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Write a function for doing an in-place shuffle of a vector.

The shuffle must be "uniform," meaning each item in the original vector must have the same probability of ending up in each spot in the final vector.

Assume that you have a function `getRandom(floor, ceiling)` for getting a random integer that is \geq floor and \leq ceiling.

Gotchas

A common first idea is to walk through the vector and swap each element with a random other element. Like so:

```
size_t getRandom(size_t floor, size_t ceiling)
{
    static random_device rdev;
    static default_random_engine generator(rdev());
    static uniform_real_distribution<double> distribution(0.0, 1.0);
    double value = distribution(generator);
    return static_cast<size_t>(value * (ceiling - floor + 1)) + floor;
}

void naiveShuffle(vector<int>& theVector)
{
    // for each index in the array
    for (size_t firstIndex = 0; firstIndex < theVector.size(); ++firstIndex) {

        // grab a random other index
        size_t secondIndex = getRandom(0, theVector.size() - 1);


        // and swap the values
        if (secondIndex != firstIndex) {
            swap(theVector[firstIndex], theVector[secondIndex]);
        }
    }
}
```

However, this does not give a uniform random distribution.

Why? We could calculate the exact probabilities of two outcomes to show they aren't the same. But the math gets a little messy. Instead, think of it this way:

Suppose our vector had 3 elements: `[a, b, c]`. This means it'll make 3 calls to `getRandom(0, 2)`. That's 3 random choices, each with 3 possibilities. So our total number of possible sets of choices is $3 * 3 * 3 = 27$. Each of these 27 sets of choices is equally probable.

But how many possible *outcomes* do we have? If you paid attention in stats class you might know the answer is $3!$, which is 6. Or you can just list them by hand and count:



- a, b, c
- a, c, b
- b, a, c
- b, c, a
- c, b, a
- c, a, b

But our function has 27 equally-probable sets of choices. 27 is not evenly divisible by 6. So some of our 6 possible *outcomes* will be achievable with more sets of choices than others.

We can do this in a single pass. $O(n)$ time and $O(1)$ space.

A common mistake is to have a mostly-uniform shuffle where an item is less likely to stay where it started than it is to end up in any given slot. Each item should have the same probability of ending up in each spot, including the spot where it starts.

Breakdown

It helps to start by ignoring the in-place requirement, then adapt the approach to work in place.

Also, the name "shuffle" can be slightly misleading—the point is to arrive at a random ordering of the items from the original vector. Don't fixate too much on preconceived notions of how you would "shuffle" e.g. a deck of cards.

How might we do this by hand?

We can simply choose a random item to be the first item in the resulting vector, then choose another random item (from the items remaining) to be the second item in the resulting vector, etc.

Assuming these choices were in fact random, this would give us a uniform shuffle. To prove it rigorously, we can show any given item a has the same probability ($\frac{1}{n}$) of ending up in any given spot.

First, some stats review: to get the probability of an outcome, you need to *multiply the probabilities of all the steps required for that outcome*. Like so:

Outcome	Steps	Probability
item #1 is a	a is picked first	$\frac{1}{n}$
item #2 is a	a not picked first, a picked second	$\frac{(n-1)}{n} * \frac{1}{(n-1)} = \frac{1}{n}$
item #3 is a	a not picked first, a not picked second, a picked third	$\frac{(n-1)}{n} * \frac{(n-2)}{(n-1)} * \frac{1}{(n-2)} = \frac{1}{n}$
item #4 is a	a not picked first, a not picked second, a not picked third, a picked fourth	$\frac{(n-1)}{n} * \frac{(n-2)}{(n-1)} * \frac{(n-3)}{(n-2)} * \frac{1}{(n-3)} = \frac{1}{n}$

So, how do we implement this in code?

If we didn't have the "in-place" requirement, we could allocate a new vector, then one-by-one take a random item from the input vector, remove it, put it in the first position in the new vector, and keep going until the input vector is empty (well, probably a *copy* of the input vector—best not to destroy the input)

How can we adapt this to be in place?

What if we make our new "random" vector simply be the *front* of our input vector?

Solution

We choose a random item to move to the first index, then we choose a random *other* item to move to the second index, etc. We "place" an item in an index by swapping it with the item currently at that index.

Crucially, once an item is placed at an index it can't be moved. So for the first index, we choose from n items, for the second index we choose from $n - 1$ items, etc.

```
size_t getRandom(size_t floor, size_t ceiling)
{
    static random_device rdev;
    static default_random_engine generator(rdev());
    static uniform_real_distribution<double> distribution(0.0, 1.0);
    double value = distribution(generator);
    return static_cast<size_t>(value * (ceiling - floor + 1)) + floor;
}

void shuffle(vector<int>& theVector)
{
    // if it's 1 or 0 items, just return
    if (theVector.size() <= 1) {
        return;
    }

    // walk through from beginning to end
    for (size_t indexWeAreChoosingFor = 0;
        indexWeAreChoosingFor < theVector.size() - 1;
        ++indexWeAreChoosingFor) {

        // choose a random not-yet-placed item to place there
        // (could also be the item currently in that spot)
        // must be an item AFTER the current item, because the stuff
        // before has all already been placed
        size_t randomChoiceIndex = getRandom(indexWeAreChoosingFor,
            theVector.size() - 1);

        // place our random choice in the spot by swapping
        if (randomChoiceIndex != indexWeAreChoosingFor) {
```

```
        swap(theVector[indexWeAreChoosingFor], theVector[randomChoiceIndex]);  
    }  
}  
}
```

This is a semi-famous algorithm known as the **Fisher-Yates shuffle** (sometimes called the Knuth shuffle).

Complexity

$O(n)$ time and $O(1)$ space.

What We Learned

Don't worry, most interviewers won't expect a candidate to know the Fisher-Yates shuffle algorithm. Instead, they'll be looking for the problem-solving skills to *derive* the algorithm, perhaps with a couple hints along the way.

They may also be looking for an understanding of why the naive solution is non-uniform (some outcomes are more likely than others). If you had trouble with that part, try walking through it again.

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