Notes on Probability and Computing

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A probability space is a measure space $(\Omega, \mathcal{F}, \mathbb{P})$ consisting of:

- the sample space Ω a set of outcomes called sample;
- the σ -algebra \mathcal{F} a family of subsets of Ω , called *events*, such that $\Omega \in \mathcal{F}$ and \mathcal{F} is closed under complements (i.e. $\forall A \in \mathcal{F}$, $\Omega \setminus A \in \mathcal{F}$) and countable unions (i.e. $\forall A_i \in \mathcal{F}$, $\bigcup_{i=1}^{\infty} A_i \in \mathcal{F}$);
- the probability function $\mathbb{P}: \mathcal{F} \to [0,1]$ such that $\mathbb{P}(\Omega) = 1$ and \mathbb{P} is σ -additive (i.e. $\mathbb{P}(\bigsqcup_{i=1}^{\infty} A_i) = \sum_{i=1}^{\infty} \mathbb{P}(A_i)$).

The motivation behind this complicated definition is that some sets are *non-measurable*, thus mathematicians developed the theory of *measure*. For instance, *Borel set* on real line forms a σ -algebra which is *generated by* open intervals. Stieltjes measures is a *Borel measure* and builds the measure foundation of *continuous probability distribution*.

Lemma 1.1 (Inclusion-exclusion principle) Let E_1, \dots, E_n be any n events. Then

$$\mathbb{P}\left(\bigcup_{i=1}^n E_i\right) = \sum_{\ell=1}^n (-1)^{\ell+1} \sum_{i_1 < i_2 < \dots < i_\ell} \mathbb{P}\left(\bigcap_{r=1}^\ell E_{i_r}\right).$$

Events E_1, E_2, \dots, E_n are mutually independent (simply called independent when k = 2) if and only if, for any subset $I \subseteq \{1, 2, \dots, k\}$, $\mathbb{P}\left(\bigcap_{i \in I} E_i\right) = \prod_{i \in I} \mathbb{P}(E_i)$. Note that events X, Y, Z, \dots are unnecessarily mutually independent when they are pairwise independent.

The conditional probability that event E occurs given that event F occurs is $\mathbb{P}(E \mid F) = \mathbb{P}(E \cap F) / \mathbb{P}(F)$ ($\mathbb{P}(F) > 0$).

Theorem 1.2 (Law of total probability) Let events $\bigsqcup_{i=1}^n E_i = \Omega$. Then we have $\mathbb{P}(B) = \sum_{i=1}^n \mathbb{P}(B \mid E_i) \cdot \mathbb{P}(E_i)$.

Theorem 1.3 (Bayes's law) Let events E_1, E_2, \dots, E_n satisfy $\bigsqcup_{i=1}^n E_i = \Omega$. Then we have

$$\mathbb{P}(E_k \mid B) = \frac{\mathbb{P}(E_k \cap B)}{\mathbb{P}(B)} = \frac{\mathbb{P}(B \mid E_k) \cdot \mathbb{P}(E_k)}{\sum_{i=1}^n \mathbb{P}(B \mid E_i) \cdot \mathbb{P}(E_i)}.$$

In the *Bayesian approach* one starts with a *prior* model, giving some initial value to the model parameters. This model is then modified, by incorporating new observations, to obtain a *posterior* model that captures the new information.

Exercise 1.6 Using mathematical induction, we have $p_{i,j} = \frac{i-1}{i+j-1} \cdot p_{i-1,j} + \frac{j-1}{i+j-1} \cdot p_{i,j-1} = \frac{i+j-2}{i+j-1} \cdot \frac{1}{i+j-2} = \frac{1}{i+j-2}$.

Exercise 1.7.b Let $F_{b_1b_2\cdots b_n}$ be the intersection of events E_i ($b_i=1$) or $\Omega\backslash E_i$ ($b_i=0$), and P_k be the sum of $\mathbb{P}(F_b)$ where b consists of k one and n-k zero. Then for every $k\geq 1$, we have $\sum_{i=1}^l (-1)^{i+1} \binom{k}{i} = 1 + (-1)^{l+1} \binom{k-1}{l} \geq 1$. Multiply both sides by P_k and sum them up. We eventually reach the desired inequality.

Exercise 1.11.b $p_3 = p_1 \cdot (1 - p_2) + (1 - p_1) \cdot p_2 \Rightarrow q_3 = 1 - 2p_3 = (1 - 2p_1)(1 - 2p_2) = q_1q_2$. Is there any underlying motivation?

Exercise 1.24 (The generalization of Karger's algorithm) Let k be the size of minimum r-way cut-sets. Considering all r-way cut-sets consisting of r-1 single vertex, the total size is $m \cdot \binom{n-2}{r-1}$ (considering each edges) with another upper bound $(m-k) \cdot \binom{n}{r-1}$. Thus,

$$m \cdot \binom{n-2}{r-1} \leq (m-k) \cdot \binom{n}{r-1} \quad \Rightarrow \quad 1 - \frac{k}{m} \geq \binom{n-2}{r-1} \binom{n}{r-1}^{-1} = \frac{(n-r+1)(n-r)}{n(n-1)}$$