

Automated Proof Checking in Introductory Discrete Mathematics Classes

by

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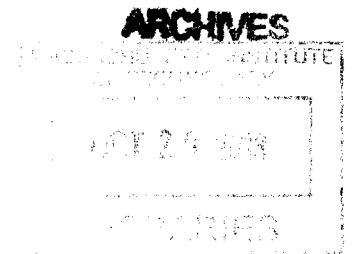
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Abstract

Mathematical rigor is an essential concept to learn in the study of computer science. In the process of learning to write math proofs, instructors are heavily involved in giving feedback about correct and incorrect proofs. Computerized feedback in this area can ease the burden on instructors and help students learn more efficiently. Several software packages exist that can verify proofs written in specific programming languages; these tools have support for some basic topics that undergraduates learn, but not all. In this thesis, we develop libraries and proof automation for introductory combinatorics and probability concepts using Coq, an interactive theorem proving language.

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Chapter 1

Introduction

Students learning about computer science, as in other scientific disciplines, must learn to articulate concepts rigorously and unambiguously. One begins to understand mathematical rigor by writing detailed proofs that consider all possibilities. For students without an extensive background in mathematics, this is a skill that needs to be gained concurrently to learning the concepts of discrete mathematics that are pervasive in computer science. As a beginner, mastering the writing of detailed proofs can be a difficult process. This requires a healthy amount of feedback as to whether the proof steps used are correct and sufficiently rigorous.

When writing code or learning similar engineering concepts, students often have automatic feedback both from compilation or interpretation of their code and from running tests written by instructors to test correctness of implementation. These kinds of feedback can aid students in quickly correcting their mistakes during the problem-solving process without having to wait for an instructor to grade or look at the work. In their mathematics classes, students do not get to enjoy such rapid feedback because exercises involve writing down justified proofs which need to be hand-checked for correctness. We seek to gain the benefits of computerized feedback in introductory undergraduate mathematics classes that teach proof techniques through the use of *automated theorem proving* software. This project focuses on concepts covered in MIT's "Mathematics for Computer Science" class, otherwise known as 6.042, with material developed in [7]. While there are many topics covered in this discrete math class, we focus specifically on combinatorics and probability.

To receive automated feedback about proof writing, students need to be able to write proof steps in a programming language that can express these steps and verify their correctness. Automated and interactive theorem proving software are a class of programming languages and tools used to write down and prove facts about mathematical and program formalizations. These languages are used primarily by specialists to prove results that require heavy amounts of casework and are intractable by hand. The first large-scale example of this is the proof of the Four-Color Theorem [4], a proof which reduced down to 1,936 machine-checked cases. Each theorem prover uses a *formal proof language* to express definitions, proofs, and results. A formal proof language contains a set of valid steps which can be used to derive results from hypotheses in incremental steps. These steps are generally expressed in terms that form the foundation of the language, similar to the axioms of Zermelo-Frenkel set theory in mathematics, which can then be used to define other higher-level concepts. Interactive theorem provers are provers where the software checks user-written proofs that are either written in the formal proof language or in a meta-language that constructs formal proofs. Such a meta-language can enable procedural writing of or searching for proofs.

Our choice of software to adapt for the writing of elementary proofs is Coq, an interactive theorem proving assistant developed by researchers at INRIA in France [8]. Coq is comprised of a base language Gallina, which can be used to express both mathematical definitions and theorems about these definitions, along with a “tactic” language L_{tac} for putting together Gallina proof terms semi-interactively. While Coq has support in its standard library for some areas of mathematical reasoning, such as classical logic and number theory, it does not yet have support for several of the concepts taught in 6.042. We have developed libraries for expressing and proving some standard types of theorems in combinatorics and probability as well as exercises that are typically seen in such an introductory discrete math class. This includes Gallina definitions of basic structures, results about them, and L_{tac} tactics to facilitate proving more related facts.

In addition to improving the process of learning problem-solving steps, we believe that interactive theorem proving can also help expose students to the concept of code verification and can help with grading of course work in online courses. Software verification is becoming increasingly important in industry as software security is becoming a more pervasive con-

cern. Having students learn about tools like Coq that are used in security verification can help them when they begin to tackle these problems. As a relation to another burgeoning area, automated proof-writing feedback is highly applicable in the realm of online education. In online classes with massive enrollment, personalized instructor feedback becomes unmanageable. Hence, automated solutions can play an important role in running these courses.

Chapter 2

Previous Work

Various classes in logic and programming type theory have been taught at other universities using Coq or some other interactive theorem proving language as a teaching assistant. As a popular example of a theorem assistant based on ML, Coq is a natural choice for advanced classes in functional programming and automated theorem proving. Nonetheless, there have also been introductory math and computer science classes taught using Coq. Many of these classes have had some degree of overlapping curricula with the discrete math taught in 6.042, but generally only to the extent that 6.042 incorporates propositional logic and teaches proof techniques. For example, Aaron Stump used Coq as a teaching assistant for classes he taught in 2006 and 2007 at Washington University in logic and discrete math. The classes covered functional programming concepts, Boolean logic, and set theory, but he did find that undergraduates were certainly able to handle the proof assistant software to write their proofs.

Unfortunately, there are many topics taught in 6.042 that do not necessarily have support yet in existing Coq libraries. Some areas, such as set theory, have sufficient implementation in the Coq standard library that only require slight improvement for use on class problem sets. Other broad topics in 6.042 are less covered in Coq libraries. These include, but are not limited to, combinatorics and probability, the two subjects treated in this research.

One investigation into making a Coq software library for teaching students was done by Frédérique Guillot in 2003, who formalized much of high-school level geometry in an effort to possibly use Coq as a teaching assistant in high school for this type of math. The general

conclusion of his paper on this effort [5] was that the Coq library itself was successful for their purposes, but that the interface for students required more work to be easily usable. In fact, a large part of his work went into this type of interface research.

There are currently a few tools for users to interface with Coq above the plain command-line interface: *CoqIDE* (see Chapter 14 of [9]) and *Proof General* [1] are both development environments which simply make it easier to interface with the frontend program *coqtop* by allowing one to perform commands such as stepping through evaluation of proofs. The software *Pcoq* [2] is a more visual-oriented interface which attempts to display formulas and enable interaction with them in a more “natural” way. Guilhot’s work involved an extension of *Pcoq* called *GeoView* for visualizing geometric statements. This line of work will likely become relevant when we begin to look at what kind of interface students will use to write their proofs using the libraries to be developed, though this step may be beyond the scope of my project.

More recently, the computer science department at Northeastern University has undertaken a field test [3] of using an interactive proof assistant in an introductory Symbolic Logic class. The software used is not Coq but rather ACL2 [6]. Regardless, the approach is still relevant to this research. While the teaching with ACL2 was popular among the six students in the trial, the instructors indeed found that sometimes generating proofs in ACL2 can be more difficult than seems to be warranted by the complexity of a given problem. One of their conclusions is particularly relevant to designing 6.042 assignments in Coq: “The course will also benefit from a large canon of carefully designed exercises that demonstrate proof principles while avoiding ACL2 subtleties along the way, such as perplexing failure output or the need to use mysterious, instructor-supplied ‘hints’.”

Chapter 3

Combinatorics

Combinatorics is a core topic that is foundational to many parts of discrete mathematics and computer science, such as probability and algorithmic analysis. Students study how to work with cardinalities and how to formulate *counting arguments*, justifications for cardinalities of sets. Proofs in typical exercises or assignments in combinatorics routinely involve counting the number of objects that satisfy some specific property. Students learn tools to simplify this task into counting sets that are easier to reason about.

3.1 Counting Example

The following exercise is an example of a simple combinatorics exercise and how a student might be expected to answer it.

Exercise 3.1.1. *Calculate the number of poker hands containing a full house.*

Proof. There is a bijection between poker hands with a full house and sequences of the form (r, s, r', s') , where

- r is the shared rank of cards in the triple
- s is the set of suits of the cards in the triple
- r' is the shared rank of cards in the pair

- s' is the set of suits of the cards in the pair

For example, we have the correspondence

$$(5\Diamond, 5\Heart, 5\Spade, 8\Club, 8\Heart) \leftrightarrow (5, \{\Diamond, \Heart, \Spade\}, 8, \{\Club, \Heart\}).$$

There are 13×12 ways to choose the ranks r and r' since they cannot be equal, and there are $\binom{4}{3}$ and $\binom{4}{2}$ ways to choose the suit subsets s and s' , respectively. Hence there are

$$13 \times \binom{4}{3} \times 12 \times \binom{4}{2} = 3744$$

possible hands with full houses. □

Underlying this proof are several basic facts about cardinalities of sets and how they can be related. The style of this proof is common to many counting arguments: in order to count the size of a set, exhibit a bijection between the set and a set that is more easily counted. Though this proof does not mention it explicitly, it makes use of the product rule.

Theorem 3.1.2. (Product Rule) *For finite sets A and B , $|A \times B| = |A| \times |B|$.*

The set of sequences (r, s, r', s') can be interpreted as the cartesian product $R^2 \times S_3 \times S_2$, where R^2 is the set of ordered pairs of distinct ranks, and S_3 and S_2 are, respectively, the sets of 3- and 2-element subsets of the set of suits. The result then follows from repeated application of the product rule.

An alternative way to think of Example 3.1.1 is using an iterative proof, refining the set in question through one choice at each step. The following result is a variation of the product rule.

Theorem 3.1.3. *For finite sets A and B and a mapping $f: A \rightarrow B$, if there exists k such that $k = |f^{-1}(b)|$ for all $b \in B$, then $|A| = k \times |B|$.*

We call such a mapping f a *k-to-1* function. Each choice in the above example can be phrased as an application of this principle. Let H be the set of poker hands with full houses, R be the set of ranks, and $f: H \rightarrow R$ be projection onto the rank shared among the triple. Then it is relatively simple to see that $|f^{-1}(r)|$ is the same regardless of the choice of $r \in R$.

We choose an arbitrary r , and then we have that $|H| = |f^{-1}(r)| \times |R|$. This reduces the problem to counting $f^{-1}(r)$, which we do in a similar fashion. Eventually, we reduce the set to the point that the free parameters are all specified and we have a set of cardinality 1.

We have found this iterative solution to be more amenable to making a concise machine-checkable proof. Here is how the proof of Example 3.1.1 is written with our Coq library.

```
Theorem full_houses : compute_size { S : Ensemble Card |
  ex_set S 5 (rank' $0 = rank' $1 ∧ rank' $0 = rank' $2 ∧
    rank' $3 = rank' $4 ∧ rank' $0 ≠ rank' $3) }.
```

To begin the proof, we have a description for what it means for a hand to have a full house. There is special notation in use here for functions and predicates of multiple arguments, including the backtick characters, which we define later in Section 3.4. The predicate `ex_set` stipulates that S is a set of 5 elements, which we can refer to as \$0 through \$4. To have a full house, three elements must have matching rank projections, and the other two elements must have a different matching rank.

Proof.

```
start_counting (card_hint (3 :: 2 :: nil)).
```

We begin with a hint to our proof environment that as we have specified it above, the hand of five cards breaks down a set of three, {\$0, \$1, \$2}, and a set of two, {\$3, \$4}. These subsets may have their elements permuted, but they will not be equal to elements of the other sets.

```
pick r as (5 ↗ rank' $0).
```

First, we pick r to be the rank of the first element, which is the rank shared by the elements in the triple. The notation $(5 \nrightarrow _)$ helps the inferencer parse this code but is not content important to the proof. This step invokes Theorem 3.1.3 with f as the specified function (rank of the first card) and B as the inferred output type of the projection (in this case **Rank**). The proof goal is then reduced to counting the cardinality k in the theorem, which here is the set of hands with full houses such that r , now assumed to be constant, is the rank of the triple. The notation `pick v as f [from T]`, after applying Theorem 3.1.3 with appropriate values, automatically tries to dispatch the supporting steps that justify its use. This procedure and the other tactics are explored in more detail in Section 3.3.

pick r' as $(5 \nrightarrow \text{rank } \$3)$ from $\{ r' \mid r' \neq r \}$.

Second, we pick r' to be the rank of the fourth element, equal to r' above, that we say comes from the set of ranks that differ from r . This form uses the specified set $\{r' \mid r' \neq r\}$ as the set B in Theorem 3.1.3.

*pick s as $(5 \nrightarrow \text{set} [\text{suit } \$0; \text{suit } \$1; \text{suit } \$2]) \text{ from } (\text{size_subset } \mathbf{Suit } 3).$
*pick s' as $(5 \nrightarrow \text{set} [\text{suit } \$3; \text{suit } \$4]) \text{ from } (\text{size_subset } \mathbf{Suit } 2).$**

We can then refine s and s' to be the sets of suits as above. After we have fully specified all of these parameters, the set we're trying to count is reduced to a one-element set. We must then write down what the fully specified element looks like, in terms of the variables we've fixed in the previous steps. However, our hypotheses do not have all the names we need. They are currently

$r : \mathbf{Rank}$
 $r' : \{ r' : \mathbf{Rank} \mid r' \neq r \}$
 $s : \text{size_subset } \mathbf{Suit } 3$
 $s' : \text{size_subset } \mathbf{Suit } 2$

We run the following tactic to produce names for the elements of the sets s and s' .

name_all.

We now have automatically generated names for these elements.

$r : \mathbf{Rank}$
 $r' : \mathbf{Rank}$
 $H : r' \neq r$
 $s : \mathbf{Ensemble } \mathbf{Suit}$
 $s' : \mathbf{Ensemble } \mathbf{Suit}$
 $s'0 : \mathbf{Suit}$
 $s'1 : \mathbf{Suit}$
 $H1 : s'0 :: s'1 :: \text{nil enumerates } s'$
 $s0 : \mathbf{Suit}$
 $s1 : \mathbf{Suit}$
 $s2 : \mathbf{Suit}$
 $H2 : s0 :: s1 :: s2 :: \text{nil enumerates } s$

We finally supply the unique set of Cards that works for these fixed variables.

unique $((r, s0) :: (r, s1) :: (r, s2) :: (r', s'0) :: (r', s'1) :: \text{nil}).$
Defined.

And this completes the proof. We can verify that we get exactly the same product as above.

Eval simpl in (proj1_sig full_houses).

$$= 1 \times 6 \times 4 \times 12 \times 13$$

: cardinality

To develop the tools used in this type of proof, we start with a formalization of the basics necessary in combinatorics.

3.2 Combinatorics Foundations

Coq has strong library support for fundamental mathematical concepts such as logic, numbers (including natural numbers, integers, rationals, and reals), and set theory. The course content of 6.042 discusses mathematics using naive set theory, which does not specify restrictions for writing down sets. As such, we let any Coq object living in `Type` signify a set. Subsets are implemented using `Ensemble` and `sig`, and are described as a predicate over a type. In this section, we lay out foundational library support for concepts presented in the combinatorics section of 6.042.

3.2.1 Functions and Relations

In order not to burden ourselves with the restriction of making all functions computable, we consider functions in the set-theoretic sense that they are a specific kind of relation between two sets. In Coq, we represent relations as follows.

Section Functions.

Variable $U \ V : \text{Type}$.
 Variable $f : U \rightarrow V \rightarrow \text{Prop}$.

The basic properties of relations are straightforward to define.

Definition function := $\forall x \ y \ y', f \ x \ y \rightarrow f \ x \ y' \rightarrow y = y'$.
 Definition injective := $\forall x \ x' \ y, f \ x \ y \rightarrow f \ x' \ y \rightarrow x = x'$.

Definition surjective := $\forall (v : V), \exists (u : U), f\ u\ v$.

Definition total := $\forall (u : U), \exists (v : V), f\ u\ v$.

We give names to various combinations of these properties. The familiar concept of surjective and injective functions is given by `surjective_fn` and `injective_fn`, respectively, but the leaner definitions of `surjector` and `injector` are actually dual to each other and are more useful in defining the relations between types that we use further on. A function is bijective if it is both surjective and injective.

Definition surjector := function \wedge surjective.

Definition injector := injective \wedge total.

Definition surjective_fn := function \wedge surjective \wedge total.

Definition injective_fn := function \wedge injective \wedge total.

Definition bijective := function \wedge injective \wedge surjective \wedge total.

The next lemmas illustrate dependencies between these composite properties.

Lemma bijective_surjector : bijective \rightarrow surjector.

Lemma bijective_injector : bijective \rightarrow injector.

Lemma bijective_surjective_fn : bijective \rightarrow surjective_fn.

Lemma bijective_injective_fn : bijective \rightarrow injective_fn.

Lemma surjector_injector_bijective : surjector \rightarrow injector \rightarrow bijective.

Lemma surjective_injective_bijective : surjective_fn \rightarrow injective_fn \rightarrow bijective.

End Functions.

The inverse of a relation is also easy to define. We prove basic results about inverses, such as the dualities between total and surjective and between function and injective that are apparent from their respective definitions.

Section Inverses.

Variable $U\ V : \text{Type}$.

Variable $f : U \rightarrow V \rightarrow \text{Prop}$.

Definition inverse : $V \rightarrow U \rightarrow \text{Prop} := (\text{fun } v\ u \Rightarrow f\ u\ v)$.

Theorem total_iff_inverse_surjective : total $f \leftrightarrow$ surjective inverse.

Theorem surjective_iff_inverse_total : surjective $f \leftrightarrow$ total inverse.

Theorem function_iff_inverse_injective : function $f \leftrightarrow$ injective inverse.

Theorem injective_iff_inverse_function : injective $f \leftrightarrow$ function inverse.

Theorem surjector_iff_inverse_injector : surjector $f \leftrightarrow$ injector inverse.

Theorem injector_iff_inverse_surjector : injector $f \leftrightarrow$ surjector inverse.

Theorem bijective_inverse : bijective $f \leftrightarrow$ bijective inverse.

End Inverses.

The composition of two functions is straightforward to define even when the functions are expressed as relations. We prove that the composition of two bijective functions is bijective, and likewise for the surjective and injective properties.

Section Compositions.

Variable $U\ V\ W : \text{Type}$.

Variable $f : U \rightarrow V \rightarrow \text{Prop}$.

Variable $g : V \rightarrow W \rightarrow \text{Prop}$.

Definition $\text{composition} : U \rightarrow W \rightarrow \text{Prop} := (\text{fun } u\ w \Rightarrow \exists v, f\ u\ v \wedge g\ v\ w)$.

Theorem $\text{bijective_composition} : \text{bijective } f \rightarrow \text{bijective } g \rightarrow \text{bijective composition}$.

Theorem $\text{injector_injector_composition} : \text{injector } f \rightarrow \text{injector } g \rightarrow \text{injector composition}$.

Theorem $\text{injective_injective_composition} :$

$\text{injective_fn } f \rightarrow \text{injective_fn } g \rightarrow \text{injective_fn composition}$.

Theorem $\text{surjector_surjector_composition} :$

$\text{surjector } f \rightarrow \text{surjector } g \rightarrow \text{surjector composition}$.

Theorem $\text{surjective_surjective_composition} :$

$\text{surjective_fn } f \rightarrow \text{surjective_fn } g \rightarrow \text{surjective_fn composition}$.

End Compositions.

We introduce a conversion from computable functions $U \rightarrow V$ to relations $U \rightarrow V \rightarrow \text{Prop}$. This makes use of an Inductive proposition, which uses constructors as the possible “proofs” of the proposition.

Section ComputableFunctions.

Variable $U\ V : \text{Type}$.

Variable $f : U \rightarrow V$.

Inductive $\text{relationize} : U \rightarrow V \rightarrow \text{Prop} :=$

$\text{relation_intro} : \forall u, \text{relationize } u\ (f\ u)$.

End ComputableFunctions.

Following the definitions in the 6.042 text [7], we define three relations between sets based on whether there exist surjective, injective, or bijective functions between them.

Section Relations.

Variable $U\ V : \text{Type}$.

Definition $\text{surj} := \exists f : U \rightarrow V \rightarrow \text{Prop}, \text{surjector } f$.

Definition $\text{inj} := \exists f : U \rightarrow V \rightarrow \text{Prop}, \text{injector } f$.

Definition $\text{bij} := \exists f : U \rightarrow V \rightarrow \text{Prop}, \text{bijective } f$.

Lemma $\text{surj_fn} : (\exists f : U \rightarrow V \rightarrow \text{Prop}, \text{surjective_fn } f) \rightarrow \text{surj}$.

Lemma inj_fn : ($\exists f : U \rightarrow V \rightarrow \text{Prop}$, injective_fn f) \rightarrow inj.

End Relations.

The relation `bij` is so integral to the discussion of cardinality that we abbreviate it with the following symbol. We also prove that it is a setoid (an equivalence relation) and register it as a setoid with Coq so that we may use some more elementary tactics with bijections, such as `symmetry`.

Infix " \approx " := `bij` (at level 70, no associativity) : *cardinal_scope*.

Section BijSetoid.

Theorem `bij_reflexive` : `reflexive _ bij`.
 Theorem `bij_symmetric` : `symmetric _ bij`.
 Theorem `bij_transitive` : `transitive _ bij`.
 Theorem `bij_setoid` : `Setoid_Theory _ bij`.

End BijSetoid.

Add *Setoid* Type `bij` `bij_setoid` as *bij_equiv*.

We end this section with several results relating `surj`, `inj`, and `bij`. The theorem `surj_inj_bij` below is nontrivial and is often referred to as the Schröder–Bernstein Theorem or the Cantor–Bernstein–Schröder Theorem. The last theorem is also nontrivial because we invoke the axiom of choice to be able to pull out an injection from the inverse of a surjection.

Section RelationResults.

Lemma `bij_surj` ($U\ V : \text{Type}$) : $U \approx V \rightarrow \text{surj } U\ V$.
 Lemma `bij_inj` ($U\ V : \text{Type}$) : $U \approx V \rightarrow \text{inj } U\ V$.
 Theorem `surj_inj_symmetric` ($U\ V : \text{Type}$) : $\text{surj } U\ V \leftrightarrow \text{inj } V\ U$.
 Theorem `surj_inj_bij` ($U\ V : \text{Type}$) : $\text{surj } U\ V \rightarrow \text{inj } U\ V \rightarrow U \approx V$.
 Corollary `inj_inj_bij` ($U\ V : \text{Type}$) : $\text{inj } U\ V \rightarrow \text{inj } V\ U \rightarrow U \approx V$.
 Corollary `surj_surj_bij` ($U\ V : \text{Type}$) : $\text{surj } U\ V \rightarrow \text{surj } V\ U \rightarrow U \approx V$.
 Theorem `inj_transitive` ($U\ V\ W : \text{Type}$) : $\text{inj } U\ V \rightarrow \text{inj } V\ W \rightarrow \text{inj } U\ W$.
 Theorem `surj_transitive` ($U\ V\ W : \text{Type}$) : $\text{surj } U\ V \rightarrow \text{surj } V\ W \rightarrow \text{surj } U\ W$.
 Theorem `injective_fn_from_surjective_fn` ($U\ V : \text{Type}$) :
 ($\exists f : U \rightarrow V \rightarrow \text{Prop}$, `surjective_fn` f) \rightarrow
 $\exists g : V \rightarrow U \rightarrow \text{Prop}$, `injective_fn` g .

End RelationResults.

3.2.2 Cardinality

The definition of cardinality for sets is integral in the discussion of combinatorics, as all things that we would like to count are expressed as sets mathematically. While for finite sets cardinality can be identified with the number of elements in a set, the definition of cardinality for infinite sets requires the concept of a bijective relationship. When phrased as facts about sets with bijections between them, proofs of cardinality laws hold equally well for finite and infinite cardinalities.

We define cardinality as a predicate on sets such that all sets that satisfy this predicate have bijections between them and the predicate is closed under bijectivity. For convenience for some later proofs, we also require that there be some set satisfying the predicate. The result is the following record type.

```
Record cardinality := {
  right_size :> Type → Prop;
  existence : ∃ S, right_size S;
  all_bij : ∀ U V, right_size U → right_size V → U ≈ V;
  closed_under_bij : ∀ U V, right_size U → U ≈ V → right_size V
}.
```

Assuming proof irrelevance, to prove equality between cardinalities it suffices to show that the `right_size` predicates match.

Lemma `cardinality_eq (c1 c2 : cardinality) : right_size c1 = right_size c2 → c1 = c2`.

It is straightforward to write a function that builds a cardinality out of a set by partially applying the relation `bij` to the given set. Proofs of closure follow from transitivity of `bij`.

Definition `cardinality_of (A : Type) : cardinality`.

```
  refine {| right_size := bij A |}; intros.
  ∃ A. auto.
  apply bij_transitive with A; auto.
  apply bij_transitive with U; auto.
Defined.
```

Following mathematical notation, we use $|A|$ to denote the cardinality of the set A .

Notation `"| A |" := (cardinality_of A) (at level 30) : cardinal_scope`.

Because of the closure law, two sets have the same cardinality if and only if they have a bijection between them.

Lemma `cardeq_bij` $A\ B : |A| = |B| \leftrightarrow A \approx B$.

An example of a cardinality that we can write down now is that of the empty set.

Definition `zero_cardinal` $:= |\mathbf{Empty_set}|$.

The general definition of inequalities between cardinals depends on the `inj` and `surj` relations defined earlier. We say that $|A| \leq |B|$ if and only if there is an injection from A to B . This yields the following definition on Coq cardinalities.

Definition `cardinal_le` ($c1\ c2 : \mathbf{cardinality}$) $:=$
 $\exists (A : \mathbf{Type}),\ c1\ A \wedge \exists (B : \mathbf{Type}),\ c2\ B \wedge \mathbf{inj}\ A\ B$.

This definition of \leq obeys the standard reflexivity and transitivity laws.

Lemma `cardinal_le_reflexive` : `reflexive _ cardinal_le`.

Lemma `cardinal_le_transitive` : `transitive _ cardinal_le`.

The other inequality relations can be defined in terms of \leq .

Definition `cardinal_ge` ($c1\ c2 : \mathbf{cardinality}$) $:= c2 \leq c1$.

Definition `cardinal_lt` ($c1\ c2 : \mathbf{cardinality}$) $:= c1 \leq c2 \wedge c1 \neq c2$.

Definition `cardinality_gt` ($c1\ c2 : \mathbf{cardinality}$) $:= c2 < c1$.

For cardinals taken from specific sets A and B , we show that \leq and \geq are equivalent to the relations `inj` and `surj`.

Theorem `cardinal_le_inj` $A\ B : |A| \leq |B| \leftrightarrow \mathbf{inj}\ A\ B$.

Theorem `cardinal_ge_surj` $A\ B : |A| \geq |B| \leftrightarrow \mathbf{surj}\ A\ B$.

3.2.3 Finite Cardinalities

Nearly all cardinalities that we deal with in combinatorics are those of finite sets. We define an injection from natural numbers to cardinalities using the “interval” of natural numbers $\{x : \mathbf{nat} \mid x < n\}$ as the prototypical n -element set. We first define shorthand notation for these finite intervals.

Definition `interval` ($n : \mathbf{nat}$) $:= \{x : \mathbf{nat} \mid x < n\}$.

Notation `"[0 - n)"` $:= (\mathbf{interval}\ n)$.

Notation `"{ 0 , 1 }"` $:= (\mathbf{interval}\ 2)$.

Definition `interval2` ($m\ n : \mathbf{nat}$) $:= \{x : \mathbf{nat} \mid m \leq x < n\}$.

Notation `"[m - n)"` $:= (\mathbf{interval2}\ m\ n)$.

We define `n_size` to be the predicate for whether a `Type` is finite with exactly `n` elements.

```
Fixpoint n_size (n : nat) (T : Type) : Prop :=
  match n with
  | 0 => T → False
  | S n' => ∃ x : T, n_size n' { x' : T | x' ≠ x }
  end.
```

We prove that `n_size` is closed under *bij*.

```
Theorem n_size_bij (n : nat) : ∀ (s1 s2 : Type), n_size n s1 → n_size n s2 → s1 ≈ s2.
Theorem n_size_closed (n : nat) : ∀ (U V : Type), n_size n U → U ≈ V → n_size n V.
```

The intervals defined above satisfy `n_size` predicates.

```
Theorem n_interval_size (n : nat) : n_size n [0, ..., n).
Theorem n_interval2_size (a b : nat) : n_size (b - a) [a, ..., b).
```

We package together the `n_size` predicate with its proofs of closure into `cardinality_n`, the injection from `nat` to `cardinality`.

```
Definition cardinality_n (n : nat) : cardinality.
  refine ({| right_size := n_size n;
             all_bij := n_size_bij n;
             closed_under_bij := n_size_closed n |}).
  ∃ [0, ..., n).
  auto.
Defined.
Coercion cardinality_n : nat → cardinality.
```

We have several easy lemmas about sets we already know that are equal to `cardinality_n` of various numbers.

```
Lemma empty_set_cardinality : zero_cardinal = 0.
Lemma interval_cardinality (n : nat) : |[0, ..., n)| = n.
Lemma interval2_cardinality (a b n : nat) : |[a, ..., b)| = b - a.
```

To show that `cardinality_n` is injective, we show inductively that there cannot be any bijection from interval *m* to interval *n* if *m* ≠ *n*. The core of this proof is to show the following lemma, constructing a bijection $\{0, \dots, a-1\} \leftrightarrow \{0, \dots, b-1\}$ out of a bijection $\{0, \dots, a\} \leftrightarrow \{0, \dots, b\}$. This is done by tying together the element that maps from *a* and the element that maps to *b*.

`Lemma interval_bijection_peel (a b : nat) :`
`interval (S a) \approx interval (S b) \rightarrow interval a \approx interval b.`
`Lemma interval_bijection_ineq (m n : nat) : m < n \rightarrow interval m $\not\approx$ interval n.`
`Theorem interval_bijection_eq (m n : nat) : interval m \approx interval n \rightarrow m = n.`
`Corollary cardinality_n_equality (m n : nat) :`
`cardinality_n m = cardinality_n n \rightarrow m = n.`

The inclusion map $i: \{0, \dots, m-1\} \hookrightarrow \{0, \dots, n-1\}$ for $m \leq n$ gives us a quick proof of the next fact.

`Theorem cardinality_n_le (m n : nat) : m \leq n \rightarrow (m \leq n)%cardinal.`

It follows that $<$ between naturals also carries over to $<$ between cardinalities.

`Theorem cardinality_n_lt (m n : nat) : m < n \rightarrow (m < n)%cardinal.`

A simple proof that a set has a certain cardinality is to exhaustively enumerate all of its elements. The following section implements this proof strategy.

Section EnumeratedTypes.

`Variable S : Type.`
`Variable enum : list S.`
`Hypothesis enum_NoDup : NoDup enum.`
`Hypothesis enum_total : $\forall (s : S), \text{In } s \text{ enum}.$`

We define the bijection between `interval (length enum)` and S using the `nth_error` function for indexing into lists.

`Definition index_enumeration (n : interval (length enum)) (s : S) :=`
`match nth_error enum (proj1_sig n) with`
`| Some x \Rightarrow x = s`
`| None \Rightarrow False`
`end.`

`Lemma index_enumeration_bijective : bijective index_enumeration.`

`Theorem enumerated_type : |S| = length enum.`

`End EnumeratedTypes.`

We can then write an L_{tac} function to use the above theorem and prove some simple examples.

`Ltac enumeration l :=`
`apply enumerated_type with (enum := l);`
`[repeat (constructor; try (simpl; intuition; discriminate)) |`
`let x := fresh in intro x; destruct x; simpl; tauto].`

Theorem unit_sz : $|\mathbf{unit}| = 1$.

enumeration (tt :: nil).

Qed.

Lemma bool_sz : $|\mathbf{bool}| = 2$.

enumeration (true :: false :: nil).

Qed.

3.2.4 Product and Sum of Cardinalities

We define the product of two cardinalities to be the following predicate.

Definition splits_as_product (c1 c2 : **cardinality**) (S : Type) :=

$\exists A : \text{Type}, \exists B : \text{Type}, S \approx (A \times B) \wedge c1\ A \wedge c2\ B$.

The \times operator on types is defined to be the Cartesian product of the two types. We see here that the set S satisfies `splits_as_product c1 c2` if it can be decomposed as the product of two sets whose cardinalities match $c1$ and $c2$. As with other cardinalities, we prove that this predicate is closed under bijections.

Lemma bij_product (A B C D : Type) : $A \approx B \rightarrow C \approx D \rightarrow A \times C \approx B \times D$.

Lemma product_cardinality_all_bij (c1 c2 : **cardinality**) (A B : Type) :

`splits_as_product c1 c2 A` \rightarrow `splits_as_product c1 c2 B` $\rightarrow A \approx B$.

Lemma product_cardinality_bij_transitive (c1 c2 : **cardinality**) (A B : Type) :

`splits_as_product c1 c2 A` $\rightarrow A \approx B \rightarrow$ `splits_as_product c1 c2 B`.

Definition product_cardinality (c1 c2 : **cardinality**) : **cardinality**.

refine { | right_size := `splits_as_product c1 c2`;

all_bij := `@product_cardinality_all_bij c1 c2`;

closed_under_bij := `@product_cardinality_bij_transitive c1 c2` | }.

...

Defined.

Infix " \times " := `product_cardinality` : *cardinal_scope*.

We can show that the product cardinality behaves as we would expect on the Cartesian product of sets. It also follows the symmetry and associativity laws that multiplication on numbers follows.

Theorem product_rule A B : $|A \times B| = |A| \times |B|$.

Theorem product_symmetric (c0 c1 : **cardinality**) : $c0 \times c1 = c1 \times c0$.

Theorem product_associative (c0 c1 c2 : **cardinality**) : $c0 \times (c1 \times c2) = (c0 \times c1) \times c2$.

Beyond following these product laws, `product_cardinality` is compatible with the product of natural numbers for cardinalities of finite sets. We show here that there is a bijection $f: \{0, \dots, m-1\} \times \{0, \dots, n-1\} \rightarrow \{0, \dots, (m-1)(n-1)\}$ defined by

$$f(a, b) = a + mb.$$

Definition `interval_product_map` ($m\ n : \mathbf{nat}$) ($x : \text{interval } m \times \text{interval } n$)
 $(y : \text{interval } (m \times n)) := (\text{proj1_sig } (\text{fst } x) + m \times \text{proj1_sig } (\text{snd } x)) \% \text{nat} = \text{proj1_sig } y.$

Lemma `interval_product_bijective` : $\forall\ m\ n$, `bijective (@interval_product_map m n)`.

Theorem `nat_product_cardinality` ($m\ n : \mathbf{nat}$) :
`cardinality_n m × cardinality_n n = cardinality_n (m × n).`

With the product cardinality defined, we can prove Theorem 3.1.3 pertaining to sets related by *k-to-1* functions. Recall that a *k-to-1* function is defined to be a function $f: U \rightarrow V$ such that for each $v \in V$, there are exactly k elements in U which map to v . We write this definition in Coq again using relations as functions.

Section Division.

Variable $U\ V : \text{Type}.$

Variable $f : U \rightarrow V \rightarrow \text{Prop}.$

Definition `k_to_1` $k := \text{total } f \wedge \text{function } f \wedge \forall\ v : V, |\{ u : U \mid f\ u\ v \}| = k.$

The theorem applies the product rule using V and a set K whose cardinality is k . We use the axiom of choice (in a dependently typed version) to instantiate the bijection between K and $\{ u : U \mid f\ u\ v \}$ for each v in V .

Theorem `division_rule` $k : \text{k_to_1 } k \rightarrow |U| = k \times |V|.$

End Division.

We can use this division rule to prove a generalized version of the product rule using dependently typed pairs rather than Cartesian product. Given a function $f : T \rightarrow \text{Type}$ that associates a `Type` to every element in T , the type `sigT f` is the type of pairs whose first element is some t in T and whose second element is of type $f\ t$. An alternative notation for `sigT` is similar to that for normal sigma types: $\{ t : T \ \& \ f\ t \}$.

Section SigTProdRule.

Variable $T : \text{Type}.$

```

Variable  $f : T \rightarrow \text{Type}$ .
Variable  $k : \text{cardinality}$ .
Hypothesis regular :  $\forall t : T, k = |f\ t|$ .
Theorem sigT_prod_rule :  $|\{ t : T \ \& \ f\ t \}| = k \times |T|$ .
End SigTProdRule.

```

This dependently typed variation leads us to define the version of the product rule which is actually used in the proof of the poker hand example at the beginning of the section. Given a subset of a type T as a predicate $P : T \rightarrow \text{Prop}$ we can use a uniform projection function Q to refine the cardinality of $\{t : T \mid P\ t\}$ into a product $|P'| \times |V|$, where P' is the preimage of a specific value under Q and V is the set of possible values for Q . In the first step of the full houses example, we use a projection function that maps a hand with a full house to the rank of the triple in that hand. Naming this function $r : H \rightarrow R$ and choosing a specific \bar{r} , the set of full house hands then has cardinality

$$|H| = |\{h \in H \mid r(h) = \bar{r}\}| \times |R|.$$

This refines the problem because we know the size $|R|$ by definition. In general, we can use any k -to-1 function in place of r .

We also have an auxiliary definition here which is the restriction of a function to a subset of its input.

```

Definition restrict  $T\ V\ (f : T \rightarrow V \rightarrow \text{Prop})\ (S : T \rightarrow \text{Prop}) :=$ 
  fun ( $t : \text{sig } S$ ) ( $v : V$ )  $\Rightarrow f\ (\text{proj1\_sig } t)\ v$ .
Notation " $(Q \mid P)$ " := ( $@\text{restrict } \_ \_ Q\ P$ ).
Section GeneralProductRule.

```

```

Variable  $T\ V : \text{Type}$ .
Variable  $P : T \rightarrow \text{Prop}$ .
Variable  $Q : T \rightarrow V \rightarrow \text{Prop}$ .

```

For convenience we let the projection function Q be defined in terms of the base set, but we require that it be only defined on the subset P and be a total function on that set.

```

Hypothesis Q_total_function : function  $(Q \mid P) \wedge \text{total } (Q \mid P)$ .
Hypothesis Q_defined_on_P :  $\forall t\ v, Q\ t\ v \rightarrow P\ t$ .

```

First, we show that the set of pairs (v, t) where $t \in Q^{-1}(v)$ is in bijection with the elements of P . This bijection is simply $(Q(t), t) \leftrightarrow t$.

Theorem projection_size : $\{v : V \ \& \ \{t : T \mid Q \ t \ v\}\} \approx \{t : T \mid P \ t\}$.

Corollary projection_cardinality :

$|\{v : V \ \& \ \{t : T \mid Q \ t \ v\}\}| = |\{t : T \mid P \ t\}|$.

Variable k : **cardinality**.

Hypothesis gen_regular : $\forall v : V, k = |\{t : T \mid Q \ t \ v\}|$.

Given that Q is k -to-1, we can apply `sigT_prod_rule` to find that the set of pairs (v, t) above has cardinality $k|V|$. This along with `projection_cardinality` gives our final rule. The reason that we don't use the earlier `k_to_1` definition here is because we are phrasing Q as a function of T rather than **sig** T , so the entire Q is not actually a total function. Avoiding the sigma type helps in reducing the boilerplate sigma projections and constructor uses needed to apply this rule.

Lemma gen_product_rule_pair : $|\{v : V \ \& \ \{t : T \mid Q \ t \ v\}\}| = k \times |V|$.

Theorem gen_product_rule : $|\{t : T \mid P \ t\}| = k \times |V|$.

End GeneralProductRule.

The sum of two cardinalities is defined analogously to product, using disjoint sum of sets rather than Cartesian product. Unsurprisingly, the $+$ operator on types in Coq is already defined to be the disjoint sum. We use the following predicate on types.

Definition splits_as_sum ($c1 \ c2$: **cardinality**) (S : Type) :=

$\exists s1 : \text{Type}, \exists s2 : \text{Type}, S \approx (s1 + s2) \wedge c1 \ s1 \wedge c2 \ s2$.

With this, we can define `sum_cardinality` as we have done with `product_cardinality`.

Lemma bij_sum ($A \ B \ C \ D$: Type) : $A \approx B \rightarrow C \approx D \rightarrow A + C \approx B + D$.

Lemma sum_cardinality_all_bij ($c1 \ c2$: **cardinality**) ($A \ B$: Type) :

$\text{splits_as_sum } c1 \ c2 \ A \rightarrow \text{splits_as_sum } c1 \ c2 \ B \rightarrow A \approx B$.

Lemma sum_cardinality_bij_transitive ($c1 \ c2$: **cardinality**) ($A \ B$: Type) :

$\text{splits_as_sum } c1 \ c2 \ A \rightarrow A \approx B \rightarrow \text{splits_as_sum } c1 \ c2 \ B$.

Definition sum_cardinality ($c1 \ c2$: **cardinality**) : **cardinality**.

refine { | right_size := splits_as_sum c1 c2;

all_bij := @sum_cardinality_all_bij c1 c2;

closed_under_bij := @sum_cardinality_bij_transitive c1 c2 }.

...

Defined.

Infix "+" := sum_cardinality : *cardinal_scope*.

Analogues to the theorems we have about `product_cardinality` also hold about `sum_cardinality`.

Theorem `sum_rule` ($s1\ s2 : \text{Type}$) : $|s1 + s2| = |s1| + |s2|$.

Theorem `sum_symmetric` ($c0\ c1 : \text{cardinality}$) : $c0 + c1 = c1 + c0$.

Theorem `sum_associative` ($c0\ c1\ c2 : \text{cardinality}$) : $c0 + (c1 + c2) = (c0 + c1) + c2$.

The map that we use similarly to the *interval_product_map* earlier is the bijection $f : \{0, \dots, m-1\} \sqcup \{0, \dots, n-1\} \rightarrow \{0, \dots, m+n-2\}$ defined as the bijection

$$\{0, \dots, m-1\} \sqcup \{0, \dots, n-1\} \approx \{0, \dots, m-1\} \sqcup \{m, \dots, m+n-2\}.$$

In code, this is

Definition `interval_sum_map` ($m\ n : \text{nat}$) ($x : \text{interval } m + \text{interval } n$)
 ($y : \text{interval } (m + n)$) := match x with
 | `inl` $a \Rightarrow \text{proj1_sig } a = \text{proj1_sig } y$
 | `inr` $b \Rightarrow (m + \text{proj1_sig } b) \% \text{nat} = \text{proj1_sig } y$
 end.

Lemma `interval_sum_bijective` : $\forall\ m\ n$, bijective (`@interval_sum_map` $m\ n$).

Theorem `nat_sum_cardinality` ($m\ n : \text{nat}$) :
`cardinality_n` $m + \text{cardinality_n } n = \text{cardinality_n } (m + n)$.

We can use `sum_cardinality` to break up a set $\{s : S \mid P\ x\}$ into two disjoint subsets based on some predicate ($Q : S \rightarrow \text{Prop}$). The initial set may or may not itself be represented as a subset.

Theorem `sum_split` ($S : \text{Type}$) ($P\ Q : S \rightarrow \text{Prop}$) :
 $|\{x \mid P\ x\}| = |\{x \mid P\ x \wedge Q\ x\}| + |\{x \mid P\ x \wedge \neg Q\ x\}|$.

Corollary `sum_split_full_set` ($S : \text{Type}$) ($Q : S \rightarrow \text{Prop}$) :
 $|S| = |\{x \mid Q\ x\}| + |\{x \mid \neg Q\ x\}|$.

We can relate `product_cardinality` to `sum_cardinality` with the normal distributivity laws. The case of $c + c = 2 \times c$ is a useful special case, but we also prove distributivity in its full generality.

Lemma `sum_cardinality_diag` ($c : \text{cardinality}$) : $c + c = 2 \times c$.

Theorem `cardinality_distrib_l` ($a\ b\ c : \text{cardinality}$) : $a \times (b + c) = a \times b + a \times c$.

Corollary `cardinality_distrib_r` ($a\ b\ c : \text{cardinality}$) : $(a + b) \times c = a \times c + b \times c$.

We may even prove some interesting facts about infinite cardinalities. For example, the set of integers \mathbf{Z} satisfies $|\mathbf{Z}| = |\mathbf{Z}| + |\mathbf{Z}|$, using the following bijection.

Lemma zip_bijection : bijective (fun (n : \mathbb{Z}) (p : $\mathbb{Z} + \mathbb{Z}$) \Rightarrow
 match p with
 | inl m \Rightarrow m + m = n
 | inr m \Rightarrow m + m + 1 = n
 end).
 Theorem Z_equals_Z_plus_Z : $|\mathbb{Z}| = |\mathbb{Z}| + |\mathbb{Z}|$.

Using the definitions of product and sum cardinality, we may also investigate the cardinalities of power sets. Since an element of Ensemble T for a type T is a subset of T , the type Ensemble T is the power set of the type T . If $|T| = n$, the power set $\mathcal{P}(T)$ has 2^n elements. First, we show that the set $\mathcal{P}(\{0, \dots, n-1\})$ has 2^n elements. This is using the standard proof by induction: the power set $\mathcal{P}(\{0, \dots, n\})$ splits in half based on whether the element n is in a given subset, and each half is in bijection with a subset of $\{0, \dots, n-1\}$. We invoke sum_cardinality_diag above (the fact that $c + c = 2 \times c$) to make use of this.

Theorem interval_power_set n : |Ensemble [0..n)| = 2^n .

After this, we can extend the theorem to all finite sets T , because if $|T| = n$ then we have a bijection between T and the discrete n -element interval. The only remaining interesting part of the proof of this power set rule is that if two sets have the same cardinality, then their power sets also have the same cardinality.

Lemma equal_power_sets T T' : $|T| = |T'| \rightarrow |\text{Ensemble } T| = |\text{Ensemble } T'|$.

Theorem power_set_rule T (n : nat) : $|T| = n \rightarrow |\text{Ensemble } T| = 2^n$.

3.2.5 Cardinalities of Subsets

We wish to write down rules about cardinalities of subsets of types, using Ensemble and sig. Given a list of elements that makes up a subset, if the list has no duplicates then the cardinality of this subset is equal to the length of the list.

Theorem cardinal_of_list A (l : list A) : NoDup l \rightarrow |sig (list_to_subset l)| = length l.

Corollary subset_length_match T (l l' : list T) :

l \cong l' \rightarrow NoDup l \rightarrow NoDup l' \rightarrow length l = length l'.

The notation $(l \cong l')$ here expands to $(\text{list_to_subset } l) = (\text{list_to_subset } l')$. Next we have several basic lemmas about empty, nonempty, and singleton sets.

Lemma empty_subset_empty $T : |\mathbf{sig} (\mathbf{Empty_set} \ T)| = 0$.

Lemma single_empty_subset $T : \forall x : \mathbf{Ensemble} \ T, |\mathbf{sig} \ x| = 0 \rightarrow \mathbf{Empty_set} \ T = x$.

Theorem single_subset $T (P : \mathbf{Ensemble} \ T) : (\exists! \ x : T, P \ x) \rightarrow |\mathbf{sig} \ P| = 1$.

Theorem no_subsets $T (P : \mathbf{Ensemble} \ T) : (\neg \exists x : T, P \ x) \rightarrow |\mathbf{sig} \ P| = 0$.

Lemma empty_type $T : |T| = 0 \rightarrow T \rightarrow \mathbf{False}$.

Lemma nonempty_type $T : \forall n, |T| = S \ n \rightarrow \exists t : T, \mathbf{True}$.

Given that there's a bijection between two sets A and B , we can find a bijection that maps some specific $a \in A$ to a given $b \in B$.

Lemma bijective_wlog $A \ B (a : A) (b : B) (f : A \rightarrow B \rightarrow \mathbf{Prop}) :$
 $\mathbf{bijective} \ f \rightarrow \exists f' : A \rightarrow B \rightarrow \mathbf{Prop}, \mathbf{bijective} \ f' \wedge f' \ a \ b$.

Removing an element from a finite subset that it's in (or from the whole set) results in a set that is smaller by one. Removing an element from a subset that it's not in does not change the size of the subset.

Theorem remove_element_subset $T (s : \mathbf{Ensemble} \ T) :$
 $\forall n, |\mathbf{sig} \ s| = S \ n \rightarrow \forall t : T, s \ t \rightarrow |\{x : T \mid s \ x \wedge x \neq t\}| = n$.

Corollary remove_element $T : \forall n, |T| = S \ n \rightarrow \forall t : T, |\{x : T \mid x \neq t\}| = n$.

Lemma remove_element_not_in_subset $T (s : \mathbf{Ensemble} \ T) :$
 $\forall n, |\mathbf{sig} \ s| = n \rightarrow \forall t, (\neg s \ t) \rightarrow |\{x \mid s \ x \wedge x \neq t\}| = n$.

We often want to consider subsets of a particular size only. For example, poker hands are subsets of the set of cards that have size 5. For this, we have the following definition.

Definition sized_subset $(T : \mathbf{Type}) (n : \mathbf{cardinality}) :=$
 $\{x : \mathbf{Ensemble} \ T \mid |\mathbf{sig} \ x| = n\}$.

Definition in_set $T \ n (S : \mathbf{sized_subset} \ T \ n) (x : T) := x \in \mathbf{proj1_sig} \ S$.

We can rephrase the Ensemble theorem *Extensionality_Ensembles* in terms of sized_subset to have a criterion for when sized_subsets are equal.

Lemma sized_subset_extensionality $T \ n (A \ B : \mathbf{sized_subset} \ T \ n) :$
 $(\forall x, \mathbf{in_set} \ A \ x \leftrightarrow \mathbf{in_set} \ B \ x) \rightarrow A = B$.

The cardinality of the type of sized_subset for a particular size is also of interest. For finite sets, this cardinality is a binomial coefficient.

Definition choose_cardinal $(T : \mathbf{Type}) (k : \mathbf{cardinality}) := |\mathbf{sized_subset} \ T \ k|$.

A simple way to define binomial coefficients in Coq is using the recursive formula

$$\binom{n}{k} = \binom{n-1}{k} + \binom{n-1}{k-1}.$$

```

Fixpoint choose_nat (n k : nat) : nat :=
  match k with
  | 0 => 1
  | S k' => match n with
            | 0 => 0
            | S n' => (choose_nat n' k' + choose_nat n' k)%nat
          end
  end.

Notation "( n 'choose' k )" := (choose_nat n k).
Notation "n !" := (fact n) (at level 15).

```

The binomial defined in this manner is equal to the common factorial definition

$$\binom{n}{k} = \frac{n!}{k!(n-k)!}.$$

Theorem choose_factorial $(n k : \text{nat}) : k \leq n \rightarrow n! = (n \text{ choose } k) \times k! \times (n - k)!$.

The major result here is that the number of k -element subsets of a set T with cardinality n is equal to $\binom{n}{k}$. The standard argument for why this cardinality is equal to the recursive definition of $\binom{n}{k}$ goes as follows. If T is empty, we are in the base case. Otherwise, we fix an element $t \in T$ and divide up the k element subsets into those which contain t and those which do not. The subsets which contain t are in bijection with subsets of $T \setminus \{t\}$ that have $k - 1$ elements and the subsets which do not contain t are in bijection with subsets of $T \setminus \{t\}$ that have k elements.

Theorem choose_equal $(n : \text{nat}) :$
 $\forall T, |T| = n \rightarrow \forall k : \text{nat}, \text{choose_cardinal } T k = (n \text{ choose } k).$

3.3 Automating Counting Proofs

To arrive at the automation of the proof steps that we see in the example from the beginning of the section, we went through applying the proof steps by hand. For brevity, we have a

slightly smaller example here that we investigate. This proof is calculating the number of three-card hands of cards there are given that there is exactly one pair in the hand. Before the proof, we have the definitions of the data types involved. Ranks and suits are defined simply as enumerated inductive types. A card is a pair with a rank and a suit.

```
Inductive Rank := RA | R2 | R3 | R4 | R5 | R6 | R7 | R8 | R9 | R10 | RJ | RQ | RK.
Inductive Suit := Clubs | Diamonds | Hearts | Spades.
Definition Card := (Rank × Suit)%type.
```

We have the two projection functions out of Card.

```
Definition rank := @fst Rank Suit.
Definition suit := @snd Rank Suit.
```

There are easy proofs of the cardinalities of **Rank** and **Suit** because they follow directly from the definitions.

```
Lemma Rank_sz : |Rank| = 13.
  enumeration (RA :: R2 :: R3 :: R4 :: R5 :: R6 :: R7 :: R8 ::
    R9 :: R10 :: RJ :: RQ :: RK :: nil).
Qed.
Lemma Suit_sz : |Suit| = 4.
  enumeration (Clubs :: Diamonds :: Hearts :: Spades :: nil).
Qed.
Hint Rewrite Rank_sz Suit_sz : cardinality.
```

We write a tactic for automatically proving the size of compound data types, using the `cardinality` hint database. This tactic knows about simplifying sets that have elements removed, binomial coefficient sizes, and sum and product cardinalities.

```
Ltac prove_size :=
  repeat rewrite product_rule;
  try (erewrite remove_element_subset; [reflexivity | prove_size | auto]);
  try (erewrite remove_element; [reflexivity | prove_size]);
  try (match goal with
    | [ ⊢ context[|sized_subset ?T ?n|] ] ⇒ fold (choose_cardinal T n)
    | _ ⇒ idtac
    end; erewrite choose_equal; [reflexivity | prove_size]);
  autorewrite with cardinality;
  reflexivity || apply nat_product_cardinality || apply nat_sum_cardinality
```

This can prove the cardinality of Card, given that the cardinalities of **Rank** and **Suit** have been entered into the hint database.

Lemma Card_sz : |Card| = 52.

 unfold Card.

 prove_size.

Qed.

Hint Rewrite Card_sz : cardinality.

Definition compute_size $S := \{ \text{size} \mid |S| = \text{size} \}$.

For cardinality proofs where we do not know the size of the set beforehand, we may phrase the size as a sigma type. A goal of `compute_size T` means that we are both computing some cardinality and proving that it equals $|T|$.

Now we arrive at the proof. We describe the “having a pair” criterion as two cards sharing a rank and the third having a different rank. The notation used, which we have seen in the example of Section 3.1, will be defined in Section 3.4.

Definition has_pair := ($\exists \text{ rank ' } \$0 = \text{rank ' } \$1 \wedge \text{rank ' } \$0 \neq \text{rank ' } \$2$)%nary.

Arguments has_pair /.

Theorem pair_size : compute_size { $S : \text{Ensemble Card} \mid \text{ex_set } S \ 3 \text{ has_pair} \}$.

The way that we express the exercise of rigorously computing a value without knowing it beforehand is captured in the definition of `compute_size`. We can reduce our goal to the proof part of the sigma type by using `eexists` to instantiate the value part with an existential variable.

Proof.

 eexists.

Here, our goal is

```
=====
|{  $S : \text{Ensemble Card} \mid$ 
   $\text{ex\_set } S \ 3 (\text{rank ' } \$0 = \text{rank ' } \$1 \wedge \text{rank ' } \$0 \neq \text{rank ' } \$2)$  } | = ?39
```

We want to apply Theorem 3.1.3, that is, use the `gen_product_rule`. Our first projection function is going to be to take the shared rank of the pair. The way we write this function is as follows.

```
erewrite gen_product_rule with (  $Q :=$ 
  (fun ( $S : \text{Ensemble Card}$ ) ( $r : \mathbf{Rank}$ )  $\Rightarrow$ 
     $\text{ex\_set } S \ 3 (\text{has\_pair} \wedge (\triangleq r) = \text{rank ' } \$0)$  )).
```

The \triangleq notation is simply a lift of a constant to a constant function and is explained in more detail in Section 3.4. Because Q is defined over the base type (here `Ensemble Card`) rather than over the subset $\{ S : \text{Ensemble Card} \mid \text{ex_set } S \text{ 3 has_pair } \}$, we have to specify in Q that its argument S satisfies the same properties that are already in the goal. However, setting this up the other way would lead to a more difficult problem: given an `Ensemble Card` that we know satisfies the `ex_set` criterion in the original goal, we cannot easily write down exactly what is the rank of the pair in that subset. The cards in the subset are only given an ordering inside of the existential covered by `ex_set`, so writing it in the above form actually allows us to use the references `$0`, etc., to write the projection function.

Applying `gen_product_rule` leaves us with four subgoals:

- The total cardinality is equal to the refined cardinality times the cardinality of the image of Q .
- The restriction of the function Q to the set we project out of (in this case, three-card hands containing one pair) is a well-defined and total function.
- Q is only defined on this set and not on other subsets of `Card`.
- Calculate the refined cardinality.

The first of these subgoals does not really require proof so much as it is asking us to relate the previous existential variable with the new existential variable that *erewrite* generated for the refined cardinality. The goal is

$$\text{=====}$$

$$?92 \times |\mathbf{Rank}| = ?39$$

We simplify the cardinality `|\mathbf{Rank}|` and then use `reflexivity` to tell Coq that this is exactly the relationship between the two existentials.

```
autorewrite with cardinality.
reflexivity.
```

In the second subgoal, we begin by splitting up the “function” and “total” propositions that come from the *Q_total_function* hypothesis.

```
simpl. split.
```

The first of these, well-definedness, is one of the most involved parts of the proof. We begin tackling this by doing some surface simplification and pulling out the hypotheses.

```
hnf. simpl.
intros (z, Hz) x y Hx Hy. simpl in *. clear Hz.
```

Our subgoal is now

```
z : Ensemble Card
x : Rank
y : Rank
Hx : ∃ x0 x1 x2 : Card,
    [x0; x1; x2] enumerates z ∧
    rank x0 = rank x1 ∧ rank x0 ≠ rank x2 ∧ x = rank x0
Hy : ∃ x x0 x1 : Card,
    [x; x0; x1] enumerates z ∧
    rank x = rank x0 ∧ rank x ≠ rank x1 ∧ y = rank x
=====
x = y
```

The notation $[x0; x1; x2]$ is simply shorthand for the list $x0 :: x1 :: x2 :: nil$. This subgoal is asking us to show that given two ranks x and y that are each the projection of Q off of a subset z , the ranks must be the same. We first reduce the complexity of the hypotheses Hx and Hy by instantiating the Cards as variables and pulling off the “enumerates” facts into their own hypotheses.

```
repeat match type of Hx with
  | ∃ -, _ ⇒ destruct Hx as (?, Hx)
end.
destruct Hx as (Henum, Hx).
repeat match type of Hy with
  | ∃ -, _ ⇒ destruct Hy as (?, Hy)
end.
destruct Hy as (Henum', Hy).
simpl in *.
```

Then we deconstruct the rest of the hypotheses and eliminate extra variables.

```
repeat (subst; intuition).
```

This cleans up the subgoal to look as follows.

```

z : Ensemble Card
x0 : Card
x1 : Card
x2 : Card
Henum : [x0; x1; x2] enumerates z
x3 : Card
x4 : Card
x5 : Card
Henum' : [x3; x4; x5] enumerates z
H : rank x0 = rank x1
H1 : rank x3 = rank x4
H3 : rank x0 = rank x2 → False
H0 : rank x3 = rank x5 → False

```

```

=====
rank x0 = rank x3

```

The interesting part of this proof is that we can tell that $[x0; x1; x2]$ is some permutation of $[x3; x4; x5]$, but they do not necessarily map one-to-one. We need to prove that if the variables have been permuted but still satisfy the set specification, then the rank of the first element is unchanged. One approach could be to consider all permutations of the list $[x0; x1; x2]$. Unfortunately, performing the resulting proof search in every branch becomes intractable for larger problems such as full poker hands of five cards. Instead, we aim to cut this search space by only considering permutations which actually satisfy `has_pair` in their ordering. In this case, the first two elements of a hand can be swapped and the hand will still satisfy `has_pair`, but no other permutations are valid. A more generalizable way of expressing this is that the equivalence relation $(\text{fun } (x \ y : \text{Card}) \Rightarrow \text{rank } x = \text{rank } y)$ partitions the hand into a set of equivalence classes. When permuting the list, elements that were equivalent must remain equivalent. Hence, a valid permutation of the list sends one equivalence class to a permutation of another class. In the case of $[x0; x1; x2]$, which partitions into $[x0; x1] ++ [x2]$, the two equivalence classes have distinct sizes so they must map to themselves.

```

assert (Heq0 : EquivPartitioned (fun x y ⇒ rank x = rank y)
                                   [x0; x2] [[x0; x1]; [x2]])
  by (hnf; split; [ go | prove_NoEquiv ]).
replace [x0; x1; x2] with (flatten ([x0; x1]; [x2])) in Henum by reflexivity.
assert (Heq1 : EquivPartitioned (fun x y ⇒ rank x = rank y)
                                   [x3; x5] [[x3; x4]; [x5]])
  by (hnf; split; [ go | prove_NoEquiv ]).

```

```

replace [x3; x4; x5] with (flatten ([[x3; x4]; [x5]])) in Henum' by reflexivity.
eapply enumerate_symmetry in Heq0; [| | simpl; eauto | apply Heq1 |; [| | auto |.
simpl in Heq0. clear Heq1.
rewrite flatten_fold_app in *.

```

After applying *enumerate_symmetry*, we have added

$$Heq0 : [[x3; x4]; [x5]] \cong [[x0; x1]; [x2]]$$

to our set of hypotheses. Recall that $l \cong l'$ means $(\text{list_to_subset } l) = (\text{list_to_subset } l')$, so the two lists are permutations of each other. This reduces our search to permutations of this list of two elements, rather than the larger list. In other cases, like the full houses example, reducing the search space from $5!$ to $2!$ possibilities is a significant difference. We have a specific tactic to do a case analysis of the different possible permutations.

```
destruct_permutations Heq0.
```

In the first case, we break up *Heq0* into

$$\begin{aligned}
H1 &: [x0; x1] \cong [x3; x4] \\
H2 &: [x2] \cong [x5]
\end{aligned}$$

This is the correct matching. In order to show that $\text{rank } x0 = \text{rank } x3$, given that $\text{rank } x0 = \text{rank } x1$ above, we look at the two cases of permutations given by *H1* and can solve them with *subst* and *intuition*.

```
destruct_permutations H1; subst; intuition.
```

In the second case, we have

$$\begin{aligned}
H1 &: [x2] \cong [x3; x4] \\
H2 &: [x0; x1] \cong [x5]
\end{aligned}$$

Two subsets can only be permutations of each other if, given that their enumerations do not have repetitions, the enumerations have the same length. We apply this logic to *H1*, and then we must prove that both lists satisfy the **NoDup** property. This is not too difficult. We have that $[x3; x4; x5]$ *enumerates* z implies **NoDup** $[x3; x4; x5]$ and that $[x3; x4]$ is a sublist of a list with no duplicates. We earlier rewrote the list $[x3; x4; x5]$ as $(\text{flatten } [[x3; x4]; [x5]])$ which lets us do this unification easily.

```
apply subset_length_match in H1; [ discriminate | | ];
```



```

destruct Henum; destruct Henum'.
eapply NoDup_in_fold_app; [ | apply H6 ]. subst; intuition.
eapply NoDup_in_fold_app; [ | apply H8 ]. subst; intuition.

```

This finishes the subgoal that Q is well-defined. The next subgoal is to prove that Q is defined on all sets that satisfy `has_pair`.

```
hnf. intros [x Hx]. simpl.
```

After some simplification, our goal is

```

x : Ensemble Card
Hx : ∃ x0 x1 x2 : Card,
    [x0; x1; x2] enumerates x ∧ rank x0 = rank x1 ∧ rank x0 ≠ rank x2
=====
∃ (v : Rank) (x0 x1 x2 : Card),
    [x0; x1; x2] enumerates x ∧
    (rank x0 = rank x1 ∧ rank x0 ≠ rank x2) ∧ v = rank x0

```

The goal is basically already in the hypothesis Hx . We instantiate the `Card` variables in the goal with the `Cards` from Hx , repeatedly using the fact

$$\exists x, \exists y, P \ x \ y \rightarrow \exists y, \exists x, P \ x \ y$$

to push the $\exists (v : \mathbf{Rank})$ back to the end. We then turn v into an existential and are able to prove the rest of goal with first-order logic.

```

destruct Hx as [c0 Hx]. apply exists_swap. ∃ c0. simpl.
destruct Hx as [c1 Hx]. apply exists_swap. ∃ c1. simpl.
destruct Hx as [c2 Hx]. apply exists_swap. ∃ c2. simpl.
eexists.
firstorder; firstorder.

```

Now we have shown that Q is defined on all sets with `has_pair`. Next we must show that these are the only sets on which it is defined.

```
intro. intro. simpl.
```

This goal has a parallel structure much like the previous, but is easier because we don't need to figure out what the rank v is.

```

t : Ensemble Card
v : Rank

```

=====

```
( $\exists$   $x$   $x0$   $x1$  : Card,
  [ $x$ ;  $x0$ ;  $x1$ ] enumerates  $t$   $\wedge$ 
  ( $\text{rank } x = \text{rank } x0 \wedge \text{rank } x \neq \text{rank } x1$ )  $\wedge v = \text{rank } x$ )  $\rightarrow$ 
 $\exists$   $x$   $x0$   $x1$  : Card,
  [ $x$ ;  $x0$ ;  $x1$ ] enumerates  $t$   $\wedge \text{rank } x = \text{rank } x0 \wedge \text{rank } x \neq \text{rank } x1$ )
```

Here we invoke Coq's tactic for solving goals that are simply first-order logic.

`firstorder.`

Finally, we arrive at our last subgoal, which is to calculate the cardinality of the refined set.

`intro r .`

Once we pull the fixed variable up into our hypotheses, we now have to calculate the number of subsets with this r fixed.

=====

```
 $\forall v$  : Rank,
?61 = |{ $t$  : Ensemble Card | ex_set  $t$  3 (has_pair  $\wedge$  ( $\trianglelefteq v$ ) = rank ' $\$0$ )} |
```

The remaining choices of projections are very similar, so we skip them here by using our automation. There are a few interesting problems that we run into in these cases that are not in this first projection. The second projection is onto ranks other than r , rather than all ranks, so we need to involve a sigma type $\{r' \mid r' \neq r\}$ that we need to `destruct` at the appropriate times. In the `sized_subset` case, we additionally run into goals similar to `[suit $x0$; suit $x1$] \cong [suit $x3$; suit $x4$]`. Here we have `[$x0$; $x1$] \cong [$x3$; $x4$]` as above, so we use the lemma `list_to_subset_map`, which shows that we can apply `map` on both sides of the permutation (\cong) expression.

```
pose (card_hint [2; 1]).
pick  $r'$  as (3  $\nrightarrow$  rank ' $\$2$ ) from { $r'$  |  $r' \neq r$ }.
pick  $s$  as (3  $\nrightarrow$  set [suit ' $\$0$ ; suit ' $\$1$  ]) from (sized_subset Suit 2).
pick  $s'$  as (3  $\nrightarrow$  suit ' $\$2$ ).
name_all.
```

Now what remains to show is that these choices of ranks and suits determine a single subset.

`symmetry. apply single_subset. simpl.`

The subgoal we have is

```

r : Rank

p := card_hint [2; 1] : PokerHint

r' : Rank

H : r' ≠ r

s : Ensemble Suit
s' : Suit
s0 : Suit

H1 : s s0

s1 : Suit

H3 : s s1



H5 : [s0; s1] enumerates s


=====
∃ ! x : Ensemble Card,
  ∃ x0 x1 x2 : Card,
    [x0; x1; x2] enumerates x ∧
    (((rank x0 = rank x1 ∧ rank x0 ≠ rank x2) ∧ r = rank x0) ∧
     r' = rank x2) ∧ s = [suit x0; suit x1] ∧
     s' = suit x2

```

Let us refer to the subset specification as *P*, so that this goal has the form $\exists ! x : \text{Ensemble Card}, P\ x$. We can write down the unique subset satisfying *P* using the fixed variables in our hypotheses.

```

∃ [(r, s0); (r, s1); (r', s')].
unfold unique. split.

```

To prove uniqueness, first we prove that the given subset does actually satisfy *P*, and then we must prove that it is the only such subset which does. In the list we have specified our elements in the same order that they appear in `has_pair`, so we use these to instantiate the next three existentials.

```

∃ (r, s0). ∃ (r, s1). ∃ (r', s'). simpl.
split.

```

The first section of the proposition is

```

[(r, s0); (r, s1); (r', s')] enumerates [(r, s0); (r, s1); (r', s')]

```

Clearly the lists are equal, so we are left with proving

```

NoDup [(r, s0); (r, s1); (r', s')]

```

This follows from the hypotheses $H : r' \neq r$ and $H5 : [s0; s1]$ *enumerates* s . We have a short tactic that encapsulates the proof of **NoDup**.

```
split; [ auto | t_NoDup ].
```

After this, the rest of the proposition follows easily. The only part that `intuition` cannot solve immediately is that $H5 : [s0; s1]$ *enumerates* s implies $s = [s0; s1]$, which we get by destructing $H5$.

```
intuition.
destruct H5. assumption.
```

The second part of the uniqueness proof is that the supplied subset is in fact the only subset which satisfies P . Given some other subset, *result*, that satisfies P , we need to prove that it is equal to ours, or that the list representation is a permutation of our list $[(r, s0); (r, s1); (r', s')]$. We start by simplifying the hypothesis (P *result*).

```
intros result H'.
destruct H' as [c0 H'].
destruct H' as [c1 H'].
destruct H' as [c2 H'].
destruct H' as [Henum H'].
destruct Henum as [Henum]. rewrite Henum. clear Henum. clear result.
repeat match goal with
  | [ H : _ ∧ _ ⊢ _ ] ⇒ destruct H; subst
  | [ H : |sig (list_to_subset _)| = _ ⊢ _ ] ⇒ clear H
end.
```

After simplification, the goal becomes

```
p := card_hint [2; 1] : PokerHint
s0 : Suit
s1 : Suit
c0 : Card
c1 : Card
c2 : Card
H0 : NoDup [c0; c1; c2]
H2 : rank c0 = rank c1
H4 : rank c0 ≠ rank c2
H1 : [suit c0; suit c1] s0
H3 : [suit c0; suit c1] s1
H5 : [s0; s1] enumerates [suit c0; suit c1]
H : rank c2 ≠ rank c0
```

=====

$[(\text{rank } c0, s0); (\text{rank } c0, s1); (\text{rank } c2, \text{suit } c2)] \cong [c0; c1; c2]$

One way to prove that these two lists are permutations of each other could be to consider all of the possible permutations. Since this case split is much slower when the size of the Ensemble is bigger, we turn again to the fact that when permuting the elements of our set, the only valid permutations come from permuting elements in appropriately defined equivalence classes. We transform this goal into a stronger one where we have to prove both that $[(\text{rank } c0, s0); (\text{rank } c0, s1)] \cong [c0; c1]$ and $[(\text{rank } c2, \text{suit } c2)] \cong [c2]$.

```
let li := constr:[2; 1] in
  match goal with
  | ⊢ ?l0 ≅ ?l1 ⇒
    replace l0 with (flatten (separate_list li l0)) by reflexivity;
    replace l1 with (flatten (separate_list li l1)) by reflexivity;
    simpl separate_list in *; apply sublist_permutation_equality; simpl
  end.
```

In order to prove the first of these, we need to leverage the hypotheses $H2 : \text{rank } c0 = \text{rank } c1$ and $H5 : [s0; s1] \text{ enumerates } [\text{suit } c0; \text{suit } c1]$. It would be possible to go through and `destruct` every `Card` in the subgoal, but such a strategy does not easily generalize to when we want to count arbitrary data types. A slightly more general principle we invoke here is the following.

Lemma *refine_permutation* $A B C (f : A \rightarrow B) (g : A \rightarrow C) (l \ l' : \text{list } A) :$
 $(\forall x \ y, f \ x = f \ y \rightarrow g \ x = g \ y \rightarrow x = y) \rightarrow$
 $\forall y, \text{Forall } (\text{fun } x \Rightarrow f \ x = f \ y) (l \ ++ \ l') \rightarrow$
 $\text{map } g \ l \cong \text{map } g \ l' \rightarrow$
 $l \cong l'.$

Given two projection functions f and g out of a type A , the fact that equality under f and equality under g together yield equality in A and the fact that all the elements of our lists are equal under f , then we can weaken our goal to $\text{map } g \ l \cong \text{map } g \ l'$. We apply this here with `rank` as f and `suit` as g .

```
constructor.
eapply (refine_permutation (hint_f p) (hint_g p)); simpl.
go.
repeat constructor; simpl; auto.
destruct H5. symmetry. assumption.
```

```

    constructor.
    eapply (refine_permutation (hint-f p) (hint-g p)); simpl.
    go.
    repeat constructor; simpl; auto.
    auto.

    constructor.
Defined.

```

This finishes an unautomated version of the proof for the number of three-card sets which have exactly one pair.

```

Eval simpl in (proj1_sig pair_size).

```

```

= 1 × 4 × 6 × 12 × 13
  : cardinality

```

Coincidentally, this is the same as the number of poker hands with full houses because $\binom{4}{1} = \binom{4}{3}$.

3.4 n -ary Function Notation

One of the issues in applying `gen_product_rule` in counting proofs is that we need to restate the subset definition in order to add conditions inside of the existential. Unfortunately, it is very difficult to use the L_{tac} term matching mechanism to match and rewrite inside of several existential qualifiers in our use case. In order to save users from having to rewrite the entire subset definition, we develop here a monadic notation for writing functions that take a number of parallel inputs. The type `(nary_fun T V n)` represents the type

$$\underbrace{T \rightarrow \cdots \rightarrow T}_n \rightarrow V.$$

```

Fixpoint nary_fun (T V : Type) (n : nat) : Type :=
  match n with
  | 0 => V
  | S n' => T → nary_fun T V n'
end.

```

An `nary_prop` is just a `Prop`-valued n -ary function.

Definition $\text{nary_prop } T \ n := \text{nary_fun } T \ \text{Prop } n$.

We have an existential quantifier for n -ary propositions. This prepends an $\exists (x : T)$ to the front of the given proposition for each input that it takes.

```
Fixpoint nary_ex T n (P : nary_prop T n) : Prop :=
  match n return nary_prop T n → Prop with
  | 0 ⇒ fun P' : nary_prop T 0 ⇒ P'
  | S n' ⇒ fun P' : nary_prop T (S n') ⇒
    ∃ x : T, nary_ex T n' (P' x)
  end P.
```

A variant of nary_ex , nary_ex_pred prepends $\exists (x : T), Q \ x \wedge ..$ for each variable, allowing an additional specification to be put on each input.

```
Fixpoint nary_ex_pred T n (Q : T → Prop) (P : nary_prop T n) : Prop :=
  match n return nary_prop T n → Prop with
  | 0 ⇒ fun P' : nary_prop T 0 ⇒ P'
  | S n' ⇒ fun P' : nary_prop T (S n') ⇒
    ∃ x : T, Q x ∧ nary_ex_pred n' Q (P' x)
  end P.
```

Functions can be lifted to be compositions with n -ary functions. We define lifts of unary and binary functions, as well as lifts of constants into constant functions.

```
Fixpoint nary_lift T V V' n (f : V → V') :
  nary_fun T V n → nary_fun T V' n :=
  match n return nary_fun T V n → nary_fun T V' n with
  | 0 ⇒ f
  | S n' ⇒ fun A x ⇒ nary_lift T n' f (A x)
  end.
```

```
Fixpoint nary_lift2 T V V' V'' n (f : V → V' → V'') :
  nary_fun T V n → nary_fun T V' n → nary_fun T V'' n :=
  match n return nary_fun T V n → nary_fun T V' n → nary_fun T V'' n with
  | 0 ⇒ f
  | S n' ⇒ fun A B x ⇒ nary_lift2 T n' f (A x) (B x)
  end.
```

```
Fixpoint nary_const T V n (v : V) : nary_fun T V n :=
  match n return nary_fun T V n with
  | 0 ⇒ v
  | S n' ⇒ (fun x ⇒ nary_const T n' v)
  end.
```

We have an indexing function, which returns exactly one of its inputs. The function

returns the m th of $n + 1$ inputs, or the last input if $m \geq n + 1$. This cannot make a nullary function because such a function has no inputs.

```
Fixpoint nary_nth (T : Type) (n m : nat) {struct m} : nary_fun T T (S n) :=
  match m with
  | 0 => fun x => nary_const T n x
  | S m' => fun x =>
    match n return (nary_fun T T n) with
    | 0 => x
    | S n' => nary_nth n' m'
    end
  end.
```

For example,

```
Eval simpl in (@nary_nth bool 5 2).
```

```
= fun _ _ x1 _ _ : bool => x1
: nary_fun bool bool 6
```

Liberal use of notation allows us to write expressions that look similar to typical Coq expressions but actually evaluate to n -ary functions. We overload the standard operators for **and**, **or**, **eq**, and **not** to be their n -ary lifts, and we introduce shorthand for lifts and the indexing function. The notation for `nary_nth` is reminiscent of shell scripting syntax, where `$0` is the first argument, `$1` is the second argument, and so on.

```
Definition nary_or T n := nary_lift2 T n or.
```

```
Definition nary_and T n := nary_lift2 T n and.
```

```
Definition nary_eq T n A := nary_lift2 T n (@eq A).
```

```
Definition nary_not T n := nary_lift T n not.
```

```
Notation "$ n" := (nary_nth _ n) (at level 5) : nary_scope.
```

```
Notation "≔ c" := (nary_const _ _ c) (at level 10) : nary_scope.
```

```
Notation "f ' x" := (nary_lift _ _ f x) (at level 10, x at next level) : nary_scope.
```

```
Notation "f " x , y" := (nary_lift2 _ _ f x y) (at level 10, y at next level) : nary_scope.
```

```
Infix "^" := (nary_and _ _ ) : nary_scope.
```

```
Infix "v" := (nary_or _ _ ) : nary_scope.
```

```
Infix "=" := (nary_eq _ _ ) : nary_scope.
```

```
Notation "~ P" := (nary_not _ _ P) : nary_scope.
```

```
Notation "a <> b" := (¬ (a = b))%nary : nary_scope.
```

In the poker hand examples, we use the following `ex_set` predicate as a way of enumerating the elements of an Ensemble so that we may supply an `nary_prop` that uses this ordering.

This uses an n -ary function that builds a list out of its arguments.

```
Fixpoint nary_list (T : Type) (n : nat) {struct n} : nary_fun T (list T) n :=
  match n return nary_fun T (list T) n with
  | 0 => nary_const _ _ nil
  | S n' => fun x => (nary_lift2 _ _ (@cons T)) ( $\triangleq$  x)%nary (nary_list _ n')
  end.
```

```
Definition ex_set T (S : Ensemble T) (n : nat) (P : nary_prop T n) :=
  nary_ex T n ((@list_enumerates_subset T)“ (nary_list T n) , ( $\triangleq$  S)  $\wedge$  P).
```

There are also notations for building lists and sets that are not simply the entire list of the arguments.

```
Notation "[ x ; .. ; y ]" := ((@cons _)“ x , .. ((@cons _)“ y , ( $\triangleq$  nil)) ..)%nary : nary_scope.
Notation "$[ x ; .. ; y ]" := ((@cons _)“ (nary_nth _ x) , ..
  ((@cons _)“ (nary_nth _ y) , ( $\triangleq$  nil)) ..)%nary : nary_scope.
Notation "set[ x ; .. ; y ]" := ((nary_lift _ _ (@list_to_subset _))
  ((@cons _)“ x , .. ((@cons _)“ y , ( $\triangleq$  nil)) ..)%nary) : nary_scope.
```

The above notations all infer the types and number of arguments for the involved n -ary functions. For the cases where we need to specify the number of arguments, we define the following notation.

```
Notation "n -/> f" := (f%nary : nary_fun _ _ n) (at level 60).
```

These are a few examples of the n -ary notations in action.

```
Eval simpl in (4  $\nrightarrow$  $[3; 0]).
```

```
= fun x _ _ x2 : ?165 => x2 :: x :: nil
: nary_fun ?165 (list ?165) 4
```

```
Eval simpl in (2  $\nrightarrow$  [mult“ ( $\triangleq$  2) , $0; mult“ ( $\triangleq$  4) , $1]).
```

```
= fun x x0 : nat => 2  $\times$  x :: 4  $\times$  x0 :: nil
: nary_fun nat (list nat) 2
```


Chapter 4

Probability

Probability theory, like counting, has a strong focus in the 6.042 curriculum. Some work has also been done to express basic concepts of probability in Coq. However, we have not begun automation for these types of problems.

Most probability exercises are similar to combinatorial exercises in that they ask for computations of values with justification. However, these values are probabilities rather than cardinalities. The premises for probability problems are generally more involved than those in combinatorics because they usually describe a probability distribution over some space of events. Questions that test knowledge of infinite probability spaces sometimes have even more complicated setups, often involving games or processes that repeat with some random chance.

4.1 Probability Example

The following simple example exercise is taken from the 6.042 textbook [7]. The proof shows the reasoning a student might be expected to explain.

Exercise 4.1.1. *What's the probability that 0 doesn't appear among k digits chosen independently and uniformly at random?*

Proof. For any given digit, the probability $\text{Pr}[\text{digit } i \text{ is nonzero}]$ is $9/10$. Since the digits are

all chosen independently the probability that they are all nonzero is

$$\Pr[\text{all digits are nonzero}] = \prod_{i=1}^k \Pr[\text{digit } i \text{ is nonzero}] = \left(\frac{9}{10}\right)^k.$$

□

We set up the problem in Coq by positing a probability distribution over the space of k -digit strings and requiring that it satisfies uniformity and independence properties.

Section IndependentEvents.

Variable $k : \text{nat}$.

Definition DigitString := ([0, ..., 10)^k)%type.

Hypothesis $D : \text{probability_space DigitString}$.

Each digit is chosen uniformly, so we assume that for each index i , the events “digit i is 0”, “digit i is 1”, etc., are uniformly distributed.

Hypothesis *uniformity* : $\forall i : [0, \dots, k), \text{uniform } _ D (\text{fun } a \ b \Rightarrow a = \text{proj1 } i \ b)$.

Writing down independence is slightly more difficult. We view the k different digits as random variables, and we assert as our hypothesis that these are all mutually independent random variables.

Definition Digit : $[0, \dots, k) \rightarrow \text{random_variable DigitString } [0, \dots, 10) := @\text{proj1 } _ _$.

Hypothesis *independence* : *mutually_independent_variables* D Digit.

The following event is the one of interest from the problem. This is an Ensemble DigitString because an event is a subset of the space over which we have a probability distribution.

Definition does_not_have_zero : Ensemble DigitString := fun (a : DigitString) \Rightarrow
 $\forall i : [0, \dots, k), \text{proj1_sig } (Digit \ i \ a) \neq 0\%nat$.

The theorem is proved by using the overall principles outlined in the written proof above. Even given we know that the digits are mutually independent, some cleverness is still required to deduce that the events “digit i is nonzero” are all independent. As defined, *mutually_independent_variables* tells us that for any specific string of digits, the probability of that string being the outcome is equal to the product of the probabilities of each digit, represented by a random variable, taking on the corresponding value. We use the following principle from the 6.042 text.

Lemma 4.1.2. *Let $Q : T \rightarrow X$ and $R : T \rightarrow X'$ be independent random variables, and assume we have functions $f : X \rightarrow Y$ and $g : X' \rightarrow Y'$. Then $f(Q)$ and $g(R)$ are independent random variables.*

Using the function `(fun (d : interval 10) => beq_nat (proj1_sig d) 0)`, which produces a boolean value for whether the digit is equal to zero, we convert the independent random variables for the digits into independent random variables for whether the digits are equal to zero. The hypothesis of uniformity gives us that one value for these random variables is $1/10$, so the other must be $9/10$. The product of the k variables gives us the result.

Theorem `Pr_of_does_not_have_zero` : `Pr D does_not_have_zero = ((9/10)^k)%Qc`.
End `IndependentEvents`.

4.2 Probability Foundations

A probability space is a probability function on a space of outcomes bundled with a proof that it is always nonnegative and that it sums to one. We choose to make this probability function take values in \mathbf{Qc} (the set of rational numbers in lowest terms) because it is much easier to compute over \mathbb{Q} than over \mathbb{R} , and nearly all probability functions we would consider in discrete math only take on rational values. The notation `(sum pr w for w : space)` is a formal sum that packages together its index set and summands, and the notation $s \mapsto v$ means that the formal sum s evaluates to v . Sums taken over finite sets can be evaluated, but we have not yet implemented evaluation of infinite series.

```
Record probability_space (space : Type) := {
  pr : space → Qc ;
  pr_nonnegative : ∀ w : space, 0 ≤ pr w ;
  pr_normalized : (sum pr w for w : space) ↦ 1
}.
```

We can define a simple example of a probability space: one which is defined on the first n naturals and where each outcome has probability $1/n$. The proof that this is a probability space is relatively straightforward using `compute`. We have a slightly roundabout method

that distinguishes between $s \mapsto v$ and `eval s` because not all sums are necessarily computable, and the notation `eval s` implicitly infers an enumeration of the index set of s . This is done using the Type Class feature in Coq.

```

Definition ex_uniform n (x : [0, ..., n)) := 1 / n.
Definition ex_space3 : probability_space [0, ..., 3).
  refine {| pr := @ex_uniform 3 |}; go.
  match goal with
  | [ ⊢ ?s ↦ ?v ] =>
    replace v with (eval s) by (compute; apply Qc_is_canon; reflexivity)
  end.
  apply finite_evaluation.
Defined.

```

We can define what it means for a space generally to be uniform, and show that our example space is indeed uniform.

```

Definition uniform_space {T} (sp : probability_space T) := ∃ q, ∀ w, pr sp w = q.
Lemma ex_space3_uniform : uniform_space (ex_space3).

```

More generally, for a finite uniform space T , the probability of each outcome happening is $1/|T|$. Here we see the use of Type Classes as the parameter ‘**Finite** T ’ is an inferrable parameter.

```

Theorem uniform_space_probability T (sp : probability_space T) ‘Finite T’ :
  uniform_space sp → ∀ t, pr sp t = 1 / (size_of T).

```

There is a distinction between *events* and *outcomes*. A probability function is defined on a space of outcomes. An event is a subset of the space of outcomes, whose probability (with a capital “P”) is defined to be the sum of the probabilities of the outcomes in the event.

Section Events.

```

Variable T : Type.
Variable sp : probability_space T.
Definition formal_Pr (E : Ensemble T) : formal_sum T Qc := sum_of E (pr sp).

```

Though we define the probability of an event as a formal sum, we would like a definition which allows us to manipulate probabilities as actual rational numbers, rather than purely formally. We can do this using Hilbert’s epsilon operator, which is an extension that instantiates an element satisfying a given property, if one exists. If one does not, the epsilon operator

returns some arbitrary element of that type (it is justified in doing so because `epsilon` requires a proof that the type is inhabited). The `epsilon` operator can be thought of as a computational version of the axiom of choice; assuming its validity allows us to deterministically instantiate an element of a set.

Definition `Pr (U : Ensemble T): Qc :=`
`epsilon (inhabits 0) (fun v => (sum (pr sp u) for u in U) ↦ v).`

If the formal sum has an evaluation, then `Pr` returns it, and sums over finite sets can always be evaluated.

Theorem `exists_Pr a (U : Ensemble T): formal_Pr U ↦ a → Pr U = a.`
Theorem `eval_Pr (U : Ensemble T) '{Finite (sig U)} : formal_Pr U ↦ Pr U.`

Formally, if we can evaluate probability over two events, then we can evaluate probability over the union of these two events. If the events are disjoint, then the probability over them is the sum of the two probabilities of the events.

Theorem `formal_Pr_disjoint_union (U V : Ensemble T) a b :`
`formal_Pr U ↦ a → formal_Pr V ↦ b →`
`Disjoint _ U V → formal_Pr (U ∪ V) ↦ a + b.`

Using the `epsilon` version of `Pr`, we can write this in a more algebraic fashion.

Theorem `Pr_disjoint_union (U V : Ensemble T) '{Finite (sig U)} '{Finite (sig V)} :`
`Disjoint _ U V → Pr U + Pr V = Pr (U ∪ V).`

As a direct corollary, we can apply the law $(U \cap V) \cup (U \cap W) = U \cap (V \cup W)$.

Corollary `Pr_intersection_sum (U V W : Ensemble T)`
`'{Finite (sig (U ∩ V))} '{Finite (sig (U ∩ W))} :`
`Disjoint _ V W → Pr (U ∩ V) + Pr (U ∩ W) = Pr (U ∩ (V ∪ W)).`

The complement law is another consequence of the disjoint union law because a set and its complement are disjoint, their union is the entire space, and the total probability over the space is 1. Here, we use U^c to denote the complement of U in T .

Theorem `Pr_complement_sum (U : Ensemble T) '{Finite (sig U)} '{Finite T} :`
`Pr U + Pr (Uc) = 1.`

This rule can be rewritten in terms of subtraction as well.

Corollary `Pr_complement (U : Ensemble T) '{Finite (sig U)} '{Finite T} :`
`Pr (Uc) = 1 - Pr U.`

There are several other probability laws in this spirit which follow from rules about operations on sets, including facts about set difference ($U \setminus V = \{x \mid x \in U \wedge x \notin V\}$) and the principle of inclusion–exclusion.

Theorem `Pr_setminus` ($U \ V : \text{Ensemble } T$) $\{\text{Finite (sig } U)\} \ \{\text{Finite (sig } V)\} :$
 $\text{Pr } (U \setminus V) = \text{Pr } U - \text{Pr } (U \cap V).$

Theorem `Pr_PIE` ($U \ V : \text{Ensemble } T$) $\{\text{Finite (sig } U)\} \ \{\text{Finite (sig } V)\} :$
 $\text{Pr } U + \text{Pr } V - \text{Pr } (U \cap V) = \text{Pr } (U \cup V).$

Theorem `Pr_monotonic` ($U \ V : \text{Ensemble } T$) : $U \subset V \rightarrow \text{Pr } U \leq \text{Pr } V.$

Lemma `Pr_empty_set` : $\text{Pr } (\text{Empty_set } _) = 0.$

Corollary `Pr_bounded_below` ($U : \text{Ensemble } T$) : $0 \leq \text{Pr } U.$

Theorem `Pr_bounded_above` ($U : \text{Ensemble } T$) $\{\text{Finite } T\} \ \{\text{Finite (sig } U)\} :$ $\text{Pr } U \leq 1.$

We also define a notion of uniform distribution of events rather than outcomes. A set of events E_j for j in some index set J are uniform if they are mutually disjoint, they cover the entire outcome space, and they all have equal probability. As with outcomes, we can show that the probability of one of J uniform events is $1/|J|$.

Definition `uniform` $\{J\} (E : J \rightarrow \text{Ensemble } T) :=$
 $(\forall i \ j, \text{Disjoint } _ (E \ i) (E \ j)) \wedge$
 $(\forall t, \exists t', \text{Ensembles.In } _ (E \ t') \ t) \wedge$
 $\exists p, \forall t', \text{Pr } (E \ t') = p.$

Theorem `uniform_probability` $\{J\} \ \{\text{Finite } J\} (E : J \rightarrow \text{Ensemble } T) :$
 $\text{uniform } E \rightarrow \forall j, \text{Pr } (E \ j) = 1 / (\text{size_of } J).$

End Events.

Another central concept in probability is conditional probability, which is straightforward to define from the definition of `Pr`.

Section Conditionals.

Variable $T : \text{Type}.$

Variable $sp : \text{probability_space } T.$

Notation `"Pr[U]"` $:= (\text{@Pr } T \ sp \ U)$ (at level 50, U at next level).

Definition `CondPr` ($U \ V : \text{Ensemble } T$) $:= \text{Pr}[U \cap V] / \text{Pr}[V].$

Notation `"Pr[U | V]"` $:= (\text{CondPr } U \ V)$
(at level 50, U at next level, V at next level).

There are many laws which hold about conditional probability as well. Note that $\text{Pr}[U \mid V]$ is not well-defined if $\text{Pr}[V]$ is zero.

Theorem CondPr_bounded ($U \ V : \text{Ensemble } T$) : $\text{Pr}[V] \neq 0 \rightarrow 0 \leq \text{Pr}[U \mid V] \leq 1$.

Theorem Bayes_rule ($U \ V : \text{Ensemble } T$) :

$$\text{Pr}[U] \neq 0 \rightarrow \text{Pr}[V] \neq 0 \rightarrow \text{Pr}[U \mid V] = \text{Pr}[V \mid U] \times \text{Pr}[U] / \text{Pr}[V].$$

Theorem Total_probability ($U \ V : \text{Ensemble } T$)

$\{ \text{Finite } T \} \{ \text{Finite (sig } U) \} \{ \text{Finite (sig } V) \} :$

$$\text{Pr}[V] \neq 0 \rightarrow \text{Pr}[V] \neq 1 \rightarrow$$

$$\text{Pr}[U] = (\text{Pr}[U \mid V] \times \text{Pr}[V]) + (\text{Pr}[U \mid V^c] \times \text{Pr}[V^c]).$$

Two events are *independent* if conditioning one on the other does not change its probability. Equivalently, they are independent if the probability of both events happening is the product of their respective probabilities.

Definition independent ($U \ V : \text{Ensemble } T$) := $\text{Pr}[U \mid V] = \text{Pr}[U]$.

Theorem independent_Intersection ($U \ V : \text{Ensemble } T$) :

$$\text{Pr}[V] \neq 0 \rightarrow (\text{independent } U \ V \leftrightarrow \text{Pr}[U \cap V] = \text{Pr}[U] \times \text{Pr}[V]).$$

As long as both events have nonzero probability, this notion of independence is commutative.

Theorem independent_comm ($U \ V : \text{Ensemble } T$) :

$$\text{Pr}[U] \neq 0 \rightarrow \text{Pr}[V] \neq 0 \rightarrow \text{independent } U \ V \rightarrow \text{independent } V \ U.$$

A set of events E_j is pairwise independent if each pair of events is independent. A set of events is mutually independent, a stronger assertion, if the intersection of any subset of them factors as the product of the probabilities of the events in that subset.

Definition pairwise_independent $\{J\} (E : J \rightarrow \text{Ensemble } T) :=$

$$\forall i \ j, i \neq j \rightarrow \text{independent } (E \ i) \ (E \ j).$$

Definition mutually_independent $\{J\} (E : J \rightarrow \text{Ensemble } T) :=$

$$\forall js : \text{list } J, \text{NoDup } js \rightarrow$$

$$\text{let events} := \text{map } E \ js \text{ in}$$

$$\text{Pr}[\text{Intersection_list } _ \text{ events}] = \text{product_over_list } (\text{Pr } sp) \ \text{events}.$$

End Conditionals.

Finally, we have the notion of random variables. These are simply functions on the outcome space, typically projections of some compound outcome.

Section RandomVariables.

Variable T : Type.

Variable sp : probability_space T .

Variable X : Type.

Definition `random_variable` := $T \rightarrow X$.

Notation "`[R = x]`" := (fun `space_elt` \Rightarrow $R \text{ space_elt} = x$)
(`R` at level 50, `x` at level 50).

This notation is the event that the random variable R takes on the value x . Random variables are independent if for any choice of values, the events that the variables take on these values are independent. There is a corresponding notion of mutual independence.

Definition `independent_variables` (`R0 R1` : `random_variable`) :=
 $\forall x0 \ x1 : X, \text{independent } sp \ [R0 = x0] \ [R1 = x1]$.

Definition `mutually_independent_variables` $\{J\}$ (`R` : $J \rightarrow \text{random_variable}$) :=
 $\forall f : J \rightarrow X, \text{mutually_independent } sp \ (\text{fun } j \Rightarrow [(R \ j) = (f \ j)])$.

End `RandomVariables`.

The principle used in the probability example at the beginning of the section can be written in the following manner, generalizing to sets of mutually independent random variables rather than pairs. The first theorem uses a different mapping for each random variable, while the second is just a specialization for when we wish to apply the same mapping to each variable.

Theorem `map_independent_variables` $\{T \ J \ X \ Y\}$ (`sp` : **probability_space** T)
(`R` : $J \rightarrow \text{random_variable } T \ X$) (`f` : $J \rightarrow X \rightarrow Y$) :
`mutually_independent_variables` `sp` `R` \rightarrow
`mutually_independent_variables` `sp` ($\text{fun } j \Rightarrow (\text{fun } t \Rightarrow (f \ j) (R \ j \ t))$).

Corollary `map_independent_variables'` $\{T \ J \ X \ Y\}$ (`sp` : **probability_space** T)
(`R` : $J \rightarrow \text{random_variable } T \ X$) (`f` : $X \rightarrow Y$) :
`mutually_independent_variables` `sp` `R` \rightarrow
`mutually_independent_variables` `sp` ($\text{fun } j \Rightarrow (\text{fun } t \Rightarrow f (R \ j \ t))$).

Chapter 5

Future Research

The work developed here can continue to be extended with more concept definitions, more exercises, and more proof automation. Making the automation increasingly robust is an area that will require heavy focus in the future. Since we eventually want different levels of detail for proofs over the course of a class like 6.042, perhaps additional abstract methodology should be considered for developing automation for different sections of the class. There may also be a need for designing an interface so that students do not have to interact directly with a Coq front end. Right now, we envision using creative definitions of notation in Coq, as with the n -ary function notation, to make it easier to write an interpreter that translates proofs of some sort into Coq code. Ideally, students could write proofs in plain English that the software would be able to verify. However, this direction is itself a difficult problem in natural language processing.

Issues related more immediately to the work developed in this project include additional topics within combinatorics and probability that merit attention and analysis. Common combinatorial concepts that remain to be implemented include uses of the pigeonhole principle and various identities involving binomial coefficients. We also want better ways to describe counting arguments in edge cases similar to examples we have developed. For instance, when counting poker hands with three of a kind we want to use “the suit of the single card with lower rank” and “the suit of the single card with higher rank” as projections, which might be easier to do with better notation. In probability, we need to combine notions of conditionals and independence to write down problems such as the classic Monty Hall problem. There

are a variety of other topics we have not yet covered in probability as well, such as infinite probability spaces, expected values, and variance.

Overall, there is considerable research and implementation left to be finished before these Coq libraries can be tested in the classroom. However, it will definitely be possible to craft Coq exercises in the spirit of many existing paper exercises, especially if we are willing to restrict the overall scope of the exercise set.

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