表示第一次试验结果为其它的点.
那么

$$P(E) = P(E|F)P(F) + P(E|S)P(S) + P(E|O)P(O).$$

显然,

$$P(E|F) = 1, \quad P(E|S) = 0, \quad P(E|O) = P(E).$$

所以

$$P(E) = P(F) + P(E)(1 - P(F) - P(S)).$$

因此

$$P(E) = \frac{P(F)}{P(F) + P(S)}.$$

Example 10. One has two boxes of matches, each having n matches, in his pocket. Each time he wants to use match, he will randomly take out a box and draw one match from it. When he finds the box he takes out is empty, find the probability that the other box has just m matches.

Solution. The desired probability is

$$\begin{aligned} P &= P(\{\text{box A is empty, box B has } m \text{ matches}\}) \\ &\quad + P(\{\text{box B is empty, box A has } m \text{ matches}\}) \\ &\stackrel{\wedge}{=} P_1 + P_2. \end{aligned}$$

Consider P_1 first. When one box is empty,

$2n + 1 - m$ drawings are considered. So

$$\begin{aligned} & \{ \text{box A is empty, box B has } m \text{ matches} \} \\ = & \{ \text{in the first } 2n + 1 - m \text{ drawings,} \\ & \text{box A is drawn at the } (2n + 1 - m)\text{-th draw} \\ & \text{and, in the first } 2n - m \text{ drawings,} \\ & \text{box A is drawn } n \text{ times,} \\ & \text{box B is drawn } n - m \text{ times} \} \end{aligned}$$

Consider it as a Bernoulli model of $2n - m + 1$ repeated independent trials, where $A = \{\text{box A is drawn}\}$ and $\overline{A} = \{\text{box B is drawn}\}$, and $p = P(A) = 1/2$.

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$$P_1 = \binom{2n - m}{n} \left(\frac{1}{2}\right)^n \left(\frac{1}{2}\right)^{n-m} \frac{1}{2}.$$

Consider it as a Bernoulli model of $2n - m + 1$ repeated independent trials, where $A = \{\text{box A is drawn}\}$ and $\bar{A} = \{\text{box B is drawn}\}$, and $p = P(A) = 1/2$. Thus

$$P_1 = \binom{2n - m}{n} \left(\frac{1}{2}\right)^n \left(\frac{1}{2}\right)^{n-m} \frac{1}{2}.$$

Similarly,

$$P_2 = \binom{2n - m}{n} \left(\frac{1}{2}\right)^n \left(\frac{1}{2}\right)^{n-m} \frac{1}{2}.$$

Hence, the desired probability is

$$\begin{aligned} P &= 2 \binom{2n-m}{n} \left(\frac{1}{2}\right)^n \left(\frac{1}{2}\right)^{n-m} \frac{1}{2} \\ &= \binom{2n-m}{n} \left(\frac{1}{2}\right)^{2n-m}. \end{aligned}$$