# Lecture Notes 2 Inductive Definitions

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Our first order of business will be to define the set of (valid) expressions of a language, introducing us to the notion of an *inductive definition* by a collection of *inference rules*. This lecture will introduce inference rules, what they mean, how to reason about them, and a few applications. The material may seem technical at first, but we will see many more examples of these concepts in the coming lectures.

Abstract syntax and inductive definitions are covered in Chapters 1 and 2 of Harper [Har16], but with a different presentation.

## **1** BNF grammars

The most concise way to define a set of expressions, and one you may have seen before, is to write a formal grammar in *Backus–Naur form* (BNF). In C311/B521 you might have written down the syntax of a simple "programming language" of booleans as follows:

Expressions e ::= true | false | (not e) | (if e e e)

In this class we are not using Racket, so we will instead write:

*Expressions* e ::= true | false | not(e) | if(e, e, e)

*Remark* 2.1. In practice, one often wants a friendlier notation for if-expressions, such as if *e* then *e* else *e*. For now we are talking about a mathematical or *abstract* conception of the syntax of a programming language, divorced from questions such as how to type it into a computer. Soon enough we will allow ourselves to write friendlier *concrete* syntax for our own sake, with the understanding that it is (in this class) a shorthand for abstract syntax.

Here *e* ranges over the set of all possible expressions, which are true, false, not(e) where *e* is an expression, and if(e, e, e) where each of these *es* is a (possibly different) expression. (Sometimes people write if(e, e', e'') to make it clearer that these *es* are not required to be the same.)

*Remark* 2.2. More precisely, we are defining a single grammatical category of *expressions*, corresponding to the *nonterminal e* on the left; there are two *terminal* symbols, namely true and false, and the vertical bars indicate alternatives.

Exercise 2.3. Are the following valid expressions?

- true yes
- not(true) yes
- 0 no
- if(true,false) no
- not no(!)
- if(true, false, ∅) − no

Note that the clauses of the grammar not only indicate what *is* a valid term but also what *isn't*: everything else.

**Definition 2.4.** *e* is a valid expression if and only if:

- *e* = true or
- *e* = false or
- e = not(e') where e' is a valid expression or
- $e = if(e_1, e_2, e_3)$  where  $e_1, e_2, e_3$  are valid expressions.

The "only if" here will be crucial to our ability to reason about the properties of valid expressions.

*Remark* 2.5. The HtDP-heads among you may imagine something like the following data definition:

; An Exp is one of: ; - true ; - false ; - (make-not Exp) ; - (make-if Exp Exp Exp)

## 2 Inference rules

To formally explain why if(not(true), false, true) is a valid expression, we might argue as follows:

- if(not(true), false, true) is an expression because
  - not(true) is an expression because
    - true is an expression
  - false is an expression
  - true is an expression

The outermost step of the argument is an instance of the general fact that  $if(e_1, e_2, e_3)$  is an expression if  $e_1$  is an expression,  $e_2$  is an expression, and  $e_3$  is an expression; in this case e is not(true) and so seeing that it is an expression relies in turn on the general fact that not(e) is an expression if e is an expression, and that true is an expression.

Let us introduce some terminology and notation to simplify this argument. We will write

e exp

for the assertion that *e* is an expression. We call exp a *judgment*, in the sense that this assertion "judges" that *e* is indeed an expression. (We can also call it an *assertion* or a *predicate*.)

There are four ways to judge e to be an expression. One is that not(e) is an expression if e is an expression. We write this as an *inference rule* with one premise  $(e \exp)$  and one conclusion  $(not(e) \exp)$ .

$$\frac{e \exp}{\operatorname{not}(e) \exp}$$

Two of the remaining three inference rules have no premises, while the final one has three premises.

$$\frac{1}{\text{true exp}} \qquad \frac{1}{\text{false exp}} \qquad \frac{e_1 \exp e_2 \exp e_3 \exp e_3 \exp e_2}{\text{if}(e_1, e_2, e_3) \exp e_3}$$

Remark 2.6. An inference rule with zero premises is called an axiom.

One benefit of inference rule notation is that we can chain these rules together:

true exp		
not(true) exp	false exp	true exp
if(not(tr	ue), false, true	e) exp

*Remark* 2.7. Chaining together inference rules as above forms a *derivation tree* whose root (at the bottom) can be concluded from its leaves/premises (at the top). If every leaf is an axiom then there are no remaining premises, the conclusion holds unconditionally, and we say that the derivation is *closed*.

Schematic picture of a derivation with wedges indicating subderivations.

*Exercise* 2.8. Write a (closed) derivation of not(not(true)) exp.

*Remark* 2.9. Note that we can instantiate e',  $e_1$ ,  $e_2$ ,  $e_3$  in the rules for not and if with any valid expression. In that way, those rules really stand for an infinite family of rules of the form "if *[blank]* is an expression then not(*[blank]*) is an expression," whereas the rules for true and false really are just singular rules.

Once again, crucially, we take the four inference rules above as specifying not only when something *can* be judged to be an expression but also when something *cannot*.

**Definition 2.10.** The judgment *e* exp holds if and only there is a closed derivation tree with conclusion *e* exp, built only out of the rules:

		<i>e</i> exp	$e_1 \exp$	$e_2 \exp$	$e_3 \exp$
true exp	false exp	not(e) exp	if	$(e_1,e_2,e_3)$	exp

This is essentially just a rephrasing of Definition 2.4.

*Remark* 2.11. We can read inference rules from top to bottom: "if e exp then not(e) exp." But we can also read them from bottom to top: "to show that not(e) exp, we can show that e exp." But importantly, inference rules are more than just random implications: they are always the *defining clauses of a judgment*.

#### 2.1 Aside: abstract syntax trees

If we turn the derivation of if(not(true), false, true) exp upside-down, we can see it as an *abstract syntax tree*, where each node corresponds to the rule being invoked (equivalently, the topmost constructor of the expression), and its child trees correspond to the list of immediate subexpressions.



**Definition 2.12.** *e* is a valid abstract syntax tree if and only if:



Again, this is essentially just a rephrasing of Definitions 2.4 and 2.10. *Remark* 2.13. These trees show up in actual implementations of programming languages, sometimes called *abstract syntax trees* or *parse trees*, as the data structures generated by parsers.

#### 2.2 More inductive structures

Inference rules are a powerful tool that will allow us to define a great many "inductive" structures, structures that are defined by a collection of (potentially self-referential) clauses and no others.

*Example* 2.14. We define the natural numbers  $\mathbb{N}$  as follows:

Natural numbers n ::= zero | suc(n)

where the "successor" of *n* represents n + 1. Equivalently, the judgment *n* nat is defined by the collection of inference rules:

 $\frac{n \text{ nat}}{\text{zero nat}}$   $\frac{suc(n) \text{ nat}}{suc(n) \text{ suc}(n)}$ 

*Example* 2.15. We could add natural numbers to our programming language:

Equivalently,

		<i>e</i> exp	$e_1 \exp$	$e_2 \exp$	$e_3 \exp$
true exp false exp		$not(e) exp$ $if(e_1, e_2, e_3)$		$(e_1,e_2,e_3)$	exp
		<i>e</i> exp		<i>e</i> exp	
Ze	ero exp	suc(e) exp	zer	o?( <i>e</i> ) exp	

We can also define multiple grammatical categories / judgments simultaneously with reference to one another.

*Example* 2.16. We define lists of natural numbers as follows:

Natural numbersn ::=zero | suc(n)Lists of naturals $\ell ::=$  $empty | cons(n, \ell)$ 

Equivalently,

	<i>n</i> nat		<i>n</i> nat	ℓ natlist
				()
zero nat	suc(n) nat	empty natlist	cons(n	, <i>t</i> ) natiist

*Example 2.17.* We define even and odd natural numbers as follows:

Evens 
$$e ::= zero | suc(o)$$
  
Odds  $o ::= suc(e)$ 

Equivalently,

	<i>o</i> odd	<i>e</i> even
zero even	$\overline{\operatorname{suc}(o) \operatorname{even}}$	$\overline{suc(o)  odd}$

*Remark* 2.18. BNF grammars will not be able to fully capture most of the languages we consider throughout the semester, because they have an infinite number of grammatical categories (types) and/or binding (variables); for these features we will need to use inference rules. However, BNF grammars will remain a convenient way to summarize some systems of inference rules.

So far all of our judgments have been *unary*: assertions about a single expression (often, that it is well-formed). We can also define *binary* judgments that make a joint assertion about two things.

Example 2.19. The binary judgment

#### *n* doubled Is n'

is defined by the following inference rules:

	n doubledIs $n'$
zero doubledIs zero	suc(n) doubledIs $suc(suc(n'))$

*Exercise* 2.20. There is exactly one concrete n for which we can derive

suc(zero) doubledIs n

What is that *n*? Write the derivation of the above judgment.

#### 2.3 Reasoning about inductive structures

**Lemma 2.21.** If *n* even then suc(suc(n)) even.

*Proof.* Suppose we have a derivation of *n* even. Then we can add two more rules to the bottom of that derivation to form a derivation of suc(suc(n)) even as required:

$$\frac{\frac{1}{n \text{ even}}}{\frac{\text{suc}(n) \text{ odd}}{\text{suc}(\text{suc}(n)) \text{ even}}}$$

So far we have used the "if" direction of a number of inductive definitions (as in Definitions 2.4, 2.10 and 2.12) to show that certain judgments hold, and we have used the "only if" direction to show that certain judgments do *not* hold (e.g.,  $\emptyset$  is not an expression). We can also use the "only if" direction to prove positive properties of judgments.

**Lemma 2.22** (Inversion for suc). If suc(m) nat then m nat.

*Proof.* The judgment *n* nat holds *only if* there is a derivation of *n* nat built out of repeated applications of the two rules

	<i>n</i> nat
zero nat	$\overline{\operatorname{suc}(n)}$ nat

In particular, any derivation of suc(m) nat must end with an application of one of these two rules. It cannot end with an application of the first rule, because then it would be a derivation of zero nat. So it must end with the second rule applied to a complete derivation of its premise *m* nat. From this we conclude *m* nat.

*Remark* 2.23. The above argument is an instance of a common reasoning pattern called *rule induction*, which we will make more precise shortly.

*Remark* 2.24. Whereas the second rule says that *m* nat is a *sufficient* condition for concluding that suc(m) nat, Lemma 2.22 states that it is in fact a *necessary* condition: that rule is the *only* way to conclude that suc(m) nat. In other words, we can "run the rule backwards," which is why it's called an *inversion lemma*.

## 3 Least sets

Before giving any more examples of rule induction, I want to pause and derive it mathematically from the "if and only if" characterization of inductive structures.

Let us write Exp for the set of valid expressions: the set of *e* for which *e* exp holds. Recall that by Definition 2.10,  $e \in Exp$  if and only if we have a derivation tree with conclusion *e* exp built only out of the following rules:

		<i>e</i> exp	$e_1 \exp$	$e_2 \exp$	$e_3 \exp$
true exp	false exp	not(e) exp	if	$(e_1, e_2, e_3) \epsilon$	exp

Let us refer to these four rules as *E*.

**Definition 2.25.** A set *X* is *E*-closed if the following four properties hold:

- true  $\in X$ ,
- false  $\in X$ ,
- $not(x) \in X$  whenever  $x \in X$ , and
- $if(x_1, x_2, x_3)$  whenever  $x_1, x_2, x_3 \in X$ .

**Theorem 2.26.** The set of expressions Exp is the least *E*-closed set; that is, (1) Exp is *E*-closed, and (2) for any *E*-closed set *X*, we have  $\text{Exp} \subseteq X$ .

#### Proof.

1. First we prove that Exp is *E*-closed. There are derivations of true exp and false exp, so by definition we have true, false  $\in$  Exp. Using the not rule, whenever we have a derivation of *e* exp we can build a derivation of not(*e*) exp. Finally, using the if rule, whenever we have derivations of  $e_1$  exp and  $e_2$  exp and  $e_3$  exp we can build a derivation of  $if(e_1, e_2, e_3)$ .

2. Suppose that X is *E*-closed and that we have a derivation of *e* exp; we translate that derivation into a proof that  $e \in X$ . Our derivation of *e* exp must end with one of the four rules of *E*. If it ends in true exp, then we must prove true  $\in X$ , which is true by *E*-closure. (Similarly for false exp.)

Alternatively, it could end with the following rule:

$$\frac{\frac{\vdots}{e \exp}}{\operatorname{not}(e) \exp}$$

In that case, we have at hand a derivation tree of *e* exp which is shorter than the tree we started with, and we can recursively translate that derivation into a proof that  $e \in X$ . Once that process is completed, we can apply *E*-closure to conclude that  $not(e) \in X$  as well. (The case for if is similar.)

*Example* 2.27. For a concrete example of the above proof, suppose that X is *E*-closed and show that  $if(not(true), false, true) \in X$  by converting each inference rule in the derivation below to an application of *E*-closure:

true exp		
not(true) exp	false exp	true exp
if(not(tr	ue), false, true	e) exp

*Remark* 2.28. The *E*-closure of Exp corresponds to the "if" direction of our definition: if we can build a derivation out of the rules of *E*, then we have a valid expression. The leastness corresponds to the "only if" direction: Exp contains only things built out of the rules of *E*. If Exp contained additional expression(s) *not* built out of the rules, those expressions would not be in every *E*-closed set.

Let's look at a more familiar example next.

**Definition 2.29.** A set *X* is *N*-closed if  $0 \in X$  and  $n \in X \implies n+1 \in X$ .

**Theorem 2.30.** The natural numbers  $\mathbb{N}$  are the least N-closed set.

Examples of other *N*-closed sets include  $\mathbb{Z}$ ,  $\mathbb{R}$ ,  $\mathbb{N} \cup \{-1\}$ , ... all of which satisfy the necessary closure conditions but include superfluous elements.

The characterization of  $\mathbb N$  as the least  $N\text{-}{\rm closed}$  set implies the principle of mathematical induction.

**Theorem 2.31** (Mathematical induction). To prove that P(n) holds for all  $n \in \mathbb{N}$ , it suffices to show that P(0) and  $\forall n \in \mathbb{N}.P(n) \implies P(n+1)$ .

*Proof.* Define  $\Sigma_P = \{n \in \mathbb{N} \mid P(n)\}$ . Clearly  $\Sigma_P \subseteq \mathbb{N}$  by construction. The hypotheses imply that  $\Sigma_P$  is *N*-closed, so by Theorem 2.30 we have  $\mathbb{N} \subseteq \Sigma_P$ . Hence  $\mathbb{N} = \Sigma_P$  and thus P(n) for all n.

Similarly, our characterization of Exp as the least *E*-closed set implies a principle of *rule induction*: to prove a property *P* for all expressions, we must show that the subset of expressions satisfying *P* is *E*-closed. That is, the property *P* is preserved by every inference rule.

**Theorem 2.32** (Rule induction for Exp). To prove that a property P(e) holds for all  $e \in Exp$ , it suffices to show that:

- P(true),
- P(false),
- for every e, if P(e) then P(not(e)), and
- for every  $e_1, e_2, e_3$ , if  $P(e_1), P(e_2)$ , and  $P(e_3)$ , then  $P(if(e_1, e_2, e_3))$ .

*Proof.* Define  $\Sigma_P = \{e \in \mathsf{Exp} \mid P(n)\}$ . Clearly  $\Sigma_P \subseteq \mathsf{Exp}$  by construction. The hypotheses imply that  $\Sigma_P$  is *E*-closed, so by Theorem 2.26 we have  $\mathsf{Exp} \subseteq \Sigma_P$ . Hence  $\mathsf{Exp} = \Sigma_P$  and thus P(e) for all  $e \in \mathsf{Exp}$ .

## 4 Rule induction

**Theorem 2.33** (Rule induction for nat). To prove that a property P(n) holds for all n nat, it suffices to show that:

- P(zero) and
- for every n, if P(n) then P(suc(n)).

**Lemma 2.34.** If n nat then there exists some n' such that n doubled ls n'.

*Proof.* We prove P(n) = "there exists some n' such that n doubled n'" by rule induction. We must show that P is closed under the rules for the nat judgment, of which there are two.

• The first rule is:

#### zero nat

We must show that *P* holds in this case, i.e. that P(zero) holds, i.e. that there exists some n' such that zero doubledIs n'. We choose n' = zero, because:

#### zero doubledIs zero

• The second rule is:

## *n* nat

## suc(n) nat

We must show that if the premise satisfies *P* (this is called our *inductive hypothesis*), then the conclusion satisfies *P*. That is, we must show that for all *n*, if *P*(*n*) holds then *P*(suc(*n*)) holds. Unfolding the definition of *P*, our inductive hypothesis *P*(*n*) is that there exists *n*' such that *n* doubledls *n*'. We must show that *P*(suc(*n*)) holds, which is to say that there exists *n*'' such that suc(*n*) doubledls *n*''. We choose *n*'' = suc(suc(*n*')) because:

$$\frac{n \text{ doubledIs } n'}{\text{suc}(n) \text{ doubledIs suc}(\text{suc}(n'))} \square$$

Inversion Lemma 2.22 also follows from rule induction, for the property P(n) = "*n* nat, and if *n* is of the form suc(*n*') for some *n*', then *n*' nat."

**Theorem 2.35** (Rule induction for doubledIs). To prove that a property P(n, n') holds for all n doubledIs n', it suffices to show that:

- P(zero, zero) and
- for all n, n', if P(n, n') then P(suc(n), suc(suc(n'))).

**Lemma 2.36.** If *n* doubledIs *n'* then *n'* even.

*Proof.* We prove P(n, n') = "n' even" by rule induction. There are two cases:

• Case 1:

#### zero doubledIs zero

We must prove that the conclusion satisfies P, i.e. P(zero, zero), or that zero even. This is an axiom.

• Case 2:

$$\frac{n \text{ doubledIs } n'}{\operatorname{suc}(n) \text{ doubledIs } \operatorname{suc}(\operatorname{suc}(n'))}$$

We must prove that *P* is closed under this rule: in other words, if P(n, n') holds then  $P(\operatorname{suc}(n), \operatorname{suc}(\operatorname{suc}(n')))$  holds. Our inductive hypothesis P(n, n') is that n' even, and we must prove that  $\operatorname{suc}(\operatorname{suc}(n'))$  even. We already proved this in Lemma 2.21.

**Lemma 2.37.** If *n* doubled is n' and *n* doubled is n'' then n' = n''.

*Proof.* We prove P(n, n') = "for all n'' for which n doubled n'', we have n' = n''" by rule induction (on n doubled n'). There are two ways that the derivation of n doubled n' can end:

• Case 1:

#### zero doubledIs zero

Show that P(zero, zero): if zero doubledIs n'', then zero = n''. This follows by inversion, because the only rule that could derive such a thing is

#### zero doubledIs zero

Therefore n'' = zero, which is what we wanted to show.

• Case 2:

*n* doubled Is n'

suc(n) doubledIs suc(suc(n'))

Prove P(n, n') implies P(suc(n), suc(suc(n'))). The latter property is: if suc(n) doubled n'', then suc(suc(n')) = n''. Our inductive hypothesis is the analogous statement about the premise: if *n* doubled m then n' = m.

By inversion on the supposed derivation of suc(n) doubled is n'', only one rule applies:

 $\frac{n \text{ doubledIs } m'}{\operatorname{suc}(n) \text{ doubledIs suc}(\operatorname{suc}(m'))}$ 

This tells us that n'' is of the form suc(suc(m')), so it remains to prove that suc(suc(m')) = suc(suc(n')). By *n* doubled *n'* and *n* doubled *m'* and our inductive hypothesis, m' = n'.

The above proof actually uses *nested* rule induction: we consider the two cases of *n* doubledIs n', and in each case, we consider the two cases of *n* doubledIs n'' (implicitly, in our appeal to inversion).

By Lemmas 2.34 and 2.37, for every n nat there is exactly one n' such that n doubledIs n'. In other words, doubling is in fact a *function* on natural numbers. We will often use inference rules to define functions in this way.

**Theorem 2.38** (Structural recursion for nat). *To define a function* f *on natural numbers, one must:* 

- *define* f(zero), and
- define f(suc(n)) for arbitrary n, given the value of f(n).

*Proof.* As with doubledIs, we can define a binary judgment n maps-to n' by the inference rules:

	<i>n</i> maps-to <i>n</i> ′		
zero maps-to	$\overline{\operatorname{suc}(n)}$ maps-to $(\ldots n' \ldots)$		

,

The proofs of Lemmas 2.34 and 2.37 show that for any *n* nat there is exactly one n' for which *n* maps-to n', so that maps-to is the graph of a function.

To perform rule induction on mutually-defined judgments, we need mutual induction hypotheses.

Theorem 2.39 (Rule induction for even/odd). If

- P(zero),
- for all o odd,  $Q(o) \implies P(\operatorname{suc}(o))$ , and
- for all e even,  $P(e) \implies Q(\operatorname{suc}(e))$ ,

then P(e) holds for all e even and Q(o) holds for all o odd.

**Theorem 2.40.** If e even then e nat, and if o odd then o nat.

*Proof.* We prove these by mutual rule induction.

• For the rule

#### zero even

we must show zero nat, which is true by an axiom.

• For the rule

our inductive hypothesis is o nat and we must show suc(o) nat, which holds by the suc rule for nat.

• For the rule

$$\frac{e \text{ even}}{\operatorname{suc}(o) \text{ odd}}$$

our inductive hypothesis is *e* nat and we must show suc(e) nat, which holds by the suc rule for nat.

Remark 2.41. HtDP-heads might have noticed

collection of inference rules : data definition :: rule induction : template

except that rule induction is about constructing a *proof* out of the input derivation. Structural recursion brings us full circle, using rule induction as a means of defining an ordinary function using a recursive template.

## **5** Evaluation

I didn't want to go a whole lecture without saying anything about programming languages. Going back to the boolean expression language from the beginning of the lecture, we can use inference rules to define a binary *evaluation* judgment,  $e \downarrow e'$ , which explains how to recursively simplify any expression *e*.

		$e \Downarrow true$	$e \Downarrow \texttt{false}$
true↓true	false ↓ false	$not(e) \Downarrow false$	$not(e) \Downarrow true$
$e_1 \Downarrow$	true $e_2 \Downarrow e_2'$	$e_1 \Downarrow \texttt{false}$	$e_3 \Downarrow e_3'$
if	$(e_1, e_2, e_3) \Downarrow e'_2$	$if(e_1, e_2, e_3)$	$) \Downarrow e'_3$