

Deterministic Algorithms



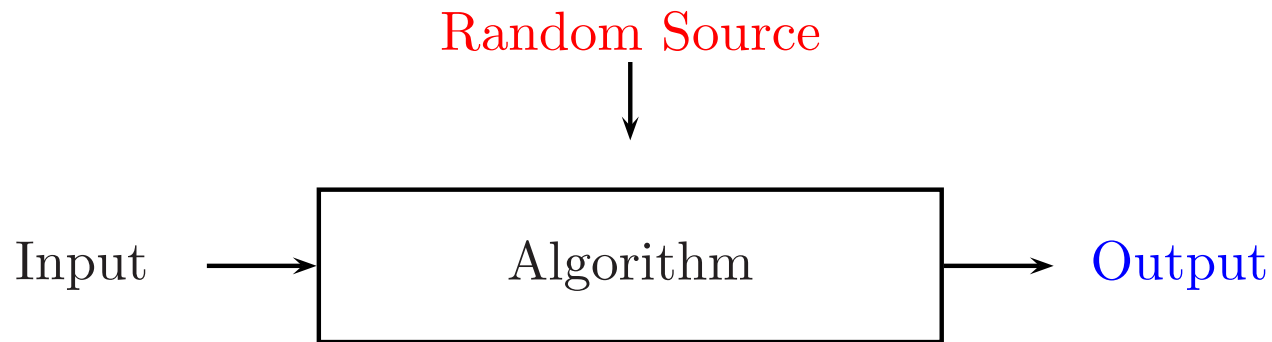
To prove that the algorithm **always** solves the problem as **quickly** as possible (in a polynomial number of steps)

Probabilistic Analysis of Deterministic Algorithms



- The **input** is assumed to be from a probability distribution.
- The **expected performace** of the algorithm is nice
- With **high probability** the algorithm **performs** well.

Randomized Algorithms



In addition to the input, the algorithm takes a source of **random numbers** and makes **random choices** during the execution.

The **running time** of a randomized algorithm may vary from run to run.

Given a problem, **we wish to design algorithms + analysis to show that the behaviour of the algorithm is likely to be good, on every input of the problem**

Probabilistic Algorithms: Monte Carlo and Las Vegas

- A **Monte Carlo** algorithm runs for a fixed number of steps and produces an answer that is correct with probability $\geq 1/2$. However, we may make the probability of error as small as we wish **amplification**.
- A **Las Vegas** algorithm always produces the correct answer; **its running time is a random variable whose expectation is bounded** (by a polynomial).

- A **Las vegas** algorithm can be made **Monte Carlo** by having it terminate with an arbitrary wrong answer, if it exceeds a time bound $t(n)$. Since Las vegas is **unlikely** to exceed its time bound, the resulting Monte Carlo is **unlikely** to give the wrong answer.
- There is not known **universal method** for making a **Monte Carlo** algorithm into a **Las Vegas** one.

Random Selection

The first randomization technique we will see is **Random Selection**:

The intuition behind this idea is that a single randomly selected individual is probably a **typical** representative of the entire population.

Therefore, random selection provides a good way to avoid selecting rare **bad** elements.

Verifying Polynomial Identities

Given polynomials $P(x)$ and $Q(x)$ we wish to verify if $P(x) \equiv Q(x)$

Recall: $P(x) \equiv Q(x)$ iff at every point x , $P(x) = Q(x)$.

Example Check if $(x + 1)(x - 2)(x + 3)(x - 4)(x + 5)(x - 6) \equiv x^6 - 7x^3 + 25$

Deterministic solution:

If P and Q are written explicitly, there is a linear algorithm (comparing the coefficients):

Transform $P(x)$ to canonical form $H(x) \equiv \sum_{i=0}^d c_i x_i$ where d is the degree of $P(x)$ or of $Q(x)$.

then $P(x) \equiv Q(x)$ iff coefficients of all monomial are equal.

But for example, Q can be given:

$$Q(x) = \prod_{i < j | i, j \neq 1} (x_i - x_j) - \prod_{i < j | i, j \neq 2} (x_i - x_j) + \prod_{i < j | i, j \neq 3} (x_i - x_j) - \dots$$
$$\pm \prod_{i < j | i, j \neq n} (x_i - x_j)$$

Randomized solution

Given $P(x)$ and $Q(x)$ of degree d :

Schwartz-Zippel's Algorithm

Choose a random integer $r \in S = [0, \dots, 100d]$

Compute $P(r)$ and $Q(r)$

if $P(r) = Q(r)$ **then** TRUE

else FALSE

For example, if $r = 2$ then $P(2) = 420$ and $Q(2) = 33$, therefore as $P(2) \neq Q(2)$ the algorithm will return FALSE.

Notice that if $P(x) \equiv Q(x)$ any value of r will yield equality

For example $(x + 1)(x - 1) \equiv x^2 - 1$.

In the case $P(x) \not\equiv Q(X)$: **A bad choice of r may lead to a wrong answer**

For ex. to check if $x^2 + 7x + 1 = (x + 2)^2$

If $r = 2$, $19 \neq 16$ so the two polynomials are different,

but $r = 1$, $9 = 9$!!

Theorem: Let $F(x) = P(x) - Q(x)$ have degree $\leq d$. If $F \neq 0$ then

$$\Pr [F(r) = 0] \leq \frac{d}{|S|} = 1\%$$

A **simple event**: a choice of r ; A **sample space**: all $[100d]$

The probability of a simple event: $\Pr [r] = \frac{1}{100d}$.

By the Fundamental Theorem of Algebra, a polynomial of degree d have at most d roots.

Therefore: a **bad** event is choosing a root of $F(X)$. There as there are at most d roots, then

$$\Pr [\text{bad event}] \leq \frac{d}{100d}.$$

- If the identity is correct, the algorithm always output the correct answer,
- If the identity is NOT correct, the algorithm outputs the **wrong** answer only if r is a root of $F(x) = P(x) - Q(x) = 0$

Amplification

We can decrease the **error probability** at the expense of increasing the run-time of the algorithm:

```
Run the algorithm  $k$  times  
if each time is TRUE output TRUE  
    else FALSE
```

The probability of a wrong answer in one run of the algorithm is $1/100$. Runs of the algorithms are **independent**. It is enough to get a correct (negative) answer in one of the runs.

The probability of error in k runs is $(\frac{1}{100})^k$.

Quicksort

Recall: Given set A , $|A| = n$, to sort:

select last key to be splitter y

compare every element in A to y to get:

$$S = \{x \in A - \{y\} | x \leq y\}, B = \{x \in A - \{y\} | x \geq y\}$$

return SyB (apply Quicksort separately to S and B)

How to choose the splitter y :

Choose the rightmost input number, in the first scanning.

7 8 2 6 5 1 3 4

2 1 3 4 5 8 7 6

2 1 3 4 5 6 8 7

1 2 3 4 5 6 7 8

1 2 3 4 5 6 7 8

Randomized Quicksort

Given set A , $|A| = n$, to sort

Rquicksort

select u.a.r. $y \in A$

compare every element in A to y to get:

$$S = \{x \in A - \{y\} | x \leq y\}, B = \{x \in A - \{y\} | x \geq y\}$$

return SyB .

Let T be the number of comparisons in a run of Rquicksort

Theorem: $\mathbf{E}[T] = O(n \log n)$.

Consider a **uniform distribution** of the instances for Rquicksort. Given n different keys, each permutation has the same probability of appearance $\frac{1}{n!}$.

Ω will consist of all possible $n!$ permutations of A .

For any $w \in \Omega$ $\Pr w = \frac{1}{n!}$.

Let a_1, a_2, \dots, a_n be **sorted** elements of A .

For $i \in [n]$ and $j > i$ define the indicator variable:

$$X_{i,j} = \begin{cases} 1 & \text{if } a_i \text{ is compared to } a_j \\ 0 & \text{otherwise} \end{cases}$$

The number of comparisons in running the algorithm is

$$T = \sum_{i=1}^n \sum_{j>i} X_{i,j}.$$

We wish to compute $\mathbf{E}[T]$

a_i is compared to a_j iff either a_i or a_j is chosen as splitter before any of the $j - i - 1$ elements in $[a_{i+1}, \dots, a_{j-1}]$.

Elements are chosen u.a.r. \Rightarrow elements in $[a_i, a_{i+1}, \dots, a_{j-1}, a_j]$ are chosen u.a.r.

$$\Pr [X_{i,j} = 1] = \frac{2}{j-i+1} \Rightarrow \mathbf{E} [X_{i,j}] = \frac{2}{j-i+1}.$$

$$\begin{aligned} \mathbf{E} [T] &= \mathbf{E} \left[\sum_{i=1}^n \sum_{j>i} X_{i,j} \right] = \sum_{i=1}^n \sum_{j>i} \mathbf{E} [X_{i,j}] \\ &= \sum_{i=1}^n \sum_{j>i} \frac{2}{j-i+1} \leq n \sum_{k=1}^n \frac{2}{k} \\ &= 2nH_n = n \log n + O(n). \end{aligned}$$

Probabilistic Analysis of Quicksort

Theorem: The expected time of Quicksort on a random input, chosen u.a.r. from all possible permutations of A is $O(n \log n)$.

Proof. Set the r.v. $X_{i,j}$ as before. As all permutations A_i, \dots, A_j have equal probability $\Pr[X_{i,j}] = \frac{2}{j-i+1} \Rightarrow \mathbf{E} \left[\sum_{i=1}^n \sum_{j>i} X_{i,j} \right] = O(n \log n)$. □

A randomized algorithm for the Minimum Cut on a graph

Let $G = (V, E)$ be an undirected graph. A **cut** in G is a set of edges whose removal separates G . The **MINCUT** problem is to find a cut of minimal size.

Given $G = (V, E)$, $|V| = n$

Karger's algorithm

while G has more than 2 vertices **do**

 select an edge $e = \{i, j\}$ uniformly at random

 contract e

output the remaining edges.

Contract E means that the i and j are merged into a single vertex, retaining all their connections to other vertices.

Notice the above algorithm:

- It is a Monte Carlo algorithm,
- $t(n) = (n - 2)O(n) = O(n^2)$.
- Always outputs a valid cut in G (why?)
- Need to analyze the probability it outputs a **minimum cut**

Let \mathcal{C} be a **particular** minimum cut, $|\mathcal{C}| = k$. We wish to evaluate the probability that no edge in \mathcal{C} is chosen in the contraction process. Let A_i be the event that \mathcal{C} survives iteration i .

Pr [A_1]: Every $j \in V$ has degree $\geq k$ (otherwise, i itself would be a min. cut), so $|E| \geq \frac{nk}{2}$ (otherwise there would be an edge with degree $< k$).

$$\therefore \mathbf{Pr} [A_1] \geq 1 - \frac{k}{(kn/2)} = 1 - \frac{2}{n} = \frac{n-2}{n}.$$

To see the prob. that \mathcal{C} survives the second iteration **given that it survived the first**, notice at second step $|E| \geq \frac{k(n-1)}{2}$, so probability of choosing an edge from \mathcal{C} at second iteration is $\leq \frac{k}{k(n-1)/2} = \frac{2}{n-1}$.

$$\therefore \mathbf{Pr} [A_2] \geq 1 - \frac{2}{n-1} = \frac{n-3}{n-1}.$$

At the i -th iteration, we have $|V| = n - i + 1$, $|\mathcal{C}| = k$, therefore $|E| \geq \frac{k(n-i+1)}{2}$.

$$\therefore \Pr [A_i | \bigcap_{j=1}^{i-1} A_j] \geq 1 - \Pr [A_i] = 1 - \frac{k}{k(n-i+1)/2} = \frac{n-i-1}{n-i+1}$$

The probability that no edge in \mathcal{C} is chosen during the whole process:

$$\Pr [\bigcap_{i=1}^{n-2} A_i] \geq \frac{n-2}{n} \times \frac{n-3}{n-1} \times \dots \times \frac{2}{4} \times \frac{1}{3} = \frac{2}{(n-1)n}.$$

Therefore the probability the algorithm gives a mistake is less or equal than $(1 - \frac{2}{n^2})$

How to reduce more this probability?. Amplify.

Repeat independently t times the algorithm, and choose the smallest cut we find. The probability that we fail to discover \mathcal{C} on all t attempts is $\leq (1 - \frac{2}{n^2})^t$. So taking $t = cn^2$, the probability of error is at most $e^{-c/2}$. Therefore, to make the probability that the algorithm fails $= 10^{-6} = e^{-14}$ we just need to do $28n^2$ repetitions.