The Magic of Probability

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GCOE/IST Citizen Lecture No. 4

Solution via Randomization

Let us start today's lecture by asking the following.

Question

Introduction I

Does probability help to design efficient algorithms?



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Introduction I

Does probability help to design efficient algorithms?

At first glance, this question may sound strange. We are very much used to design deterministic algorithms or to write deterministic programs.

- (1) The instruction is a finite text.
- (2) The computation is done step by step, where each step performs an elementary operation.
- (3) In each step of the execution of the computation it is **uniquely** determined which elementary operation we have to perform.
- (4) The next computation step depends only on the input and the intermediate results computed so far.

Looking at all the usual algorithms we know, we can say that an algorithm is a computation method having the following properties.

(1) The instruction is a finite text.

- (2) The computation is done step by step, where each step performs an elementary operation.
- (3) In each step of the execution of the computation it is uniquely determined which elementary operation we have to perform.
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If probability should help, then we have to replace Property (3) above by allowing instructions of the following type:

Flip a coin. If "head," goto i else goto j .

Such a replacement directly implies that a program may have many *different* computations when run on the same input. So, on some runs, the program may *fail* to achieve its goal, and on some runs, it may *succeed*. However, we shall make sure that the program either succeeds or fails to achieve its goal.

But in general, we have no way to tell which of these two cases did actually happen.

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Let us recall the basic definition of probability. Classically, one defines the probability of an event to be *the ratio of favorable cases to all cases*.

Example ⁻

Introduction IV

Let us consider a fair die having the usual six possible outcomes 1, 2, 3, 4, 5, 6. We consider the event that we throw an even number. Then the favorable cases are 2, 4, 6. So, the probability to throw an even number is 3/6 = 1/2.

It should be mentioned that this definition only works if all elementary events have the same probability. Therefore, we required our die to be fair.

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It should be mentioned that this definition only works if all elementary events have the same probability. Therefore, we required our die to be fair. So, in our lecture we shall ensure that each run of the probabilistic algorithm has the same probability. We shall achieve this goal by allowing only one random choice at the first stage of the algorithm. All other stages (or steps) must be deterministic.

Then our design of a probabilistic algorithm should ensure that the "overwhelming" number of runs is successful.

Now, it is time to present the problem we wish to study. This problem is taken from Juraj Hromkovič's book *Algorithmic Adventures, From Knowledge to Magic*.

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Solution via Randomization

The Problem I

Suppose we have two computers C_1 and C_2 that are very far apart. Nevertheless, we want to have on both computers the same *huge* database. Initially, on both computers we have the same database. However, the database evolves over time and every new datum is sent to both computers.

So, the changes are to be performed *simultaneously* on both computers, e.g., incorporating newly discovered genome sequences into both databases.

From time to time we wish to check whether or not both computers do have the same database. In order to simplify the presentation, we consider the contents of the databases of C_1 and C_2 as a sequence of bits, i.e., computer C_1 has $x = x_1 x_2 \cdots x_{n-1} x_n$ and computer C_2 has $y = y_1 y_2 \cdots y_{n-1} y_n$.

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Solution via Randomization

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Thus, by using a communication channel (a network) between C_1 and C_2 , we have to verify whether or not x = y (see the figure below).



Figure 1: The verification task.

To solve this task one has to design a communication protocol.



We measure the complexity of the communication by counting the number of bits exchanged between C_1 and C_2 . The bad news are that any deterministic communication protocol cannot be better (on most inputs) than to exchange n bits.

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Suppose that we have $n = 10^{16}$ (which is roughly the memory size of 25 000 DVDs). Exchanging such an amount of bits over the internet without making any error is almost unrealistic given current technology. And the communication complexity is too high. If we can transmit 10^7 many bits per second, it will take approxiamately 31 years.

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Solution via Randomization

The Problem IV

For the remaining part of the talk, it is advantageous to *interpret* the sequences $x = x_1x_2 \cdots x_{n-1}x_n$ and $y = y_1y_2 \cdots y_{n-1}y_n$, $x_i, y_i \in \{0, 1\}, i = 1, ..., n$, as numbers. That is

$$num(x) = \sum_{i=1}^{n} 2^{n-i} \cdot x_i, \text{ and}$$
$$num(y) = \sum_{i=1}^{n} 2^{n-i} \cdot y_i.$$

Example 2 $num(10001) = 1 \cdot 2^4 + 0 \cdot 2^3 + 0 \cdot 2^2 + 0 \cdot 2^1 + 1 \cdot 2^0 = 17,$ $num(11111) = 1 \cdot 2^4 + 1 \cdot 2^3 + 1 \cdot 2^2 + 1 \cdot 2^1 + 1 \cdot 2^0 = 31.$

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The Problem V

Obviously, we have x = y if and only if num(x) = num(y). We should also note that

 $\begin{array}{rll} 0 & \leqslant & num(x) \leqslant 2^n-1 \ , & \mbox{ and} \\ 0 & \leqslant & num(y) \leqslant 2^n-1 \ . \end{array}$

So, these numbers are *huge*.

We need the following notations. For every positive integer $m \ge 2$ we set

 $\begin{aligned} & \text{PRIM}(\mathfrak{m}) &= \{ p \mid p \text{ is a prime and } p \leqslant \mathfrak{m} \}, & \text{and} \\ & \text{Prim}(\mathfrak{m}) &= |\text{PRIM}(\mathfrak{m})|, \end{aligned}$

where |PRIM(m)| denotes the number of elements in the set PRIM(m).

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$$\begin{split} PRIM(\mathfrak{m}) &= \left\{ p \mid p \text{ is a prime and } p \leqslant \mathfrak{m} \right\}, \quad \text{ and } \\ Prim(\mathfrak{m}) &= \left| PRIM(\mathfrak{m}) \right|, \end{split}$$

where |PRIM(m)| denotes the number of elements in the set PRIM(m).

End

The Proble

Solution via Randomization

In the following we denote by $r = a \mod b$ the remainder of the division a : b.

Example 3

The Problem VI

Let a = 17 and b = 5; then we can write

 $17 = 3 \cdot 5 + 2$,

and therefore we obtain $2 = 17 \mod 5$.

Now we are in the position to present our *randomized communication protocol* for the comparison of num(x) and num(y).

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The Probler

Solution via Randomization

The RCP WITNESS I

The randomized communication protocol WITNESS

given: Computer C_1 and n bits $x_1x_2 \cdots x_n$ Computer C_2 and n bits $y_1y_2 \cdots y_n$

- Phase 1: C_1 chooses uniformly at random a prime p from $PRIM(n^2)$.
- Phase 2: C_1 computes the integer

 $s = num(x) \mod p$

and sends s and p in binary representation to C_2 . Phase 3: After reading s and p, the computer C_2 computes

 $q = num(y) \mod p$.

If $q \neq s$, then C₂ outputs "not equal." If q = s, then C₂ outputs "equal."

The Problem

Solution via Randomization

The RCP WITNESS II

Example 4

Let x = 01111 and y = 10110. Hence, n = 5 and

$$\begin{array}{lll} num(x) &=& 0\cdot 2^4 + 1\cdot 2^3 + 1\cdot 2^2 + 1\cdot 2^1 + 1\cdot 2^0 = 15 \ , \\ num(y) &=& 1\cdot 2^4 + 0\cdot 2^3 + 1\cdot 2^2 + 1\cdot 2^1 + 0\cdot 2^0 = 22 \ . \end{array}$$

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Also, $n^2 = 25$ and PRIM(25) = {2, 3, 5, 7, 11, 13, 17, 19, 23}. Assume that in Phase 1 the prime 5 has been chosen uniformly at random.

Then in Phase 2 the computer C_1 computes $0 = 15 \mod 5$ and sends 0 and 101 to C_2 .

In Phase 3, the computer C_2 computes $2 = 22 \mod 5$ and thus outputs "not equal."

The Problem

Solution via Randomization

The RCP WITNESS III

Note that in our example the output of C_2 is for all primes from PRIM(25) "not equal" except for p = 7; here we obtain "equal." Next, we analyze the communication cost of RCP WITNESS. As already stated, we have num(x), num(y) $\in [0, 2^n - 1]$. We have to send two numbers s and p which are by construction less than n^2 . For representing such numbers in binary we need

 $\lceil \log_2 n^2 \rceil \leqslant 2 \cdot \lceil \log_2 n \rceil$

many bits. Thus, we have to send at most $4 \cdot \lfloor \log_2 n \rfloor$ many bits.

What does this mean for $n = 10^{16}$? The best deterministic protocol must send 10^{16} many bits. Our RCP WITNESS works within

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Clearly, 256 many communication bits and 10¹⁶ many communication bits are incomparable in terms of the communication cost. For this *unbelievable* large saving of communication cost we pay by losing the certainty of always getting the correct result.

Thus, the remaining task is to ask the following.

Question

How large is the degree of unreliability?

To answer this question, we have to compute the so-called *error probability*.

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We say that a prime is good for (x, y) if it leads to the correct output in the RCP WITNESS.

Otherwise, we say that a prime is bad for (x, y).

In our example, 7 was bad for (01111, 10110) and all other primes in PRIM(25) were good for (01111, 10110).

Since each prime from $PRIM(n^2)$ has the same probability to be chosen, that means we have to estimate

 $\operatorname{Error}_{RCPW}(x, y) = \frac{\text{the number of bad primes for } (x, y)}{\operatorname{Prim}(n^2)}$

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So, we have to show that $\text{Error}_{RCPW}(x, y)$ is small for every instance (x, y) of our identity problem (see Figure 2).



Figure 2: The error probability.

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Solution via Randomization

One of the deepest and most important discoveries in number theory is that for all m > 67 we have

$$\operatorname{Prim}(\mathfrak{m}) > \frac{\mathfrak{m}}{\ln \mathfrak{m}} ,$$

where ln m denotes the *natural* logarithm of m. This is known as the Prime Number Theorem and it was only proved in 1896 independently by Hadamard and de la Vallée Poussin.

Therefore, for all $n \ge 9$ we know that

$$\Pr(n^2) > \frac{n^2}{2 \cdot \ln n} \,.$$

Now, it is best to make the following case distinction.

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The RCP WITNESS VIII

Case 1. x = y

Since x = y, we conclude that num(x) = num(y), and hence

 $s = num(x) \mod p = num(y) \mod p = q$

for all $p \in PRIM(n^2)$.

That is, in this case we have $\text{Error}_{RCPW}(x, y) = 0$ for all strings x and y such that x = y.

So, our RCP WITNESS can *only* make an error if $x \neq y$.

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Solution via Randomization

The RCP WITNESS IX

Case 2. $x \neq y$

Let p be a bad prime for (x, y). Then we have

 $s = num(x) \mod p = num(y) \mod p = q$

Thus, s = q and we can write

$$num(x) = h_x \cdot p + s$$

$$num(y) = h_y \cdot p + s.$$

Without loss of generality we assume $num(x) \ge num(y)$ and by subtracting the latter two equation, we obtain

$$diff(x, y) = num(x) - num(y) = (h_x - h_y) \cdot p,$$

that is, a bad prime must divide diff(x, y).

The Probler

Solution via Randomization

The RCP WITNESS X

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We know that $diff(x, y) < 2^n$ and that every prime $p_i > i$, where p_i is the ith prime number dividing diff(x, y), i = 1, ..., k. Thus, we obtain

$$\begin{aligned} \text{diff}(x,y) & \geqslant \quad p_1 \cdot p_2 \cdot \ldots \cdot p_k > 1 \cdot 2 \cdot \ldots \cdot k \\ & = \quad k! \; . \end{aligned}$$

Hence, we arrive at the condition $2^n > k!$. Finally, $n! > 2^n$ for $n \ge 4$ and thus k < n must hold.

So, we have seen that at most $k \leq n - 1$ primes from PRIM (n^2) could be bad.

This allows us to upperbound the error probability.

 $\operatorname{Error}_{RCPW}(x,y) = \frac{\text{the number of bad primes for } (x,y)}{\operatorname{Prim}(n^2)}$ $\leqslant \frac{n-1}{n^2/\ln n^2} \leqslant \frac{2 \cdot \ln n}{n}.$

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Concluding Remarks I

So, in our example the error probability to output "equal" for sequences x and y with $x \neq y$ is upper bounded by $(2 \cdot \ln n)/n$. For $n = 10^{16}$, this yields

$\frac{0.36841}{10^{14}}$,

which is amazingly small.

Even better, if we look into the future and consider even bigger databases, then the error probability will be even smaller.

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Solution via Randomization

Concluding Remarks II

The results obtained allow for a further improvement. As we have seen, if x = y then our RCP WITNESS *is correct* with certainty. All uncertainty is in case that $x \neq y$. Here we may wrongly output "equal." However, if we have found a prime $p \in PRIMES(n^2)$ *witnessing* that $x \neq y$ then the result is again certainly correct.

Now, if we chose ℓ many primes independently at random from PRIMES(n^2) instead of just one, then the probability **not** to find a witness for $x \neq y$ is



For our $n = 10^{16}$ this gives for l = 10 the upperbound



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Solution via Randomization

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Solution via Randomization

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Solution via Randomization

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Archimedes (287 - 212 bef. Chr.)

The Probler

Solution via Randomization

Concluding Remarks IV

They call it

$M \ A \ G \ I \ C$

The Problem

Solution via Randomization

Concluding Remarks IV

They call it

MAGIC

We call it

SCIENCE

Thank you!

The Magic of Probability

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