Counting is the task of finding the number of elements (also called the cardinality) of a given set. When the set is small, we can count its elements “by hand”. When sets are larger counting can be trickier. We’ll see some useful ways for counting the number of elements of larger sets.

Say you have a six-sided die and a two-sided coin — it comes out heads (H) or tails (T). What is the number of possible outcomes when both the die and the coin are tossed? There are 6 possible outcomes for the die and 2 for the coin, so the total number of outcomes is $6 \times 2 = 12$.

It will be useful to describe this kind of problem using the language of sets. The set $S$ of possible outcomes of the die is $S = \{1, 2, 3, 4, 5, 6\}$, so $|S| = 6$. The set $T$ of possible outcomes of the coin is $T = \{\text{H}, \text{T}\}$, so $|T| = 2$. The set of possible outcomes of the die and the coin is the product set $S \times T$:

$$S \times T = \{(1, \text{H}), (1, \text{T}), (2, \text{H}), (2, \text{T}), \ldots, (6, \text{H}), (6, \text{T})\}.$$ 

The number of elements of $S \times T$ is $|S| \cdot |T| = 6 \cdot 2 = 12$.

In general, given any two finite sets $S$ and $T$ the product set $S \times T$ consists of all ordered pairs of elements $(s, t)$ such that $s$ is in $S$ and $t$ is in $T$:

$$S \times T = \{(s, t) : s \in S \text{ and } t \in T\}.$$ 

The number of elements of $S \times T$ is the product of the number of elements of $S$ and the number of elements of $T$, i.e., $|S \times T| = |S| \cdot |T|$.

Let’s do another example: Let $R$ and $B$ be the sets of outcomes of a toss of a red and a blue six-sided die, respectively. Then $R = \{1, 2, 3, 4, 5, 6\}$ and $B = \{1, 2, 3, 4, 5, 6\}$. When both dies are tossed, the set of outcomes is $R \times B = \{(1, 1), (1, 2), \ldots, (6, 6)\}$ and the number of outcomes is $|R \times B| = |R| \cdot |B| = 36$.

In cases like this when $S = T$ we can denote the set $S \times T$ by $S^2$. (Remember that this is the power of a set, not the power of a number). The set $S^2$ has $|S|^2$ elements.

We can also take the product of more than two sets. The set $S_1 \times \cdots \times S_n$ (where $S_1, \ldots, S_n$ are finite sets) consists of all sequences $\mathbf{s} = (s_1, \ldots, s_n)$ where $s_i$ is in $S_i$ for all $i$ between 1 and $n$:

$$S_1 \times \cdots \times S_n = \{(s_1, \ldots, s_n) : s_1 \in S_1 \text{ and } \ldots \text{ and } s_n \in S_n\}.$$ 

The set $S_1 \times \cdots \times S_n$ has $|S_1| \cdots |S_n|$ elements.

When $S_1 = \cdots = S_n = S$, we write $S^n$ for $S_1 \times \cdots \times S_n$. This set has $|S|^n$ elements.

For example, the set of outcomes when 9 different six-sided dies are tossed is $\{1, 2, 3, 4, 5, 6\}^9$. This set has $6^9$ elements, so there are $6^9$ possible outcomes.

\[\text{1The order of elements in a sequence matters and there can be repetitions: For example, (1,1,2), (2,1,1), and (1,2,2) are all different sequences.}\]
1 Functions, bijections, and counting

One technique to count the number of elements of a set $S$ is to come up with a “nice” correspondence between a set $S$ and another set $T$ whose cardinality we already know.

Let’s count the number of subsets of the set $S_n = \{1, 2, 3, \ldots, n\}$.

For example, when $n = 1$, $S_1 = \{1\}$ are there are two subsets: $\emptyset$ and $\{1\}$. When $n = 2$, $S_2 = \{1, 2\}$ and there are four subsets: $\emptyset$, $\{1\}$, $\{2\}$, and $\{1, 2\}$.

I suspect that most of you can come up with a proof by induction that shows the set $S_n$ has $2^n$ elements. Instead of using induction, let me show you another way.

The set $\{0, 1\}^n$ consists of all possible $n$-bit sequences. This is a product set, so it has $2^n$ elements. We will show how to represent subsets of $S_n$ by elements of $\{0, 1\}^n$ in a unique way: Each subset of $S_n$ is represented by an $n$ bit sequence, and each $n$ bit sequence represents some subset. So their number must be the same.

Here is how the representation works: The subset $\{s_1, \ldots, s_k\}$ of $S_n$ is represented by the bit sequence that has ones in positions $s_1, \ldots, s_k$ and zeros in all the other positions. For example, if $n = 7$, the subset $\{3, 4, 6\}$ of $S_n$ is represented by the bit sequence $(0, 0, 1, 1, 0, 1, 0)$ in $\{0, 1\}^7$.

Clearly every subset of $S_n$ is represented by some $n$ bit sequence. It is also true that every $n$ bit sequence represents a subset: The sequence $(b_1, \ldots, b_n)$ represents the set of all $i$ between 1 and $n$ such that $b_i = 1$. For example, the sequence $(0, 0, 1, 1, 0, 1, 1)$ represents the set $\{3, 4, 6, 7\}$. We showed a “one-to-one” correspondence between the subsets of $S_n$ and the elements of $\{0, 1\}^n$ ($n$-bit sequences), so their number must be the same.

As this is our first argument of this type, let us explicitly write out the correspondence between the elements of $S_2$ and the sequences in $\{0, 1\}^2$:

$\emptyset \leftrightarrow (0, 0)$ \quad $\{1\} \leftrightarrow (1, 0)$ \quad $\{2\} \leftrightarrow (0, 1)$ \quad $\{1, 2\} \leftrightarrow (1, 1)$.

It is useful to have a language to describe these correspondences between sets. For this we need to talk about functions.

**Functions** A function $f$ from a set $X$ to a set $Y$ associates to every element $x$ in $X$ an element $f(x)$ in $Y$. For example, $f(0) = 0$, $f(1) = 1$, $f(2) = 0$, $f(3) = 3$, $f(4) = 0$ is a function from set $X = \{0, 1, 2, 3, 4\}$ to the set $Y = \{0, 1, 2, 3\}$.

One way to specify a function is to list the values on all elements $x$ in $X$:

<table>
<thead>
<tr>
<th>$x$:</th>
<th>0</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
</tr>
</thead>
<tbody>
<tr>
<td>$f(x)$:</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>3</td>
<td>0</td>
</tr>
</tbody>
</table>

or by means of a bipartite graph $G_f$ whose vertices\footnote{We abuse notation here and use the same label for vertices in $X$ and vertices in $Y$, but we think of them as different.} are partitioned into $X$ and $Y$ and whose edges are those $\{x, y\}$ such that $f(x) = y$: 
A more convenient way is to describe the function using logic. This can be done in many ways, for example by giving a formula for calculating $f$:

$$f(x) = \begin{cases} 
  x, & \text{if } x \text{ is odd} \\
  0, & \text{if } x \text{ is even}
\end{cases} \quad x \in \{0, 1, 2, 3, 4\}$$

an algorithm for calculating $f$:

Function $f(x)$, where $x \in \{0, 1, 2, 3, 4\}$:

Let $b = x \mod 2$.

Output $b \cdot x$.

or, say, by giving a recurrence:

$$f(x) = (x \mod 2) \cdot (f(x - 2) + 2) \text{ for } x \geq 2, \quad f(0) = 0, f(1) = 1.$$

No matter which way you describe a function, you must make sure that $f(x)$ is well-defined for every $x$ in $X$: This value is specified and it is specified uniquely. For example, here is a bad way to define a function:

$$f(x) = y \quad \text{where } x \in \{0, 1, 2, 3, 4\} \text{ and } y^2 = x.$$ 

this specification is consistent with both $f(1) = 1$ and $f(1) = -1$, so it does not uniquely describe the value $f(1)$.

We write $f: X \to Y$ for “a function from set $X$ to set $Y$”.

A function $f: X \to Y$ is an injective if distinct elements in $x$ are mapped to distinct elements in $Y$. More precisely, $f$ is injective if for every pair of elements $x$ and $x'$ in $X$ such that $x \neq x'$, we have $f(x) \neq f(x')$. The above function is not injective because $0 \neq 2$ but $f(0) = f(2)$.

A function $f: X \to Y$ is surjective if every element $y$ in $Y$ is mapped to by some $x$ in $X$. More precisely, $f$ is surjective if for every $y$ in $Y$ there exists $x$ in $X$ such that $f(x) = y$. The above function is not surjective because $2 \in Y$ but nothing maps to it.

Let’s assume $X$ and $Y$ are finite sets. When $f$ is injective, the edges of the graph $G_f$ are a matching and they match all the vertices of $X$, so $Y$ must be at least as large as $X$, that is $|Y| \geq |X|$. When $f$ is surjective, then every element of $Y$ can be matched to some element of $X$ (by taking an arbitrary edge incident to it), so $Y$ can be at most as large as $X$, that is $|Y| \leq |X|$.

We say $f$ is bijective if it is injective and surjective. When $X, Y$ are finite and $f$ is bijective, the edges of $G_f$ form a perfect matching between $X$ and $Y$, so $|X| = |Y|$.
This is why bijective functions are useful for counting: If we know $|X|$ and can come up with a bijective $f: X \rightarrow Y$, then we immediately get that $|Y| = |X|$. The tricky part is coming up with $f$ and showing that it is bijective — namely, both injective and surjective.

Let’s rework our previous example in this language. Let $X = \{0, 1\}^n$ and $Y$ be the set of all subsets of $\{1, \ldots, n\}$. Let $f: X \rightarrow Y$ be the function that on input $(b_1, \ldots, b_n)$ outputs the set of indices $i$ such that $b_i = 1$, namely $f((b_1, \ldots, b_n)) = \{i: b_i = 1\}$.

**Theorem 1.** $f$ is a bijective function.

*Proof.* First we prove $f$ is injective: Different bit sequences map to different sets. Suppose two bit sequences $x$ and $x'$ are different. Then they must differ in one of their positions, say position $i$. Then the element $i$ is in one of the sets $f(x), f(x')$ but not in the other, so $f(x) \neq f(x')$.

Now we show $f$ is surjective. Given any subset $S$ of $Y$, we’ll show there exists a bit sequence $x$ such that $f(x) = S$. Let $x$ be the $n$-bit sequence that has a 1 in position $i$ if $i$ is in $S$ and 0 if $i$ is not in $S$. Then $f(x) = S$. □

**Corollary 2.** The number of subsets of $\{1, \ldots, n\}$ is $2^n$.

*Proof.* By Theorem 1, there is a bijection from elements of $\{0, 1\}^n$ to subsets of $\{1, \ldots, n\}$. Therefore the number of subsets of $\{1, \ldots, n\}$ equals the number of elements of $\{0, 1\}^n$, which is $2^n$. □

2 The sum rule

The sum rule says that if $A_1, \ldots, A_n$ are disjoint sets then

$$|A_1 \cup \cdots \cup A_n| = |A_1| + \cdots + |A_n|.$$  

This rule is useful when the set whose number of elements we want to count can be written as a disjoint union of simpler sets.

For example, if you have 10 red balls, 7 blue balls, and 4 red balls, then the total number of balls you have is $10 + 7 + 4 = 21$. You could have done this in first grade. So why do we need sets and unions? Sets help us break complicated counting problems into simpler ones.

Suppose you need to choose a password between 6 and 8 symbols out of which the first symbol must be a letter (lowercase or uppercase) and the remaining ones must be letters or digits. How many possible passwords are there?

Let $P$ be the set of possible passwords, $L$ be the set of letters, and $S$ be the set of symbols (letters or digits). The English alphabet has 26 letters, each of which can be lowercase or uppercase, so $|L| = 52$. The set $S$ contains all the letters plus the ten digits, so $|S| = 62$.

To count the number of passwords $|P|$, it makes sense to “decompose” the set $P$ in terms of the simpler sets $L$ and $S$. Here is how we do it. First, $P$ is a disjoint union of six, seven, and eight
letter passwords — let’s denote these sets by $P_6$, $P_7$, and $P_8$. Therefore

$$|P| = |P_6| + |P_7| + |P_8|.$$ 

The set $P_6$ is a product set: It consists of all sequences of a letter followed by five symbols, so $P_6 = L \times S \times S \times S \times S \times S = L \times S^5$. Similarly, $P_7 = L \times S^6$ and $P_8 = L \times S^7$. Therefore

$$|P| = |L \times S^5| + |L \times S^6| + |L \times S^7|$$

$$= |L| \cdot |S|^5 + |L| \cdot |S|^6 + |L| \cdot |S|^7$$

$$= 52 \cdot 62^5 + 52 \cdot 62^6 + 52 \cdot 62^7$$

$$\approx 1.8 \cdot 10^{14}.$$ 

3 The general product rule and permutations

You toss a blue six-sided die and a red six-sided die. How many outcomes are there in which the face values of the two dice come out different?

The set of possible outcomes is not a product set, but we can still reason like this: There are 6 possibilities for the face value of the first die. For each one of these six possibilities, we are interested in the outcomes in which the second die has a different face value. No matter what the toss of the first die came out to be, there are always five possible choices for the toss of the second die that make their face values different. Therefore the total number of outcomes is $6 \times 5 = 30$.

This is an instance of the general product rule: Given a set $S$ of sequences of length $k$, where

- There are $n_1$ possible first entries,
- There are $n_2$ possible second entries for each first entry
- There are $n_3$ possible third entries for each combination of first and second entries, and so on up to $n_k$

the size of $S$ is $n_1 \cdot n_2 \cdot \cdots \cdot n_k$.

A permutation of a set $S$ is a sequence that contains every element of $S$ exactly once. For example, the permutations of the set $\{1, 2, 3\}$ are the sequences $(1, 2, 3)$, $(1, 3, 2)$, $(2, 1, 3)$, $(2, 3, 1)$, $(3, 1, 2)$, $(3, 2, 1)$.

How many permutations does an $n$ element set have? The first entry in the permutation sequence can be chosen in $n$ possible ways. Each such choice leaves out $n - 1$ possibilities for the second element. Each combination of choices for the first two elements leaves out $n - 2$ combinations for the third element. Continuing like this, we get that the number of permutations of an $n$ element set is

$$n \cdot (n - 1) \cdot (n - 2) \cdots 1 = n!.$$ 

Let’s do another example of the product rule. In how many ways can you place three different pieces on an $8 \times 8$ chessboard — a bishop, a knight, and a pawn — so that no two pieces share a
row or a column? The position of the three pieces is specified by a six numbers \((br, bc, kr, kc, pr, pc)\). The first number \(br\) indicates the bishop’s row, the second number \(bc\) indicates the bishop’s column, and so on.

We can count the number of allowed sequences with the generalized product rule. There are 8 possibilities for the bishop’s row \(br\). For each such choice, there are 8 possibilities for the bishop’s column \(bc\). Once the bishop’s position was chosen, there are 7 choices for the knight’s row \(kr\) (any one but \(br\)) and (for each of them) 7 choices from \(kc\) (any one but \(bc\)). Once the bishop and the knight were positioned, there are 6 choices for \(pr\) and (for each of them) 6 choices of \(pc\). The total number of configurations is \(8 \cdot 8 \cdot 7 \cdot 7 \cdot 6 \cdot 6 = (8 \cdot 7 \cdot 6)^2\).

4 The pigeonhole principle

The pigeonhole principle says that if you toss more pigeons into fewer holes, at least two pigeons will land in the same hole. If \(X\) is the set of pigeons, \(Y\) is the set of holes, and \(f : X \rightarrow Y\) is the function that tells you which pigeon goes into which hole, we get the following mathematical statement:

**Theorem 3** (The pigeonhole principle). For any two finite sets \(X\) and \(Y\) such that \(|X| > |Y|\) and any function \(f : X \rightarrow Y\), there exist inputs \(x \neq x'\) such that \(f(x) = f(x')\).

**Proof.** We prove the contrapositive. If for all inputs \(x \neq x'\), \(f(x) \neq f(x')\), then \(f\) is injective so \(|Y| \geq |X|\). \(\square\)

The pigeonhole principle quickly answers questions that may otherwise look impossibly difficult. Let’s do a few examples.

**Example 1.** In a room of 400 people, there must be two that have the same birthday.

Here, \(X\) is the set of people in the room, \(Y\) is the set of 365 days in the year, and \(f(x) = y\) if person \(x\) was born on day \(y\) of the year. By the pigeonhole principle, there must be a pair of people \(x \neq x'\) such that \(f(x) = f(x')\), that is they are born on the same day of the year.

**Example 2.** Among any 10 points in a unit \((1 \times 1)\) square, there must be a pair that are within distance \(\sqrt{2}/3 \approx 0.471\).

Divide the unit square evenly into nine \(\frac{1}{3} \times \frac{1}{3}\) blocks. Let \(X\) be the set of ten points, \(Y\) be the set of nine blocks, and \(f(x) = y\) if point \(x\) falls inside block \(y\). By the pigeonhole principle, at least two points must fall into the same block. The diameter of this block is \(\frac{1}{3}\sqrt{2}\), so the distance between the two points within the block cannot be larger than this number.

Here is an example of a possible configuration. The two vertices connected by a line fall within the same block, so they must be within a distance of \(\frac{1}{3}\sqrt{2}\).
Example 3. You have a set $A$ of 8 integers, each of them between 0 and 30. There are two different subsets of $S$ so that the sum of the numbers in each subset is the same.

Let $X$ be the set of all subsets of $A$, $Y$ be the set $\{0, 1, \ldots, 240\}$, and $f: X \rightarrow Y$ be the function that on input $S$, outputs the sum of all integers in $S$. Since there are 8 numbers between 0 and 30, this sum is always a number between 0 and $8 \cdot 30 = 240$.

The number of subsets of $A$ is $2^8 = 256$. On the other hand, $|Y| = 241$. By the pigeonhole principle, two different subsets $S_1, S_2$ of $A$ must have the same sum.

We can even make the stronger conclusion that there exists two disjoint subsets of $S$ with the same sum: If $S_1$ and $S_2$ are not disjoint, take away the elements in their intersection from both of them. This changes the value of both sums by the same amount, so the sums stay equal but the sets are now disjoint.

There is a more general variant of the pigeonhole principle that is sometimes useful.

**Theorem 4** (Generalized pigeonhole principle). For every integer $k$ every pair of sets $X, Y$ such that $|X| > k|Y|$ and every function $f: X \rightarrow Y$, there exist $k + 1$ distinct elements $x_1, \ldots, x_{k+1}$ of $X$ such that $f(x_1) = \cdots = f(x_{k+1})$.

**Proof.** We prove the contrapositive. Suppose that for every $y \in Y$, there are at most $k$ distinct elements of $x$ that map to it. Order the elements of $X$ in some way from smallest to largest. Let $f': X \rightarrow Y \times \{1, \ldots, k\}$ be the function such that $f'(x) = (y, i)$ if $x$ is the $i$-th smallest element of $X$ among those that $f(x) = y$. Then the function $f$ is injective, so $|Y \times \{1, \ldots, k\}| \geq |X|$. By the product rule, $|Y \times \{1, \ldots, k\}| = k|Y|$ and so $|Y| \geq k|X|$.

**Example 4.** In a group of 1500 people, there must be three people of the same gender that have the same birthday.

Since $1500 > 4 \cdot 365$, by the generalized pigeonhole principle there must be at least five people in the group that have the same birthday. Let $X$ be these five people, and $f: X \rightarrow \{\text{male, female}\}$ be the map that assigns each person their gender. By the generalized pigeonhole principle, there are at least three people among the five that have the same gender.

Here is another solution. Let $X$ be the group of 1500 people, $Y$ be the product set $\{1, \ldots, 365\} \times \{\text{male, female}\}$, and $g: X \rightarrow Y$ be the function that assigns to each person $x$ in the group their birthday and their gender. Since $|Y| = 365 \cdot 2 = 730$ and $1500 > 2 \cdot 730$, by the generalized pigeonhole principle there are three people in the group with the same birthday and the same gender.
5 Comparison based sorting

In Lecture 8 we showed that the Merge Sort procedure makes $n \log n - n + 1$ pairwise comparisons when sorting any sequence of length $n$ (assuming $n$ is a power of two). Is there a different procedure that performs fewer comparisons?

We will argue that any procedure for sorting a sequence of numbers that only relies on comparing pairs of numbers (and not on other things, like the values of these numbers) cannot do substantially better. No matter what the procedure does, there exists at least one sequence of length $n$ on which the procedure performs $\Omega(n \log n)$ comparisons.

How can we even reason about a sorting procedure without knowing how it works? The main difficulty here is to come up with a good model of sorting procedures. Once we do that, the calculations won’t be hard. Instead of sorting, let’s first talk about a simpler but related example — the game of twenty questions.

Twenty questions The game of twenty questions involves two players, Alice and Bob. Alice thinks of an integer between 0 and $N - 1$ and writes it on a piece of paper. Bob and Alice then alternate in asking and answering questions about this number. Bob can ask any question as long as it has a yes/no answer and Alice always answers Bob’s questions truthfully. How many questions does Bob need to ask in order to find out Alice’s secret?

Here is a simple strategy for Bob, assuming $N$ is a power of two, say $N = 2^q$. Represent each integer between 0 and $N - 1$ uniquely by its base 2 expansion $(b_1, \ldots, b_q)\footnote{The binary-to-integer conversion $f((b_1, \ldots, b_q)) = b_1 + 2b_2 + \cdots + 2^{q-1}b_{q-1}$ is a bijective function from $\{0,1\}^n$ to the set $\{0, \ldots, N - 1\}$, so each number $x \in \{1, \ldots, N\}$ is uniquely represented by a bit sequence $(b_1, \ldots, b_q)$ such that $f((b_1, \ldots, b_q)) = x$.}$ For each $i$ ranging from 1 to $q$, ask the question “Is $b_i = 1$”? Alice’s answer to this question determines the value of $b_i$ (if yes $b_i = 1$, if no $b_i = 0$) so after $q$ questions Bob finds the binary representation of Alice’s secret. With this strategy, Bob asks $q = \log N$ questions.

Let us now argue that no matter which strategy Bob uses, there is at least one possible secret for Alice that makes Bob ask at least $\log N$ questions. Assume, for contradiction, that Bob can guess Alice’s secret with certainty after asking $q < \log N$ questions. If we represent the answers to Bob’s questions by a $\{\text{yes}, \text{no}\}$ sequence of length $q$, then the number of possible answer sequences is $|\{\text{yes, no}\}^q| = 2^q$. As there are $N$ possible secrets for Alice and $N > 2^q$, by the pigeonhole principle at least two of Alice’s possible secrets would result in the same sequence of answers to Bob’s questions. After asking $q$ questions and seeing these answers, Bob will not know which of these two is Alice’s real secret so he cannot guess it with certainty.

We model the execution of a comparison-based sorting procedure as a “twenty questions” game between the input sequence (Alice) and the sorting procedure (Bob). The procedure only asks questions of the type “is entry $i$ in the sequence smaller than entry $j$?” which have yes/no answers.

Let’s do a small example. Suppose we have some sorting procedure — which we call Bob — that was given the sequence $(a_1, a_2, a_3)$ to sort. We’ll assume $a_1, a_2, a_3$ are all distinct. Initially, Bob
does not know anything about this sequence so he asks a question like “is \( a_1 < a_2 \)?” Say the answer to this question is “yes”. Then he asks for example “Is \( a_2 < a_3 \)?” If the answer to this question is also “yes”, then Bob has enough information to determine the correct order of the sequence, namely \( a_1 < a_2 < a_3 \).

Now suppose to the first question “Is \( a_1 < a_2 \)” is “yes” but the answer to “Is \( a_2 < a_3 \)” is “no”. There are two possible orderings of the sequence that are consistent with these answers: \( a_1 < a_2 < a_3 \) and \( a_2 < a_1 < a_3 \). Bob cannot sort the sequence just based on this information.

Let’s fix \( a_1, \ldots, a_n \) to be any \( n \) distinct values to sort and let \( S \) be the set of all permutations of \( \{ a_1, \ldots, a_n \} \). Bob could be given any of these permutations as an input sequence. The answers to his first \( q \) comparison questions are \{yes, no\} sequences of length \( q \). If \( |S| > 2^q \), then by the pigeonhole principle at least two input sequences lead to the same sequence of answers. Bob cannot distinguish which one of these two is his input, and so he cannot sort it properly.

We conclude that if \( S \) has strictly more than \( 2^q \) elements, then not all sequences of \( n \) elements can be sorted after \( q \) comparisons. The set \( S \) consists of the permutations of \( \{ a_1, \ldots, a_n \} \), so it has size \( n! \). Sorting with \( q \) comparisons is impossible as long as \( n! > 2^q \); in other words sorting requires \( \log n! \) comparisons.

What does the number \( \log_2 n! \) look like? We can write

\[
\log n! = \frac{\ln n!}{\ln 2} = \frac{1}{\ln 2} \ln(1 \cdot 2 \cdots n) = \frac{1}{\ln 2} (\ln 1 + \ln 2 + \cdots + \ln n).
\]

This sum is perfect for the integral method.

Lower bounding the sum by the related integral minus the error term gives

\[
\ln 1 + \ln 2 + \cdots + \ln n \geq \int_1^{n+1} \ln x \, dx - \ln(n + 1)
\]

The antiderivative of \( \ln x \) is \( x \ln x - x \), so

\[
\ln 1 + \ln 2 + \cdots + \ln n \geq (n + 1) \ln(n + 1) - (n + 1) - \ln(n + 1).
\]

Therefore any procedure for comparison-based sorting requires at least

\[
(n + 1) \log(n + 1) - \frac{1}{\ln 2} (n + 1) - \log(n + 1) = \Omega(n \log n)
\]

comparisons on at least one input sequence.
References

This lecture is based on Chapters 7 and 11 of the text *Mathematics for Computer Science* by E. Lehman, T. Leighton, and A. Meyer.