

Axioms of Probability

Let (Ω, A) be a measurable space of events.

(i) for a Event $E \in A$.

$$P(E) \geq 0 \quad (\text{non-negativity})$$

(ii) for two events $E_1, E_2 \in A$.

$$E_1 \subseteq E_2$$

$$P(E_1) \leq P(E_2) \geq 0$$

$$(iii) \quad P(\Omega) = 1$$

$$(iv) \quad P\left(\bigcup_{i=1}^{\infty} E_i\right) = \sum_{i=1}^{\infty} P(E_i)$$

$$(v) \quad \text{for } E, \text{ and } E'$$
$$P(E) = 1 - P(E')$$

Conditional Probability.

$$P(A/B)$$

(i) A and B are independent.
$$P(A/B) = P(A)$$

(ii) A and B are dependent
$$P(A/B) = \frac{P(B/A) \cdot P(A)}{P(B)}$$

(iii) A and B are disjoint/mutually exclusive

$$P(A/B) = 0$$

(iv) Product rule.

$$\begin{aligned} P(A, B) &= P(A/B) \cdot P(B) \\ &= P(B/A) \cdot P(A) \end{aligned}$$

Generalizing
~~P~~

$$P(A_1, A_2, \dots, A_n) = P(A_n/A_1, \dots, A_{n-1}) \cdot P(A_1, \dots, A_{n-1})$$

Bayes' Rule

$$P(A/B) = \frac{P(B/A) \cdot P(A)}{P(B)}$$

$P(A)$ = Prior Probability of hypothesis

$P(B)$ = Evidence

$P(B/A)$ = Likelihood

$P(A/B)$ = Posterior

Odd of A/B

$$\begin{aligned} O(A/B) &= \frac{P(A/B)}{P(\neg A/B)} \\ &= \frac{P(A/B)}{1 - P(A/B)} \end{aligned}$$

Probability Ranking Principle

Let

$$D = \{d_1, d_2, \dots, d_n\} \quad |D| = n$$

q is a fixed query.

As, Per Probability Ranking Principle (PRP)

for a given ' q ' we can rank documents as decreasing order of probability of relevance.

$$\bigwedge_{d_i=1}^n \text{rank decreasing } P[R=1/d, q]$$

$$P[R=1 / \vec{\pi}, \vec{q}]$$

$$P[R=1 / \langle t_1, \dots, t_n \rangle, \vec{q}]$$

This ranking of documents are same if we use probability of odd

that is

$$O[R/\vec{\pi}, \vec{q}] = \frac{P[R=1 / \vec{\pi}, \vec{q}]}{P[R=0 / \vec{\pi}, \vec{q}]}$$

$$O[R/\vec{\pi}, \vec{q}] = \frac{P[R=1/\vec{q}] \cdot P[\vec{\pi}/R=1, \vec{q}]}{P(\vec{\pi}/\vec{q})}$$

$$\div \frac{P[R=0/\vec{q}] \cdot P[\vec{\pi}/R=0, \vec{q}]}{P(\vec{\pi}/\vec{q})}$$

Here, $P(\vec{\pi}/\vec{q})$ and $P[R=1/\vec{q}]/P[R=0/\vec{q}]$ is constant as per Assumption

Hence

$$\begin{aligned} O[R/\vec{\pi}, \vec{q}] &= \frac{P[\vec{\pi}/R=1, \vec{q}]}{P[\vec{\pi}/R=0, \vec{q}]} \\ &= \prod_{t=1}^N \frac{P[\pi_t/R=1, \vec{q}]}{P[\pi_t/R=0, \vec{q}]} \cdot O(R/\vec{q}) \end{aligned}$$

Let

$$p_t = P(\pi_t=1/R=1, \vec{q})$$

$$u_t = P(\pi_t=1/R=0, \vec{q})$$

So,

	$R=1$	$R=0$
$x_t=1$	p_t	u_t
$x_t=0$	$1-p_t$	$1-u_t$

$$O[R/\vec{x}, \vec{v}] = \bar{\lambda} \frac{p_t}{t=x_t=q_t=1} u_t \cdot \bar{\lambda} \frac{1-p_t}{t=x_t=q_t=1} 1-u_t$$

Now

$$\bar{\lambda} \frac{1-p_t}{t=x_t=0, q_t=1} 1-u_t \text{ is a constant}$$

$$O[R/\vec{x}, \vec{v}] = \bar{\lambda} \frac{p_t(1-u_t)}{t=x_t=q_t=1} u_t(1-p_t)$$

$$\bar{\lambda} \frac{1-p_t}{t=q_t=1} 1-u_t$$

This odd $[R=1/\vec{x}, \vec{v}]$ can be used as a Retrieval Status Value.

$$RSV = \bar{\lambda} \frac{p_t(1-u_t)}{t=x_t=q_t=1} u_t(1-p_t)$$

$$= \bar{\lambda} \frac{p_t/(1-p_t)}{t=x_t=q_t=1} u_t/(1-u_t)$$

We can use logarithm to transform this into summation.

Estimating P_t and U_t .

- Let N be the # of Doc in the corpus.
- R be the # relevant documents in our sample
- n_t be the number of documents containing t .
- r_t be the # of ^{relevant} Doc. that contain t

$$P_t = \frac{r_t}{R}$$

$$U_t = \frac{n_t - r_t}{N - R}$$

Lidstone Smoothing ($\lambda = 0.5$)

$$P_t = \frac{r_t + 0.5}{R + 1}$$

$$U_t = \frac{n_t - r_t + 0.5}{N - R + 1}$$

Example