# Lecture: Syntax, Inference Rules, and Our First Example PL Programming Languages (CSCI 3300)

Prof. Harley Eades (heades@gru.edu).

## 1 Introduction to the course

Programming languages (PLs) are ubiquitous in computer science. Any student of computation has used some programming language whether they a computer scientist, software engineer, or information technologist. This then implies that the PLs are the most important artifacts in the computational sciences.

One would find it hard to argue that PLs are important, but one could argue that we already have a lot of PLs, and it seems as if these are enough, but is this true? I claim that it is not the case. Computational devices are ubiquitous in our society. They power our power plants, our water supply, virtually every consumer product is built using some computational device, in addition they power our automobiles, our air planes, and our medical devies. It is safe to say that we put our vary lives in the hands of computational devices, which implies that we put our lives in the hands of the programmers who designed the software systems powering such devies. What happens when a software bug finds its way into the safety critical devices that we depend on so dearly?

In 2010 there were a approximately a hundred reports made to the National Highway Traffic Safety Administration of potential problems with the braking system of the 2010 Toyota Prius [?]. The problem was that the anti-lock braking system would experience a "short delay" when the brakes where pressed by the driver of the vehicle [?]. This actually caused some crashes. Toyota found that this short delay was the result of a software bug, and was able to repair the the vehicles using a software update [?]. Another incident where substantial harm was caused was in 2002 where two planes collided over Überlingen in Germany. A cargo plane operated by DHL collided with a passenger flight holding fifty-one passengers. Air-traffic control did not notice the intersecting traffic until less than a minute before the collision. It was officially ruled by the German Federal Bureau of Aircraft Accidents Investigation that the on-board collision detection software was indeed faulty [?].

Avi Rubin is a professor at John Hopkins university and has recently shown that pacemakers can be remotely hacked. It may come to the surprise to computer scientists, but pacemakers have wifi access now. Avi has shown that this connection can be indeed hacked, and could be used by the attacker to cause a fatal heart attack. These hacks are indeed due to software vulnerabilities – that is bugs.

How can we prevent these disastrous problems? Currently, we using testing. Programmers come up with a huge number of test inputs to a program, and then they run the program on each one of these tests. If the program behaves across all the tests, then we consider the program as being correct. Does this catch all bugs? No! In fact, no amount of testing could ever **prove** the absence of all bugs! If the programmer is not clever enough, then they could indeed miss some corner case. This is evident in the examples above. So is it possible to be sure a program contains no bugs? The only way to be sure a program contains no bugs is by mathematically proving that it meets some specification. Every programmer uses some criteria to write a program, and a programs specification corresponds to this criteria transformed into a list of mathematical formulas. Then if one can prove that these formulas are all true with respect to the program, then one can be sure with mathematical certainty that it is correct. That is bug free. The joy of this approach is that if one of the formulas cannot be proven, then a bug exists. Furthermore, this implies that the bug would be found during development instead of in production. So for example, the Toyota Prius bug in the

breaking system would have been found before all those crashes, and costly recalls. So why is it the case that companies use testing instead of mathematical proof? The reason is that current programming languages are not amendable to mathematical proofs. There are just to many complex interactions.

Lets consider an example. Suppose I say a function squared takes in an int and outputs an int, then we know something about squared, we know that if we take and apply it to, say, 2, then we know that it will give us back an int, perhaps, 4. Is it guaranteed that this is all it does? Suppose we are given the squared function and it is written in C# do we know that it does nothing more than compute some integer? The answer is most definitely no. In C# the function squared could ask for some user input, print data to the screen, and even make some network transactions or ask for the time of day. Hidden effects like these are all called side effects. So in a programming language like C#, C, C++, Java, PHP, Python, and many more mainstream PLs we cannot really be sure what squared is doing. Thus making it very difficult to reason about these programs. So can we fix this problem? Can we design a PL that is more amendable to mathematical reasoning?

I claim – as do many other PL researchers – that we can prevent software bugs by mathematical reasoning by adopting new more rigorously studied programming languages. In fact, designing these types of PLs is the very driving force of modern day PL research. To design these types of programming languages we must revisit the very foundations of programming languages, and rebuild from the ground up. It turns out that by adopting the notion of a pure strongly typed PL we can rule out a large number of software bugs during development without using testing, and before lives are taken!

A pure strongly typed PL is one in which there are either no side effects at all or all side effects are accounted for in the type system. Consider the squared function again. The type of this function can be written as  $int \rightarrow int$  which means taken an int as input, and then return an int as output. Now suppose squared prints out some data to the screen, then in a pure strongly typed PL the type becomes  $int \rightarrow IO(int)$ . Here IO(int) means the function returns an int and conducts some input-output effect. This allows us to take into account side effects when reasoning about the program. So how does all this work? Well that is part of the topic of this course.

Another part of the major topic of this course is on pure strongly typed **functional** PLs. Functional programming languages differ from the PLs like C# or Java by having a mathematical foundation. The sole datatype is a function, and in fact, the only thing a functional PL contains are functions. It turns out that we can do a lot with this paradigm. PLs like C#, Java, and C are known as imperative and object oriented languages. These have a foundation in an actual machine like a Turing Machine, while, functional PLs have their foundation in mathematics. However, it turns out that they are equivalent in power. One major part of this course is to understand the design and implementation of functional PLs. Functional programming languages are also great for exploring different design features, and we will use them to study features like parametric polymorphism, and recursion. By the end of this course each student will understand how different each PL can be, and be able to better differentiate between the various PLs when choosing one for a particular project.

So a natural question is then is this all theoretical? What baring does any of this have on the real world? A lot! The industry is already taking steps to push the old style of programming out, and adopt new PLs that are highly motivated by the pure strongly typed functional PLs. A great example is Apple's new PL called Swift. They have decided to completely abandon all Objective C code from all their IOS devices. Swift is based on strongly typed functional programming because of its brevity of code, and the strong correctness guarantees. Another example, is Mozilla's Rust programming language which is a systems programming language which is designed for speed. A final example is the Haskell programming language. Haskell has paved the way for functional programming language research and much of the features of Swift and Rust where first studied in Haskell, and Haskell is used in the industry, especially, the financial sector. Lastly, Haskell is the last major topic of this course.

## 2 Grammars

The human interface to computational devices are PLs, but the interface to PLs is syntax. So how to we formally define the syntax of programming languages? We use a mathematical device called a grammar. Now instead of going into the fascinating area of formal language theory – which is where grammars were studied extensively – we will instead focus on several examples to learn how grammars work.

We can consider the syntax of programming languages to represent a formal language. A formal language is a set words generated from a set of symbols. Now consider the question, how can we generate all of these words? This is the very goal of a grammar.

S	$\rightarrow$	aE
E	$\rightarrow$	$\boldsymbol{a}$
E	$\rightarrow$	aO
0	$\rightarrow$	aE

A grammar consists of one or more productions denoted  $A \to s$  where we call A the non-terminal of the production. Note that the lefthand side of a production is always a non-terminal, and not a composition of terminals and non-terminals. Each production generates either a symbol in the language – also called a non-terminal – or is the combination of terminals (symbols of the language) and non-terminals. So the righthand side of a production is a composition of non-terminals and terminals, but it may also only consist of non-terminals. In the example above I have put boxes around each production, colored the non-terminals red, and the terminals blue. The non-terminal S is a special non-terminal called the start non-terminal. The start non-terminal is always used first no matter what. A grammar may also have several productions associated with the start non-terminal.

Once we have a grammar we can use it to generate any word in its corresponding language. We generate words using a grammar by constructing a derivation. Lets consider an example:

A derivation always begins with one of the productions whose lefthand side is the start symbol. Then we proceed by simply choosing one of the non-terminals on the lefthand side – it does not matter which is chosen or in what order – and replacing it with the lefthand side of one of its associated productions. We repeat this until all non-terminals have been replaced by terminals. So every derivation ends with a word in the language the grammar generates.

This is great, but now ask the question, what if we are given a word, say, this word

, is it possible to test to see if it is in the language  $L_{\rm even}$ ? Now for our simple example we could just tally up the number of a's and if its even, then accept, otherwise we reject the above word, but if this were a more complex language, like say, a PL, then that would not work. It turns out that we can say, well, if there exists a derivation that ends with the above word, then it is a member of the language  $L_{\rm even}$  otherwise it is not. In fact, there are known algorithms for doing just that. In fact, this is the basic operation for parsers, and when using a PL you obtain a syntax error, the compiler took the input and tried to derive using the grammar of the PL. So grammars are the driving force behind syntax, and parser technology.

Lets consider another more interesting example:

$$\begin{array}{ccc} S & \rightarrow & () \\ S & \rightarrow & (S) \\ S & \rightarrow & SS \end{array}$$

What language does this grammar generate? Its the language of balanced parenthesis. A few example words would then be (), (), (), (), (), (), (), and (), (), (), (), (), and (), (), (), and (), and (), (),

I would like to mention one notational convention I am using. Notice that if I want to have multiple productions for a symbol, like S in the previous example, I write multiple productions with the same non-terminal on the lefthand side. There is a shorthand convention that we will use where we combine these all into a single production. For example, the previous grammar can be written like this:

$$S \rightarrow () | (S) | SS$$

Here the | symbol should be read as "or."

Finally, lets see how we can put all this together to formally specify the syntax of a programming language. The following grammar is a grammar of a toy programming language:

```
\begin{array}{ll} \text{(booleans)} & b \to \mathsf{true} \mid \mathsf{false} \mid b \land b \mid b \lor b \mid \neg b \mid e > e \\ \text{(expressions)} & e \to 0 \mid e+1 \mid e*e \mid e+e \mid e-e \\ \text{(command)} & S \to \mathsf{skip} \mid x := e \mid \mathsf{if} \, b \, \mathsf{then} \, S \, \mathsf{else} \, S \mid \mathsf{while} \, b \, \mathsf{do} \, S \end{array}
```

This is slightly more complex, but the same rules we have been discussing above hold. An example derivation is:

```
\begin{array}{ll} S & \Rightarrow & \text{if } b \text{ then } S \text{ else } S \\ & \Rightarrow & \text{if } e > e \text{ then } S \text{ else } S \\ & \Rightarrow & \text{if } e + 1 > e \text{ then } S \text{ else } S \\ & \Rightarrow & \text{if } 0 + 1 > e \text{ then } S \text{ else } S \\ & \Rightarrow & \text{if } 0 + 1 > e + 1 \text{ then } S \text{ else } S \\ & \Rightarrow & \text{if } 0 + 1 > e + 1 \text{ then } S \text{ else } S \\ & \Rightarrow & \text{if } 0 + 1 > e + 1 \text{ then } S \text{ else } S \\ & \Rightarrow & \text{if } 0 + 1 > 0 + 1 + 1 \text{ then } S \text{ else } S \\ & \Rightarrow & \text{if } 0 + 1 > 0 + 1 + 1 \text{ then } \text{ while } b \text{ do } S \text{ else } S \\ & \Rightarrow & \text{if } 0 + 1 > 0 + 1 + 1 \text{ then } \text{ while true do skip else } S \\ & \Rightarrow & \text{if } 0 + 1 > 0 + 1 + 1 \text{ then } \text{ while true do skip else } S \\ & \Rightarrow & \text{if } 0 + 1 > 0 + 1 + 1 \text{ then } \text{ while true do skip else skip} \end{array}
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At this point we have seen basically all that we will need to know about grammars for this course and the large majority of PL research. However, the previous grammar we saw for the toy PL is a bit more formal than we will need.

So we now look at an alternative presentation of the same grammar:

```
\begin{array}{ll} \text{(booleans)} & b ::= \mathsf{true} \mid \mathsf{false} \mid b \wedge b \mid b \vee b \mid \neg b \mid e > e \\ \text{(expressions)} & e ::= 0 \mid e+1 \mid e*e \mid e+e \mid e-e \\ \text{(command)} & c ::= \mathsf{skip} \mid x := e \mid \mathsf{if} \, b \, \mathsf{then} \, c \, \mathsf{else} \, c \mid \mathsf{while} \, b \, \mathsf{do} \, c \end{array}
```

This is called the Backus-Naur Form (BNF) of the formal grammar above. In PL we usually do not indicate a start non-terminal, but rather we give the grammar as a means of understanding how to derive legal expressions of the language.

# 3 Iffy-Lang

In this section I would like to introduce the PL Iffy-Lang. Its sytax is as follows:

(Bool) 
$$b ::= 1 \mid 0 \mid b_1 \wedge b_2 \mid b_1 \vee b_2 \mid \text{if } b \text{ then } b_1 \text{ else } b_2$$

Now the above syntax allows us to write small programs like:

if 
$$b$$
 then  $0$  else  $1$ 

However, the syntax does not describe how to actually compute with such programs. So how do we run them?

The operational behavior of a program is often described by a set of rules called the reduction rules. The following rules describe the operational behavior of Iffy-Lang:

One should read these rules from the top down as an "if-statement", that is, the rule:

$$\frac{P_1 \quad \cdots \quad P_n}{C}$$
 NAME

should be read as if  $P_1, \dots, P_n$  hold, then C holds. So for example the rule:

$$\frac{b \rightsquigarrow b'}{(\text{if } b \text{ then } b_1 \text{ else } b_2) \rightsquigarrow (\text{if } b' \text{ then } b_1 \text{ else } b_2)} \text{ }^{\text{IF}}$$

says, if  $b \rightsquigarrow b'$  holds, then (if b then  $b_1$  else  $b_2$ )  $\rightsquigarrow$  (if b' then  $b_1$  else  $b_2$ ) holds.

Now the symbol  $\rightsquigarrow$  should be read as "reduces to", hence, we read  $b \rightsquigarrow b'$  as "b reduces to b'". The rules above can be seen as defining what it means for a boolean b to reduce to b'.

The previous set of inference rules tell us how to reduce an Iffy program for one step only. The following rules generalize this to allow for multiple steps:

$$\frac{b_1 \leadsto b_2}{b_1 \leadsto b_2} \quad \text{Refl} \quad \frac{b_1 \leadsto b_2}{b_1 \leadsto b_2} \quad \text{Step} \quad \frac{b_1 \leadsto^* b_2 \quad b_2 \leadsto^* b_3}{b_1 \leadsto^* b_3} \quad \text{Mult}$$

The judgment  $b \rightsquigarrow^* b'$  is called multi-step reduction.

# 4 Functional Iffy

The Iffy PL as it is defined in the previous section is not very powerful. It only allows for defining very basic boolean programs, and does not encourage abstraction. In this section we extend Iffy with functions and global definitions. This will allows us to construct reusable functions on the booleans, and thus allowing us to use abstraction, and even define new boolean operators that are not currently allowed in Iffy. Furthermore, we can define full Iffy programs.

Adding new structures to programming languages is a huge task, and it gets bigger as the PL gets more complicated. We will see here that when adding functions to Iffy we have to add the syntax, and then add proper reduction rules capturing the operational meaning of functions. The latter point will require some non-trivial tools.

# 5 Syntax of Functional Iffy

The first step in this extension is adding the syntax. The syntax for Functional Iffy is as follows:

```
\begin{array}{lll} \text{(Programs)} & P & ::= & x := b \mid P_1 \, \mathsf{NEWLINE} \, P_2 \\ \text{(Booleans)} & b & ::= & x \mid 1 \mid 0 \mid b_1 \wedge b_2 \mid b_1 \vee b_2 \mid \mathsf{if} \, b \, \mathsf{then} \, b_1 \, \mathsf{else} \, b_2 \mid \mathsf{fun} \, x \Rightarrow b \mid \mathsf{app} \, b_1 \, b_2 \end{array}
```

This syntax gives us more control and a lot more power. To define a global definition we use the syntax x := b, where x is the name of the program we are defining, and b is the actual boolean program. The name x can be any string following the Haskell variable naming convention. We will use the convention that every globally defined name must be defined before it can be used. As an example we can define the negation operator on booleans as follows:

$$\mathsf{not} := \mathsf{fun}\, x \Rightarrow \mathsf{if}\, x \, \mathsf{then}\, 0 \, \mathsf{else}\, 1$$

The following program defines the De Morgan's dual of conjunction:

```
\mathsf{not} := \mathsf{fun}\, x \Rightarrow \mathsf{if}\, x \, \mathsf{then}\, 0 \, \mathsf{else}\, 1
\mathsf{dmconj} := \mathsf{fun}\, x \Rightarrow \mathsf{fun}\, y \Rightarrow \mathsf{app}\, \mathsf{not}\, ((\mathsf{app}\, \mathsf{not}\, x) \wedge (\mathsf{app}\, \mathsf{not}\, y))
```

The syntax for functions is  $\operatorname{fun} x \Rightarrow b$  where x is the name of the input and b is the body of the function. These are called anonymous unary functions. Given a function  $b_1$  we can apply it to an argument using the syntax  $\operatorname{app} b_1 b_2$  where  $b_2$  is any other program. Thus,  $\operatorname{app} \operatorname{not} 0$  is equivalent to  $\operatorname{app} (\operatorname{fun} x \Rightarrow \operatorname{if} x \operatorname{then} 0 \operatorname{else} 1) 0$ . Function application is the locus of computation, and is where all the fun happens during reduction.

#### 5.1 Evaluation of Functional Iffy Programs

At this point we arrive at the difficult part of defining the operational behavior of our new additions. The question we should be asking ourselves is how does  $\operatorname{\mathsf{app}} b_1 b_2$  compute? Suppose we have the identity function  $\operatorname{\mathsf{fun}} x \Rightarrow x$ . This function simply echos back whatever input we give it. So for example,  $\operatorname{\mathsf{app}} (\operatorname{\mathsf{fun}} x \Rightarrow x) 0$  should reduce to 0, but how do we capture this in a rule?

**Parse Trees.** An extremely useful tool when thinking about syntax of programming languages is the notion of a parse tree. The following definition defines how to compute the parse tree of a given Iffy boolean program:

**Definition 1.** The following meta-function defines how to compute the parse tree of a given boolean program by structural recursion on the input program:

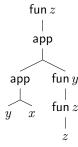
We call pT b a meta-function because it operators outside of the language, and is defined on programs, as opposed to being a program itself. This meta-function is defined by **structural recursion** on the input program because each recursive call is done on a subpiece of the input program. This style of definition is typical of the definitions in programming language design.

Free and bound variables. Consider the syntax for functions  $\operatorname{fun} x \Rightarrow b$ . The body of the function, b, may then depend on the input variable x. This may seem like an obvious point, but there are now some non-trivial programs that can cause confusion. An example of one is the program  $\operatorname{app}(\operatorname{fun} x \Rightarrow x) x$ , and another one is  $\operatorname{fun} x \Rightarrow \operatorname{fun} x \Rightarrow x$ . The input variable to a function, for example x in  $\operatorname{fun} x \Rightarrow b$ , is called a bound variable where its scope is limited to the body of the function b. One should think of these as local variables. A bound variable is always associated with a binder which is the "fun" in the syntax of functions. Consider the first example above,  $\operatorname{app}(\operatorname{fun} x \Rightarrow x) x$ , the x in the body of the function is bound, but the x in the argument position is not bound. Unbound variables are called **free variables**. In the second example,  $\operatorname{fun} x \Rightarrow x$ , the x in the body of the inner function is bound to the second binder, and not the first.

Parse trees make it a lot easier to see which variables are bound, and their associated binders. First, construct the parse tree of a program, and then trace the tree in reverse starting with a variable – the leaf of the tree – then the first binder we hit where the input variable matches the variable we started with is the variables binder, and we conclude that the variable is bound, otherwise, we did not find a binder, and we conclude it is a free variable. For example, consider the following parse tree:

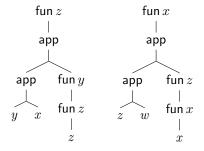


If we place our finger on the x leaf, and then trace backwards through the tree we will hit fun x, and then we can conclude that x is bound. If we attempt the same procedure for the leaf labeled y we never hit a binder matching y, and thus, y is a free variable. Consider a second example:



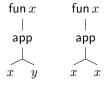
In the previous example the variables y and x are both free, while the variable z is bound to the inner-most binder fun z and not the top most one.

Notice that renaming variables in a program does not change its operational behavior. Consider the two parse trees:



These two denote the exact same program. We say that they are equivalent up to renaming of variables. The process of renaming variables is called  $\alpha$ -conversion.

Keep in mind that when renaming variables the status of the varible – bound or free – must be the same. As an example the following two programs are **not the same**:



At this point we can define a bound variable as a variable underneath a binder labeled with its name, otherwise we call the variable bound.

There is an alternative way of determining when a variable is bound or free by defining the set of all free variables of a program:

```
\begin{array}{lll} \mathsf{FV}(0) & = & \emptyset \\ \mathsf{FV}(1) & = & \emptyset \\ \mathsf{FV}(x) & = & \{x\} \\ \mathsf{FV}(b_1 \wedge b_2) & = & \mathsf{FV}(b_1) \cup \mathsf{FV}(b_2) \\ \mathsf{FV}(b_1 \vee b_2) & = & \mathsf{FV}(b_1) \cup \mathsf{FV}(b_2) \\ \mathsf{FV}(\text{if } b \text{ then } b_1 \text{ else } b_2) & = & \mathsf{FV}(b) \cup \mathsf{FV}(b_1) \cup \mathsf{FV}(b_2) \\ \mathsf{FV}(\text{fun } x \Rightarrow b) & = & \mathsf{FV}(b) - \{x\} \\ \mathsf{FV}(\text{app } b_1 \ b_2) & = & \mathsf{FV}(b_1) \cup \mathsf{FV}(b_2) \end{array}
```

Then a variable x is free in the program b iff  $x \in FV(b)$ . Otherwise we say it is bound.

Similarly, we can construct the set of all bound variables:

```
\begin{array}{lll} \mathsf{BV}(0) & = & \emptyset \\ \mathsf{BV}(1) & = & \emptyset \\ \mathsf{BV}(x) & = & \emptyset \\ \mathsf{BV}(b_1 \wedge b_2) & = & \mathsf{BV}(b_1) \cup \mathsf{BV}(b_2) \\ \mathsf{BV}(b_1 \vee b_2) & = & \mathsf{BV}(b_1) \cup \mathsf{BV}(b_2) \\ \mathsf{BV}(\text{if } b \text{ then } b_1 \text{ else } b_2) & = & \mathsf{BV}(b) \cup \mathsf{BV}(b_1) \cup \mathsf{BV}(b_2) \\ \mathsf{BV}(\text{fun } x \Rightarrow b) & = & \mathsf{BV}(b) \cup \{x\} \\ \mathsf{BV}(\text{app } b_1 \ b_2) & = & \mathsf{BV}(b_1) \cup \mathsf{BV}(b_2) \end{array}
```

Now a variable x is bound in a program b iff  $x \in BV(b)$ . Otherwise we say it is free.

**Substitution.** Our ultimate goal is to be able to define the evaluation rules for function application. That is, if we have an expression  $app(fun x \Rightarrow x) 0$  how to we define the evaluation rules so that this expression evaluates to the expression 0?

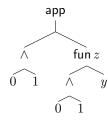
To answer this question we need to define a new meta-function called **capture-avoiding substitution**. This is an operation that takes as input two Iffy programs, say  $b_1$  and  $b_2$ , and the name of a free variable, say x, that may or not be free in  $b_2$ , and then outputs a new Iffy program where every occurrence of x in  $b_2$  is replaced by  $b_1$ . Consider an example, an Iffy program in parse-tree form is defined as follows:

$$\begin{array}{c}
\mathsf{app} \\
x & \mathsf{fun} z \\
x & y
\end{array}$$

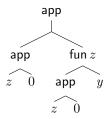
Then substituting the program 1 for x in the above program yields:

$$\begin{array}{c|c}
\text{app} \\
\hline
1 & \text{fun } z \\
\hline
1 & y
\end{array}$$

If we replace x with program  $0 \wedge 1$  then we obtain the following:

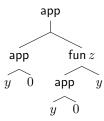


Now if we replace x with the program app z 0, note that z is a free variable, we obtain the following:



This does not seem quite right, because the variable z was free in the program we were replacing x with, and so it should always remain free after the substitution, but at the second location of x the variable z has become bound. This is called **variable capture** and is something that we must prevent.

To prevent variable capture first rename all the variables in  $b_1$  so that they are distinct from the variables in  $b_2$ . This is called  $\alpha$ -conversion. Then do the substitution. So if we replace x with the program app z 0, we first change z to a new name, say y, resulting in app y 0, and then do the substitution obtaining:



We can see that every occurrence of y is free in the final result.

At this point we define the substitution function on programs.

**Definition 2.** Capture-avoiding substitution is defined as follows where it is assumed that the variable names in  $b_1$  are distinct from the variable names in  $b_2$ .

$$\begin{array}{lll} [b_1/x]0 & = & 0 \\ [b_1/x]1 & = & 1 \\ [b_1/x]x & = & b_1 \\ [b_1/x]y & = & y \\ \hline & where \ x \neq y \\ [b_1/x](b_2 \wedge b_3) & = & ([b_1/x]b_2) \wedge ([b_1/x]b_3) \\ [b_1/x](b_2 \vee b_3) & = & ([b_1/x]b_2) \vee ([b_1/x]b_3) \\ \hline & [b_1/x](\text{if } b \text{ then } b_2 \text{ else } b_3) & = & \text{if } ([b_1/x]b) \text{ then } ([b_1/x]b_2) \text{ else } ([b_1/x]b_3) \\ \hline & [b_1/x](\text{fun } x \Rightarrow b_2) & = & \text{fun } x \Rightarrow b_2 \\ \hline & [b_1/x](\text{fun } y \Rightarrow b_2) & = & \text{fun } y \Rightarrow [b_1/x]b_2 \\ \hline & where \ x \neq y \end{array}$$

Here is an example of using this function:

```
\begin{array}{ll} [0/x](\operatorname{fun} y \Rightarrow \operatorname{if} x \operatorname{then} 1 \operatorname{else} (x \vee y)) &=& \operatorname{fun} y \Rightarrow [0/x](\operatorname{if} x \operatorname{then} 1 \operatorname{else} (x \vee y)) \\ &=& \operatorname{fun} y \Rightarrow \operatorname{if} \left([0/x]x\right) \operatorname{then} \left([0/x]1\right) \operatorname{else} \left([0/x](x \vee y)\right) \\ &=& \operatorname{fun} y \Rightarrow \operatorname{if} 0 \operatorname{then} \left([0/x]1\right) \operatorname{else} \left([0/x](x \vee y)\right) \\ &=& \operatorname{fun} y \Rightarrow \operatorname{if} 0 \operatorname{then} 1 \operatorname{else} \left([0/x](x \vee y)\right) \\ &=& \operatorname{fun} y \Rightarrow \operatorname{if} 0 \operatorname{then} 1 \operatorname{else} \left(\left([0/x]x\right) \vee \left([0/x]y\right)\right) \\ &=& \operatorname{fun} y \Rightarrow \operatorname{if} 0 \operatorname{then} 1 \operatorname{else} \left(0 \vee \left([0/x]y\right)\right) \\ &=& \operatorname{fun} y \Rightarrow \operatorname{if} 0 \operatorname{then} 1 \operatorname{else} \left(0 \vee \left([0/x]y\right)\right) \\ &=& \operatorname{fun} y \Rightarrow \operatorname{if} 0 \operatorname{then} 1 \operatorname{else} \left(0 \vee \left([0/x]y\right)\right) \\ \end{array}
```

Evaluation. We have finally arrived at the main event evaluation of functional Iffy! Suppose we were asked

to run the Iffy program  $\operatorname{\mathsf{app}}(\operatorname{\mathsf{fun}} x \Rightarrow x) \, 0$ , how would we do this? It turns out that a special rule called the  $\beta$ -rule can be defined that allows us to simplify a program of the form  $\operatorname{\mathsf{app}}(\operatorname{\mathsf{fun}} x \Rightarrow b) \, b'$  for any programs b and b'. A program of this form is called a  $\beta$ -reducible expression ( $\beta$ -redex or redex).

To simplify a  $\beta$ -redex we will use capture-avoiding substitution. The following inference rule defines evaluation for  $\beta$ -redexes:

$$\overline{\mathsf{app}\,(\mathsf{fun}\,x\Rightarrow b)\,b'\leadsto [b'/x]b}\quad \mathsf{Beta}$$

This rule says that in one step we can evaluate a  $\beta$ -redex into a substitution where we replace the variable, x, in b with the argument the function is being applied to b'. Thus, function application is just substitution or replacement. Note that on the left-hand side x is bound in b, but when we move to the right-hand side we rip out the application and the binder, and thus, x changes from a bound variable to a free variable which can be replaced using substitution. In this rule we write [b'/x]b because we do not yet know what b' and b are, but when evaluating a particular example we will know what they are, and thus, [b/x]b should be replaced with the actual output of the function.

We can now state the final single-step evaluation rules for functional Iffy:

The multiset reduction rules do not change. They remain completely the same.

We now give several example derivations:

#### Example 3.

$$\frac{}{\mathsf{app}\,(\mathsf{fun}\,x\Rightarrow x)\,0\leadsto 0}\;\mathsf{Beta}$$

The previous example is a complete and valid derivation. Why is this? It is valid, because [0/x]x = 0, and thus, the previous derivation applies the  $\beta$ -rule.

#### Example 4.

$$\frac{\mathsf{app}\,(\mathsf{fun}\,x\Rightarrow\mathsf{if}\,x\,\mathsf{then}\,0\,\mathsf{else}\,1)\,(0\wedge1)\rightsquigarrow\mathsf{if}\,(0\wedge1)\,\mathsf{then}\,0\,\mathsf{else}\,1}{\mathsf{app}\,(\mathsf{fun}\,x\Rightarrow\mathsf{if}\,x\,\mathsf{then}\,0\,\mathsf{else}\,1)\,(0\wedge1)\rightsquigarrow^*\mathsf{if}\,(0\wedge1)\,\mathsf{then}\,0\,\mathsf{else}\,1}\,\,\mathsf{Step}} \\ \frac{D}{\mathsf{app}\,(\mathsf{fun}\,x\Rightarrow\mathsf{if}\,x\,\mathsf{then}\,0\,\mathsf{else}\,1)\,(0\wedge1)\rightsquigarrow^*1}$$

D:

$$\frac{\frac{1}{0 \land 1 \leadsto 0} \text{ AndFalse1}}{\frac{\text{if } (0 \land 1) \text{ then } 0 \text{ else } 1 \leadsto \text{if } 0 \text{ then } 0 \text{ else } 1}{\text{if } (0 \land 1) \text{ then } 0 \text{ else } 1 \leadsto^* \text{if } 0 \text{ then } 0 \text{ else } 1} \text{ Step}} \underbrace{\text{if } 0 \text{ then } 0 \text{ else } 1 \leadsto^* 1}_{\text{if } (0 \land 1) \text{ then } 0 \text{ else } 1 \leadsto^* 1} \text{ IFFalse}}_{\text{MULT}}$$

Note that  $[0 \wedge 1/x]$  (if x then 0 else 1) = if  $(0 \wedge 1)$  then 0 else 1.

### Example 5.

$$\frac{\operatorname{app}\left(\operatorname{fun} x\Rightarrow \operatorname{fun} y\Rightarrow \operatorname{if} x \operatorname{then}\left(x\wedge y\right) \operatorname{else}\left(x\vee y\right)\right) 0 \leadsto \operatorname{fun} y\Rightarrow \operatorname{if} 0 \operatorname{then}\left(0\wedge y\right) \operatorname{else}\left(0\vee y\right)}{\operatorname{app}\left(\operatorname{app}\left(\operatorname{fun} x\Rightarrow \operatorname{fun} y\Rightarrow \operatorname{if} x \operatorname{then}\left(x\wedge y\right) \operatorname{else}\left(x\vee y\right)\right) 0\right) 1 \leadsto \operatorname{app}\left(\operatorname{fun} y\Rightarrow \operatorname{if} 0 \operatorname{then}\left(0\wedge y\right) \operatorname{else}\left(0\vee y\right)\right) 1} \xrightarrow{\operatorname{APP1}} \operatorname{APP1} \\ \operatorname{app}\left(\operatorname{app}\left(\operatorname{fun} x\Rightarrow \operatorname{fun} y\Rightarrow \operatorname{if} x \operatorname{then}\left(x\wedge y\right) \operatorname{else}\left(x\vee y\right)\right) 0\right) 1 \leadsto^* \operatorname{app}\left(\operatorname{fun} y\Rightarrow \operatorname{if} 0 \operatorname{then}\left(0\wedge y\right) \operatorname{else}\left(0\vee y\right)\right) 1} \xrightarrow{\operatorname{APP1}} \operatorname{D_1} \\ \operatorname{app}\left(\operatorname{app}\left(\operatorname{fun} x\Rightarrow \operatorname{fun} y\Rightarrow \operatorname{if} x \operatorname{then}\left(x\wedge y\right) \operatorname{else}\left(x\vee y\right)\right) 0\right) 1 \leadsto^* 1} \operatorname{MULT}$$

 $D_1$ :

$$\frac{\mathsf{app}\,(\mathsf{fun}\,y\Rightarrow\mathsf{if}\,0\,\mathsf{then}\,(0\wedge y)\,\mathsf{else}\,(0\vee y))\,1\rightsquigarrow\mathsf{if}\,0\,\mathsf{then}\,(0\wedge 1)\,\mathsf{else}\,(0\vee 1)}{\mathsf{app}\,(\mathsf{fun}\,y\Rightarrow\mathsf{if}\,0\,\mathsf{then}\,(0\wedge y)\,\mathsf{else}\,(0\vee y))\,1\rightsquigarrow^*\mathsf{if}\,0\,\mathsf{then}\,(0\wedge 1)\,\mathsf{else}\,(0\vee 1)}\,\,\mathsf{STEP}} \\ \mathsf{app}\,(\mathsf{fun}\,y\Rightarrow\mathsf{if}\,0\,\mathsf{then}\,(0\wedge y)\,\mathsf{else}\,(0\vee y))\,1\rightsquigarrow^*\,1}\,\,\mathsf{MULT}$$

 $D_2$ :

$$\frac{ \frac{\text{if 0 then } (0 \wedge 1) \text{ else } (0 \vee 1) \rightsquigarrow 0 \vee 1}{\text{if 0 then } (0 \wedge 1) \text{ else } (0 \vee 1) \rightsquigarrow^* 0 \vee 1} } \underbrace{\text{STEP}}_{\text{STEP}} \underbrace{ \frac{0 \vee 1 \rightsquigarrow 1}{0 \vee 1 \rightsquigarrow^* 1} } \underbrace{\text{ORTRUE2}}_{\text{O } \vee 1 \rightsquigarrow^* 1} \underbrace{\text{STEP}}_{\text{MULT}}$$

Rewrite Rules. So far we have used the evaluation rules to show that one program evaluates into another. Is it possible to evaluate the program without knowing what the output would be? That is, given a program b is it possible to simply run it? Yes, we can, and we can define rules, called **rewrite rules**, that allow us to do this.

The set of rewrite rules for functional Iffy are defined as follows:

$$\frac{1}{(1 \land 1) \rightsquigarrow^* 1} \text{ AndTrue} \qquad \frac{1}{(0 \land 1) \rightsquigarrow^* 0} \text{ AndFalse1} \qquad \frac{1}{(1 \land 0) \rightsquigarrow^* 0} \text{ AndFalse2}$$

$$\frac{1}{(0 \land 0) \rightsquigarrow^* 0} \text{ AndFalse} \qquad \frac{1}{(1 \lor 1) \rightsquigarrow^* 1} \text{ OrTrue} \qquad \frac{1}{(1 \lor 0) \rightsquigarrow^* 1} \text{ OrTrue1} \qquad \frac{1}{(0 \lor 1) \rightsquigarrow^* 1} \text{ OrTrue1}$$

$$\frac{1}{(0 \lor 0) \rightsquigarrow^* 0} \text{ OrFalse} \qquad \frac{1}{(0 \lor 1) \rightsquigarrow^* 1} \text{ orther } b_1 \text{ else } b_2 \rightsquigarrow^* b_1 \text{ IfTrue} \qquad \frac{1}{(0 \lor 1) \rightsquigarrow^* b_2} \text{ IfFalse}$$

$$\frac{1}{(0 \lor 0) \rightsquigarrow^* 0} \text{ OrFalse} \qquad \frac{1}{(0 \lor 0) \rightsquigarrow^* 0} \text{ orther } b_1 \text{ else } b_2 \rightsquigarrow^* b_2 \text{ orther } b_1 \text{ else } b_2 \rightsquigarrow^* b_2 \text{ orther } b_2 \text{ orther }$$

They are the same as all the axioms of the single-step reduction rules. We will use these rules to pattern match on subprograms of larger programs against the left-hand side of the rules above, and then replace the matching subprogram with the right-hand side. However, we first must understand what a subprogram.

**Definition 6.** A program  $b_1$  is a subprogram of  $b_2$  if and only if at least one of the following is true:

- $b_1$  is literally  $b_2$
- If  $b_2 = b_3 \wedge b_4$  for some programs  $b_3$  and  $b_4$ , and  $b_1$  is one of these two programs
- If  $b_2 = b_3 \vee b_4$  for some programs  $b_3$  and  $b_4$ , and  $b_1$  is one of these two programs
- If  $b_2 = \text{if } b_3 \text{ then } b_4 \text{ else } b_5 \text{ for some programs } b_3, b_4, \text{ and } b_5, \text{ and } b_1 \text{ is one of these three programs}$
- If  $b_2 = \text{fun } x \Rightarrow b_3$  for some program  $b_3$ , and  $b_1$  is  $b_3$
- If  $b_2 = app b_3 b_4$  for some programs  $b_3$  and  $b_4$ , and  $b_1$  is one of these two programs

The program 0 is a subprogram of  $0, 0 \lor x$  is a subprogram of  $\operatorname{fun} x \Rightarrow (0 \lor x)$ . Similarly,  $\operatorname{app} y (\operatorname{fun} x \Rightarrow x)$  is a subprogram of  $0 \land (\operatorname{app} y (\operatorname{fun} x \Rightarrow x))$ . Some none subprograms would be  $\operatorname{app}$  and  $\operatorname{fun} x$  which are not subprograms of any programs.

Now if a subprogram matches the left-hand side of the set of axioms listed above, then we are allowed to replace that subprogram with the right-hand side. An example will best illustrate how we are going to write this down (then we will discuss this more in class):

$$\frac{\operatorname{app}\left(\operatorname{app}\left(\operatorname{fun}x\Rightarrow\operatorname{fun}y\Rightarrow\operatorname{if}x\operatorname{then}\left(x\wedge y\right)\operatorname{else}\left(x\vee y\right)\right)0\right)1}{\underset{\text{DRT}_{\text{NUE2}}}{\overset{\operatorname{Betta}}{\sim}}} \underbrace{\frac{\operatorname{app}\left(\operatorname{fun}y\Rightarrow\operatorname{if}0\operatorname{then}\left(0\wedge y\right)\operatorname{else}\left(0\vee y\right)\right)1}{\operatorname{if}0\operatorname{then}\left(0\wedge 1\right)\operatorname{else}\left(0\vee 1\right)}}_{\underset{\text{ORT}_{\text{RUE2}}}{\overset{\operatorname{Betta}}{\sim}}} \underbrace{\frac{\operatorname{app}\left(\operatorname{fun}y\Rightarrow\operatorname{if}0\operatorname{then}\left(0\wedge y\right)\operatorname{else}\left(0\vee y\right)\right)1}{\operatorname{o}\vee 1}}_{\underset{\text{ORT}_{\text{RUE2}}}{\overset{\operatorname{Betta}}{\sim}}}$$

First, notice how much shorter these are, but also note that the order in which we simplified the above does not matter. For example, we could have done the following:

```
\operatorname{app}\left(\operatorname{\underline{app}}\left(\operatorname{fun} x\Rightarrow\operatorname{fun} y\Rightarrow\operatorname{if} x\operatorname{then}\left(x\wedge y\right)\operatorname{else}\left(x\vee y\right)\right)0\right)1 \quad \begin{array}{c} \overset{\operatorname{Beta}}{\leadsto} \\ \overset{\operatorname{IfFALSE}}{\leadsto} \\ \overset{\operatorname{Beta}}{\bowtie} \\ \overset{\operatorname{Beta}}{\bowtie} \\ \overset{\operatorname{ORTRUE2}}{\bowtie} \\ \overset{\operatorname{ORTRUE2}}{\leadsto} \\ \end{array} \quad \begin{array}{c} \operatorname{app}\left(\operatorname{fun} y\Rightarrow\operatorname{\underline{if}} 0\operatorname{then}\left(0\wedge y\right)\operatorname{else}\left(0\vee y\right)\right)1\\ \\ \overset{\operatorname{ORTRUE2}}{\bowtie} \\ \overset{\operatorname{ORTRUE2}}{\leadsto} \\ \end{array} \quad \begin{array}{c} 1\\ \end{array}
```

When rewriting instead of deriving the format above must be adhered to on homework assignments.