



2 (a) In the extreme case when $Pr(B) = 1$

A, B are statistically independent but $\text{conf}(A \rightarrow B) = 1$

On the other hand, $Pr(B)$ is related to $S(B)$, so $\text{lift}(A \rightarrow B)$ and $\text{conv}(A \rightarrow B)$ are able to capture the information of $Pr(B)$.

(b) i. $\text{conf}(A \rightarrow B)$

When $Pr(B) = 1$

$$\text{conf}(A \rightarrow B) = \text{conf}(B \rightarrow A)$$

$$\text{iff } 1 = Pr(A|B)$$

$$\text{iff } Pr(A) = 1$$

so $\text{conf}(A \rightarrow B)$ is not always symmetric.

ii $\text{lift}(A \rightarrow B)$

$$\text{lift}(A \rightarrow B) = \frac{\text{conf}(A \rightarrow B)}{S(B)} = \frac{S(A \cup B)}{S(A) \cdot S(B)} = \frac{\text{conf}(B \rightarrow A)}{S(A)} = \text{lift}(B \rightarrow A)$$

iii $\text{conv}(A \rightarrow B)$

$$\text{conv}(A \rightarrow B) = \frac{1 - S(B)}{1 - \frac{S(A \cup B)}{S(A)}} = \frac{S(A) - S(A) \cdot S(B)}{S(A) - S(A \cup B)} = 1 + \frac{S(A \cup B) - S(A) \cdot S(B)}{S(A) - S(A \cup B)}$$

When $S(A) \neq S(B)$ and $S(A \cup B) \neq S(A) \cdot S(B)$

$$\text{conv}(A \rightarrow B) \neq \text{conv}(B \rightarrow A)$$

(c) i $\text{conf}(A \rightarrow B)$

For perfect implication $A \rightarrow B$,

$$\text{conf}(A \rightarrow B) = 1 = \text{maximum achievable value}$$

ii $\text{lift}(A \rightarrow B)$

For perfect implication $A \rightarrow B$,

$$\text{lift}(A \rightarrow B) = \frac{1}{s(B)} \neq N \text{ (maximum achievable value)}$$

in general,

so lift is not desirable.

iii $\text{conv}(A \rightarrow B)$

For perfect implication $A \rightarrow B$,

$$\text{conv}(A \rightarrow B) = \frac{1 - s(B)}{0} = +\infty \text{ (ignore } \frac{0}{0} \text{ cases)}$$

so conv is desirable.

(d)(e) See the codes

Q3 (a) Let E_i be the event that the i -th non-zero column is not chosen.

$$\begin{aligned} \text{Then } P(E_1, \dots, E_m) &= P(E_1) \cdot P(E_2|E_1) \cdot \dots \cdot P(E_m|E_1, \dots, E_{m-1}) \\ &\leq P(E_1) \cdot P(E_2) \cdot \dots \cdot P(E_m) \\ &= (P(E_1))^m = \left(\frac{n-k}{n}\right)^m \end{aligned}$$

Note that $P(E_2|E_1) \leq P(E_2)$ (and similar inequalities) holds naturally in this case.

$$(b) \quad \left(\frac{n-k}{n}\right)^m = \left(1 - \frac{k}{n}\right)^m = \left(1 - \frac{1}{\frac{n}{k}}\right)^m$$

We can make $m \approx 10 \frac{n}{k}$

$$\text{i.e. } k \approx 10 \frac{n}{m}$$

$$\text{then } \left(1 - \frac{1}{\frac{n}{k}}\right)^m \approx \left(1 - \frac{1}{\frac{n}{k}}\right)^{10 \frac{n}{k}} \approx e^{-10}$$

(c)	S_1	S_2	Jaccard similarity = $\frac{1}{2}$
	1	0	random cyclic permutation gives $\frac{2}{3}$
	0	0	
	1	1	

Q4. (a) By Markov's inequality

$$\Pr \left[\sum_{j=1}^L |T \cap W_j| \geq 3L \right] \leq \frac{E \left[\sum_{j=1}^L |T \cap W_j| \right]}{3L}$$

$$= \frac{E[|T \cap W_1|]}{3}$$

Since \mathcal{H} is

$(\lambda, c\lambda, P_1, P_2)$ -sensitive

$$\leq \frac{n P_2^k}{3}$$

$$= \frac{n \cdot \frac{1}{n}}{3} = \frac{1}{3}$$

$$(b) \Pr [\forall 1 \leq j \leq L, g_j(x^*) \neq g_j(z)] \leq (1 - P_1^k)^L$$

$$= (1 - P_1^{\log_{1/P_2}(n)})^{n^{\frac{\log(\frac{1}{P_1})}{\log(\frac{1}{P_2})}}}$$

$$= \left(1 - \frac{1}{P_1^{\log_n(\frac{1}{P_2})}} \right)^{P_1^{\log_n(\frac{1}{P_2})}} < \frac{1}{e}$$

$$(c) \text{ By (a) (b), with probability } \geq (1 - \frac{1}{3} - \frac{1}{e})$$

we have $< 3L$ points in T chosen in step 3

and at least 1 point not in T chosen in step 3.

(d) See the codes