

Randomly Permuting Arrays,
More Fun with Indicator Random
Variables

CS255

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Outline

- Finishing Up The Hiring Problem
- Randomly Permuting Arrays
- More uses of Indicator Random Variables

Finishing up the Hiring Problem

- Last day we analyzed the Hire-Assistant algorithm assuming the inputs were ordered according to a random uniform permutation.
- We can use coin tosses (hence, a randomized algorithm to ensure this situation).

Randomized-Hire-Assistant(n)

1. randomly permute the list of candidates
 2. $best \leftarrow$ dummy candidate
 3. for $i \leftarrow 1$ to n
 4. do interview of candidate i
 5. if candidate i is better than $best$
 6. then $best \leftarrow i$
 7. hire candidate i
- So we need a way to generate a random permutation.

Randomly Permuting Arrays (Method 1)

- Idea:
 - start with non-permuted list: $A = \langle 1, 2, 3, 4 \rangle$.
 - Generate random priorities: $\langle 36, 3, 97, 19 \rangle$
 - Sort the elements of A according to these priorities to get $B = \langle 2, 4, 1, 3 \rangle$

- In more detail:

Permute-By-Sorting(A)

1. $n \leftarrow \text{length}[A]$
2. for $i \leftarrow 1$ to n
3. do $P[i] = \text{Random}(1, n^3)$
4. sort A , using P as sort keys
5. return A .

Analyzing Method 1

Lemma: Procedure **Permute-By-Sorting** produces a uniform random permutation of the input, assuming that the priorities are distinct.

Proof: Let $\sigma:[1..n] \rightarrow [1..n]$ be a permutation, $\sigma(i)$ being where i goes under this permutation. Let X_i to be the indicator that $A[i]$ receives the $\sigma(i)$ th smallest priority. That is, it indicates that i will be mapped correctly after sorting by priorities. So if X_i holds then after sorting the element original value i stored in $A[i]$ gets mapped to $A[\sigma(i)]$. By the definition of conditional probability, $\Pr\{Y|X\} = \Pr\{X \cap Y\}/\Pr\{X\}$, so $\Pr\{X \cap Y\} = \Pr\{X\} * \Pr\{Y|X\}$. Using this, we have $\Pr\{X_1 \cap \dots \cap X_n\} =$

$\Pr\{X_1 \cap \dots \cap X_{n-1}\} * \Pr\{X_n | X_1 \cap \dots \cap X_{n-1}\}$. Continuing to expand, we get:

$$\Pr\{X_1 \cap X_2 \cap \dots \cap X_n\} = \Pr\{X_1\} \cdot \Pr\{X_2|X_1\} \cdot \dots \cdot \Pr\{X_n|X_{n-1} \cap \dots \cap X_1\}$$

We can now fill in some of these values:

$\Pr\{X_1\} = 1/n =$ probability that one priority chosen out of n is $\sigma(1)$ th smallest.

$\Pr\{X_i|X_1 \cap \dots \cap X_{i-1}\} = 1/(n - i + 1) =$ since of the remaining elements $i, i+1, \dots, n$, each is equally likely to be the $\sigma(i)$ th smallest.

So $\Pr\{X_1 \cap \dots \cap X_n\} = 1/n * 1/(n-1) * \dots * 1/2 * 1/1 = 1/n!$.

As σ was arbitrary, any permutation is equally likely.

More on Method 1

- What do we do if the priorities aren't all distinct?
- Well, we just try again and draw a new list of priorities.
- What's the likelihood this bad situation happens?

Claim: The probability that all the priorities is unique is at least $1 - 1/n$.

Proof: Let X_i be the indicator that the i th priority was unique. Again,

$$\begin{aligned} Pr\{X_1 \cap X_2 \cap \dots \cap X_n\} &= Pr\{X_1\} \cdot Pr\{X_2|X_1\} \dots Pr\{X_n|X_{n-1} \cap \dots \cap X_1\} \\ &= \frac{n^3}{n^3} \frac{n^3 - 1}{n^3} \dots \frac{n^3 - (n - 1)}{n^3} \\ &\geq \frac{n^3 - n}{n^3} \frac{n^3 - n}{n^3} \dots \frac{n^3 - n}{n^3} \\ &= \left(1 - \frac{1}{n^2}\right)^{n-1} \geq 1 - \frac{n-1}{n^2} \text{ since } (1 - a)(1 - b) > (1 - a - b) \text{ if } a, b \text{ nonnegative} \\ &> 1 - 1/n \end{aligned}$$

Randomly Permuting Arrays (Method 2)

Randomize-In-Place(A)

1. $n \leftarrow \text{length}[A]$
2. for $i \leftarrow 1$ to n
3. do swap($A[i]$, $A[\text{Random}(i,n)]$)

Analysis of Method 2

Lemma: Just prior to the i th iteration of the for loop, for each possible $(i-1)$ -permutation, the subarray $A[1..i-1]$ contains this permutation with probability $(n-i+1)!/n!$

Proof: By induction on i .

Base case: When $i=1$, $A[1..0]$ is the empty array. It is supposed to contain a given 0-permutation with probability $(n-1+1)!/n! = n!/n! = 1$. As a 0-permutation has no elements and there is only one of them this is true. For the

Induction step: Assume just before the i th iteration, each $(i-1)$ -permutation occurs in the $A[1..i-1]$ with probability $(n-i+1)!/n!$. A particular, i -permutation $\langle x_1, \dots, x_{i-1}, x_i \rangle$ consists of an $(i-1)$ -permutation followed by x_i . By the induction hypothesis, the probability of the i -permutation is thus

$$[(n-i+1)!/n!] * \Pr\{A[i] = x_i | A[1..i-1] = \langle x_1, \dots, x_{i-1} \rangle\}.$$

The second factor is $1/(n-i+1)$ since by line 3 of Randomize-in-Place, x_i is chosen at random from $A[i..n]$. So the probability of the i -permutation is $(n-i+1)!/n! * (1/(n-i+1)) = (n-i)!/n!$ as desired.

More Analysis of Method2

Theorem: Randomize-In-Place produces a uniformly chosen random permutation.

Proof: The program could generate any n -permutation. Further it terminates just before its $(n+1)$ st iteration and thus by the lemma generates a given random n -permutation with probability:

$$(n - (n+1) + 1)!/n! = 0!/n! = 1/n! \text{ as desired.}$$

The Birthday Problem

- How many people must there be in a room before there is a 50% chance that two were born on the same day of the year?
- Let b_1, b_2, \dots, b_k be IDs for people in the room and their birthday are independent random events.
- Let n be the number of days in a year. (i.e., $n=365$). Let r be the r th day of year.
- Assume $\Pr\{\text{birthday}(b_i) = r\} = 1/n$.
- $\Pr\{\text{birthday}(b_i) = r \text{ and } \text{birthday}(b_j) = r\} = \Pr\{\text{birthday}(b_i) = r\} * \Pr\{\text{birthday}(b_j) = r\} = 1/n^2$.
- $$\begin{aligned} \Pr\{b_i = b_j\} &= \sum_{r=1}^n \Pr\{\text{birthday}(b_i) = r \text{ and } \text{birthday}(b_j) = r\} \\ &= \sum_{i=1}^n (1/n^2) \\ &= 1/n. \end{aligned}$$

More on the Birthday Problem

- To determine the odds of whether at least two out of the k people have matching birthday, we look at the complementary event: What are the odds that no-one shares a birthday?
- Let A_i indicate that for no $j < i$, do b_j and b_i have the same birthday.
- Let $B_1 = A_1$ and $B_{i+1} = A_{i+1} \cap B_i$.
- So $\Pr\{B_k\} = \Pr\{B_{k-1}\} * \Pr\{A_k | B_{k-1}\}$
 $= \Pr\{B_1\} \Pr\{A_2 | B_1\} * \dots * \Pr\{A_k | B_{k-1}\}$
 $= 1 (1 - 1/n)(1 - 2/n) \dots (1 - (k-1)/n)$

Now can use $1+x \leq e^x$ to get this is less than

$$e^{-1/n} e^{-2/n} * \dots * e^{-(k-1)/n} = e^{-(1/n)*(1+2+\dots+(k-1))} = e^{-k(k-1)/2n}$$

which is less than $1/2$ if $-k(k-1)/2n \leq -\ln 2$. Solving for k using the quadratic formula, this implies $k \geq [1 + (1 + (8 \ln 2) * n)^{1/2}] / 2$. When $n=365$, $k \geq 23$.