

Counting

The next major topic of the course is probability theory. Suppose you toss a fair coin a thousand times. How likely is it that you get exactly 500 heads? And what about 1000 heads? It turns out that the chances of 500 heads are roughly 2.5%, whereas the chances of 1000 heads are so infinitesimally small that we may as well say that it is impossible. But before you can learn to compute or estimate odds or probabilities you must learn to count! That is the subject of this note.

1 Counting Sequences

We start by considering a simple scenario. We pick $k \leq n$ elements from an n -element set $S = \{1, 2, \dots, n\}$ one at a time while removing the sampled element from S . This type of sampling is called *sampling without replacement*. We wish to count the number of different ways to do this, taking into account the order in which the elements are picked. For example, when $k = 2$, picking 1 and then 2 is considered a different outcome from picking 2 followed by 1. Another way to ask the question is this: we wish to form an ordered sequence of k distinct elements, where each element is picked from the set S . How many different such ordered sequences are there?

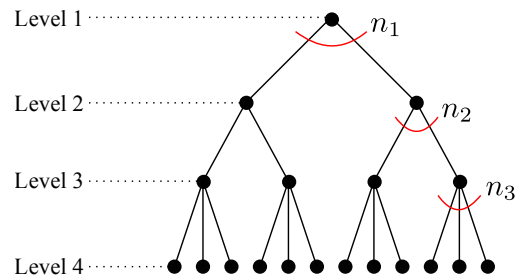
If we were dealing cards, the set would be $S = \{1, \dots, 52\}$, where each number represents a card in a deck of 52 cards. Picking an element of S in this case refers to dealing one card. Note that once a card is dealt, it is no longer in the deck and so it cannot be dealt again. So the hand of k cards that are dealt consists of k distinct elements from the set S .

For the first card, it is easy to see that we have 52 distinct choices. But now the available choices for the second card depend upon what card we picked first. The crucial observation is that regardless of which card we picked first, there are exactly 51 choices for the second card. So the total number of ways of choosing the first two cards is 52×51 . Reasoning in the same way, there are exactly 50 choices for the third card, regardless of our choices for the first two cards. It follows that there are exactly $52 \times 51 \times 50$ sequences of three cards. In general, the number of sequences of k cards is $52 \times 51 \times \dots \times [52 - (k - 1)]$.

This is an example of the First Rule of Counting:

First Rule of Counting: If an object can be made by a succession of k choices, where there are n_1 ways of making the first choice, and *for every* way of making the first choice there are n_2 ways of making the second choice, and *for every* way of making the first and second choice there are n_3 ways of making the third choice, and so on up to the n_k -th choice, then the total number of distinct objects that can be made in this way is the product $n_1 \times n_2 \times n_3 \times \dots \times n_k$.

Here is another way of picturing the First Rule of Counting. Consider the following tree:



It has branching factor n_1 at the root, n_2 at every vertex at the second level, \dots , n_k at every vertex at the k -th level. Each path from the root to a vertex at level $k + 1$ (i.e., a leaf vertex) represents one possible way of making the object by making a succession of k choices. So the number of distinct objects that can be made is equal to the number of leaves in the tree. Moreover, the number of leaves in the tree is the product $n_1 \times n_2 \times n_3 \times \dots \times n_k$. For example, if $n_1 = 2$, $n_2 = 2$, and $n_3 = 3$, then there are 12 leaves (i.e., outcomes).

2 Counting Sets

Consider a slightly different question. We would like to pick k distinct elements of $S = \{1, 2, \dots, n\}$ (i.e., without repetition), but we do not care about the order in which we picked the k elements. For example, picking elements $1, \dots, k$ is considered the same outcome as picking elements $2, \dots, k$ and picking 1 as the last (k -th element). Now how many ways are there to choose these elements to obtain the same outcome?

When dealing a hand of cards, say a poker hand, it is often more natural to count the number of distinct hands (i.e., the set of 5 cards dealt in the hand), rather than the order in which they were dealt. As we have seen in the section above, if we are considering order, there are $52 \cdot 51 \cdot 50 \cdot 49 \cdot 48 = \frac{52!}{47!}$ outcomes. But how many distinct hands of 5 cards are there? Here is another way of asking the question: each such 5 card hand is just a subset of S of cardinality 5. So we are asking how many 5 element subsets of S are there?

Here is a clever trick for counting the number of distinct subsets of S with exactly 5 elements. Create a bin corresponding to each such 5 element subset. Now take all the sequences of 5 cards and distribute them into these bins in the natural way: each sequence gets placed in the bin corresponding to the set of 5 elements in the sequence. Thus, if the sequence is $(2, 1, 3, 5, 4)$, then it is placed in the bin labeled $\{1, 2, 3, 4, 5\}$. How many sequences are placed in each bin? The answer is exactly $5!$, since there are exactly $5!$ different ways to order 5 cards.

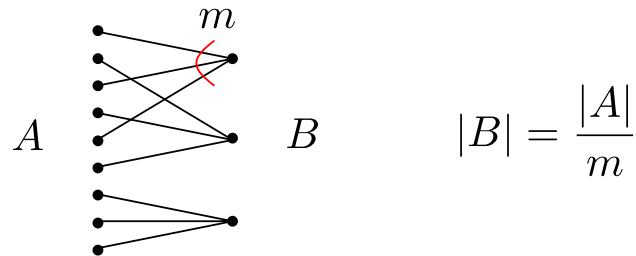
Recall that our goal was to compute the number of 5 element subsets, which now corresponds to the number of bins. We know that there are $\frac{52!}{47!}$ distinct 5-card sequences, and there are $5!$ sequences placed in each bin. The total number of bins is therefore $\frac{52!}{47!5!}$.

This quantity $\frac{n!}{(n-k)!k!}$ is used so often that there is special notation for it: $\binom{n}{k}$, pronounced “ n choose k ,” and it’s known as the *binomial coefficient*. This is the number of ways of picking k distinct elements from S , where the order of placement does not matter. Equivalently, it’s the number of ways of choosing k objects out of a total of n distinct objects, where the order of the choices does not matter.

The trick we used above is actually our Second Rule of Counting:

Second Rule of Counting: Assume an object is made by a succession of choices, and the order in which the choices are made does not matter. Let A be the set of ordered objects and let B be the set of unordered objects. If there exists an m -to-1 function f from A to B , we can count the number of ordered objects (pretending that the order matters) and divide by m (the number of ordered objects per unordered objects) to obtain $|B|$, the number of unordered objects.

Note that we are assuming that the number of ordered objects is the same for every unordered object; the rule cannot be applied otherwise. Here is another way of picturing the Second Rule of Counting:



The function f simply places the ordered outcomes into bins corresponding to the unordered outcomes. In our poker hand example, f will map $5!$ elements in the domain of the function (the set of ordered 5 card outcomes) to one element in the range (the set of 5-element subsets of S). The number of elements in the range of the function is therefore $\frac{52!}{47!5!}$.

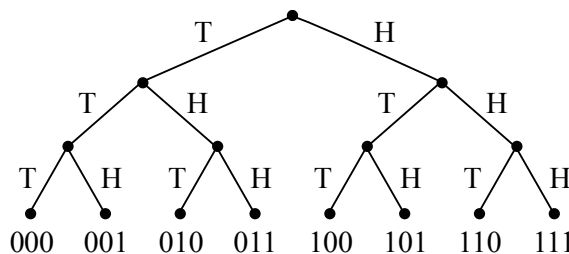
3 Sampling with Replacement

Sometimes we wish to consider a different scenario of sampling where after we choose an element from $S = \{1, 2, \dots, n\}$, we throw it back into S so that we can choose it again. Assume that we are still picking k elements out of S one at a time and that order matters. How many distinct sequences of k elements can we obtain under this new sampling scheme? We can use the First Rule of Counting. Since we have n choices in each trial, $n_1 = n_2 = \dots = n_k = n$. Then we have a grand total of n^k sequences.

This type of sampling is called *sampling with replacement*; multiple trials can have the same outcome. Card dealing is a type of *sampling without replacement*, since two trials cannot both have the same outcome (one card cannot be dealt twice).

3.1 Coin Tosses

Let us return to coin tosses. How many different outcomes are there if we toss a coin k times? By an outcome, we mean a length- k sequence consisting of heads and tails. Suppose we represent each outcome by a binary string of length k where 0 corresponds to a tail (T) and 1 a head (H). For example, the string 001 would represent an outcome of 2 tails followed by a head. The following tree illustrates the experiment for $k = 3$, with the leaves in 1-to-1 correspondence with the set of possible outcomes:



Here $S = \{0, 1\}$. This is a case of sampling with replacement; multiple coin flips could result in tails (we could pick the element 0 from the set S in multiple trials). Order also matters; e.g., strings 001 and 010 are considered different outcomes. By the reasoning above (using the First Rule of Counting) we have a total of $n^k = 2^k$ distinct outcomes, since $n = 2$.

3.2 Rolling Dice

Let's say we roll two dice, so $k = 2$ and $S = \{1, 2, 3, 4, 5, 6\}$. How many possible outcomes are there? In this setting, ordering matters; obtaining 1 with the first die and 2 with the second is different from obtaining 2 with the first and 1 with the second. We are sampling with replacement, since we can obtain the same result on both dice.

The setting is therefore the same as the coin flipping example above (order matters and we are sampling with replacement), so we can use the First Rule of Counting in the same manner. The number of distinct outcomes is therefore $n^2 = 6^2 = 36$.

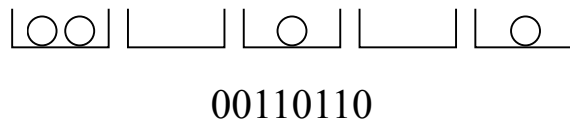
3.3 Sampling with Replacement, but where Order Does Not Matter

Say you have unlimited quantities of apples, bananas and oranges. You want to select 5 pieces of fruit to make a fruit salad. How many ways are there to do this? In this example, $S = \{1, 2, 3\}$, where 1 represents apples, 2 represents bananas, and 3 represents oranges. $k = 5$ since we wish to select 5 pieces of fruit. Ordering does not matter; selecting an apple followed by a banana will lead to the same salad as a banana followed by an apple.

This scenario is trickier to analyze. It is natural to apply the Second Rule of Counting because order does not matter. So we first pretend that order matters, and then the number of ordered objects is 3^5 as discussed above. How many ordered options are there for every unordered option? The problem is that this number differs depending on which unordered object we are considering. Let's say the unordered object is an outcome with 5 bananas. There is only one such ordered outcome. But if we are considering 4 bananas and 1 apple, there are 5 such ordered outcomes (represented as 12222, 21222, 22122, 22212, 22221).

Now that we see the Second Rule of Counting will not help, can we look at this problem in a different way? Let us first generalize back to our original setting: we have a set $S = \{1, 2, \dots, n\}$ and we would like to know how many ways there are to choose multisets (sets with repetition) of size k . Remarkably, we can model this problem in terms of binary strings.

Assume we have 1 bin for each element of S , so n bins in total. For example, if we selected 2 apples and 1 banana, bin 1 would have 2 elements and bin 2 would have 1 element. In order to count the number of multisets, we need to count how many different ways there are to fill these bins with k elements. We don't care about the order of the bins themselves, just how many of the k elements each bin contains. Let's represent each of the k elements by a 0 in the binary string, and separations between bins by a 1. Consider the following picture:



This would be a sample placement where $S = \{1, \dots, 5\}$ and $k = 4$. Counting the number of multisets is now equivalent to counting the number of placements of the k 0's. We have just reduced what seemed like a very complex problem to a question about a binary string, simply by looking at it from a different perspective!

How many ways can we choose these locations? The length of our binary string is $k + n - 1$, and we are choosing which k locations should contain 0's. The remaining $n - 1$ locations will contain 1's. Once we pick a location for one zero, we cannot pick it again; repetition is not allowed. Picking location 1 followed by location 2 is the same as picking location 2 followed by location 1, so ordering does not matter. It follows that all we wish to compute is the number of ways of picking k elements from $k + n - 1$ elements, without

replacement and where the order of placement does not matter. This is given by $\binom{n+k-1}{k}$, as discussed in the Counting Sets section above. This is therefore the number of ways in which we can choose multisets of size k from set S .

Returning to our example above, the number of ways of picking 5 pieces of fruit is exactly $\binom{3+5-1}{5} = \binom{7}{5}$.

Notice that we started with a problem which seemed very different from previous examples, but, by viewing it from a certain perspective, we were able to use previous techniques (those used in counting sets) to find a solution! This is key to many combinatorial arguments, as we will explore further in the next section.

4 Combinatorial Proofs

Combinatorial arguments are interesting because they rely on intuitive counting arguments rather than algebraic manipulation. We can prove complex facts, such as $\binom{n}{k+1} = \binom{n-1}{k} + \binom{n-2}{k} + \dots + \binom{k}{k}$. You can directly verify this identity by algebraic manipulation. But you can also do this by interpreting what each side means as a combinatorial process. The left hand side is just the number of ways of choosing a subset X with $k+1$ elements from a set $S = \{1, \dots, n\}$. Let us think about a different process that results in the choice of a subset with $k+1$ elements. We start by picking the lowest-numbered element of X . If we picked 1, to finish constructing X , we must now choose k elements out of the remaining $n-1$ elements of S greater than 1; this can be done in $\binom{n-1}{k}$ ways. If instead the lowest-numbered element we picked is 2, then we have to choose k elements from the remaining $n-2$ elements of S greater than 2, which can be done in $\binom{n-2}{k}$ ways. Moreover all these subsets are distinct from those where the lowest-numbered element was 1. So we should add the number of ways of choosing each to the grand total. Proceeding in this way, we split up the process into cases according to the first (i.e., lowest-numbered) object we select, to obtain:

$$\text{First element selected is } \left\{ \begin{array}{ll} \text{element 1,} & \text{leading to } \binom{n-1}{k} \text{ remaining ways} \\ \text{element 2,} & \text{leading to } \binom{n-2}{k} \text{ remaining ways} \\ \text{element 3,} & \text{leading to } \binom{n-3}{k} \text{ remaining ways} \\ \vdots & \\ \text{element } (n-k), & \text{leading to } \binom{k}{k} \text{ remaining ways} \end{array} \right.$$

(Note that the lowest-numbered object we select cannot be higher than $n-k$ as we have to select $k+1$ distinct objects.)

The last combinatorial proof we will do is the following: $\binom{n}{0} + \binom{n}{1} + \dots + \binom{n}{n} = 2^n$. To see this, imagine that we have a set S with n elements. On the left hand side, the i -th term counts the number of ways of choosing a subset of S of size exactly i ; so the sum on the left hand side counts the total number of subsets (of any size) of S .

We claim that the right hand side (2^n) does indeed also count the total number of subsets. To see this, just identify a subset with an n -bit vector, where in each position j we put a 1 if the j -th element is in the subset, and a 0 otherwise. So the number of subsets is equal to the number of n -bit vectors, which is 2^n (there are 2 options for each bit). Let us look at an example, where $S = \{1, 2, 3\}$ (so $n = 3$). Enumerate all $2^3 = 8$ possible subsets of S : $\emptyset, \{1\}, \{2\}, \{3\}, \{1, 2\}, \{1, 3\}, \{2, 3\}, \{1, 2, 3\}$. The term $\binom{3}{0}$ counts the number of ways to choose a subset of S with 0 elements; there is only one such subset, namely the empty set $\emptyset = \{\}$. There are $\binom{3}{1} = 3$ ways of choosing a subset with 1 element, $\binom{3}{2} = 3$ ways of choosing a subset with 2 elements, and $\binom{3}{3} = 1$ way of choosing a subset with 3 elements (namely, the subset consisting of every element of S). Summing, we get $1 + 3 + 3 + 1 = 8$, as expected.