Logic and Proofs

1.1 Propositional Logic

Definition: proposition

A proposition is a declarative statement that is either true or false, but not both.

Example. The following are propositions:

- (1) Columbus is the capital of Ohio. (True)
- (2) 2 + 2 = 5. (False)
- (3) A circle with radius r has area πr^2 . (True)
- (4) The number $\sqrt{2}$ is rational. (False)
- (5) Every square is a rectangle. (True)
- (6) Every rectangle is a square. (False)

We often refer to statements with letters (typically, p, q, r, s, ...). The truth value of a proposition is denoted T (if true) or F (if false). Our goal will be to study how the truth value of propositions varies when combined with logical operators and other propositions to form compound propositions.

Definition: negation

Let p be a proposition. The negation of p, denoted $\neg p$, is the statement "It is not the case that p". The truth value of $\neg p$ is the opposite of p.

We read $\neg p$ as "not p". (Note that some authors use $\sim p$ or \overline{p} to denote $\neg p$.)

Example. Consider the negation of each proposition above:

- (1) Columbus is *not* the capital of Ohio. (False)
- (2) $2 + 2 \neq 5$. (True)
- (3) A circle with radius r does not have area πr^2 . (False)
- (4) The number $\sqrt{2}$ is not rational. (True)
- (5) Every square is *not* a rectangle. (False)
- (6) Every rectangle is *not* a square. (True)

We can represent the relationships between the different possibilities for the truth value of p and its negation in a $truth\ table$:

p	$\neg p$
T	F
F	T
· .	

Definition: conjunction, disjunction

Let p and q be propositions.

The *conjunction* of p and q, denoted $p \wedge q$, is the proposition "p and q". The conjunction $p \wedge q$ is true when both p and q are true, and false otherwise.

The disjunction of p and q, denoted $p \vee q$, is the proposition "p or q". The disjunction $p \vee q$ is true when p and q are false, and true otherwise.

Example. (1) The proposition

Columbus is the capital of Ohio and $2+2 \neq 5$

is true because both individual propositions are true, even though the two statements have nothing to do with one another.

(2) The proposition

Every square is a rectangle and every rectangle is a square

is false because one of the individual propositions is false.

(3) The proposition

The number $\sqrt{2}$ is rational or every square is a rectangle

is true because the second statement is true.

(4) The proposition

Every rectangle is a square or 2 + 2 = 5

is false because both propositions are false.

Always, or in mathematics is defined as above. This is known as the *inclusive or*. Do not confuse this with $exclusive\ or$, defined below.

Definition: exclusive or

Let p and q be propositions. The *exclusive or* of p and q, denoted $p \oplus q$, is the proposition that is true if exactly one of p and q is true and is false otherwise.

Since there are four possibilities of pairs (p,q), then the truth table for $p \land q$, $p \lor q$, or $p \oplus q$ should have four rows as shown below:

p	q	$p \wedge q$	$p \lor q$	$p \oplus q$
T	T	T	T	F
T	F	F	T	T
F	T	F	T	T
F	F	F	F	F

We now move on to defining conditional statements, which are critical in stating mathematical facts that depend on a hypothesis.

Definition: conditional statement, hypothesis, conclusion

Let p and q be propositions. The *conditional statement* $p \to q$ is the proposition "if p, then q." The conditional statement is false when p is true and q is false, and true otherwise. Given $p \to q$, we call p the *hypothesis* and q the *conclusion*.

Consider the following example by an imaginary politician:

If I am elected, then I will lower taxes.

Only in the instance in which the politician is elected but *does not* lower taxes is the statement clearly false. If the politician is not elected, there is no expectation put on the conclusion.

There are many other ways to phrase a conditional statement. Two are p only if q, as well as p implies q (see the text for many more examples). Importantly, in this instance we say that p is a sufficient condition for q, and we say that q is a necessary condition for p.

p	q	$p \rightarrow q$
T	T	T
T	F	F
F	T	T
F	F	T

This is a little awkward to many students at first. Note in the last two rows, p is false, and so the conditional statement is $vacuously\ true$.

From a conditional statement $p \to q$ one can form several other conditional statements:

Converse: $q \to p$

Inverse: $\neg p \rightarrow \neg q$

Contrapositive: $\neg q \rightarrow \neg p$

Example. Consider the statement:

I eat oatmeal for breakfast whenever it is Tuesday.

This can be rephrased as $p \to q$:

If it is Tuesday, then I eat oatmeal for breakfast.

Write the converse, inverse, and contrapositive of this statement.

We now look at combining conditional statements.

Definition: biconditional

Let p and q be propositions. The *biconditional statement* $p \leftrightarrow q$ is the proposition "p if and only if q." The binconditional statement $p \leftrightarrow q$ is true when p and q have the same truth values, and is false otherwise.

Example. Consider the statement

You can take the flight if and only if you buy a ticket.

Some use "iff" in place of "if and only if", though this is frowned upon in formal mathematics.

p	q	$p \leftrightarrow q$
T	T	T
T	F	F
F	T	F
F	F	T

Example. Show that $p \leftrightarrow q$ has the same truth value as $(p \to q) \land (q \to p)$.

We conclude by noting the precedence of logical operators:

Operator	Precedence
	1
٨	2
V	3
\rightarrow	4
\leftrightarrow	5

However, especially in cases of disjunction versus conjunction, it is best to make use of parentheses to distinguish hierarchy.

1.2 Applications of propositional logic

We begin by discussing how to translate english sentences into propositional logic.

Example. Consider the statement:

You can see the movie only if you are over 18 years old or you have the permission of a parent.

The three statements here are

m: You can see the movie

e: You are over 18 years old

p: You have the permission of a parent

As a compound conditional statement, this translates to $m \to (e \lor p)$.

Translating a natural language into propositional logic is a necessary component of creating system specifications for hardware and software. When there are several specifications, they must be consistent and not lead to a contradiction.

Example. Consider the following system specifications:

- Whenever the system software is being upgraded, users cannot access the file system.
- If the users can access the file system, then they can save new files.
- If users cannot save new files, then the system software is not being upgraded.

The three statements here are:

p: The system software is being upgraded

q: Users can access the file system

r: Users can save new files

The system specifications translate to $p \to \neg q$, $q \to r$, and $\neg r \to \neg p$.

We could create a truth table and verify that there is choice of truth value for p, q, r so that all three statements are true. We could make a truth table to check this, or use deduction.

Suppose p is true. The first statement requires $\neg q$ to be true, so q is false. Now the last statement requires that $\neg r$ is false, so r is true. If q is false and r is true, then the middle statement is true. Hence, the system is consistent.

Boolean searches use propositional logic. Here it is important that we make use of parentheses to distinguish hierarchy.

Example. Suppose we want a web search about hiking in Virginia, but not in West Virginia. We would search: (HIKING AND VIRGINIA) NOT WEST.

1.3 Propositional equivalences

There are two types of compound propositions that are important going forward to name.

Definition: tautology, contradiction, contingency

A compound proposition that is always true is called a *tautology*. A compound proposition that is always false is called a *contradiction*. A compound proposition that is not a tautology or a contradiction is called a *contingency*.

Example. Let p be a proposition. Then $p \vee \neg p$ is a tautology while $p \wedge \neg p$ is a contradiction.

In logical reasoning, it is important to be able to compare logical statements. Oftentimes, we can replace one logical statement with another.

Definition: logically equivalent

The compound propositions p and q are logically equivalent if $p \leftrightarrow q$ is a tautology. We denote logically equivalent compound propositions by $p \equiv q$.

Instead of showing directly that $p \leftrightarrow q$, it is often easier to show that their columns in a truth table are the same.

Example. The following equivalences are known as De Morgan's Laws (for logic). Let p and q be statements. Then

$$(1) \neg (p \land q) \equiv (\neg p) \lor (\neg q)$$

(2)
$$\neg (p \lor q) \equiv (\neg p) \land (\neg q)$$

We will verify the second. The first is left as an exercise.

De Morgan's laws can be useful in negating statements in natural language. Note that these extend to more than two statements (write the extended forms).

Example. Consider the following statement:

Evelyn likes art and playing the piano

The two statements are

p: Evelyn likes art q: Evelyn likes playing the piano

Since the negation of $p \wedge q$ is $\neg p \vee \neg q$, then the negation is

Evelyn does not like art, or Evelyn does not like playing the piano

Example. Show that $p \to q$ and $\neg p \lor q$ are logically equivalent. This is known as the *conditional-disjunction equivalence*.

There are many other such rules listed in your textbook (see Section 1.3). Besides De Morgan's laws, you are not expected to know these all by heart. However, you should be able to verify all of them. (Display these laws on screen.)

Example. Show that $p \lor (q \land r) \equiv (p \lor q) \land (p \lor r)$. This is the distributive law of disjunction over conjunction.

Example. We can combine our rules to make formal arguments that do not require truth tables. For example, we will use the above rules to show that $\neg(p \to q) \equiv p \land \neg q$:

$$\neg(p \to q) \equiv \neg(\neg p \lor q)$$
 conditional-disjunction equivalence
$$\equiv \neg(\neg p) \land \neg q$$
 De Morgan's Law
$$\equiv p \land \neg q$$
 double negation law

Example. We will show that $(p \land q) \rightarrow (p \lor q)$ is a tautology using our laws.

$$(p \wedge q) \to (p \vee q) \equiv \neg (p \wedge q) \vee (p \vee q) \qquad \text{see above}$$

$$\equiv (\neg p \vee \neg q) \vee (p \vee q) \qquad \text{De Morgan}$$

$$\equiv (\neg p \vee p) \vee (\neg q \vee q) \qquad \text{associative and commutative}$$

$$\equiv T \vee T \qquad \text{see above}$$

$$\equiv T \qquad \text{domination law}$$

1.4 Predicates and quantifiers

Quantifiers allow us to make statements involving variables. This is known as *predicate logic*. For example, the mathematical statement "x > 3" does not have a truth value *until* we choose a value for x. Similarly, the statement "Team x won the game" is meaningless until we identify "Team x".

Predicates are used frequently in programming to decide whether to execute a command. Consider the following line of code:

if
$$x > 0$$
, then $x := x + 1$

Definition: propositional function, subject, predicate

A propositional function $P(x_1, x_2, ..., x_n)$ is a proposition that depends on the variables $x_1, 2_2, ..., x_n$, known as *subjects*. The property that must be decided for P is called the *predicate*.

Example. If P(x) is the statement that "x > 3", then P(4) is true but P(2) is false.

If Q(x,y) denotes the statement "x=y+3", then Q(4,1) is true but Q(0,0) is false.

Oftentimes we will want to consider a propositional function not just at one value, but at many or all values in the range of considered values, called the *domain*. Alternatively, we may want to decide if it is true for any value of x. To make such statements, we use quantifiers.

Definition: universal quantifier, counterexample

The universal quantification of P(x), denoted $\forall x P(x)$, is the statement

P(x) for all values of x in the domain

Here \forall is called the *universal quantifier*. A value of x in the domain for which P(x) is false is called a *counterexample* to $\forall x P(x)$.

There are many other ways to express "for all". Other common ones are "for every", "for each", and "for arbitrary". It is best to avoid "for any" because it can be ambiguous.

Example. Let P(x) be the statement "x + 1 > x" where the domain is all real numbers. This is true for all x, so the universal quantifier $\forall x P(x)$ is true.

Let Q(x) be the statement "x < 2" where the domain is all real numbers. Since Q(3) is false, then x = 3 is a counterexample for $\forall x Q(x)$. Hence $\forall x Q(x)$ is false.

Generally we treat all domains as nonempty. If the domain is empty, then $\forall x P(x)$ is true for any propositional function P(x). The domain is important. For example, in the example above, if we restricted the domain to negative integers, then $\forall x Q(x)$ is true.

In formal mathematics, where proofs are written in prose, it is not customary too use the symbols \forall and \exists , except in set notation. When we move on to writing proofs, I will enforce this, but for logical discourse it is allowed.

Definition: existential quantifier

The existential quantification of P(x), denoted $\exists x P(x)$, is the proposition

There exists an element x in the domain such that P(x)

Here \exists is called the *existential quantifier*.

We might use instead of "there exists" the phrases "for some", "for at least one", or "there is".

Example. Let P(x) denote the statement x > 3 where the domain consists of all real numbers. Then P(4) is true so $\exists x P(x)$ is true.

Let Q(x) denote the statement x = x + 1 where the domain consists of all real numbers. Then no x satisfies Q(x), so $\exists x Q(x)$ is false.

Note that this statement requires only existence, not uniqueness.

Definition: uniqueness quantifier

The notation $\exists !xP(x)$ is the proposition

There exists a unique x in the domain such that P(x)

Here $\exists!$ is called the uniqueness quantifier.

Example. Let P(x) denote the statement x-1=0 where the domain consists of all real numbers. Then P(1) is true but P(x) is false for all $x \neq 1$. So $\exists! x P(x)$ is true.

Let Q(x) denote the statement $x^2 = 1$ where the domain consists of all real numbers. Then Q(1) and Q(-1) are both true, so $\exists ! x Q(x)$ is false.

In a natural way, we can combine our quantifiers with our logical operators. Note that \forall and \exists have higher precedence than those from propositional calculus.

We now study when these propositions are logically equivalent.

Definition: logical equivalence

Statements involving predicates and quantifiers are logically equivalent if and only if they have the same truth value no matter which predicates are substituted into these statements and which domain of discourse is used for the variables in these propositional functions.

We use the notation $S \equiv T$ to indicate that statements involving predicates and quantifiers are equivalent.

Example. We will show that $\forall x (P(x) \land Q(x))$ and $\forall P(x) \land \forall Q(x)$ are logically equivalent.

Assume we have a common domain and let a be in that domain. Then $P(a) \wedge Q(a)$ is true. But this implies that P(a) is true and that Q(a) is true. But our choice of a was arbitrary. So, $\forall x P(x)$ is true and $\forall x Q(x)$ is true. Thus, $\forall P(x) \wedge \forall Q(x)$ is true.

Now suppose that $\forall P(x) \land \forall Q(x)$ is true. Thus, $\forall P(x)$ is true and $\forall Q(x)$ is true. Hence, if a is in the domain, then P(a) is true and Q(a) is true. Thus, $P(a) \land Q(a)$ is true. Since this holds for all a in the domain, then $\forall x (P(x) \land Q(x))$ is true.

We now show how negation plays with propositional functions.

Example. We will show that $\neg \forall x P(x) \equiv \exists x \neg P(x)$.

Suppose the statement $\neg \forall x P(x)$ is true. This implies that $\forall x P(x)$ is false, which implies that there is some counterexample to $\forall x P(x)$. Let y be the counterexample. That is, P(y) is false. But this implies that $\neg P(y)$ is true. So $\exists x \neg P(x)$ is true. The converse is similar.

This leads to DeMorgan's Laws for quantifiers

$$\neg(\forall x P(x)) \equiv \exists x, \neg P(x)$$

$$\neg(\exists x P(x)) \equiv \forall x, \neg P(x)$$

Sometimes translating sentences from natural language using predicates and quantifiers is nontrivial.

Example. (1) Someone in your class has seen the movie Borat.

There is a student x in this class having the property that x has seen Borat.

(2) All the students in this class go to Miami.

For every student x in this class, x goes to Miami.

(3) There is an integer that squares to 9

There exists an integer n having the property that $n^2 = 9$.

(Note the domain of this statement is all integers, not all real numbers.)

(4) Every integer that is not odd is even.

Let E(n) be the propositional function that is true when n is even. Let O(n) be the propositional function that is true when n is odd.

$$\forall n \in \mathbb{Z}, \neg O(n) \rightarrow \neg E(n)$$

The statement $n \in \mathbb{Z}$ means that n is an element of the set \mathbb{Z} , which is the set of integers. We will discuss this notation more in coming sections.

1.5 NESTED QUANTIFIERS

We can combine quantifiers to make more complex statements.

Example. Let P(x,y) be the statement x+y=0 and consider the quantification $\forall x \exists y P(x,y)$ where the domain is all real numbers.

The is is the proposition

For all real numbers x, there exists a real number y such that x + y = 0

This is considered a nested quantifier because we could set Q(x) to be the statement $\exists y P(x,y)$ and then the statement above becomes $\forall x Q(x)$.

This statement is true if we can pick an arbitrary element x and show that Q(x) is true. But now for Q(x) to be true, we have to show that there exists some y that makes P(x, y) true. Obviously picking y = -x makes this true.

Example. Let P(x,y) be the statement x + y = y + x where the domain is all real numbers. The quantification $\forall x \forall y P(x,y)$ is the statement

For all real numbers x, for all real numbers y, x + y = y + x

This is the commutative law of addition for real numbers. Note that the truth value is equivalent to that $\forall y \forall x P(x, y)$, so the order of two universal quantifiers may be interchanged.

Example. Consider the first example. The quantification $\exists y \forall x P(x, y)$ is *false*. Therefore, one may not in general interchange a universal quantifier with that of an existential quantifier.

Example. Consider the statement,

Every real number except zero has a multiplicative inverse

Stated another way

For every real number x, if $x \neq 0$, then there exists a real number y such that xy = 1We can state this using quantifiers as

$$\forall x((x \neq 0) \rightarrow \exists y(xy = 1))$$

1.7 Introduction to Proofs

In mathematical writing, a *theorem* is a statement that can be shown to be true. Typically, a *proposition* is a less important theorem (but this is often a matter of taste). A *lemma* is typically a smaller statement used in proving a more important proof. A *corollary* is a consequence of another theorem. A *conjecture* is a proposed theorem (without proof) based on some established evidence.

Theorems and other statements are verified using a *proof*, which is a logical argument (typically written in natural language) that establishes the validity of the statement. Proofs rely on axioms, such as the laws of real numbers, as well as other theorems that have already been justified. In this section we will discuss some methods of proof as well as consider some simple examples.

Example. Theorems are typically stated as conditionals (or biconditionals), but are not always stated in terms of universal quantifiers, so it is important for approaching a proof to be able to translate. The statement

If n is an odd integer, then n^2 is odd

means

For all integers n, if n is odd, then n^2 is odd

This statement is of the form $\forall n(P(n) \to Q(n))$, where P(n) is the statement n is odd and Q(n) is the statement that n^2 is odd. The domain is the set of integers. So, we begin by choosing an arbitrary integer n and assuming P(n) is true. We then try to establish that Q(n) is true.

To prove a conditional $p \to q$, one can use *direct proof* in which p is assumed to be true, then use logical inference to establish that q is true.

Definition: even, odd, parity

An integer n is *even* if there exists an integer k such that n = 2k, and n is *odd* if there exists an integer k such that n = 2k + 1. Two integers have the same parity when both are even or both are odd, and they have *opposite parity* when one is even and one is odd.

Example. Prove: "If n is an odd integer, then n^2 is odd."

Proof. Suppose n is an odd integer. By the definition of odd integers, n = 2k + 1 for some integer k. By algebra, $n^2 = (2k + 1)^2 = 4k^2 + 4k + 1 = 2(2k^2 + 2k) + 1$. Since $2k^2 + 2k$ is another integer, then n^2 is odd by the definition of odd integers.

Example. Prove that the sum of two even integers is even.

Before we begin, we translate as a conditional:

If m and n are even integers, then m + n is an even integer

Let P(n) be the statement that n is even. We are trying to prove

$$\forall m \forall n ((P(m) \land P(n)) \rightarrow P(m+n))$$

So, we choose arbitrary integers m and n and assume both are even. Then try to conclude that P(m+n) is even.

Proof. Suppose m and n are even integers. Then m=2k and $n=2\ell$ for integers k and ℓ . (Note we did not choose k for both since that would imply m=n and this is not a hypothesis.) Then $m+n=2k+2\ell=2(k+\ell)$, which is even by definition.

Definition: perfect square

An integer a is a perfect square if $a = b^2$ for some integer b.

Example. Prove: "If m and n are both perfect squares, then nm is a perfect square."

Another proof method is *proof by contraposition*. Recall that the statement $p \to q$ is logically equivalent to $\neg q \to \neg p$. So, in contrast to direct proof, here we assume q is false and try to prove that p is false.

Example. Prove that if a is an integer and a^2 is even, then a is even.

Proof. Suppose that
$$a$$
 is odd (i.e., a is not even). Then $a = 2k + 1$ for some integer k . By algebra, $a^2 = (2k + 1)^2 = 4k^2 + 4k + 1 = 2(2k^2 + 2k) + 1$, so a^2 is odd.

Example. Prove that if x, y, and z are integers and x + y + z is odd, then at least one of x, y, and z is odd.

The negation of the conclusion is that all three are even. So we use proof by contraposition.

Proof. Suppose x, y, z are even. Then x = 2a, y = 2b, and z = 2c for integers a, b, c. Then by algebra, x + y + z = 2a + 2b + 2c = 2(a + b + c) which is even by definition.

Recall that the statement $p \to q$ is vacuously true when p is true. For example, the statement

If n is an integer with $10 \le n \le 15$ is a perfect square, then n is also a perfect cube is vacuously true because there are no perfect squares between 10 and 15.

Definition: rational, irrational

The real number r is rational if there exist integers p and q with $q \neq 0$ such that r = p/q. A real number that is not rational is called *irrational*.

Suppose we want to prove a statement p is true (not necessarily a conditional). Perhaps we can find a contradiction q (i.e., $q = r \land \neg r$) such that $\neg p \to q$ is true. Since q is false, then this implies that $\neg p$ is false, so p is true.

Example. Prove that $\sqrt{2}$ is irrational.

Let p be the statement " $\sqrt{2}$ is irrational". We assume p is false and try to arrive at a contradiction.

Proof. Suppose $\sqrt{2}$ is not irrational, so $\sqrt{2}$ is rational. Then by definition, $\sqrt{2} = a/b$ for integers a and b, $b \neq 0$. Assume a and b have no common factors (that is, a/b is expressed in lowest terms). Squaring both sides of $\sqrt{2} = a/b$ gives $2 = a^2/b^2$. Hence, $2b^2 = a^2$. This implies a^2 is even, so a is even. That is, a = 2c for some integer c. Then we have $2b^2 = 4c^2$, so $b^2 = 2c^2$. That is, b is also even. But then a and b have a common factor (of 2), a contradiction.

In the proof we assumed $\neg p$, or the statement that $\sqrt{2} = a/b$ is rational, and proved q = a and b have a common factor is true. But q is false, so $\neg p$ is false. That is, p is true.

Here is another classical example. Recall that an integer n > 1 is *prime* if its only factors are 1 and itself. (We do *not* consider 1 to be a prime.)

Example. Prove that there are infinitely many primes.

Suppose there are finitely many primes. We list them p_1, p_2, \ldots, p_n . The product $p = p_1 p_2 \cdots p_n$ is divisible by each of the p_i , so p+1 is not divisible by any. Hence, p+1 is prime, a contradiction.

Recall that statements of the form $\forall x P(x)$ can be proven false simply by finding a counterexample.

Example. Consider the statement, "If n^2 is positive, then n is positive"

Let P(n) be the statement that n is positive, and let Q(n) be the statement that n^2 is positive. The given conditional is for the form $\forall n(Q(n) \to P(n))$. If we consider n = -1, then Q(-1) is true but P(-1) is false, so the conditional is false.

In writing proofs, students should be careful not to fall into one of many traps. For example, a common mistake is to assume the conclusion of the statement. Another, which is similar is called circular reasoning, also called begging the question.

Example. Prove the statement, "If n^2 is even, then n is even".

Proof. Suppose n^2 is even. Then $n^2 = 2k$ for some integer k. Let $n = 2\ell$ for some integer ℓ . Then n is even by definition.

This "proof" is incorrect because instead of reasoning why n is even, we simply concluded (out of nowhere!) and equivalent statement for n being even.

1.8 Proof Methods and Strategy

In this section we examine other methods and strategies for approaching mathematical proof. Other strategies will come later.

Proof by cases is useful when there are only *finitely many* cases to consider. That is, we are proving a statement of the form $(p_1 \lor p_2 \lor \cdots \lor p_n) \to q$, which is equivalent to

$$(p_1 \to q) \land (p_2 \to q) \land \dots \land (p_n \to q).$$

Be careful to not apply this strategy when there are *infinitely many* cases.

Example. Prove that 10 is not the square of a positive integer.

Proof. We have two cases: $1 \le x \le 3$ and $x \ge 4$. In the first case, we check directly that $1^2 = 1 \ne 10$, $2^2 = 4 \ne 10$, and $3^2 = 9 \ne 10$. Now if $x \ge 4$, then $x^2 \ge 4^2 \ge 10$. The result follows.

Example. If n is an integer, then $n^2 \ge n$. Consider the cases that $n = 0, n \ge 1$, and $n \le -1$.

Example. Given real numbers x and y, the notation $\min(x, y)$ means the *minimum* of the two numbers. Let a, b, c be real numbers. Then $\min(a, \min(b, c)) = \min(\min(a, b), c)$.

Proof. There are many cases to consider. We will show a few but you should work out the rest. Suppose $a \le b \le c$. Then

$$\min(a, \min(b, c)) = \min(a, b) = a$$
 and $\min(\min(a, b), c) = \min(a, c) = a$.

Similarly, the case $a \le c \le b$ gives

$$\min(a, \min(b, c)) = \min(a, c) = a$$
 and $\min(\min(a, b), c) = \min(a, c) = a$.

There are four remaining cases which should be checked, but the reasoning is the same. \Box

One way we could simplify the previous proof is to use without loss of generality (WLOG). This is useful when we observe that many cases follow by the same logic, perhaps by switching variables.

Example. If x and y are integers and both xy and x + y are even, then both x and y are even.

Proof. We use proof by contraposition. Suppose (at least) one of x or y is odd. Without loss of generality, we may assume that x is odd. (We can do this because the case that y is odd is identical.) We will show that either xy or x + y is odd.

Write x = 2k + 1 for some integer k. Now we have two cases: y is even or y is odd. If y is even, then $y = 2\ell$ for some integer ℓ and so $x + y = 2(k + \ell) + 1$ is odd. On the other hand, if y is odd, then $y = 2\ell + 1$ for some integer ℓ and so $xy = 2(2k\ell + k + \ell) + 1$ is odd. The proof follows from exhaustion.

We now turn to existence proofs, that is, proving statements of the form $\exists x P(x)$. There are two types of these: constructive (where we present a concrete example) and nonconstructive (prove that an element must exist without explicitly describing it).

Example. There exists an integer that can be written as the sum of two positive cubes in two different way.

Proof. Ramanujan's number 1729 is an example because $1729 = 10^3 + 9^3 = 12^3 + 1^2$. (In fact, it is the smallest such number.)

Example. Prove that either $2 \cdot 10^{500} + 15$ or $2 \cdot 10^{500} + 16$ is not a perfect square.

Proof. If $2 \cdot 10^{500} + 15$ is not a perfect square, then we are done. So, suppose it is a perfect square. Then $2 \cdot 10^{500} + 15 = a^2$ for some integer a. This implies that $2 \cdot 10^{500} + 16 = a^2 + 1$. But there are not two consecutive perfect squares.

Notice in this proof we did not actually compute whether or not one of them is, or is not, a perfect square, so this proof is nonconstructive.

At times we will be asked to prove more than just existence, but existence and uniqueness. To prove a uniqueness statement, assume there are two, and prove that they are in fact the same.

Example. Given a real number x there exists unique numbers n and ϵ such that $x = n + \epsilon$, n is an integer, and $0 \le \epsilon < 1$.

Proof. Set n = [x], the integer or "whole" part of x. Then clearly $\epsilon = x - n$ satisfies $0 \le \epsilon < 1$. This proves existence.

To prove uniqueness, suppose there is another pair n' and ϵ' such that $x = n' + \epsilon'$, n' is an integer, and $0 \le \epsilon' < 1$. Without loss of generality, assume that $\epsilon \ge \epsilon'$. Then we have $n + \epsilon = n' + \epsilon'$ so $\epsilon - \epsilon' = n' - n$. But then n' - n is an integer while $0 \le \epsilon - \epsilon' < 1$. Hence, $\epsilon - \epsilon' = 0$, so $\epsilon = \epsilon'$ and n = n'.

Sets and Functions

2.1 Sets

Definition: Set, elements

A set is an unordered collection of distinct objects, called elements.

A set is said to contain its elements. We write $a \in A$ to say that a is an element of the set A. Conversely, we write $a \notin A$ to say that a is not an element of the set A.

When a set is small, we can list the elements directly using braces. For example, the set of vowels in the English language could be denoted $V = \{a, e, i, o, u\}$. However, recall that a set is by definition unordered. So it is not important which order we list the vowels in.

When a set is too big to list its elements, we list *some* of the elements then use ellipses (...) to indicate that the pattern continues. For examples, the postive integers less than 100 could be written $\{1, 2, 3, \dots, 99\}$. It is important that the pattern is clear to the reader. If we had only written $\{1, \ldots, 99\}$, then we could mean all of the numbers between 1 and 99, or just the odds.

Example. Important sets to know are the following:

- (1) the natural numbers $\mathbb{N} = \{0, 1, 2, 3, \ldots\}$
- (2) the integers, $\mathbb{Z} = \{\dots, -2, -1, 0, 1, 2, \dots\}$
- (3) the real numbers, \mathbb{R}
- (4) the complex numbers, \mathbb{C}

We use a superscript + to denote the positive elements in any of these sets. So the positive real numbers is denoted \mathbb{R}^+ .

Another way to list a set is to use set builder notation. This has the form

$$\{x \in D \mid x \text{ has property } P\}$$

where D is the domain from which we can choose x. If the domain is understood by context, then it is often omitted. For example, the set $\{1,3,5,7\}$ could be written in set builder notation as

$$\{x \in \mathbb{Z}^+ \mid x \text{ is odd and } x < 10\}.$$

As another example, consider the rational numbers

$$\mathbb{Q} = \left\{ \frac{p}{q} : p, q \in \mathbb{Z}, q \neq 0 \right\}.$$

We will also use interval notation for subsets of \mathbb{R} :

$$[a, b] = \{x \mid a \le x \le b\}$$

$$[a, b) = \{x \mid a \le x < b\}$$

$$(a, b] = \{x \mid a < x \le b\}$$

$$(a, b) = \{x \mid a < x < b\}.$$

Definition: Equal sets

Two sets are *equal* if they have the same elements.

In logical form, we say A and B are equal if and only if $\forall x (x \in A \leftrightarrow x \in B)$. We write A = B when A and B are equal sets. For example, the sets $\{1,3,5\}$ and $\{5,1,3\}$ are equal.

Definition: Empty set, singleton set

The set with no elements is called the *empty set*. A set with only one element is called a *singleton set*.

We denote the empty set by \emptyset . Do not confuse this with $\{\emptyset\}$, which is a singleton set containing one element, the empty set.

Definition: Subset

The set A is a subset of a set B if and only if every element of A is also an element of B.

We use the notation $A \subseteq B$ to indicate that A is a subset of B. In this context we also say that B is a *superset* if A. In logical form, $A \subseteq B$ is equivalent to the quantification $\forall x (x \in A \to x \in B)$. Hence, to show that A is a subset of B, it suffices to choose an arbitrary element of A and justify why it is in B. Conversely, to show that A is *not* a subset of B (denoted $A \not\subset B$) we need only find one element in A that is not in B.

Example. $\emptyset \subseteq \mathbb{N} \subseteq \mathbb{Z} \subseteq \mathbb{Q} \subseteq \mathbb{R} \subseteq \mathbb{C}$.

Theorem 1: Common subsets

The empty set is a subset of every set. A set is always a subset of itself.

Proof. Let S be a set. We claim $\emptyset \subseteq S$. That is, we wish to show that $\forall x (x \in \emptyset \to x \in S)$. For any x, the statement $x \in \emptyset$ is false because \emptyset contains no elements. Thus, $x \in \emptyset \to x \in S$ is true.

Next we claim $S \subseteq S$. That is, we claim $\forall x (x \in S \to x \in S)$. Rewriting the conditional we have $x \in S \to x \in S \equiv \neg (x \in S) \lor (x \in S)$, which is a tautology.

We say A is a proper subset of B if A is a subset of B and $A \neq B$. In this case, the textbook uses the notation $A \subset B$, though it is more common (in my opinion) to write $A \subsetneq B$.

To show that sets A and B are equal, it suffices to show that $A \subseteq B$ and $B \subseteq A$. That is, we have the equivalence,

$$\forall x(x \in A \leftrightarrow x \in B) \equiv \forall x(x \in A \rightarrow x \in B) \land \forall x(x \in B \rightarrow x \in A)$$

Definition: Finite set, infinite set, cardinality

Let S be a set. If there are exactly n distinct elements in S, where N is a nonnegative integer, then S is a *finite set* of *cardinality* n. We denote the cardinality of S by |S|. A set that is not finite is said to be *infinite*.

Example. Let A be the set of odd positive integers less than 10, then A is finite and |A| = 5. On the other hand, the set of positive integers is finite. The empty set has $|\emptyset| = 0$.

Definition: Power set

Given a set S, the power set of S is the set of all subsets of S, denoted $\mathcal{P}(S)$.

Example. Let $S = \{0, 1, 2\}$. Then

$$\mathcal{P}(S) = \{\emptyset, \{0\}, \{1\}, \{2\}, \{0, 1\}, \{0, 2\}, \{1, 2\}, \{0, 1, 2\}\}\$$

Recall that an ordered n-tuple (a_1, a_2, \ldots, a_n) is the ordered collection that has a_1 as its first element, a_2 as its second, etc. This is in contrast to a set, where order does not matter.

Definition: Cartesian product

Let A and B be sets. The Cartesian product of A and B is the set of all ordered pairs (a, b), where $a \in A$ and $b \in B$.

We denote the Cartesian product by $A \times B = \{(a, b) \mid a \in A \land b \in B\}$. We call a subset of $A \times B$ a relation. We will study these in some depth later in the course.

Example. The Cartesian product of $A = \{1, 2\}$ and $B = \{a, b, c\}$ is

$$A \times B = \{(1, a), (1, b), (1, c), (2, a), (2, b), (2, c)\}.$$

We can of course extend the notion of the Cartesian product to several sets A_1, A_2, \ldots, A_n :

$$A_1 \times A_2 \times \cdots \times A_n = \{(a_1, a_2, \dots, a_n) \mid a_i \in A_i \text{ for } i = 1, 2, \dots, n\}.$$

2.2 Set Operations

We now study operations on sets, which are closely related to logical operators.

Definition: Union, intersection, complement

Let A and B be sets. The *union* of A and B, denoted $A \cup B$ is the set that contains those elements that are either in A or in B, or in both. The *intersection* of A and B, denoted $A \cap B$ is the set that contains those elements that are either in both A and B. The *difference* of A and B, denoted A - B, is the set containing those elements that are in A but not in B.

(Draw Venn diagrams for each of these.) Note that A-B is sometimes denoted $A \setminus B$. In set builder notation, these are

$$A \cup B = \{x : x \in A \lor x \in B\}, \quad A \cap B = \{x : x \in A \land x \in B\}, \quad A - B = \{x : x \in A \land x \notin B\}$$

Example. Let
$$A = \{1, 3, 5\}$$
, $B = \{1, 2, 3\}$. Then $A \cup B = \{1, 2, 3, 5\}$, $A \cap B = \{1, 3\}$, $A - B = \{2\}$.

We say two sets are *disjoint* if their intersection is the empty set. For example, the set of even integers and the set of odd integers are disjoint. The next result gives a general way to count the number of elements in a union.

Inclusion-Exclusion Principle

Let A and B be sets. Then $|A \cup B| = |A| + |B| - |A \cap B|$.

A universal set for sets A and B is any set containing both of them. In most cases, there will be a natural choice for U. For example, if A and B are both sets of integers, then we may take $U = \mathbb{Z}$.

Definition: Complement

Let A and be a subset of a universal set U. The complement of A (in U) is $\overline{A} = U - A$.

Some authors use A' or A^c for the complement. Remember this definition depends on the universal set U. If A and B are subsets of a universal set U, then $A - B = A \cap \overline{B}$.

Example. If U is the English alphabet and V is the set of vowels, then \overline{V} is the set of consonants.

We now turn to verifying several set identities, which is similar to verifying logical equivalences.

Example. We will prove the first De Morgan's law for set identities: $\overline{A \cap B} = \overline{A} \cup \overline{B}$:

$$\overline{A \cap B} = \{x \mid x \notin (A \cap B)\} = \{x \mid \neg(x \in (A \cap B))\} = \{x \mid \neg(x \in A \land x \in B)\}$$
$$= \{x \mid \neg(x \in A) \lor \neg(x \in B)\} = \{x \mid x \notin A \lor x \notin B\} = \{x \mid x \in \overline{A} \lor x \in \overline{B}\}$$
$$= \{x \mid x \in \overline{A} \cup \overline{B}\} = \overline{A} \cup \overline{B}.$$

Alternatively, to prove two sets are equal, we can show that each side is a subset of the other side.

Example. Prove the distributive law for union: $A \cup (B \cap C) = (A \cup B) \cap (A \cup C)$.

Proof. Let $x \in A \cup (B \cap C)$. Then $x \in A$ or $x \in B \cap C$. If $x \in A$, then $x \in A \cup B$ and $x \in A \cup C$, so $x \in (A \cup B) \cap (A \cup C)$. If $x \in B \cap C$, then $x \in B$ so $x \in A \cup B$, and $x \in C$ so $x \in A \cup C$. Again we have $x \in (A \cup B) \cap (A \cup C)$. Thus, in either case we have $A \cup (B \cap C) \in (A \cup B) \cap (A \cup C)$.

Suppose $y \in (A \cup B) \cap (A \cup C)$. Then $y \in A \cup B$ and $y \in A \cup C$. Suppose $y \in A$, then $y \in A \cup (B \cap C)$. Now suppose $y \notin A$. Since $y \in A \cup B$, then $y \in B$. Similarly, since $y \in A \cup C$, then $y \in C$. Thus, $y \in B \cap C$ so again we have $y \in A \cup (B \cap C)$. Thus, in either case we have $(A \cup B) \cap (A \cup C) \subset A \cup (B \cap C)$.

Finally, we can now prove new identities by using our existing identities.

Example. Let A, B and C be sets. Show that (A - B) - C = (A - C) - (B - C).

$$(A-C)-(B-C)=(A\cap \overline{C})\cap \overline{(B\cap \overline{C})} \qquad \text{equivalent definition of set difference}$$

$$=(A\cap \overline{C})\cap (\overline{B}\cup C) \qquad \text{De Morgan's Law}$$

$$=((A\cap \overline{C})\cap \overline{B})\cup ((A\cap \overline{C})\cap C) \qquad \text{distributive property}$$

$$=((A\cap \overline{B})\cap \overline{C})\cup (A\cap (\overline{C}\cap C)) \qquad \text{associative/commutative laws for intersection}$$

$$=((A\cap \overline{B})\cap \overline{C})\cup (A\cap \emptyset) \qquad \text{negation law}$$

$$=((A\cap \overline{B})\cap \overline{C})\cup \emptyset \qquad \text{domination law}$$

$$=(A\cap \overline{B})\cap \overline{C} \qquad \text{identity law}$$

$$=(A-B)-C \qquad \text{equivalent definition of set difference}$$

We can extend union and intersection to several sets. Let A_1, A_2, \ldots, A_n be a collection of n subsets (of a universal set U). Then

$$\bigcup_{i=1}^{n} A_i = A_1 \cup A_2 \cup \dots \cup A_n$$

$$\bigcap_{i=1}^{n} A_i = A_1 \cap A_2 \cap \dots \cap A_n$$

This can also be extended to an infinite number of sets.

2.3 Functions

Definition: Function

Let A and B be nonempty sets. A function f from A to B is an assignment of exactly one element of B to each element of A.

Our notation for a function f from A to B is $f:A\to B$. If $a\in A$, then f(a)=b where b is the unique element of B assigned by the function f to the element $a\in A$. We say f maps A to B.

One can also define functions in terms of relations. Recall that a relation R is a subset of $A \times B$. Then R is a function if there is exactly one ordered pair (a, b) for each element $a \in A$.

Definition: domain, codomain, range/image

If f is a function from A to B, we say A is the *domain* of f and B is the *codomain* of f. If f(a) = b, then we say b is the *image* of a and a is a *preimage* of b. The *range* (or *image*) of f is the set of all images of elements of A.

We were careful to say that b is the image but a is a preimage, as preimages need not be unique.

Example. One example of a function g would be to assign each student in a class their current grade. So, the domain is the set of students in the class and the codomain is the set $\{A, B, C, D, F\}$. So g(Bob) = F, because Bob has never shown up for this class.

Example. Let $f: \mathbb{Z} \to \mathbb{Z}$ be defined by $f(x) = x^2$. The domain of f is \mathbb{Z} and the codomain is \mathbb{Z} . The image of f is the set of all integers that are perfect squares $\{0, 1, 4, 9, \dots\}$.

A function is real-valued if its codomain is \mathbb{R} , and integer-valued if its codomain is \mathbb{Z} .

Example (Pointwise addition and multiplication). Let f_1 and f_2 be functions from A to \mathbb{R} . Then $f_1 + f_2$ and $f_1 f_2$ are also functions from A to \mathbb{R} defined for all $x \in A$ by

$$(f_1 + f_2)(x) = f_1(x) + f_2(x),$$
 $(f_1 f_2)(x) = f_1(x) f_2(x).$

We now generalize the notion of image of a function.

Definition: Image of a subset

Let $f:A\to B$ be a function and let $S\subseteq A$. The *image* of S under f is the subset B that consists of the images of the elements of S.

In the setup of the definition, we denote the image of S by

$$f(S) = \{t \mid \exists s \in S(t = f(s))\} = \{f(s) \mid s \in S\}.$$

The image of the function f is then $f(A) = \{f(a) : a \in A\}.$

Definition: Injective, surjective, bijective

Let $f: A \to B$ be a function.

- The function f is said to be *one-to-one* (or *injective*) if and only if f(a) = f(b) implies a = b for all $a, b \in A$.
- The function f is said to be *onto* (or *surjective*) if and only if for every $b \in B$ there exists an element $a \in A$ such that f(a) = b.
- \bullet The function f is said to be a *one-to-one correspondence* (or *bijective*) if it is both one-to-one and onto.

(Various examples using arrows.)

Example. The function $f: \mathbb{R} \to \mathbb{R}$ given by $f(x) = x^2$ is not one-to-one or onto. However, if we restrict the domain $f: \mathbb{R}^+ \to \mathbb{R}^+$ then it is one-to-one. To justify this we follow the definition. Let $a, b \in \mathbb{R}^+$ such that f(a) = f(b). Then $a^2 = b^2$. Taking square roots and noting that the domain consists only of positive real numbers we have a = b.

By restricting the codomain, the function also becomes onto. To justify this, again we follow the definition. Let $r \in \mathbb{R}^+$. We need to find a preimage for r. Note that $\sqrt{r} \in \mathbb{R}^+$ and $f(\sqrt{r}) = (\sqrt{r})^2 = r$.

Example. Let A be a set. The *identity map* (on A) is the function $\iota_A : A \to A$ given by $\iota_A(x) = x$ for all $x \in A$. This function is a bijection.

We saw above that we can combine two functions with the same domain and codomain (\mathbb{R} or \mathbb{Z}) by pointwise addition or multiplication. There is a more natural way to combine functions but again we must be careful with domain and codomain.

Definition: Composition of functions

Let $g:A\to B$ and $f:B\to C$ be functions. The composition $f\circ g:A\to B$ is defined by the rule

$$(f \circ g)(a) = f(g(a))$$
 for all $a \in A$.

It is critical that the codomain of g be (included in) the domain of f. For the reason, even when $f \circ g$ is defined then $g \circ f$ may not be.

(Draw composition diagram)

Also note that, even when both are defined, generally we have that $f \circ g \neq g \circ f$. (That is, function composition is *noncommutative*.)

Example. Consider $f(x) = x^2$ and g(x) = x + 1 (both functions defined $\mathbb{R} \to \mathbb{R}$).

The notion of composition is inherently tied to that of invertibility of functions, though that will not be our definition.

Definition: Inverse function

Let $f: A \to B$ be a bijective function. The *inverse function* of f is the function that assigns to an element $b \in B$ the unique element $a \in A$ such that f(a) = b.

Our notation for the inverse function of f is f^{-1} . Hence, $f^{-1}(b) = a$ when f(a) = b. Then f^{-1} is a function from B to A/

Example. (1) Let $f: \mathbb{Z} \to \mathbb{Z}$ be given by f(x) = x + 1. This function is bijective and so it has an inverse. If we choose $y \in \mathbb{Z}$, then f(y - 1) = (y - 1) + 1 = y. Hence, the inverse function of f is $f^{-1}: \mathbb{Z} \to \mathbb{Z}$ given by $f^{-1}(y) = y - 1$.

- (2) The inverse of $g: \mathbb{R}^+ \to \mathbb{R}^+$ given by $g(x) = x^2$ has inverse function $g^{-1}(y) = \sqrt{y}$.
- (3) The function $h: \mathbb{R}^+ \to \mathbb{R}^+$ given by $h(x) = e^x$ has inverse function $h^{-1}: \mathbb{R}^+ \to \mathbb{R}^+$ given by $h^{-1}(y) = \ln(y)$.

Note that if $f: A \to B$ is bijective with inverse function $f^{-1}: B \to A$, then for any $a \in A$ and $b \in B$ we have

$$(f^{-1} \circ f)(a) = f^{-1}(f(a)) = a$$

 $(f \circ f^{-1})(b) = f(f^{-1}(b)) = b.$

Thus, $f^{-1} \circ f = \iota_A$ and $f \circ f^{-1} = \iota_B$.

Definition: Graph of a function

Let $f: A \to B$ be a function. The graph of f is the subset $\{(a,b): a \in A, b = f(a)\}$.

2.4 Sequences

We normally think of a sequence as an (infinite) ordered set. However, it will be convenient to describe them more generally.

Definition: Sequence

A sequence is a function from a subset of the integers to a set S. We use the notation a_n to denote the image of the integer n. We call a_n a term of the sequence.

Typically our domain (index set) will either be \mathbb{N} or \mathbb{Z}^+ , but sometimes we will want something else. We denote our sequence by $\{a_n\}$.

Example. Consider the sequence $\{a_n\}$ where $a_n = 1/n$. Then the sequence begins

$$1, \frac{1}{2}, \frac{1}{3}, \frac{1}{4}, \dots$$

Definition: Geometric progression

Let $a, r \in \mathbb{R}$. A geometric progression (or geometric sequence) is a sequence of the form

$$a, ar, ar^2, \ldots, ar^n, \ldots$$

We call a the *initial term* and r the *common ratio*.

The function in a geometric progression is $f(x) = ar^x$. Our indexing set is \mathbb{N} (starts with n = 0).

Example. The sequence $\{b_n\}$ with $b_n = \frac{1}{2}(-2)^n$ is a geometric progression $(a = 1/2, r = (-2)^n)$. The sequence that begins,

$$\frac{1}{2}$$
, -1 , 2 , -4 , . . .

Example. Suppose we have a geometric progression whose first term is 4! = 64 and third term is 96. Thus, $a_0 = a = 64$ and $a_3 = ar^2 = 96$. Thus, $64r^2 = 96$ so $r^2 = 3/2$. Hence, $r = \sqrt{3/2}$.

Definition: Arithmetic progression

Let $a, d \in \mathbb{R}$. An arithmetic progression (or arithmetic sequence) is a sequence of the form

$$a, a+d, a+2d, \ldots, a+nd, \ldots$$

We call a the *initial term* and d the *common difference*.

Example. The sequence $\{s_n\}$ with $s_n = 2 - 5n$ is an arithmetic progression (a = 2, d = -5). The sequence begins with $2, -3, -8, -13, \ldots$

Example. Suppose an arithmetic progression has first term 5 and fourth term 14. That is, $a_0 = 5$ and $a_3 = 14$ (careful with indexing). Then we have 5 + d(3) = 14. Solving gives d = 3. Hence, the arithmetic progression is $\{a_n\}$ with $a_n = 5 + 3n$.

Many important sequences can be defined recursively.

Definition: Recurrence relation

A recurrence relation for the sequence $\{a_n\}$ is an equation that expresses a_n in terms of one or more of the previous terms of the sequence. A sequence is called a *solution* of a recurrence relation if its terms satisfy the recurrence relation.

We call the first few terms necessary to start the recurrence relation the *initial conditions*.

Example. (1) Let $\{a_n\}$ be a sequence that satisfies the recurrence relation $a_n = a_{n-1} + 2$ and suppose $a_0 = 5$. Then the sequence begins

$$5, 7, 9, 11, \dots$$

- (2) Recall that for a positive integer n, the factorial of n is $n! = n \cdot (n-1) \cdot (n-2) \dots 1$. The sequence $\{a_n\}$ where $a_n = n!$ can be defined recursively by $a_1 = 1$ and $a_n = na_{n-1}$ for n > 1.
- (3) Suppose we have a sequence $\{a_n\}$ where $a_0=3$ and for n>1, $a_n=2^{a_{n-1}}$. Then

$$a_1 = 2^{a_0} = 2^1 = 2$$
 and $a_2 = 2^{a_1} = 2^2 = 4$.

(4) The Fibonnaci sequence $\{f_n\}$ is defined by the initial conditions $f_0 = 0$, $f_1 = 1$, and satisfies the recurrence relation $f_n = f_{n-1} + f_{n-2}$. The sequence begins

$$0, 1, 1, 2, 3, 5, 8, 13, 21, \dots$$

Some recurrence relations can be resolved to determine a *closed formula* for the terms a_n that does not depend on earlier terms in the sequence.

Example. Consider our sequence that satisfies the recurrence relation $a_n = a_{n-1} + 2$ and suppose $a_0 = 5$. We could simply write down the first few terms and see that $a_n = 5 + 2n$.

Given a sequence, we can add the terms¹. The notation we will use for this is called *summation notation* (or *sigma notation*). Suppose we have the following terms in a sequence $\{a_n\}$: $a_m, a_{m+1}, \ldots, a_n$. We express the sum as

$$\sum_{j=m}^{n} a_{j} = a_{m} + a_{m+1} + \dots + a_{n}.$$

The variable j is the index of summation, m is the lower limit, and n is the upper limit.

¹Students have undoubtably seen this in the context of Riemann sums. Those who have taken Calc II have seen infinite series but we will only be dealing with finite sums.

Example. Suppose we have the sequence $\{a_n\}$ where $a_n = n^2$. The sum of the first 5 terms is

$$\sum_{i=1}^{5} j^2 = 1^2 + 2^2 + 3^2 + 4^2 + 5^2 = 1 + 4 + 9 + 16 + 25 = 55.$$

Wouldn't it be nice if there were easy formula to compute such a sum? Hmm....

Example. Recall our geometric progression $\{b_n\}$ with $b_n = \frac{1}{2}(-2)^n$. (Recall here a = 1/2, $r = (-2)^n$). Then the sum of the first 4 terms is

$$\sum_{i=0}^{3} = \frac{1}{2} + -1 + 2 + -4 = -5/2.$$

(Only if time...)

There is a nice formula for computing the sum of the first (n+1) terms of a geometric progression. First note that if r=1, then the sum $\sum_{j=0}^{n} ar^{j}$ is just $a+a+\cdots+a=(n+1)a$. Suppose $r\neq 1$ and set

$$S_n = \sum_{j=0}^n ar^j.$$

Then by reindexing we have

$$rS_n = \sum_{j=0}^n ar^{j+1} = \sum_{j=1}^{n+1} ar^j.$$

So these two sums are the same except for the first summand of S_n and the last summand of rS_n . This implies that

$$S_n - rS_n = a - ar^{n+1}$$

 $(1 - r)S_n = a(1 - r^{n+1})$
 $S_n = \frac{a(1 - r^{n+1})}{1 - r}$.

Now check this with the previous example.

2.5 Cardinality of Sets

We now discuss how one compare the "sizes" of infinite set.

Definition: Cardinality

Nonempty sets A and B are said to have the same *cardinality* if there exists a bijection $f: A \to B$.

When two sets have the same cardinality, we write |A| = |B|.

Recall that when a set S is finite, say with n elements, then there is a bijection $f:\{1,\ldots,n\}\to S$ and we write |S|=n (or |S|=n). So cardinality is really only interesting with regards to infinite sets. But, not all infinities are created equal.

If there exists an injective function $f: A \to B$, then we write $|A| \le |B|$. If there exists an injection but no bijection, then we write |A| < |B|. Similarly, if there exists a surjective function $f: A \to B$, then $|B| \le |A|$.

Definition: Countable, uncountable

A set that is either finite or has the same cardinality as the positive integers \mathbb{N} is said to be *countable*. A set that is not countable is said to be *uncountable*.

We write $|\mathbb{N}| = \aleph_0$ ("aleph null").

Example. Here we demonstrate several countable sets.

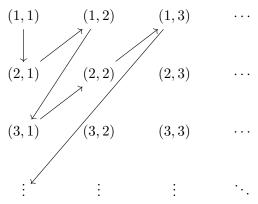
- (2) The positive integers \mathbb{Z}^+ (starting at 0) are countable. Define a map $f: \mathbb{N} \to \mathbb{Z}^+$ by f(n) = n+1. (Hilbert's Grand Hotel)
- (1) The positive even integers, E^+ , are countable. Define a map $f: \mathbb{Z}^+ \to E^+$ by f(n) = 2n. It is clear that this function is both injective and surjective. Hence it is bijective. A similar proof works for the positive odd integers.
- (3) The integers \mathbb{Z} are countable. We could set up a formal correspondence, or we could simply note that we can list the integers in a specified order:

$$0, 1, -1, 2, -2, 3, -3, \dots$$

For those interested, we could define a bijection $f: \mathbb{N} \to \mathbb{Z}$ in the following way:

$$f(n) = \begin{cases} -\frac{n}{2} & \text{if } n \text{ is even} \\ \frac{n+1}{2} & \text{if } n \text{ is odd} \end{cases}$$

Example. We will show that \mathbb{Q}^+ , the set of positive rationals, is countable. To do this, we set up an ordering



Example. On the other hand, \mathbb{R} is not countably infinite. To do this, we use the decimal expansion of a real number.

Let $f: \mathbb{N} \to \mathbb{R}$ be a function. Choose an element $b = b_0.b_1b_2b_3b_4... \in \mathbb{R}$ in the following way:

- Choose b_0 so that it is different from the whole part of f(0).
- Choose b_1 so that is is different from the first decimal place of f(1).
- Choose b_2 so that is different from the second decimal place of f(2).
- And so on...
- Choose b_k so that is is different from the kth decimal place of f(k).

Then for every n, b differs from f(n) in at least one decimal place, so $f(n) \neq b$ for every n. That is, f is not surjective. Since f was chosen arbitrarily, this implies that there is no surjective function (and hence no bijective function) $\mathbb{N} \to \mathbb{R}$. However, there is an obvious injective map $\mathbb{N} \to \mathbb{R}$ (just send each element of \mathbb{N} to itself in \mathbb{R}), so $|\mathbb{N}| < |\mathbb{R}$.

Induction and Recursion

5.1 MATHEMATICAL INDUCTION

Induction is a mathematical tool used to prove statements about the natural numbers (or sometimes about the integers). It is related in a very close way to the idea of recursion in programming. We will discuss this relationship but first we will discuss the mechanics of mathematical induction.

Suppose that we can prove that it is possible to reach a staircase. That is, we can somehow get to the first step. Then we prove that from *any* step, we are able to get to the next step. By showing these two things we have now shown that it is possible to climb the entire staircase, no matter how many steps there are (possible infinite!).

Principle of Mathematical Induction

To prove that P(n) is true for all positive integers n, where P(n) is a propositional function, we complete two steps:

Basis step (base case): We verify that P(1) is true.

Inductive step: We show that the conditional statement $P(k) \to P(k+1)$ is true for all positive integers k.

Be careful reading this principle too literally. What the basis step is really saying is that we are proving our "first case" is true. Depending on the situation, that might be k = 1, but it might be k = 0 or some other value. The inductive step is not assuming that P(k) is true for all k but rather for some k. We are assuming that we are somewhere (arbitrary) on the staircase, and proving that we can reach the next step. This assumption (that P(k) is true) is called the *inductive hypothesis*.

Example. We will prove that the sum of the first n positive integers is n(n+1)/2.

For n a positive integer, let P(n) be the proposition that $1 + 2 + \cdots + n = n(n+1)/2$. Since 1 = 1(1+1)/2, then P(1) is true. (This completes the basis step.)

Now suppose that for some positive integer k, P(k) is true. That is, $1 + 2 + \cdots + k = k/(k+1)/2$ (this is our inductive hypothesis). Now we must show that P(k+1) is true. The key observation here is that P(k+1) is the sum of the first k integers, so it is the sum of the first k integers

plus (k+1). Using our inductive hypothesis and algebra we have

$$1+2+\cdots+(k+1)=(1+2+\cdots+k)+(k+1)$$

$$=\left(\frac{k(k+1)}{2}\right)+(k+1) \text{ (by the inductive hypothesis)}$$

$$=\frac{k(k+1)+2(k+1)}{2}$$

$$=\frac{(k+1)(k+2)}{2},$$

which is the statement for P(k+1). Since we have now verified P(k+1), the given statement is true by the principle of mathematical induction.

It is important to indicate where we use the inductive hypothesis in our proof.

Example. Suppose we want a formula for the sum of the first n positive odd integers. We check the first few cases:

$$1 = 1,$$
 $1 + 3 = 4,$ $1 + 3 + 5 = 9,$ $1 + 3 + 5 + 7 = 16.$

From this, we might conjecture that the sum of the first n positive odd integers is

$$1+3+5+\cdots+(2n-1)=n^2$$
.

We now prove this conjecture using induction.

For n a positive integer, let P(n) be the proposition that the sum of the first n positive integers is n^2 . Since $1 = 1^2$, then P(1) is true. (This completes the basis step.)

Now suppose that for some positive odd integer k, P(k) is true. That is, $1+2+\cdots+(2k-1)=k^2$. Now we must show that P(k+1) is true. Using our inductive hypothesis and algebra we have

$$1 + 2 + \dots + (2k + 1) = (1 + 2 + \dots + (2k - 1)) + (2k + 1)$$

= $k^2 + 2k + 1$ (by the inductive hypothesis)
= $(k + 1)^2$,

which is the statement for P(k+1). Since we have now verified P(k+1), the given statement is true by the principle of mathematical induction.

In some cases, we may use summation notation to simplify an expression.

Example. Prove that for every positive integer n, $\sum_{j=1}^{n} j 2^{j} = (n-1)2^{n+1} + 2$.

Let P(n) be the proposition that $\sum_{j=1}^{n} j 2^{j} = (n-1)2^{n+1} + 2$. If n=1 then we have

$$\sum_{j=1}^{1} j2^{j} = 1 \cdot 2^{1} = 2 = (1-1)2^{2} + 2.$$

Hence, P(1) is true.

Now suppose that P(k) is true for some positive integer k. Then we have

$$\sum_{j=1}^{k+1} j 2^j = (k+1)2^{k+1} + \sum_{j=1}^k j 2^j$$

$$= (k+1)2^{k+1} + \left((k-1)2^{k+1} + 2\right) \text{ (by the inductive hypothesis)}$$

$$= 2^{k+1} \left((k+1) + (k-1)\right) + 2$$

$$= 2^{k+1} (2k) + 2$$

$$= k2^{k+2} + 2.$$

which is the statement for P(k+1). Since we have now verified P(k+1), the given statement is true by the principle of mathematical induction.

Induction can be used to verify inequalities, along with a host of other statements about the integers.

Example. Prove that $3^n < n!$ if n is an integer greater than 6.

Let P(n) be the proposition that $3^n < n!$. (Note that this is actually false for $n \ge 6$.) One checks (using a calculator, presumably) that $3^7 = 2187$ and 7! = 5040, so P(7) is true.

Now suppose that P(k) is true for some $k \geq 7$ (that is, $3^k < k!$). Since 3 < k+1, then we have

$$3^{k+1} = 3^k \cdot 3$$

 $< k! \cdot 3$ (by the inductive hypothesis)
 $< k! \cdot (k+1)$
 $= (k+1)!$

This is exactly the statement for P(k+1). Since we have now verified P(k+1), the given statement is true by the principle of mathematical induction.

Our next example uses sets instead of numerical values.

Example. Prove that if A_1, A_2, \ldots, A_n and B are sets, then

$$(A_1 \cap A_2 \cap \cdots \cap A_n) \cup B = (A_1 \cup B) \cap (A_2 \cup B) \cap \cdots \cap (A_n \cup B)$$

Let P(n) be the statement that the above identity holds for the positive integer n. Note that n = 1 is trivial. We will use n = 2 as our basis step. In this case, the distributive law for sets gives

$$(A_1 \cap A_2) \cup B = (A_1 \cup B) \cap (A_2 \cup B).$$

That is, P(2) is true.

Assume P(k) is true for some integer $k \geq 2$. Then

$$(A_1 \cap A_2 \cap \cdots \cap A_{k+1}) \cup B = ((A_1 \cap A_2 \cap \cdots \cap A_k) \cap A_{k+1}) \cup B$$

$$= ((A_1 \cap A_2 \cap \cdots \cap A_k) \cup B) \cap (A_{k+1} \cup B) \quad \text{(distributive law)}$$

$$= ((A_1 \cup B) \cap (A_2 \cup B) \cap \cdots \cap (A_k \cup B)) \cap (A_{k+1} \cup B) \quad \text{(inductive hypothesis)}$$

$$= (A_1 \cup B) \cap (A_2 \cup B) \cap \cdots \cap (A_{k+1} \cup B) \quad \text{(associative law)}.$$

This is the statement for P(k+1). Hence, the given statement is true by the principle of mathematical induction.

A slicker way of writing the above identity is as

$$\left(\bigcap_{j=1}^{n} A_j\right) \cup B = \bigcap_{j=1}^{n} (A_j \cup B).$$

Finally, we consider an example involving divisibility.

Example. Prove that 6 divides $n^3 - n$ whenever n is a nonnegative integer.

(Here we take our base case to be n = 0.)

For n a nonnegative integer, let P(n) be the statement that 6 divides $n^3 - n$. Clearly 6 divides $0^3 - 0 = 0$, so P(0) is true.

Assume P(k) is true for some integer $k \ge 0$. That is, 6 divides $k^3 - k$. Then

$$(k+1)^3 - (k+1) = (k^3 + 3k^2 + 3k + 1) - (k+1) = (k^3 - k) + 3(k^2 + k).$$

Regardless of whether k is even or odd, $k^2 + k$ is even. That is, 2 divides $k^2 + k$. But then 6 divides $3(k^2 + k)$. By the inductive hypothesis, 6 divides $k^3 - k$. Hence, 6 divides $(k^3 - k) + 3(k^2 + k)$ so 6 divides $(k + 1)^3 - (k + 1)$. Thus, P(k + 1) is true. Hence, the given statement is true by the principle of mathematical induction.

5.2 Strong induction

We now consider a variation on induction, called *strong induction*. We return to the staircase analogy to illustrate the difference between (regular) induction and strong induction. Instead of assuming that we are on step k, we assume that we have climbed from step 1 to step k.

Strong Induction

To prove that P(n) is true for all positive integers n, where P(n) is a propositional function, we complete two steps:

Basis step (base case): We verify that P(1) is true.

Inductive step: We show that the conditional statement $[P(1) \land P(2) \land \cdots \land P(k)] \rightarrow P(k+1)$ is true for all positive integers k.

Recall that a positive integer p is prime if the only factors of p are 1 and p itself. An integer that is not prime is composite.

Example. Prove that if n > 1 is an integer, then n is a product of primes.

Let P(n) be the proposition that n is a product of primes. Then P(2) is true because 2 is itself prime (so it is a product of one prime). Now let k be some integer $k \geq 2$ and assume that $P(2), P(3), \ldots, P(k)$ are true. We must verify P(k+1). There are two cases. The first is that k+1 is itself prime. Then there is nothing to prove and P(k+1) is true. Otherwise, k+1 is composite. But then k+1=ab where a,b are integers satisfying 1 < a,b < k+1. Since P(a) and P(b) are true by the (strong) inductive hypothesis, then a and b are products of primes. Thus k+1 is a product of two products of primes, so it is itself a product of primes.

Though it is not proved above, this decomposition of an integer into a product of primes is unique up to rearranging the factors.

Example. Match game.

Example. Suppose we have 4-cent stamps and 7-cent stamps. Let P(n) be the statement that a postage of n cents can be made using only these stamps. We claim P(n) is true for $n \ge 18$. First we prove P(n) for n = 18, 19, 20, 21.

$$18 = 7(2) + 4(1), \quad 19 = 7(1) + 4(3), \quad 20 = 7(0) + 4(5), \quad 21 = 7(3) + 4(0).$$

This proves our base cases. Now suppose that P(k) is true for some $k \geq 21$. We claim that it is true for k+1. Note that $18 \leq (k+1)-4 < k+1$. Since P(k-3) is true by the inductive hypothesis, then we can write k-3 using some combination of 4-cent and 7-cent stamps. Thus, we need only one more 4-cent stamp to produce k+1. This proves P(k+1), so the result holds by strong induction.

5.3 RECURSION

We have already discussed recursive sequences. Due to strong induction, we can now recursively define functions after specifying initial values. For example, given a sequence $\{a_n\}$, one can define the summation

$$\sum_{k=0}^{n} a_k$$

recursively by setting $S_0 = a_0$ and $S_m = S_{m-1} + a_m$ for $m \ge 1$.

A recursively defined function is a function whose domain is the set of nonnegative integers and is defined recursively as follows:

Basis step: Specify the value of the function at zero.

Recursive step: Give a rule for finding its value at an integer from its values at smaller integers.

One of our favorite examples of a recursive sequence/function is the Fibonacci sequence $f_0, f_1, f_2, ...$ where $f_0 = 0$ and $f_1 = 1$. Let $\phi = (1 + \sqrt{5})/2$ (the golden ratio).

Example. For $n \geq 3$, prove that $f_n > \phi^{n-2}$.

Let P(n) be the statement that $f_n > \phi^{n-2}$. Since

$$f_3 = 2 > \phi$$
 and $f_4 = 3 > (3 + \sqrt{5})/2 = \phi^2$

then P(3) and P(4) are true.

We proceed using strong induction. That is, assume that $P(3), P(4), \ldots, P(k)$ are true for some $k \geq 4$. Note that ϕ is a solution to the equation $x^2 - x - 1$. That is $\phi^2 = \phi + 1$. Hence,

$$\phi^{(k+1)-2} = \phi^{k-1} = \phi^2 \phi^{k-3}$$

$$= (\phi + 1)\phi^{k-3}$$

$$= \phi^{k-2} + \phi^{k-3}$$

$$< f_k + f_{k-1} \quad \text{(by the inductive hypothesis)}$$

$$= f_{k+1}.$$

Hence, P(k+1) is true and so the result holds by strong induction.

Example. Let $f: \mathbb{Z}^+ \to \mathbb{Z}^+$ be a function defined by f(1) = 1, and f(n) = 2f(n-1) for $n \ge 1$, so f is defined recursively. This is well-defined because for every $n \in \mathbb{Z}^+$, we can determine the successive value of f(n) and the output is an element of \mathbb{N} . We write the first few values of this function:

$$f(1) = 1$$
, $f(2) = 2f(1) = 2$, $f(3) = 2f(2) = 4$, $f(4) = 2f(3) = 8$.

So we see that for $n \ge 1$, f could be expressed by the closed formula $f(n) = 2^{n-1}$. We can prove that this formula is correct by induction.

Recursion can be used to define sets. For example, let Σ be a set (called an *alphabet*), then Σ^* is the set of *strings* over the alphabet Σ . This is defined recursively. The basis step is the empty string $\lambda \in \Sigma^*$. The inductive step is to say that if $x \in \Sigma^*$ and $y \in \Sigma^*$, then $xy \in \Sigma^*$ (that is, the concatenation of x and y). For example example, if $\Sigma = \{0, 1\}$, then Σ^* is the set of bit strings.

Example. Define a set S by the basis step $3 \in S$ and the recursive step is the rule: "If $x \in S$ and $y \in S$, then $x + y \in S$." We claim this defines the set of positive multiples of 3.

Let A be the set of positive multiples of 3. We claim A = S. We will prove this by showing that $A \subset S$ and $S \subset A$. Both directions will rely on induction.

Let $3n \in A$ and let P(n) be the statement that $3n \in S$. Then P(1) is true by hypothesis. Assume P(k) is true for some $k \geq 1$. Since $3k \in S$ by the inductive hypothesis and $3 \in S$ by our given hypothesis, then $3(k+1) = 3k + 3 \in S$ by the inductive rule for S. Thus, P(k+1) is true. Thus, $A \subseteq S$.

Now we show the other inclusion. Our basis step in this case is simply to notice that $3 \in A$. Now assume that all elements $k \in S$ are also in A, for some $k \geq 3$. By the recursive definition, k+1=x+y for $x,y \in S$. But then x,y < k+1, so by the inductive hypothesis, 3 divides x and 3 divides y. Thus, 3 divides x+y and so $x+1 \in A$. This shows $x \in A$.

Though we will study graphs in more detail later, this is a convenient opportunity to introduce a special type. A *graph* consists of vertices and edges which connect pairs of vertices.

Example. In this example we define rooted trees.

The basis step is to define a single vertex r to be a rooted tree. Recursively we now suppose that T_1, T_2, \ldots, T_n are rooted trees with roots r_1, r_2, \ldots, r_n respectively. The graph formed by starting a root r, which is not in any of the rooted trees T_1, T_2, \ldots, T_n , and adding an edge from r to each of the vertices r_1, r_2, \ldots, r_n is also a rooted tree.

Counting

6.1 The basics of counting

In this chapter we introduce, formally, the study of Combinatorics. This area of math studies (complicated) counting problems, as well as connections between various types of counting problems. Some parts of this you are undoubtably aware of (e.g., binomial coefficients and basic probability) but other rules will be less familiar. We begin with two basic rules. These should not be confused with rules from calculus which happen to have the same name (context matters!).

The Product Rule

Suppose that a procedure can be broken down into two tasks. If there are n_1 ways to do the first task and there are n_2 ways to do the second task, then there are n_1n_2 ways to do the procedure.

Example. Suppose we have 8 boxes which can each hold one ball, and we have two balls. How many ways are there to assign a ball to a box? For the first ball, there are 8 choices. Then for the second ball there are only 7 choices. Hence there are $8 \cdot 7 = 56$ total ways to assign balls to boxes.

The product rule scales inductively. If a procedure can be broken down into m tasks and for task k there are n_k ways to perform that task, then there are $n_1\dot{n}_2\dots n_m$ ways to do the procedure.

Example. Consider the last example but with four balls. Continuing our logic from before, there are $8 \cdot 7 \cdot 6 \cdot 5 = 1680$ ways to assign the four balls to different boxes.

Example. Suppose a license plate is made up of three letters followed by three digits (0-9). How many possible license plates combinations are there?

There are 26 choices for each letter and there are 10 choices for each digit. Since we allow repeats, there are $26 \cdot 26 \cdot 26 \cdot 10 \cdot 10 \cdot 10 = 17,576,000$ possible combinations.

The Sum Rule

If a task can be done either one of n_1 ways or in one of n_2 ways, where none of the set of n_1 ways is the same as any of the of n_2 ways, then there are $n + 1 + n_2$ ways to do the task.

Example. Suppose a representative for the Greek council¹ from either the sorority $\Gamma\Gamma\Gamma$ or the fraternity $\Upsilon\Upsilon\Upsilon$. There are 37 members of $\Gamma\Gamma\Gamma$ and 52 members of $\Upsilon\Upsilon\Upsilon$. Hence, there are 52 + 37 = 89 ways to choose a representative.

1

¹Totally a real thing, right?

Example. A certain password is either eight or nine characters long. Each character is either an upper case letter or a digit (0-9). Each password must contain at least one letter and one digit. How many passwords are possible.

Let P_8 be the number of passwords that are eight characters long. Ignoring the extra rule, there are 36^8 possible passwords. On the other hand, there are 26^8 passwords that contain *only* letters. There are 10^8 passwords that contain *only* digits. These two sets have nothing in common. Thus,

$$P_8 = 36^8 - 26^8 - 10^8 = 2,612,182,842,880.$$

A similar computation shows that

$$P_9 = 36^9 - 26^9 - 10^9 = 96,129,452,989,440.$$

Thus, the total number of passwords is

$$P_8 + P_9 = 98,741,635,832,320.$$

We can state these two rules in terms of sets. Let A_1, A_2, \ldots, A_m be finite sets. Then the number of elements in the cartesian product of these sets is

$$|A_1 \times A_2 \times \cdots \times A_m| = |A_1| \cdot |A_2| \cdot \cdots \cdot |A_m|.$$

Note that choosing an element in the cartesian product involves choosing exactly one element from each set. One the other hand, if A_1, A_2, \ldots, A_m are pairwise disjoint $(A_i \cap A_j = \emptyset \text{ for } i \neq j)$ finite sets, then

$$|A_1 \cup A_2 \cup \cdots \cup A_m| = |A_1| + |A_2| + \cdots + |A_m|$$

Now we generalize the sum rule.

The Subtraction Rule

If a task can be done either one of n_1 ways or in one of n_2 ways, then the number of ways to do the task is $n + 1 + n_2$ minus the number of ways to do the task that are common to both ways.

This is merely a combinatorial retelling of the *principle of inclusion-exclusion*. If A_1 and A_2 are finite sets, then

$$|A_1 \cup A_2| = |A_1| + |A_2| - |A_1 \cap A_2|.$$

Example. How many bit strings of length six start or end with 1?

A bit string consists of either a 0 or 1, and so there are $2^6 = 64$ total bit strings of length six. If a bit string starts with 1, then there are $2^5 = 32$ choices for the remaining bits. Similarly, there are 32 strings that end with 1. If a bit string starts and ends with 1, then there are $2^4 = 16$ choices for the remaining four bits. Consequently, there are (32 + 32) - 16 = 48 bit strings that start or end with 1.

The Division Rule

There are n/d ways to do a task if it can be done using a procedure carried out in n ways, and for every way w, exactly d of the n ways correspond to way w.

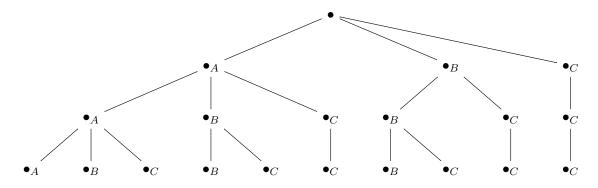
Stated in terms of sets, if a finite set A is the union of n pairwise disjoint subsets each with d elements, then n = |A|/d.

Example. Four people are to be seated around a circular table (with four seats). Two seatings are considered the same when each person has the same left neighbor and the same right neighbor. How many distinct seatings are possible.

Label one seat 1 and proceed labeling 2,3,4 clockwise around the table. There are four ways to seat one of the people in seat 1, three ways to seat someone in seat 2, two ways to seat someone in seat 1, and then only one choice for the last seat. Thus, there are 4! = 24 possible seatings, ignoring the rule. However, each of the four choices for seat 1 leads to the same arrangements, since we only distinguish arrangements when the neighbors are both different. There were four choices for seat 1, and so there are 24/4 = 6 possible seatings.

A final strategy, that we will discuss, for solving counting problems is to use a tree diagram. In this diagram, the leaves represent the possible outcomes.

Example. A passcode is a string from the alphabet $\{A, B, C\}$. Repeated letters are allowed but letters must appear alphabetically (e.g., a B may not appear before an A). How many passcodes are possible?



We can read off from the graph that there are 10 possibilities. These are

 $\{AAA, AAB, AAC, ABB, ABC, ACC, BBB, BBC, BCC, CCC\}.$

6.2 PIGEONHOLE PRINCIPLE

Imagine we have k pigeonholes and k + 1 pigeons. If those k + 1 pigeons fly into the pigeonholes, then at least one of the pigeonholes contains more than one pigeon. This may seem exceptionally trivial, but there are significant applications of this fact.

The Pigeonhole Principle

If k is a positive integer and k+1 or more objects are put into k boxes, then there is at least one box containing two or more of the objects.

Proof. We use proof by contraposition. Suppose no box contains more than one object. Then there are at most k objects.

For example, in any group of 367 people, there must be at least two with the same birthday. In any group of 27 English words, there must be at least two that begin with the same letter. In a class of 102 people, at least two students will receive the same grade on an exam.

Another application of this principle is to functions.

Corollary 1

A function from a set with k+1 or more elements to a set with k elements is not one-to-one.

The contrapositive of this statement is also useful. If $f: A \to B$ a one-to-one function between finite sets, then $|A| \leq |B|$.

The Generalized Pigeonhole Principle

If N objects are placed into k boxes, then there is at least one box that contains at least $\lceil N/k \rceil$ objects.

Proof. Suppose no box contains more than $\lceil N/k \rceil$ objects. Note that $\lceil N/k \rceil < (N/k) + 1$. Then the total number of objects is at most

$$k\left(\lceil N/k \rceil - 1\right) < k\left(\left(\frac{N}{k} + 1\right) - 1\right) = N.$$

For example, suppose we have 100 people together in a room. Since there are only 12 months, then at least 9 people were born in the same month.

As another example, suppose we have a standard deck of 52 cards. If we select cards at random, how many must we select to guarantee we have at least 3 of the same suit. Let N be the number of cards we need to select. The four "boxes" in this case are the four suits. So what we want is $\lceil N/4 \rceil \geq 3$. The smallest positive integer that makes this true is N=9.

We now study some applications of the (generalized) pigeonhole principle to sequences. A sequence $\{a_n\}$ (indexed by \mathbb{Z}^+) is said to be *increasing* if $a_i \geq a_{i-1}$ for all i and *strictly increasing* if $a_i > a_{i-1}$ for all i. Similarly, we say $\{a_n\}$ is *decreasing* if $a_i \leq a_{i-1}$ for all i and *strictly decreasing* if $a_i < a_{i-1}$ for all i.

A subsequence of $\{a_n\}$ is a sequence of the form $a_{i_1}, a_{i_2}, a_{i_3}$ where $1 \le i_1 < i_2 < i_3 < \dots$ That is, a subsequence is a collection of elements in $\{a_n\}$ in the same order that they appear in the original sequence.

Example. Suppose we have 101 people of standing in a line, all of whom have different heights. We want to find 11 people in the order they are standing in the line with heights that are increasing or decreasing.

We let a_i denote the height of the *i*th person in line, so that our sequence $a_1, a_2, \ldots, a_{101}$ represents the line of people. We are looking for a subsequence of $\{a_n\}$ that is either (strictly) increasing or (strictly) decreasing. (The parentheses are there because, as everyone has different heights, then in this context they mean the same thing.)

To each a_k , we associate the pair (i_k, d_k) where i_k is the length of the longest increasing sequence starting at a_k , and d_k is the length of the longest decreasing sequence starting at a_k .

Suppose (by way of contradiction) that there exists no increasing or decreasing sequences of length 11. Then $1 \le i_k, d_k \le 10$ for all k. By the product rule, this implies that there are $10 \cdot 10$ possible ordered pairs for the (i_k, d_k) . But there are actually 101 such ordered pairs, so by the pigeonhole principle, two of these are equal. That is, there are some terms a_s and a_t with s < t where $i_s = i_t$ and $d_s = d_t$. We claim this is impossible.

Because the terms are distinct, we have either $a_s < a_t$ or $a_s > a_t$. If $a_s < a_t$, then, because $i_s = i_t$, we can form an increasing sequence by taking a_s followed by the increasing sequence of length i_t starting at a_t . But then this increasing sequence starts at a_s has length $i_t + 1 = i_s + 1$. But this contradicts the definition of i_s . A similar contradiction arrives by assuming $a_s > a_t$ but using decreasing sequences.

We conclude then that there must be a subsequence of $\{a_n\}$ of length (at least) 11 which is either increasing or decreasing.

This extended example is actually (essentially) a proof (in disguise) of the following fact.

Theorem 2

Every sequence of $n^2 + 1$ distinct real numbers contains a subsequence of length n + 1 that is either strictly increasing or decreasing.

6.3 Permutations and Combinations

Example. Suppose in a group of 10 people, an executive committee is to be formed consisting of a President, Vice President, and a Secretary. How many different committee's can be formed.

It is important to note that order matters. There are 10 possibilities for President, but once this person has been chosen, there are only 9 possibilities for Vice President, then 8 for Secretary. By the product rule, there are $10 \cdot 9 \cdot 8 = 720$ possible committees.

Definition: Permutation, r-permutation

A permutation of a set of distinct objects is an ordered arrangement of those objects. An ordered arrangement of r elements of a set is called an r-permutation.

We denote the number of r-permutations of a set with n elements by P(n,r). Note that P(n,0) = 1 for any $n \ge 0$.

Example. Given a set $\{a, b, c, d\}$ of 4 elements, we have

$$P(4,4) = 4! = 24$$
, $P(4,3) = 4 \cdot 3 \cdot 2 = 24$, $P(4,2) = 4 \cdot 3 = 12$, $P(4,1) = 4$.

Theorem 3

If n is a positive integer and r is an integer with $1 \le r \le n$, then

$$P(n,r) = n(n-1)(n-2)\dots(n-r+1).$$

Corollary 4

If n and r are integers with $0 \le r \le n$, then $P(n,r) = \frac{n!}{(n-r)!}$

Example. How many permutations of ABCDEFGH contain the string ABC?

We treat the string ABC as a single character, so we must determine the permutations of 6 objects, which is 6! = 720.

We now consider an example where order does not matter.

Example. Suppose the math department has a faculty of 10 professors. Three need to be chosen for the awards and honors committee. How many different groups of three professors can we form?

If order matters, then there are P(10,3) = 720 such committees. But we have overcounted. For a particular committee, there are 3! = 6 permutations of that committee which are the same (because order does not matter). Hence, by the division rule, the actual number of committees is 720/6 = 120.

Definition: r-combination

An r-combination of elements of a set is an unordered selection of r elements from the set.

We denote the number of r-combinations of a set with n distinct elements by C(n,r). This is also denoted by the binomial coefficient, $\binom{n}{r}$.

Example. Given a set $\{a, b, c, d\}$ of 4 elements, we have

$$C(4,4) = 1$$
, $C(4,3) = 4$, $C(4,2) = 6$, $C(4,1) = 4$, $C(4,0) = 1$.

Theorem 5

The number of r-combinations of a set with n elements, where $0 \le r \le n$, is

$$C(n,r) = \frac{n!}{r!(n-r)!}.$$

Proof. By the division rule,

$$C(n,r) = \frac{P(n,r)}{P(r,r)} = \frac{n!/(n-r)!}{r!(r-r)!} = \frac{n!}{r!(n-r)!}.$$

Example. In the faculty example, we have

$$C(10,3) = \frac{10!}{3!7!} = \frac{10 \cdot 9 \cdot 8}{6} = 120.$$

Example. How many five-card poker hands can be dealt from a standard deck of of 52 cards?

We have

$$C(52,5) = \frac{52!}{5!47!} = \frac{52 \cdot 51 \cdot 50 \cdot 49 \cdot 48}{5 \cdot 4 \cdot 3 \cdot 2 \cdot 1} = 2,598,960.$$

Note that this is the same as C(52,47) by symmetry.

The following can be realized by switching the terms in the denominator of C(n,r) and noting that n-(n-r)=r.

Corollary 6

Let n and r be integers with $0 \le r \le n$. Then C(n,r) = C(n,n-r).

A combinatorial proof of an identity is a proof that uses counting arguments to prove that both sides of the identity count the same objects but in different ways (double counting proof), or a proof that is based on showing that there is a bijective between the sets of objects counted by the two sides of the identity (bijective proofs).

For example, we could use a bijective proof to see that C(n,r) = C(n,n-r). Let S be a set with n elements. Define a map f on the power set of S by $f(A) = \overline{A}$. This map is a bijection on $\mathcal{P}(S)$ since f(f(A)) = A. If A has r elements, then \overline{A} has n - r elements.

On the other hand, we could use a double counting proof. Choosing a subset A of S by specifying the elements that are in A is the same as specifying the elements that are not in A. Thus, counting the number of subsets with r elements is the same counting the number of subsets with r elements.

Example. The same school, with 10 math faculty, has 8 computer science faculty. How many ways are there to choose two professors from each committee to be on a discrete math committee.

There are C(10, 2) ways to choose math faculty and C(8, 2) ways to choose computer science faculty. Hence, by the product rule, the total is,

$$C(10,2) \cdot C(8,2) = \frac{10!}{2!8!} \cdot \frac{8!}{2!6!} = 45 \cdot 28 = 1260.$$

6.4 Binomial Theorem

A binomial is an expression of the form x + y. The binomial theorem gives a mechanism for representing powers of this expression. Recall that $\binom{n}{j}$ is the same as C(n, j).

Theorem 7: The Binomial Theorem

Let x and y be variables and let n be a nonnegative integer. Then

$$(x+y)^n = \sum_{j=0}^n \binom{n}{j} x^{n-j} y^j.$$

Writing out the first few terms, we have

$$(x+y)^n = \sum_{j=0}^n \binom{n}{j} x^{n-j} y^j = \binom{n}{0} x^n + \binom{n}{1} x^{n-1} y + \dots + \binom{n}{n-1} x y^{n-1} + \binom{n}{n} y^n.$$

The proof of this theorem is surprisingly simple. We count how many times $x^{n-j}y^j$ shows up in the product

$$(x+y)^n = (x+y)(x+y)\dots(x+y) \quad \text{(n times)}.$$

Notice that a term in the product is obtained by choosing either an x or a y from each binomial factor. Choosing n-j x's from n possible choices is $\binom{n}{n-j}$ and this is equal to $\binom{n}{j}$.

Example. When n=3 we have

$$(x+y)^3 = {3 \choose 0}x^3 + {3 \choose 1}x^2y + {3 \choose 2}xy^2 + {3 \choose 3}y^3$$
$$= x^3 + 3x^2y + 3xy^2 + y^3.$$

We can use this to find a specific coefficient even when writing out the entire expansion is unrealistic.

Example. The coefficient of x^4y^{16} in $(x+y)^{20}$ is

$$\binom{20}{16} = \frac{20!}{16!4!} = 4845.$$

Now consider the coefficient of x^4y^{16} in $(2x-3y)^{20}$. Observe that (2x-3y)=(2x+(-3y)) so the binomial theorem here gives

$$(x+y)^{20} = \sum_{j=0}^{20} {20 \choose j} (2x)^{n-j} (-3y)^j.$$

Hence, the coefficient of x^4y^{16} is

$$\binom{20}{16}(2)^4(-3)^{16} = 3,336,981,811,920.$$

The binomial theorem leads to some other very interesting identities.

Corollary 8

Let n be a nonnegative integer. Then

$$\sum_{j=0}^{n} \binom{n}{j} = 2^{n}.$$

Proof. By the binomial theorem with x = 1 and y = 1 we have

$$2^{n} = (1+1)^{n} = \sum_{j=0}^{n} \binom{n}{j} 1^{n-j} 1^{j} = \sum_{j=0}^{n} \binom{n}{j}.$$

Corollary 9

Let n be a positive integer. Then

$$\sum_{j=0}^{n} (-1)^{j} \binom{n}{j} = 0.$$

Proof. By the binomial theorem with x=1 and y=-1 we have

$$0 = 0^n = (1 + (-1))^n = \sum_{j=0}^n \binom{n}{j} 1^{n-j} (-1)^j = \sum_{j=0}^n \binom{n}{j} (-1)^j.$$

6.5 Generalized Permutations and Combinations

In this section we generalize our counting methods to include repetition.

The next theorem we have already seen. Suppose we want a bit string of length r. Then for each spot we can choose a 0 or a 1. So the total number is 2^r . (Order matters here.)

Theorem 10

The number of r-permutations of a set of n objects with repetition allowed is n^r .

Now we discuss combinations.

Example. Suppose at a school there are at least 3 faculty in Math, CSE, and Stat. How many committees of three faculty can be formed if multiple are allowed from the same department.

Order does not matter here, so we are considering a combination as opposed to a permutation. One way to do this would be to list all of the options, we use M,C,S to refer to the respective departments. By listing we find that there are 10 possibilities:

MMM, MMC, MMS, MCC, MCS, MSS, CCC, CCS, CSS, SSS

Theorem 11

The number of r-combinations from a set with n elements when repetition is allowed is

$$C(n+r-1,r) = C(n+r-1,n-1).$$

Sketch. We are choosing r elements from a set of n elements where order does not matter and repetition is allowed. We represent the r elements by stars *. (For example, 6 elements would have 6 stars: ******.) Since order does not matter, we may assume the stars are in order from the "first" element to the "last element". We put a bar every time there is a switch from one element to another.

(For example, if we have 6 elements from a set with four consisting of two of the first element, one of the second, none of the third, and three of the fourth, we would represent as **|*||***.)

Thus, we actually have n-1+r choices total of where to place stars and bars, but out of these we are choosing r positions for the stars (or, equivalently, choosing n-1 positions for the bars).

Example. If we have a set of 5 (unordered) elements. The number of ways to choose 3 of them with repetition is

$$C(5+3-1,3) = C(7,3) = \frac{7!}{3!4!} = 35.$$

Example. How many solutions does the equation

$$x_1 + x_2 + x_3 + x_4 = 14$$

have, where the x_i are nonnegative integers? How many if the x_i are required to be positive.

One such solution would be 1 + 1 + 1 + 11. It should be clear that counting these directly would be quite exhausting! We should think about this a different way.

Using our "stars-and-bars" idea above, we can think about the stars representing 14 elements. Each placement of 4-1=3 bars then gives a way of adding four numbers to get 14. Obviously order does not matter and repetition is allowed. So the total would be

$$C(4+14-1,4-1) = C(17,3) = \frac{17!}{14!3!} = \frac{17 \cdot 16 \cdot 15}{6} = 680.$$

If the x_i are required to be positive (i.e., $x_i \ge 1$), then we make a replacement. Set $y_i = x_i - 1$ for each i, so that $y_i \ge 0$ for all i. Then

$$y_1 + y_2 + y_3 + y_4 = (x_1 - 1) + (x_2 - 1) + (x_3 - 1) + (x_4 - 1) = (x_1 + x_2 + x_3 + x_4) - 4 = 14 - 4 = 10.$$

So, we have reduced this to a problem very similar to the above. Then we have

$$C(4+10-1,4-1) = \frac{13!}{10!3!} = \frac{13 \cdot 12 \cdot 11}{3 \cdot 2 \cdot 1} = 286.$$

Suppose we have the word SAMPLE. How many strings can we make by reordering the letters? This is a simple problem because all of the letters are distinct. So we are just asking for the number of permutations from the set $\{S, A, M, P, L, E\}$. This is just P(6, 6) = 6! = 720.

Now suppose we have the word SUCCESS. This has repetition. Any order of the three S's would be the same string. We can put the three S's in the seven positions in C(7,3) ways. Subsequently, we can choose the positions for the two C's in C(4,2) ways. Then we have C(2,1) ways to place the U leaving only C(1,1) = 1 way to place the E. This gives a total of

$$C(7,3)C(4,2)C(2,1)C(1,1) = 420.$$

Theorem 12

The number of different permutations of n objects, where there are n_i indistinguishable objects of type k, is

$$\frac{n!}{n_1!n_2!\dots n_k!}.$$

Example. We already know that for one person, there are C(52, 5) ways to deal out a poker hand (because order of cards do not matter). Suppose now that we have four people playing poker. How many ways can we deal hands of 5 cards to the four people. After we have dealt the hand to the first person, then there are only 47 cards left, so there are C(47,5) ways to deal the cards to the second person. Continuing in this way we find that

$$C(52,5)C(47,5)C(42,5)C(37,5) = \frac{52!}{5!5!5!5!32!}$$

ways to deal cards to all of the players.

This example is one of distributing "distinguishable objects" (distinct cards) into "distinguishable boxes" (players' hands). Counting these is essentially the same as counting in the last theorem.

Theorem 13

The number of ways to distribute n distinguishable objects into k distinguishable boxes so that n_i objects are placed into box i is

$$\frac{n!}{n_1!n_2!\dots n_k!}.$$

Using this in the last example we get $\frac{52!}{5!5!5!5!37!}$ notice that the last box is actually the remainder of the deck where we "place" the remainder of the cards.

If the objects themselves are indistinguishable, then this is no different from how we counted above. A set of n-indistinguishable objects is different than a set of n elements. On the other hand, if the objects are distinguishable but the boxes are not, then the solution is much more difficult and there is no simple closed formula.

Example. How many ways are there to put four math professors into three (indistinguishable) offices?

We count. Let $\{A, B, C, D\}$ denote the four professors. One option is to put all four in them all in the same office: $\{\{A, B, C, D\}\}$. Another option is to put one of them alone and the other three together:

$$\{\{A\}, \{B, C, D\}\}, \{\{B\}, \{A, C, D\}\}, \{\{C\}, \{A, B, D\}\}, \{\{D\}, \{A, B, C\}\}.$$

We could also have two in one office and two in another. There are three ways to do this. Or we could put two together and the other two in separate offices. In total, there are fourteen ways to do this.

While you are not required to learn/memorize this formula, the numbers that arise in these problems are called $Stirling\ numbers$ (of the second kind). The number of ways to put n distinguishable objects into j indistinguishable boxes is

$$S(n,j) = \frac{1}{j!} \sum_{i=0}^{j-1} (-1)^i \binom{j}{i} (j-i)^n.$$

Example. How many ways can we put five copies of the same book into three identical boxes?

This is counting *partitions*:

$$5, 4+1, 3+2, 3+1+1, 2+2+1, 2+1+1+1, 1+1+1+1+1+1$$

Relations

9.1 Relations and their properties

Definition: Binary relation

Let A and B be sets. A binary relation from A to B is a subset of $A \times B$.

If $R \subset A \times B$ is a relation, then we use the notation aRb to mean $(a,b) \in R$ and we say "a is related to b". We write $a \not R b$ to mean $(a,b) \notin R$.

For example, we might let A be the set of US capitals and B US states. Then let R to be pairs (a, b) if a is the capital of state b. So (Columbus, Ohio) is in R as is (Frankfort, Kentucky). However, (Indianapolis, Wisconsin) would not be in R.

A function is a special kind of relation. A function is a relation $R \subset A \times B$ if there is some element $(a,b) \in R$ for every $a \in A$, and $(a,b),(a,b') \in R$ implies b=b'. Using function notation we define f by f(a) = b.

Definition: Relation on a set

A relation on a set A is a relation from A to A.

Another way to state the above is that a relation on a set A is a subset of $A \times A$.

Example. Define a relation on the set $A = \{1, 2, 3, 4, 5, 6\}$ by $R = \{(a, b) \mid a \text{ divides } b\}$. Then

$$R = \{(1,1), (1,2), (1,3), (1,4), (1,5), (1,6), (2,2), (2,4), (2,6), (3,3), (3,6), (4,4), (5,5), (6,6)\}$$

We could represent this graphically by making two columns 1-6 and drawing an arrow if elements are related.

Now we can consider special properties of relations.

Definition: Reflexive, symmetric, antisymmetric

Let R be a relation on a set A

- (1) R is called *reflexive* if $(a, a) \in R$ for every element $a \in A$.
- (2) R is called *symmetric* if $(b, a) \in R$ whenever $(a, b) \in R$ for all $a, b \in A$.
- (3) R is called antisymmetric if $(a,b) \in R$ and $(b,a) \in R$ implies a=b for all $a,b \in A$.
- (4) R is called *transitive* if whenever $(a,b) \in R$ and $(b,c) \in R$, then $(a,c) \in R$ for all $a,b,c \in A$.

Can state in terms of quantifiers.

Example. Consider each relation R on the set \mathbb{R} .

- (1) $\{(x,y) \mid x+y=0\}$ This is not reflexive since $(1,1) \notin R$. It is symmetric. It is not antisymmetric since $(1,-1), (-1,1) \in R$ but $1 \neq -1$. It is not transitive.
- (2) $\{(x,y) \mid x = \pm y\}$
- (3) $\{(x,y) \mid x-y \text{ is rational}\}$
- (4) $\{(x,y) \mid xy \ge 0\}$
- (5) $\{(x,y) \mid xy = 0\}$
- (6) $\{(x,y) \mid x=1\}$
- (7) $\{(x,y) \mid x=1 \text{ or } y=1\}$

Let R_1 and R_2 be two relations on $A \times B$. Then we can combine relations using our usual set operations.

Example. Let

$$R_1 = \{(1,2), (2,3), (3,4)\}$$

$$R_2 = \{(1,1), (1,2), (2,1), (2,2), (2,3), (3,1), (3,2), (3,3), (3,4)\}$$

be relations from $\{1,2,3\}$ to $\{1,2,3,4\}$. Determine $R_1 \cup R_2$, $R_1 \cap R_2$, $R_1 \oplus R_2$ (exclusive or), $R_1 - R_2$, and $R_2 - R_1$.

Definition: Composite relation

Let R be a relation from a set A to a set B and S a relation from B to a set C. The *composite* of R and S is the relation consisting of ordered pairs (a,c) where $a \in A$ and $c \in C$, and for which there exists an element $b \in B$ such that $(a,b) \in R$ and $(b,c) \in S$. We denote this composition by $S \circ R$.

Example. Let R be the relation $\{(1,2), (1,3), (2,3), (2,4), (3,1)\}$ and let S be the relation $\{(2,1), (3,1), (3,2), (4,2)\}$ Find $S \circ R$.

9.4 Closure of relations

If R is a relation on a set A, then it may or may not have some property P, such as reflexivity, symmetry, or transitivity. If R does not have a property P, we would like to find the smallest relation containing R that has property P.

Definition: Closure

If R is a relation on a set A, then the *closure* of R with respect to P, if it exists, is the relation S on A with property P that contains R and is a subset of every subset of $A \times A$ containing R with property P.

Example. Let $A = \{1, 2, 3\}$ and consider the relation $R = \{(1, 1), (1, 2), (2, 1), (3, 2)\}$. Then R is not reflexive because it is missing (2, 2) and (3, 3). The reflexive closure of R is

$$\{(1,1),(1,2),(2,1),(2,2),(3,2),(3,3)\}$$

The relation R is not symmetric either because it is missing (2,3). The symmetric closure of R is

$$\{(1,1),(1,2),(2,1),(2,3),(3,2)\}$$

Finally, we note that R is not transitive because it is missing (2,2) and (3,1). The transitive closure of R is

$$\{(1,1),(1,2),(2,1),(2,2),(3,1),(3,2)\}$$

Remember in all of these examples we are obtaining sets and so order of elements is not important.

We will now discuss our strategy for finding these closures.

Let R be a relation on a set A. The diagonal relation on A is

$$\Delta = \{(a, a) \mid a \in A\}.$$

Then the reflexive closure of R on A is $R \cup \Delta$.

The inverse relation of R is

$$R^{-1} = \{ (b, a) \mid (a, b) \in R \}.$$

The symmetric closure is then $R \cup R^{-1}$.

To discuss transitive closure it is useful to first discuss directed graphs related to relations. We can form a directed graph from R by marking all elements of A vertices and drawing an arrow from a to b if $(a,b) \in R$.

(Draw graph for previous relation.)

We denote by R^n the pairs (a, b) such that there is a path of length n from a to b. The connectivity relation R^* consists of the pairs (a, b) such that there is a path of at least one from a to b in R. The transitivity closure of R equals the connectivity relation R^* .

9.5 Equivalence relations

Definition: Equivalence relation, equivalent

A relation on a set A is called an *equivalence relation* if it is reflexive, symmetric, and transitive. Two elements a and b that are related by an equivalence relation are called *equivalent*.

You have undoubtably seen this before, even if it wasn't made explicit. For example, triangle congruence is an equivalence relation, as is triangle similarity. Those who have taken linear algebra have seen matrix similarity, which is also an equivalence relation.

We use the notation $a \sim b$ to indicate that a and b are equivalent relations when R is understood. If it is not clear which relation we are referring to, we may write $a \sim_R b$.

Example. Consider the relation R on \mathbb{Z} defined by aRb if $a=\pm b$. Then R is an equivalence relation. In other notation, we write $a \sim a$ and $a \sim -a$.

Recall that we say that an integer a divides an integer b if b is a multiple of a. We write $a \mid b$ in this situation.

Example. The relation $R = \{(a, b) \mid a \text{ divides } b\}$ is not an equivalence relation since it is not symmetric. For example, $2 \mid 4$ but $4 \nmid 2$.

Example. Let A be a set and let $f: A \to A$ be a function. Define a relation R on A by $(x, y) \in R$ if and only if f(x) = f(y). Then R is an equivalence relation. Two element $x, y \in A$ are equivalent $(x \sim y)$ if they have the same image under f.

Definition: Equivalence class

Let R be an equivalence relation of a set A. The equivalence class of $a \in A$ is the set

$$[a]_R = \{b \in A \mid (a, b) \in R\} = \{b \in A \mid a \sim b\}.$$

Again, we drop the subscript R if the relation is understood. Colloquially, we can define the equivalence class of a as the set of all elements that are related to a. Note that $[a]_R$ is never empty since the relation is reflexive and so $a \sim a$. Also, if $b \in [a]_R$, then by symmetry and transitivity we see that $[a]_R = [b]_R$. We call any element $b \in [a]_R$ a representative of $[a]_R$.

Example. Let m be an integer with m > 1. (If this feels too abstract, just let m = 5 throughout, or your favorite integer.) Define a relation R on \mathbb{Z} by aRb if a - b is divisible by m. We claim R in an equivalence relation.

Let $a \in \mathbb{Z}$. Since a - a = 0 and m divides 0, then aRa. Thus, R is symmetric.

Suppose aRb. Then m divides a-b so a-b=km for some integer k. But them b-a=-(a-b)=(-k)m, so m divides b-a. That is, bRa so R is symmetric.

Finally, suppose aRb and bRc. Then a-b=km and $b-c=\ell m$ for some integers k and ℓ . Then $a-c=(a-b)+(b-c)=(k+\ell)m$, so m divides a-c. That is, aRc so R is transitive. Thus, R is an equivalence relation.

This particular equivalence relation is known as *congruence mod m*. In place of the \sim notation, we write $a \equiv b \pmod{m}$ to indicate that $a \sim b$ under this relation (that is, a - b is a multiple of m). This particular equivalence relation is very important in number theory and abstract algebra.

We can list explicitly the congruence classes of an element under congruence for a particular m.

Example. Let R be congruence mod 5. The equivalence classes are then

$$[0] = {\ldots, -10, -5, 0, 5, 10, \ldots}$$

$$[1] = {\ldots, -9, -4, 1, 6, 11, \ldots}$$

$$[2] = {\ldots, -8, -3, 2, 7, 12, \ldots}$$

$$[3] = {\ldots, -7, -2, 3, 8, 13, \ldots}$$

$$[4] = {\ldots, -6, -1, 4, 8, 14, \ldots}.$$

We observe that every integer belongs to exactly one of these equivalence classes. Another way to express this phenomenon is that *equivalence classes partition a set*.

Theorem 1

Let R be an equivalence relation on a set A and let $a, b \in A$. The following are equivalent:

- (i) $(a,b) \in R$
- (ii) [a] = [b]
- (iii) $[a] \cap [b] \neq \emptyset$.

Now let us summarize what this theorem says. Since R is reflexive, then every element belongs to at least one equivalence class. Since (iii) \Rightarrow (ii), then if [a] and [b] have anything in common, then they are the same. Thus, equivalence classes are either equal or disjoint. Thus, the equivalence classes split the set A into pairwise disjoint subsets. This is known as partitioning sets. While one may define partitions independently, it turns out that every partition defines an equivalence class, so in a sense, partitions and equivalence relations are the same thing.

Now we study another special kind of relation.

Definition: Partial ordering, poset

A relation R on a set S is call a partial ordering (or partial order) if it is reflexive, antisymmetric, and transitive. A set S together with a partial order R is called a partially ordered set (or poset), and is denoted (S, R). Members of S are called elements of the poset.

Example. One of the most recognizable partial orders is \leq on \mathbb{Z} . Since $a \leq a$, then \leq is reflexive. If $a \leq b$ and $b \leq a$, then a = b, so \geq is antisymmetric. Finally, if $a \leq b$ and $b \leq c$, then $a \leq c$, so \leq is transitive. Hence, (\mathbb{Z}, \leq) is a poset.

One can similarly see that "divides" is partial order on \mathbb{Z}^+ .

Example. The inclusion relation \subseteq is a partial ordering on the power set of a set S.

We observe in the last relation that not all elements are comparable. For example, even though $\{1,2\}$ and $\{1,3\}$ are in P(S), it does not make sense to write $\{1,2\} \subseteq \{1,3\}$ or $\{1,3\} \subseteq \{1,2\}$. We use the notation $a \leq b$ to indicate that $(a,b) \in R$.

Definition: Comparable, incomparable

The elements a and b of a poset (S, \preccurlyeq) are called *comparable* if either $a \preccurlyeq b$ or $b \preccurlyeq a$. When a and b are elements such that neither $a \preccurlyeq b$ nor $b \preccurlyeq a$, then a and b are called *incomparable*.

In $(\mathbb{Z}^+, |)$, the numbers 3 and 9 are comparable because 3 | 9 but 5 and 7 are incomparable because $5 \nmid 7$ and $7 \nmid 5$.

Here we might recognize that in the poset (\mathbb{Z}, \leq) , every pair of element is comparable.

Definition: Totally ordered set, total order

If (S, \preceq) is a poset and every two elements of S are comparable, S is called a *totally ordered* set (or *linearly ordered set*), and \preceq is called a *total order* (or *linear order*). A totally ordered set is also called a chain.

Hence, (\mathbb{Z}, \leq) is a totally ordered set.

Definition: Well-ordered set

A poset (S, \preccurlyeq) is a well-ordered set if \preccurlyeq is a total ordering and every nonempty subset of S has a least element.

The set (\mathbb{Z}, \leq) is not well-ordered. However, the set (\mathbb{Z}^+, \leq) is well-ordered.

Example. Define a relation \leq on $\mathbb{Z}^+ \times \mathbb{Z}^+$ by $(a_1, a_2) \leq (b_1, b_2)$ if $a_1 < b_1$ or if $a_1 = b_1$ and $a_2 \leq b_2$. This is known as *lexicographic ordering*. The relation \leq makes $\mathbb{Z}^+ \times \mathbb{Z}^+$ into a well-ordered set.

(Show lattice of $\mathbb{Z}^+ \times \mathbb{Z}^+$ and how to compare points.)

This extends easily to $\mathbb{Z}^+ \times \mathbb{Z}^+ = (\mathbb{Z}^+)^n$. Notice that this is the same type of ordering used in a dictionary.

We can extend the ideal of a lexicographic ordering to any pair of posets. Let (A_1, \preccurlyeq_1) and (A_2, \preccurlyeq_2) be posets. Define a partial order \preccurlyeq on $A_1 \times A_2$ by $(a_1, a_2) \preccurlyeq (b_1, b_2)$ if either $a_1 \preccurlyeq_1 b_1$ or if $a_1 = b_1$ and $a_2 \preccurlyeq_2 b_2$.

A $Hasse\ diagram$ for a poset is the directed graph for the relation R associated to the poset. We drew this previously when considering the transitive closure of a relation. However, we perform some reductions on the diagram to simplify. In particular, we remove any all loops at a vertex, and all arrows that are implied by the transitivity condition. Finally, if we assume that arrows are pointed upwards, then we can remove the direction on the arrows.

(Draw the Hasse diagram for $\{(a,b) \mid a \text{ divides } b\}$ on $\{1,2,3,4,6,8,12\}$.)