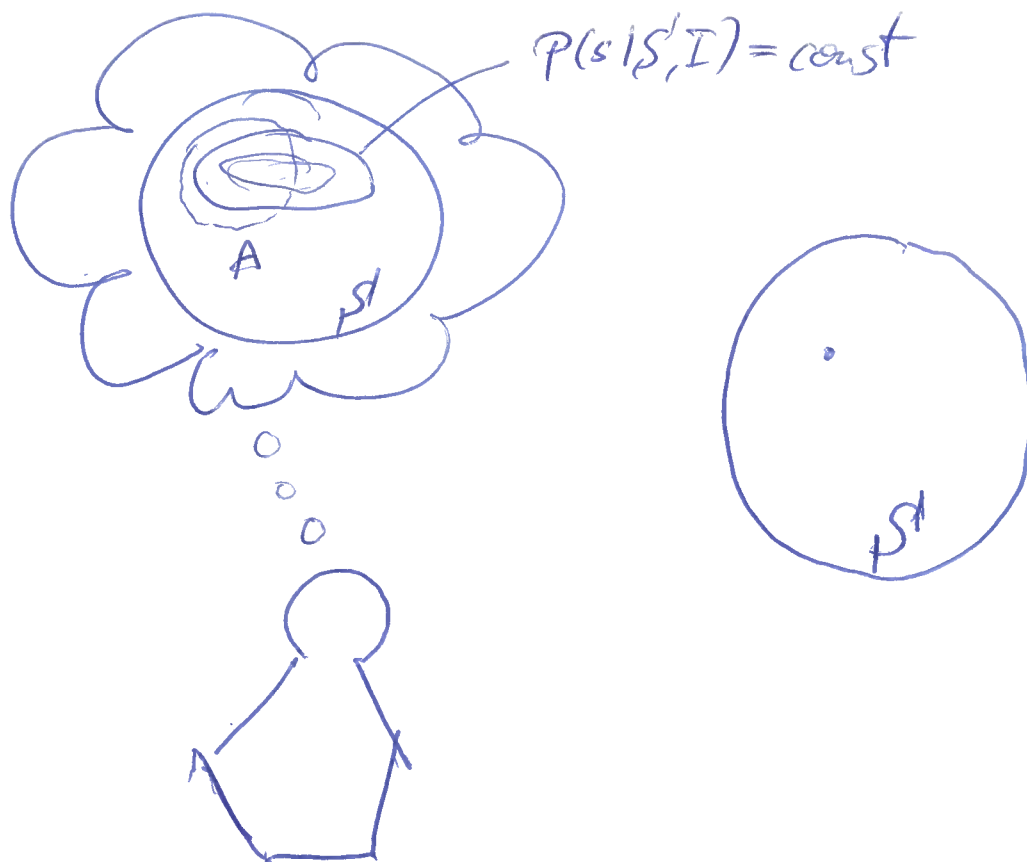




agent = intelligent system
 (human, animal, computer program, AI,
 cell, gene regulation network, ...)

Probability

-3-

$P(A|B)$ = degree of belief in A
assuming B is the case $\in [0,1]$

Frequency
 $F(A|B)$ = frequency of A happening
(per time, volume, etc) assuming
B is the case
$$= \frac{\# (AB) \text{ happens}}{\# B \text{ happens}} \in [0,1]$$

$$\begin{aligned} A, B &= \\ AB &= A \text{ and } B \\ &= A \cap B \end{aligned}$$

Note: The condition can only be omitted if it is clear from the context. Many, if not most paradoxes in Probability theory arise from confusion about the condition!

Calibration

$$P(A|B, F(A|B)=f) = f$$

If a frequency is known, ^{for AB} it is also the probability for AB.

Product rule:

$$\begin{aligned} P(A|B|C) &= P(A|BC)P(B|C) \\ &= P(B|AC)P(A|C) \\ \Rightarrow P(A|BC) &= \frac{P(B|AC)P(A|C)}{P(B|C)} \end{aligned}$$

e.g. Bayes theorem

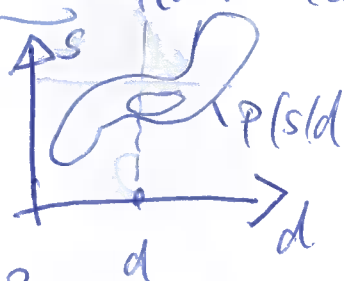
A = some variable $s \in S$ has a specific value, $s = s^*$ ($= \mathbb{I}_{s=s^*}$)

B = some data $d \in D$ has a specific value, $d = d^*$ ($= \mathbb{I}_{d=d^*}$)

C = all the prior knowledge on s and its relation to d , basically specifying $P(s) := P(s_{\text{real}} = s | C)$

and $P(s|d) := P(d_{\text{real}} = d | AC)$
 (likelihood) $P(d|s)P(s)$ / prior B $P(d|s) = \frac{P(d,s)}{P(s)}$

$$P(s|d) = \frac{P(d|s)P(s)}{P(d)} \quad \text{evidence}$$



How to calculate the evidence?

We need the sum rule!

Sum rule

$$P(A|C) + P(\bar{A}|C) = 1 \quad -5-$$

$$\bar{A} = \neg A = \text{not } A \quad (\text{cancel } A)$$

marginalization

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

$$P(A, \bar{B}|C) = P(\bar{B}|A, C) P(A|C)$$

$$P(A, B|C) + P(A, \bar{B}|C) = \underbrace{[P(B|A, C) + P(\bar{B}|A, C)]}_{1 \text{ (sum rule)}} P(A|C)$$

$$= P(A|C)$$

extension: $\{B_i\}_{i=1}^n$ mutually exclusive $[P(B_i, B_j|I) = 0 \text{ for } i \neq j]$

$A+B = A \vee B = A \text{ or } B$ and exhaustive $[P(B_1 + \dots + B_n|I) = 1]$
in I

$$\Rightarrow P(A|I) = \sum_{i=1}^n P(A, B_i|I)$$

generalized sum rule:

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

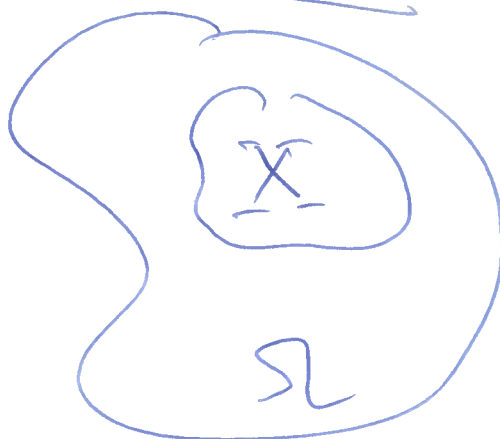
$|I$
 $\uparrow$
condition for
all probabilities
in an equation

Probability density functions

$I: X \subset \Omega, x \in \Omega$

$$P(x \in \bar{X} | I)$$

$$= \int_{\bar{X}} dx P(x|I)$$



product rule, sum rule, marginalization
all apply

Laplace's principle of insufficient reasoning

Let $\{B_i\}_{i=1}^n$ be mutually exclusive and exhaustive in I , and I symmetric w.r.t. to the B_i 's

$$\Rightarrow P(B_i | I) = P(B_j | I) = \frac{1}{n}$$

uniform distribution

Convention

data d : known quantity

signal s : unknown quantity of interest

noise u : " " that is not of interest