

# Random Permutations

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CS 165



Trees with snow on branches, "Half Dome, Apple Orchard, Yosemite," 1933. Ansel Adams. U.S. government image. U.S. National Archives and Records Administration.

# Generating Random Permutations

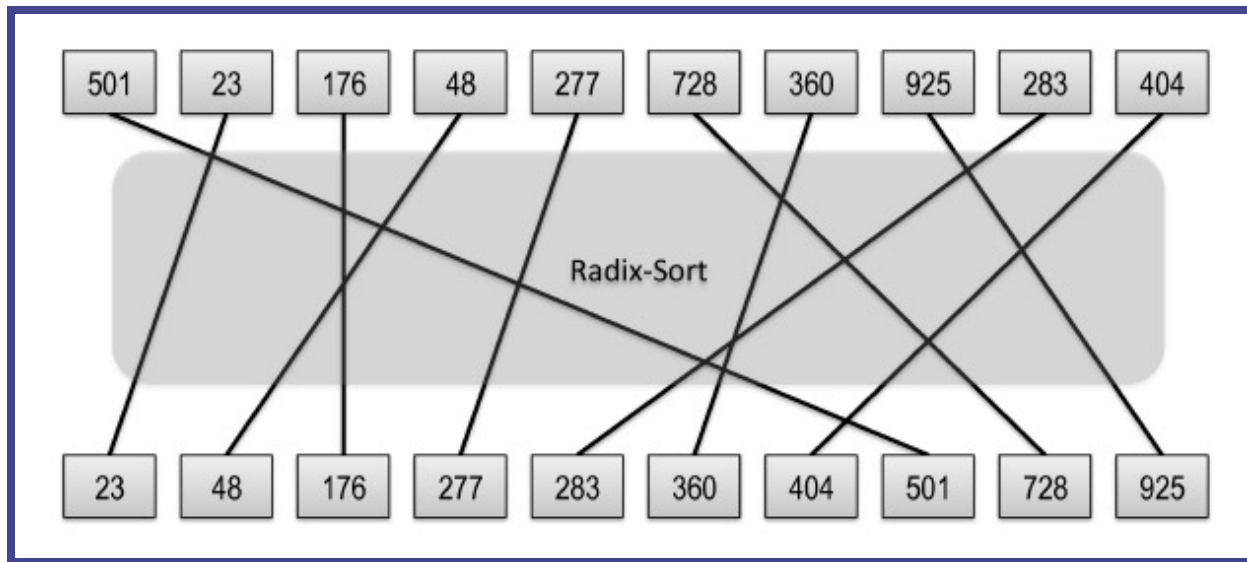
- The input to the random permutation problem is a list,  $X = (x_1, x_2, \dots, x_n)$ , of  $n$  elements, which could stand for playing cards or any other objects we want to randomly permute.
- The output is a reordering of the elements of  $X$ , done in a way so that all permutations of  $X$  are equally likely.
- We can use a function, **random**( $k$ ), which returns an integer in the range  $[0, k - 1]$  chosen uniformly and independently at random.

# Applications: Simple Algorithms and Card Games

- ❑ A **randomized algorithm** is an algorithm whose behavior depends, in part, on the outcomes of random choices or the values of random bits.
- ❑ The main advantage of using randomization in algorithm design is that the results are often simple and efficient.
- ❑ In addition, there are some problems that need randomization for them to work effectively.
- ❑ For instance, consider the problem common in computer games involving playing cards—that of randomly shuffling a deck of cards so that all possible orderings are equally likely.

# Algorithm 1: Random Sort

- This algorithm simply chooses a random number for each element in  $X$  and sorts the elements using these values as keys.



# Analysis of Random-Sort

- To see that every permutation is equally likely to be output by the random-sort method, note that each element,  $x_i$ , in  $X$  has an equal probability,  $1/n$ , of having its random  $r_i$  value be the smallest.
- Thus, each element in  $X$  has equal probability of  $1/n$  of being the first element in the permutation.
- Applying this reasoning recursively, implies that the permutation that is output has the following probability of being chosen:

$$\left(\frac{1}{n}\right) \cdot \left(\frac{1}{n-1}\right) \cdots \left(\frac{1}{2}\right) \cdot \left(\frac{1}{1}\right) = \frac{1}{n!}$$

- That is, each permutation is equally likely to be output.
- There is a small probability that this algorithm will fail, however, if the random values are not unique.

# Fisher-Yates Shuffling

- There is a different algorithm, known as the Fisher-Yates algorithm, which always succeeds.

**Algorithm** FisherYates( $X$ ):

*Input:* An array,  $X$ , of  $n$  elements, indexed from position 0 to  $n - 1$

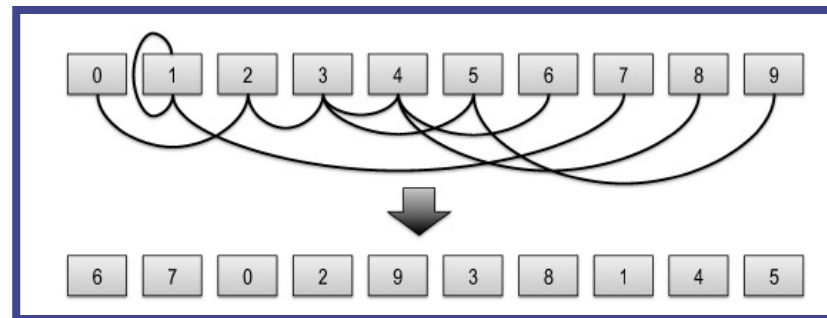
*Output:* A permutation of  $X$  so that all permutations are equally likely

**for**  $k = n - 1$  **downto** 1 **do**

    Let  $j \leftarrow \text{random}(k + 1)$       //  $j$  is a random integer in  $[0, k]$

    Swap  $X[k]$  and  $X[j]$       // This may “swap”  $X[k]$  with itself, if  $j = k$

**return**  $X$



# Analysis of Fisher-Yates

- This algorithm considers the items in the array one at time from the end and swaps each element with an element in the array from that point to the beginning.
- Notice that each element has an equal probability, of  $1/n$ , of being chosen as the last element in the array  $X$  (including the element that starts out in that position).
- Applying this analysis recursively, we see that the output permutation has probability

$$\left(\frac{1}{n}\right) \cdot \left(\frac{1}{n-1}\right) \cdots \left(\frac{1}{2}\right) \cdot \left(\frac{1}{1}\right) = \frac{1}{n!}$$

- That is, each permutation is equally likely.