

1. let a, M be sets of events

2. $P(a|M)$ conditional probability of a given M

$$P(a|M) = \frac{P(a \cap M)}{P(M)}$$

part of a included in M



3. Special cases A: $a \cap M = \emptyset$ then $P(a|M) = 0$

~~$P(a|M)$~~

B: $a \subset M$ then $a \cap M = a$ and $P(a|M) = \frac{P(a)}{P(M)}$

C: $a \supset M$ then $a \cap M = M$ and $P(a|M) = 1$

4. Dice example probability that the throw will be 3 if it is odd

$$P(3|\text{odd}) = \frac{P(3 \cap \text{odd})}{P(\text{odd})} = \frac{1/6}{1/2} = \frac{1}{3}$$

total probability Theorem:

given a_i such that $a_i \cap a_j = \emptyset$ (mutually exclusive) and $\sum_{i=1}^N a_i = S$ (everything) then

$$P(B) = \sum_{i=1}^N P(B, a_i)$$

6. Bayes Theorem, given a_i as in 5

$$P(a_i|B) = \frac{P(B|a_i)P(a_i)}{\sum_j P(B|a_j)P(a_j)}$$

7. Special case Suppose $P(a_i) / P(a_j) \approx 1$ for all i, j
and that $P(B|a_i) \gg \sum_{n=2}^N P(B|a_n)$

Then

$$P(a_i|B) = \frac{P(B|a_i)P(a_i)}{P(B|a_i)P(a_i) + \dots} \approx 1$$

8. Example of Bayes Theorem.

2 Boxes, ~~Box 1~~ each w/ 1000 cards. Box 1 has 999 white and 1 red, box 2 has 999 red and one white. A random card is extracted from a random ~~each~~ box, and its ~~color~~ white. what is prob. it came from box 1.

B_1 and B_2 are elements of Box 1 and Box 2 respectively.
W and R are white and red, respectively

$$P(B_1 | W) = \frac{999}{1000} \quad P(B_1 | R) = \frac{1}{1000} \quad P(B_1) = \frac{1}{2} = P(B_2)$$

$$P(B_2 | W) = \frac{1}{1000} \quad P(B_2 | R) = \frac{999}{1000} \quad P(W) = \frac{1}{2} \quad P(R) = \frac{1}{2}$$

$$P(W | B_1) = \frac{P(B_1 | W) P(W)}{P(B_1 | R) P(R) + P(B_1 | W) P(W)} = \frac{\frac{999}{1000} \cdot \frac{1}{2}}{\frac{1}{1000} \cdot \frac{1}{2} + \frac{999}{1000} \cdot \frac{1}{2}} \approx 1$$

Random variable \hat{x} where \mathcal{M} is a condition on x

$P(x | \mathcal{M})$ probability that r.v. \hat{x} is between $(x, x+\delta x)$ and that $x \in \mathcal{M}$.

10. Cumulative probability

$$F_x(x | \mathcal{M}) = P(\hat{x} \leq x | \mathcal{M}) = \frac{P(\hat{x} \leq x, \mathcal{M})}{P(\mathcal{M})}$$

note $P(x | \mathcal{M}) = \frac{d}{dx} F(x, \mathcal{M})$

11. Special case $\mathcal{M} : \{\hat{x} \leq a\}$ given that its less than a , what's the probability its less than x ?

$$F(x | \mathcal{M}) = P(\hat{x} \leq x | \hat{x} \leq a) = \frac{P(\hat{x} \leq x, x \leq a)}{P(x \leq a)}$$

but assume ~~so~~ so $P(\hat{x} \leq x, \hat{x} \leq a) = P(\hat{x} \leq x)$

$$P(\hat{x} \leq x | \hat{x} \leq a) = \frac{P(\hat{x} \leq x)}{P(\hat{x} \leq a)} = \frac{F(x)}{F(a)}$$

Then density f is

$$f(x | x \leq a) = \begin{cases} f(x) / F(a) = \frac{f(x)}{\int_0^a f(x) dx} & x < a \\ 0 & x > a \end{cases}$$

12. Failure rates.

\hat{x} is a r.v. that gives the time of failure of a system, and has density $f(x)$.

A, compute $f(\hat{x}=x | \hat{x} > t)$, the conditional distribution assuming that nothing's failed up to time t ($t < x$)

$$F(x | \hat{x} > t) = \frac{P(\hat{x} \leq x \cap \hat{x} > t)}{P(\hat{x} > t)} = \frac{F(x) - F(t)}{1 - F(t)}$$



computing density gives

$$f(x | \hat{x} > t) = \frac{f(x)}{\int_t^{\infty} f(x) dx} = \frac{F'(x)}{1 - F(t)}$$

B, conditional failure rate $\beta(t) = f(t | \hat{x} > t)$

$$\beta(t) = \frac{F'(t)}{1 - F(t)} \quad \text{integrates to}$$

$$-\ln[1 - F(t)] = \int_0^t \beta(x) dx$$

$$f(t) = \beta(t) e^{-\int_0^t \beta(x) dx}$$

~~note $\beta(t) = a e^{-at}$ then $f(t) =$~~

13. Bayes Thm for r.v.'s

$$f(x | a) = \frac{P(a | \hat{x}=x) f(x)}{\int_{-\infty}^{\infty} P(a | \hat{x}=x) f(x) dx}$$