

4/5 Permutations and Combinations § 6.2

Def'n A permutation of n distinct elements x_1, x_2, \dots, x_n is an ordering of the elements, i.e., a list of the elements where each x_i appears exactly once.

E.g. There are 6 permutations of A, B, C :

ABC ACB BAC BCA CAB CBA

Also makes sense to define \Rightarrow Recall that for a positive integer $n \geq 1$, we defined n factorial as $n! = n \times (n-1) \times (n-2) \times \dots \times 3 \times 2 \times 1$

$0! = 1$

Theorem The # of permutations of n elements is $n!$

Pf: Create a permutation by choosing 1st element in 1st, then 2nd, ..., down to n^{th} . There are n choices for 1st.

Then there are $(n-1)$ choices for 2nd (since 1st is not available) $(n-2)$ choices for 3rd, etc., down to 1 choice for n^{th} .

By mult. principle, gives $n \times (n-1) \times (n-2) \times \dots \times 1 = n!$ total. \square

We can also do a slightly more general thing:

Def'n An r -permutation of x_1, \dots, x_n is a length r list of elements in x_1, \dots, x_n where each appears at most once. (We need $r \leq n$ for such a list to exist.)

E.g. There are 12 2-permutations of A, B, C, D :

AB AC AD BA BC BD CA CB CD DA DB DC.

We use $P(n, r) := \#$ of r -permutations of n elt. set.

Thm $P(n, r) = n \times (n-1) \times \dots \times (n-r+1) = \frac{n!}{(n-r)!}$

Pf: Same as proof for usual permutations,

but stop after the r^{th} step. \square

We often want to count unordered collections of given size.
Def'n An r-combination of x_1, \dots, x_n is a length r unordered collection of elements in x_1, \dots, x_n ,
 i.e., a size r subset of $\{x_1, \dots, x_n\}$.

E.g. There are 6 2-combinations of A, B, C, D:
 $\{A, B\}$ $\{A, C\}$ $\{A, D\}$ $\{B, C\}$ $\{B, D\}$ $\{C, D\}$

Let $C(n, r) = \#$ r-combinations of n element set
 (common notation $\binom{n}{r}$ - read 'n choose r' - used too...)

How to give a formula for $C(n, r)$?

We can create an r-permutation of x_1, \dots, x_n as follows:

1. Pick one of the $C(n, r)$ r-combinations, call it $\{y_1, \dots, y_r\} \subseteq \{x_1, \dots, x_n\}$
2. Choose one of the $r!$ permutations of y_1, \dots, y_r .

E.g. To make a 2-permutation of A, B, C, D, we first pick one of the 6 2-combinations, and then choose one of the $2! = 2$ ways to permute its letters:

$\{A, B\}$	$\{A, C\}$	$\{A, D\}$	$\{B, C\}$	$\{B, D\}$	$\{C, D\}$
↙ ↘	↙ ↘	↙ ↘	↙ ↘	↙ ↘	↙ ↘
AB BA	AC CA	AD DA	Bc CB	BD DB	CD DC

By the multiplication principle, this means

of ways to make r-permutation of x_1, \dots, x_n = # ways to make r-combination of x_1, \dots, x_n x # of permutations of r things,

i.e. $P(n, r) = C(n, r) \times r!$

\Rightarrow Theorem $C(n, r) = \frac{P(n, r)}{r!} = \frac{n!}{(n-r)! r!}$

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E.g. We saw there were 6 2-combinations of A, B, C, D,
and $C(4, 2) = \frac{4!}{2!2!} = \frac{4 \times 3 \times 2 \times 1}{2 \times 1 \times 2 \times 1} = 6$. ✓

We will have a lot more to say about these $C(n, r)$
in a little bit. Here is just a taste:

Exercise Show $\sum_{r=0}^n C(n, r) = 2^n$.

Hint: Imagine choosing an arbitrary subset of $\{1, 2, \dots, n\}$
by first choosing its size r .

E.g. A standard deck of cards has 52 cards in it:

- there are 4 suits: spades ♠, hearts ♥, clubs ♣, diamonds ♦
- there are 13 ranks: 2-10 and J Q K A

for a total of $4 \times 13 = 52$ different cards.

A poker hand consists of 5 of these 52 cards.

- Q: 1) How many poker hands are there?
2) How many poker hands have cards of
all the same suit (this is called a "flush")?

A: 1) Since a poker hand is an unordered subset of size 5
from 52 elements, there are $C(52, 5) = 2,598,960$
different poker hands

2) To make a flush, first pick the suit of all the cards,
then select 5 of the 13 ranks for the hand

$$\Rightarrow 4 \times C(13, 5) = 4 \times 1287 = 5,148 \text{ flushes.}$$

This means $\approx 0.2\%$ of hands are flushes (very rare!)

4/14 Generalized Permutations §6.3

There are $n!$ permutations of n distinct letters:

ABC ACB BAC BCA CAB CBA

But what about rearrangements of a word with repeated letters?

E.g. How many ways are there to rearrange the letters in MISSISSIPPI ?

Some of the $11!$ permutations will "be the same", so the answer is something less than $11!$.

Let's start with something easier:

how to count rearrangements of AAA BBB BB.

A rearrangement is 8 letters, 3 of them A's, 5 B's:

— — — — —
 ↑ ↑ ↑

Of the 8 positions for letters, we can select any 3 for A's, and then the 5 B's must go in the other positions:

B B A B A B A B

We are choosing 3 spots out of 8, which gives $C(8, 3) = 8! / (3! \cdot 5!) = 56$ total rearrangements.

For MISSISSIPPI, we can do similarly, but in more steps.

We have 11 spots, choose 4 of them for the I's:

— I — I I — I — — — — — $C(11, 4)$

Then from remaining 7 spots, choose 4 for the S's:

— S S S S — — — — — $C(7, 4)$

Then from remaining 3 spots, choose 2 for the P's:

P P I S S S — — — — — $C(3, 2)$

The M goes in remaining spot in $C(1, 1)$ ways.

All together, there are $C(11, 4) \cdot C(7, 4) \cdot C(3, 2) \cdot C(1, 1)$

$$= \frac{11!}{4! 7!} \cdot \frac{7!}{4! 3!} \cdot \frac{3!}{2! 1!} \cdot \frac{1!}{1! 0!} = \frac{11!}{4! 4! 2! 1!}$$

= 34,650 rearrangements of MISSISSIPPI.

Theorem For a word which has m different kinds of letters, with n_1 of the 1st letter, n_2 of the 2nd letter, ... and n_m of the m^{th} letter, so $n = n_1 + n_2 + \dots + n_m$ total letters, the # of rearrangements = $n! / (n_1! \cdot n_2! \cdot n_3! \dots \cdot n_m!)$.

Pf: Same as what we just explained! ▣

Eg. MISSISSIPPI $\Rightarrow n=11$, $n_1=4$ I's, $n_2=4$ P's, $n_3=2$ S's, $n_4=1$ M
So that # rearrangements = $11! / (4! 4! 2! 1!)$.

Notice that if all letters are distinct, so $n_1 = n_2 = \dots = n_m = 1$, we get usual $n! / (1! \cdot 1! \dots 1!) = n!$ permutations, and the more repeated letters we have, the more we have to divide $n!$ by to account for the repeat.

In fact, another way to think about the formula:

if we put subscripts (or 'colors') on repeated letters, like:

$M_1 I_1 S_1 S_2 I_2 S_3 S_4 I_3 P_1 P_2 I_4$

then all these letters become 'distinct', so that there are $n!$ ($=11!$) different permutations of the subscripted letters. And then...

Given any rearrangement (without subscripts), there are $n_1! n_2! \dots n_m!$ ($=4! 4! 2! 1!$) ways to put subscripts on all the repeated letters.

So dividing $n!$ by $n_1! n_2! \dots n_m!$ gives us

the number of ways of rearranging the letters (similar to how dividing $P(n, r)$ by $r!$ gave $C(n, r)$). ▣
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4/17 Generalized combinations § 6.3

Last class we saw how to deal with repeats in permutations. What about combinations where we allow repeats?

E.g.: At a bagel shop they have four flavors of bagels: plain, sesame, everything, & cinnamon raisin

You want to buy 13 bagels (= a baker's dozen).

How many ways are there to do this?

If we had to pick 13 distinct flavors of bagels, this would be a $C(n, k)$ combinations problem.

But of course we can repeat flavors in our purchase.

There is a very nice trick for these kinds of problems called "stars and bars", where we represent a bagel purchase by a picture that looks like this:

* * * | * * | * * * * * | * * *
plain sesame everything cinnamon raisin

This means that we buy 3 plain, 2 sesame, 5 everything, and 3 cinnamon raisin bagels.

Any pattern of 13 *'s ('stars') and 3 |'s ('bars') gives us a bagel purchase; the *'s represent the bagels, with the |'s serving as 'dividers' between bins representing the 4 flavors.

So to count bagel purchases, we just need to count patterns of 13 *'s and 3 |'s.

But this is exactly the word rearrangement problem, where we saw the answer is:

$$C(16, 13) = \frac{16!}{13! 3!} = \underline{560}.$$

In general, we have the following formula for counting combinations with repeats allowed:

Theorem The number of ways to select k things from m options, allowing selecting an option multiple times is $C(k+m-1, k) = C(k+m-1, m-1) = \frac{(k+m-1)!}{(m-1)! k!}$.

(Notice that we always have $C(n, k) = C(n, n-k)$.)

E.g. You have 11 identical candies to give to 3 children. How many different ways can you distribute the candies?

Idea: represent a candy distribution like:

4 candies		2 candies		5 candies
* * * *		* *		* * * * *
1 st child		2 nd child		3 rd child

This 'stars and bars' trick shows it is the same as the bagel problem, and so there are $C(11+2, 11) = \frac{13!}{11! \cdot 2!} = \underline{78}$ ways to give candies.

Q: What if we are required to give each child at least one candy?

A: First give each child one candy. This leaves $(11-3) = 8$ candies which can be distributed arbitrarily in $C(8+2, 8) = \frac{10!}{8! \cdot 2!} = \underline{45}$ ways.

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4/19 Binomial coefficients and the Binomial Theorem

We start with an algebra exercise:

$$\begin{aligned}(a+b)^n &= (a+b)(a+b)(a+b) \\ &= a a a + a a b + a b a + a b b \\ &\quad + b a a + b a b + b b a + b b b \\ &= \underline{a^3} + \underline{3} a^2 b + \underline{3} a b^2 + \underline{b^3}\end{aligned}$$

where a and b can be any numbers (or variables)

What's the significance of this sequence 1, 3, 3, 1?

If we expanded:

$$(a+b)^4 = \dots = a^4 + 4a^3b + 6a^2b^2 + 4ab^3 + b^4$$

We'd get the coefficient sequence 1, 4, 6, 4, 1.

And in general we have ...

Theorem (Binomial Theorem)

$$(a+b)^n = \sum_{k=0}^n C(n, k) a^{n-k} b^k.$$

Pf. Imagine expanding $(a+b)^n$:

$$\underbrace{(a+b)}_1 \underbrace{(a+b)}_2 \dots \underbrace{(a+b)}_n \leftarrow \begin{array}{l} n \text{ terms} \\ \text{in total} \end{array}$$

If we want to make a term of $a^{n-k}b^k$ from these multiplications, we have to choose the "b" part from exactly k of the $(a+b)$'s and the "a" part from the $n-k$ other $(a+b)$'s.

Thus, the number of ways to do this is the # of ways to choose k positions from n , which by definition is $C(n, k) = \frac{n!}{k!(n-k)!}$. \square

Note! In this context, also use notation $\binom{n}{k} = C(n, k)$ for the "n choose k" numbers: $\sum_{k=0}^n \binom{n}{k} a^{n-k} b^k = (a+b)^n$

The $\binom{n}{k}$ are also called binomial coefficients.

Using the binomial theorem, we can give short proofs of some identities we've already seen, like:

Theorem $\sum_{k=0}^n C(n, k) = 2^n$

Pf: Bin. Thm. says $\sum_{k=0}^n C(n, k) = (a+b)^n$

Let $a=1$ and $b=1 \Rightarrow \sum_{k=0}^n C(n, k) 1^{n-k} 1^k = (1+1)^n$

$\Rightarrow \sum_{k=0}^n C(n, k) = 2^n$ \square

What about the alternating sum of the $C(n, k)$'s?

Ex. $C(3, 0) - C(3, 1) + C(3, 2) - C(3, 3)$

$= 1 - 3 + 3 - 1 = 0$

or $C(4, 0) - C(4, 1) + C(4, 2) - C(4, 3) + C(4, 4)$

$= 1 - 4 + 6 - 4 + 1 = 0$

Theorem For $n \geq 1$, $\sum_{k=0}^n (-1)^k C(n, k) = 0$.

Proof: Let $b=-1$ and $a=1$ in the Binomial Theorem:

$\sum_{k=0}^n (-1)^k C(n, k) = (1-1)^n = 0^n = 0$ \square

NOTE: $C(0, 0) = \frac{0!}{0!0!} = 1$, so for $n=0$ we have

$\sum_{k=0}^n (-1)^k C(n, k) = C(0, 0) = 1$,

which means we should interpret 0^0 as 1.

The HW has other identities for the $C(n, k)$ which can be proved using the Binomial Theorem...

4/21 Pascal's Triangle §6.7

The Binomial Theorem $(x+y)^n = \sum_{k=0}^n C(n,k) x^k y^{n-k}$
 suggests that we should look at the sequence
 $C(n,0), C(n,1), C(n,2), \dots, C(n,n)$ in a "row"
 Actually, we can put all these rows together
 into an infinite triangular array:

$$\begin{array}{cccc} & & C(0,0) & \\ & & C(1,0) & C(1,1) \\ & & C(2,0) & C(2,1) & C(2,2) \\ & & C(3,0) & C(3,1) & C(3,2) & C(3,3) \\ & & & & & \vdots \end{array}$$

Notice how we put each row a half step to the left
 of the row above it, so the "centers" are the same.
 Filling in the values of these $C(n,k)$ gives:

$$\begin{array}{ccccccc} & & & & 1 & & \\ & & & & 1 & & 1 \\ & & & 1 & 2 & 1 & \\ & & 1 & 3 & 3 & 1 & \\ & 1 & 4 & 6 & 4 & 1 & \\ & 1 & 5 & 10 & 10 & 5 & 1 \\ - & 1 & 6 & 15 & 20 & 15 & 6 & 1 \\ & & & & & & & \vdots \end{array}$$

This array of $C(n,k)$'s is called Pascal's triangle.
 Many of the results about binomial coefficients
 we've seen before are visible in Pascal's triangle:

- $\sum_{k=0}^n C(n,k) = 2^n$ means sum of n^{th} row
 of Pascal's triangle is 2^n
- $C(n,k) = C(n,n-k)$ means Pascal's triangle is
symmetric across vertical axis.

Pattern of even vs. odd entries also very interesting:
 See the HW problem about this...

The following recurrence for $C(n, k)$ is very useful:

Theorem (Pascal's Identity)

$$C(n+1, k) = C(n, k) + C(n, k-1) \text{ for all } 1 \leq k \leq n.$$

Note: This means each entry in Pascal's triangle is the sum of the two entries above it:

e.g.

$$\begin{array}{cccccc} 1 & 5 & 10 & 10 & 5 & 1 \\ & 6 & \nearrow 15 & \searrow 20 & 15 & 6 & 1 \end{array}$$

Together with $C(n, 0) = C(n, n) = 1$ on outside this lets us repeatedly fill in all of the triangle.

Pf of Pascal's Identity: Let's do a combinatorial proof.

$C(n+1, k)$ is the # of size k subsets of $\{1, 2, \dots, n+1\}$.

Let's show that $C(n, k) + C(n, k-1)$ is also this #.

Let S be size k subset of $\{1, 2, \dots, n+1\}$.

If $n+1 \notin S$, then S is also a size k subset of $\{1, 2, \dots, n\}$, which are counted by $C(n, k)$.

If $n+1 \in S$, then $S \setminus \{n+1\}$ is a size $(k-1)$ subset of $\{1, 2, \dots, n\}$, counted by $C(n, k-1)$.

So there is a bijective correspondence between size k subsets of $\{1, 2, \dots, n+1\}$ (counted by $C(n+1, k)$)

and size k or $(k-1)$ subsets of $\{1, 2, \dots, n\}$

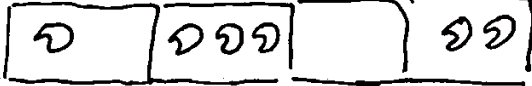
(counted by $C(n, k) + C(n, k-1)$ by addition principle).

Thus $C(n+1, k) = C(n, k) + C(n, k-1)$, as claimed. \square

4/24 The Pigeonhole Principle §6.8

So far we've considered the problem of counting the number of discrete objects satisfying certain conditions. But sometimes we just want to show at least one exists. The Pigeonhole Principle is good for this:

Theorem If you put n pigeons into k holes, and $k < n$, then at least one hole has at least 2 pigeons.

E.g.:  ← 6 pigeons in 4 holes
⇒ at least one hole has at least two pigeons

The trick when using the pigeonhole principle is to figure out what should be the "pigeons" and what the "holes."

E.g.: If there are at least 367 people in a room, then there must be at least two people who share a birthday ("twins").

Here the "holes" are the calendar dates, and the "pigeons" are the people.

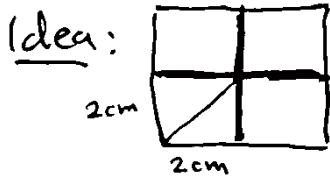
There are only 366 different dates (remember: Feb. 29) so with 367 people there must be a "collision" of birthdays.

Notice: The Pigeonhole Principle is "non-constructive": it doesn't tell us which people share a birthday or which birthday is shared...

Also, doesn't necessarily reflect typical behavior:

E.g.: with only 23 people, >50% chance of shared birthday, and with 50 people, >97% chance!

E.g. Show that if you put 5 dots on a $4\text{ cm} \times 4\text{ cm}$ square, at least two dots are within 3 cm of each other.



← Break $4\text{ cm} \times 4\text{ cm}$ square into four $2\text{ cm} \times 2\text{ cm}$ sub-squares.

Then, by Pigeonhole Principle, at least two of the 5 dots are in same sub-square.

And the maximum distance of two points in a $2\text{ cm} \times 2\text{ cm}$ square is the length of the diagonal $= 2 \cdot \sqrt{2}\text{ cm} \approx 2 \cdot 1.4\text{ cm} < 3\text{ cm}$. ✓

Let's show a more sophisticated example related to divisibility of integers:

Two integers are coprime if they have no common factor (# that divides them) bigger than 1.

E.g. 2 and 6 are not coprime since both divisible by 2.
9 and 15 are not coprime since both divisible by 3.
But 4 and 15 are coprime since they have no common factor.

Theorem If S is a subset of $\{1, 2, 3, \dots, 20\}$ of size ≥ 11 , then there are two numbers a and b in S such that a and b are coprime.

Note: Not true for S of size = 10 since

$\{2, 4, 6, 8, 10, 12, 14, 16, 18, 20\}$

has all #'s with two as a factor (even #'s), so no two #'s in S are coprime.

Proof: We first need the following lemma:

Lemma For any positive integer n , the numbers n and $n+1$ are coprime.

Proof: Suppose $r > 1$ is a factor (divisor) of n .

Then $n+1 \equiv 1 \pmod{r}$, meaning the remainder when dividing $n+1$ by r is 1 . So $n+1$ is not divisible by r . Thus n and $n+1$ have no common factors. \square

Next, we use the pigeonhole principle:

Let the "holes" be pairs of consecutive #'s:

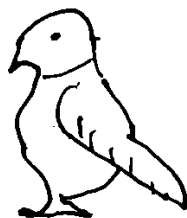
$\{1, 2\}, \{3, 4\}, \{5, 6\}, \dots, \{19, 20\}$

These are 10 holes. So if the subset S has size at least 11, it has two #'s in the same hole.

By the previous lemma, those #'s are coprime. \square

As you can see from these examples, even though the statement of the pigeonhole principle is very simple, figuring out how to apply it to a given problem can require a lot of creativity, and it can lead to unexpected results!

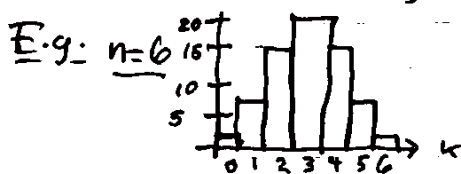
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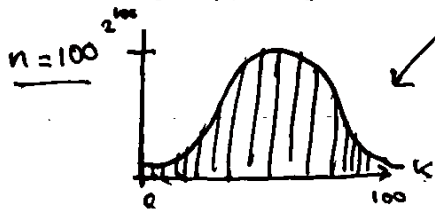
4/26 Further Topics in Discrete Math

We finished the material from the textbook. If you liked this class, here are further topics you could learn about:

Discrete probability Theory: We talked a little about probabilities with poker hands. One of the most important results in probability is visible in Pascal's triangle: Consider plotting n^{th} row as a histogram:



$C(n, k)$ for $k=0$ to n :
1, 6, 15, 20, 15, 6, 1



for big values of n ,
the histogram will approach a smooth shape called the "Bell curve" or "Gaussian Curve"

This shape tells you how many heads you can expect to see if you flip a fair coin n times. It is a "universal shape" in probability, statistics, and the sciences.

Generating functions: For a sequence of combinatorial numbers, its generating function is a way of recording the sequence in a polynomial or power series:

We have already seen a very important example of a generating function with the Binomial Theorem:

$$(1+x)^n = \sum_{k=0}^n C(n, k) x^k$$

We keep track of the #'s $C(n, k)$ in the polynomial $(1+x)^n$.

For an ∞ sequence of #'s, we get a power series instead.
Recall the Fibonacci numbers 1, 1, 2, 3, 5, 8, ...
defined by $F_1 = 1$, $F_2 = 1$, and $F_n = F_{n-1} + F_{n-2}$ for $n > 2$.

$$\text{Then: } \sum_{n=0}^{\infty} F_n x^n = \frac{x}{1-x-x^2} \quad \leftarrow \text{think "Taylor series" from calculus.}$$

This perspective is very powerful in that we can apply techniques from algebra and calculus to understand combinatorial problems:
e.g: radius of convergence is related to growth rate of coefficients.

Graph theory: Graphs consist of vertices (dots \bullet) and edges ($\overset{u}{\curvearrowright} \overset{v}{\curvearrowleft}$) between the vertices.

They are pictures like this:



We have already used graphs to represent functions and relations, but graphs are very versatile structures that can model all kinds of things; e.g. social networks.

There is a lot that can be said about both the typical and extremal structure of graphs!

Thank you all for being excellent students this semester! If you ever want to talk more about math, don't hesitate to send me an email or knock on my office door...