

Notes on Probability and Computing

Hao Su

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Chapter 1

Events and Probability

Eg 1.1 Consider the following procedure that verifies if $F(x) \equiv G(x)$, where $F(x)$ is given as a product $F(x) = \prod_{i=1}^d (x - a_i)$ and $G(x)$ is given in its canonical form: assume that the maximum exponent of x in $F(x)$ and $G(x)$ is d , choose an integer r uniformly at random in the range $\{1, \dots, 100d\}$, and decide based on if $F(x) = G(x)$. Should an error occur then $F(x) \not\equiv G(x)$ and r is a root of $F(x) - G(x) = 0$, whose degree is no larger than d and thus has no more than d roots. Hence the chance of a wrong answer produced by this procedure is no more than $1/100$. \diamond

Defn 1.2 A *probability space* has three components:

1. a sample space Ω , which is the set of all possible outcomes of the random process modeled by the probability space;
2. a family of sets \mathcal{F} representing the allowable events, where each set in \mathcal{F} is a subset of the sample space; and
3. a probability function $\Pr : \mathcal{F} \rightarrow \mathbb{R}$ satisfying 1.3.

An element of Ω is called a *simple* or *elementary* event. \diamond

In a discrete probability space $\mathcal{F} = 2^\Omega$, and \Pr is uniquely defined by the probabilities of the simple events. The events need to be *measurable*, thus $\in \mathcal{F}$ and \mathcal{F} should be closed under complement and union and intersection of countably many sets (a σ -algebra).

Defn 1.3 A *probability function* is any function $\Pr : \mathcal{F} \rightarrow \mathbb{R}$ that satisfies the following conditions:

1. for any event E , $0 \leq \Pr(E) \leq 1$;
2. $\Pr(\Omega) = 1$; and

3. for any finite or countably infinite sequence of pairwise mutually disjoint events E_1, E_2, E_3, \dots ,

$$\Pr\left(\bigcup E_i\right) = \sum \Pr(E_i). \quad \diamond$$

Lem 1.4 For any two events E_1 and E_2 ,

$$\begin{aligned} \Pr(E_1 \cup E_2) &= \Pr(E_1 - E_1 \cap E_2) + \Pr(E_2 - E_1 \cap E_2) + \Pr(E_1 \cap E_2) \\ &= \Pr(E_1 - E_1 \cap E_2) + \Pr(E_2) \\ &= \Pr(E_1) + \Pr(E_2) - \Pr(E_1 \cap E_2) \end{aligned} \quad \diamond$$

Draw venn diagrams to utilize 1.3.

Lem 1.5 *Union Bound* For any countable sequence of events E_1, E_2, \dots ,

$$\Pr\left(\bigcup E_i\right) \leq \sum \Pr(E_i). \quad \diamond$$

Lem 1.6 *Inclusion-Exclusion Principle* Let E_1, \dots, E_n be any n events. Then

$$\begin{aligned} \Pr\left(\bigcup_{i=1}^n E_i\right) &= \sum_{i=1}^n \Pr(E_i) - \sum_{i_1 < i_2} \Pr(E_{i_1} \cap E_{i_2}) + \sum_{i_1 < i_2 < i_3} \Pr(E_{i_1} \cap E_{i_2} \cap E_{i_3}) \\ &\quad - \dots + (-1)^{l+1} \sum_{i_1 < i_2 < \dots < i_l} \Pr\left(\bigcap_{r=1}^l E_{i_r}\right) + \dots \end{aligned} \quad \diamond$$

Defn 1.7 Two events E and F are *independent* iff

$$\Pr(E \cap F) = \Pr(E) \cdot \Pr(F).$$

Events E_1, E_2, \dots, E_k are *mutually independent* iff for any $I \subseteq [1, k]$,

$$\Pr\left(\bigcap_{i \in I} E_i\right) = \prod_{i \in I} \Pr(E_i). \quad \diamond$$

Defn 1.8 The *conditional probability* that E occurs given that F occurs is

$$\Pr(E|F) = \frac{\Pr(E \cap F)}{\Pr(F)}.$$

and is well defined only if $\Pr(F) > 0$. \diamond

This is very intuitive, looking for the probability of $E \cap F$ within the set of events defined by F . When E and F are independent and $\Pr(F) \neq 0$, we have

$$\Pr(E|F) = \frac{\Pr(E \cap F)}{\Pr(F)} = \frac{\Pr(E) \Pr(F)}{\Pr(F)} = \Pr(E).$$

We may sample with or without replacement in 1.1, and sampling Without replacement seems a little better. We may opt for sampling with replacement yet, for easier analysis and implementation.

Eg 1.9 Consider three $n \times n$ matrices \mathbf{A} , \mathbf{B} , and \mathbf{C} , and assume we are working over integers modulo 2. We want to verify whether $\mathbf{AB} = \mathbf{C}$. Multiplying \mathbf{A} and \mathbf{B} takes roughly $\Theta(n^{2.37})$ operations. Instead we may choose a random vector $\bar{r} = (r_1, r_2, \dots, r_n) \in \{0, 1\}^n$ and verify if $\mathbf{A}(\mathbf{B}\bar{r}) \neq \mathbf{C}\bar{r}$, which should take $\Theta(n^2)$ time in an obvious way. \diamond