Titolo

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Chapter 1

Intro

In 2001 Pym and Harland publish a paper [2] where they propose a new way to tackle the problem of splitting sequents during linear logic proof search using boolean constraints. The paper proposes a new calculus for linear logic that associates to each formula a boolean variable, and enforces linearity by constraints on said variables. This way the complexity shifts from choosing the right set of formulas to prove a certain branch, to solving for boolean assignment – a problem for which there are much more sophisticated algorithmus.

We examine the efficiency of this method and we compare it to other provers for different substets of linear logic.

1.1 Linear logic

Linear logic is a logic proposed by Jean-Yves Girard in his seminal paper of 1987 [1]. The distintive trait of this logic is that its formulae cannot be copied (called weakening) or discarded (called contraction), but instead they are consumed. And a certain sequent it true if and only if all its formulae get consumed exactly once. For this reason this logic is sometimes called a logic of resources, in the same way classical logic is a logic of truths and intuitionistic logic is a logic of proofs.

In linear logic each connective of classical logic is doubled. To better see this let's analyze classic conjuction, this can be defined as

$$\frac{\Delta \vdash \phi_2, \Gamma \quad \Delta \vdash \phi_1, \Gamma}{\Delta \vdash \phi_1 \land \phi_2, \Gamma} \quad \frac{\Delta'' \vdash \phi_2, \Gamma'' \quad \Delta' \vdash \phi_1, \Gamma'}{\Delta', \Delta'' \vdash \phi_1 \land \phi_2, \Gamma', \Gamma''}$$

On the other hand, these two rules are not equivalent in linear logic, since the former implies some weakening and contraction. This is exactly the reason why in linear logic all connectives have two versions: an additive one – where the two branches keep the same context, and a multiplicative one – where the context gets partioned between the two branches. Obviously the constants \top and \bot also have two versions. We have that

	Additive	Multiplicative
\wedge	&	\otimes
\vee	\oplus	239
Т	Т	1
\perp	0	\perp

It is the multiplicative side which brings the most complexity. The action of partitioning the context – called splitting – implies an exponential number of attempts to find which subset of the multiset is right for a certain branch.

Linear logic defined as of right now, albeit having the added complexity of splitting, is nonetheless decidable: since formulae are finite and they cannot be copied, it is possible to explore all the possibilities. To make linear logic as strong as classical logic two new connectives are added: $!\phi$ and $!\phi$ – called respectively bang and why-not. These are called exponentials and their purpose is to localize uses of contraction and weakening. For example, formulas marked with ! can be used any number of times.

1.2 State of the art

Most forward provers for classic linear logic use some combination of focusing and normalization to structure their proofs, with the notable exception of llprover not using normalization. We confront our prover with two other provers: llprover (1997,) and APLL (circa 2019,).

Usually the splitting is handled in two ways: trying every partition possible, or using something called the method of input/output. The latter tries to do one branch of the proof of a multiplicative, and then feeds the remaining formulae in the sequent of the other branch.

We now give a deeper look at the provers we confront with.

1.2.1 APLL

APLL is the underlying prover of click&collect. It provides 4 different searches – forward and backwards for classic and intuitionistic linear logic. We will focus on the backwards algorithm for classic linear logic.

Before diving into the analysis of this prover, we give the definition of why-not height

Definition 1 (Why-not height). Why-not height is the maximum number of nested "why-not"s in a formlua, or

$$wnh(\phi) = \begin{cases} 0 & \text{if } \phi \in \{\bot, \top, 1, 0\} \\ max(wnh(\phi_1), wnh(\phi_2)) & \text{if } \phi \in \{\phi_1 \otimes \phi_2, \phi_1 \ \ \forall \phi_2, \phi_1 \oplus \phi_2, \phi_1 \& \phi_2\} \\ wnh(\phi_1) & \text{if } \phi \in \{\phi_1^{\bot}, !\phi_1\} \\ 1 + wnh(\phi_1) & \text{if } \phi \in \{?\phi_1\} \end{cases}$$

This measure is used in their particular way of dealing with uncontrained formulae, but is also used as a way to decide which branch to prove first for any normal operator.

The program is written in OCaML and implements a standard focused proof search on normalized formulae as seen in [3]. In this section we will illustrate two noteworthy characteristics of its implementation:

• Sequent splitting when encountering a tensor is done by generating all the numbers up to $2^{|\Delta|}$ – where Δ is the sequent – and using the bit representation of those to create the two subsets. This can be seen in the function split_list, which in turn calls split_list_aux

where the argument k is the number that determines the decomposition of the sequent. This function is called recursively when a tensor is encountered during proof search, starting at $k = 2^{|\Delta|}$ and decreasing by one at each iteration

As we will see in 4.2 this implementation choice will result in a degradation of performance on formulae with a high number of multiplicatives.

- This prover does not use a simple limit to the number of applications of the contraction rule in a branch, instead an initially empty queue of unrestricted formulae (select_d2) and a counter (max_d2) are kept during the search. Two cases arise:
 - if select_d2 = [] and max_d2 > 0 then the sequent of unrestricted formulae is taken, negative terms are filtered out and it is sorted based on why-not height.

```
1 (* ... *)
2 if select_d2 = [] then begin
3 (if max_d2 = 0 then (bl := true; raise NoValue));
4 let select_d2 ' =
5 sort_whynot (List.filter (fun x -> not (is_neg x))
```

```
(Set_formula.elements theta)) in
f select_d2' = [] then None
else
papply_d2 select_d2' (max_d2 - 1) end
(* ... *)
```

This new list of unrestricted formule becomes the new select_d2. Otherwise if select_d2 is still empty after being refilled (line 7) or if max_d2 is 0 (line 3) the branch fails.

if select_d2 is not empty, then the first formula in the queue is extracted and added to the working set. If the branch fails the formula gets discarded and the next one in the queue is tried.

The main purpose of this whole process is to avoid infinite loops that always contract on the same formula. Instead all the formulae are tried one by one, starting from the simplest (lower why-not height).

The counter max_d2 is a local bound, since decreasing it in a branch does not affect other branches.

1.2.2 llprover

11prover is a prover by Naoyuki Tamura. Where APLL had different provers for implicative and classical linear logic, this prover encodes all the rules as the same predicate rule/6, using the first argument as a selector for the system. Using classical logic as the system uses all the rules, included the ones for implicative linear logic. For this reason the prover does not implement normalization.

Another particular characteristic of llprover is that it uses a local bound with iterative deepening, so in the benchmarks for formulae which need a lot of contractions, it will perform slightly worse.

1.3 Why Prolog

Prolog as a language and as an environment has been historically tied to automated theorem proving for its ability to express these kind of algorithms naturally. One particularly conventient characteristic of Prolog is its automatic managing of backtracking, in most other languages we would have had to use exceptions to walk down the stack, or a queue of unfinished computations, which would have made the code much less readable.

Most Prolog implementations also support CLP or constraint logic programming. This allows to have constraints referencing some attributes of variables in the body of clauses, in our case we use $\text{CLP}(\mathcal{B})$ for boolean constraints and an handy interface to a sat-solver. The library exposes operators to compose boolean formulas made of prolog variables

```
X = (X = = 1),

Y = (X * X)
```

and to check the satisfaiability of said formulas

$$\begin{array}{l} ?{\rm{ - \ \, sat}\,(X \ * \ Y)}\,. \\ X \, = \, Y, \ Y \, = \, 1\,. \end{array}$$

One other characteristic of Prolog which revealed to be very handy for our prover is unification. Using this we didn't have to explicitly propagate the solutions of the SAT-solver, which instead were automatically propagated between branches.

Chapter 2

The focused calculus

Before describing the calculus we must give some definitions

Definition 2 (Annotated formula). Given a formula ϕ defined as in Figure 2.1 and a boolean expression e defined as in Figure 2.2, an *annotated formula* is simply a term

$$af(\phi, e)$$

that associates the formula to the expression. We denote

$$\exp(\operatorname{af}(\phi, e)) = e$$

and then extend this notation to sequents, such that $\exp(\Delta)$ is the set of all boolean expressions of Δ .

As seen in Figure 2.1 and 2.2 we will usually use ϕ to refer to formulas, x to refer to boolean variables, and e to refer to boolean expressions (conjuctions of variables).

The purpose of putting formulae and variables together in the annotated formula is twofold:

- the actions taken on the formula determine the constraints that will be generated, and these depend on the variables associated to said formula;
- after the constraints are solved we can query the assignement of the variables and find out if the associated formula is used or not in a certain branch of a proof.

Figure 2.1: Linear logic connectives

$$x ::= x_i \mid \overline{x_i}$$
 (Variable)
 $e ::= x \land e \mid x$ (Expression)

Figure 2.2: Definition of a boolean variable and expression

Definition 3 (Consuming formulae). Given a boolean expression e as in Figure 2.2, we use the following notation

- "e used" to state that the formula associated to the boolean expression e gets consumed in this branch of the proof, this corresponds to saying the constraint e is true;
- " $e \neg used$ " to state that the formula associated to the boolean variable e does not get consumed in this branch of the proof, this corresponds to saying the constraint e is not true.

We then extends these predicates to sequents

$$\Delta$$
 used = $\{e_2 \text{ used } | e_2 \in \exp(\Delta)\}$
 $\Delta \neg \text{used} = \{e_2 \neg \text{used } | e_2 \in \exp(\Delta)\}$

Definition 4 (Members of the sequent). Given any sequent this can be in either two forms:

• focused or in the synchronous phase, written:

$$\vdash \Psi : \Delta \Downarrow \phi \parallel \Lambda : V$$

• in the asynchronous phase, written:

$$\vdash \Psi : \Delta \Uparrow \Phi \parallel \Lambda : V$$

These two have more or less the same members, which are

- the set Ψ of unrestricted formulae, or all formulae that can be freely discarded or duplicated;
- the multisets Δ and Φ of linear (annotated) formulae, these are respectively the formulas "put to the side" and the formulae which are being "worked on" during a certain moment of the asynchronous phase;
- the set Λ of constraints, which is to be interpreted as the conjunction of its members, and the set V which represents a propagated solution. Furthermore if Λ is satisfiable we write $\Lambda \downarrow V$, so:

$$\Lambda \downarrow V \iff \bigwedge_{e \in \Lambda} e \text{ is } sat \text{ by } V$$

V is in and of itself a boolean expression as in Figure 2.2, in fact it can be seen as the conjunction of the variables assigned to true and the negation of

the variables assigned to false. As such it can be used as a constraint, stating that a certain solution must be respected in a new one.

This approach to constraints helps to make the flow of the variables and solutions through the proof tree more explicit and clear and leaves no ambiguity to where the constraints should be checked.

The choice of letters is mainly a mnemonic or visual one, constraints Λ "go-up" the proof tree and solutions V "come down" from the leaves.

Definition 5 (Splitting a sequent). Given a sequent of annotated formulae Δ we define the operation of splitting it as a function

$$\operatorname{split}(\Delta) \mapsto (\Delta^L, \Delta^R)$$

where, given a set X of new variable names x_i for each formula $\phi_i \in \Delta$

$$\Delta^{L} = \{ \operatorname{af}(\phi_{i}, x_{i} \wedge e_{i}) \mid i \in \{1, \dots, n\} \}$$

$$\Delta^{R} = \{ \operatorname{af}(\phi_{i}, \overline{x_{i}} \wedge e_{i}) \mid i \in \{1, \dots, n\} \}$$

with n the cardinality of Δ , and ϕ_i and e_i respectively the formula and the variable of the i-eth annotated formula in Δ using an arbitrary order.

It is worth noting that the variables $x_i \in X$ used for Δ^L and Δ^R must be the same, this condition is necessary for the mechanism ensuring that a formula used on the left side of a tensor proof is not used on the right side and viceversa.

With a slight abuse of notation we will write split(Δ)_L and split(Δ)_R respectively as the left projection and the right projection of the pair (Δ _L, Δ _R).

As a small example for clarity, given the sequent

$$\Delta = \operatorname{af}(a \otimes b, x_1), \operatorname{af}(c^{\perp}, x_2)$$

this is split into

$$\operatorname{split}(\Delta)_L \mapsto \operatorname{af}(a \otimes b, x_3 \wedge x_1), \operatorname{af}(c^{\perp}, x_4 \wedge x_2)$$
$$\operatorname{split}(\Delta)_R \mapsto \operatorname{af}(a \otimes b, \overline{x_3} \wedge x_1), \operatorname{af}(c^{\perp}, \overline{x_4} \wedge x_2)$$

Definition 6. Lastly given a formula ϕ we define the following predicates

• ϕ asy is a predicate that's true only when ϕ is an asynchronous formula, which are

$$\phi ::= \phi \ \% \ \phi \ | \ \phi \ \& \ \phi \ | ?\phi \ | \ \top \ | \ \bot$$

• ϕ neg lit is a predicate that's true only when ϕ is a negative literal, in our implementation negative literals are atoms, and positive literals are negated atoms.

We are now ready to present the full calculus.

Figure 2.3: The complete boused constraint calculus

Chapter 3

Implementation

3.1 Formula transformations

Before beginning the proof a sequent passes through a number of transformations. These transformations both preprocess the sequent to a more convenient form, and also add information about the subformulae.

As a first transformation the sequent gets normalized into a sequent in negated normal form (NNF). NNF is the form where all negations are pushed down to atoms and all linear implications $(-\circ)$ are expanded into pars (??) using the following tautology

$$a \multimap b \Leftrightarrow a^{\perp} \Re b$$

Normalization is a common technique – used in all the provers we compare with. The process is composed of just two steps

- the left sequent is negated and appended to the right sequent, implemented by the predicate negate_premises/3;
- 2. the predicate nnf/2, which encodes the DeMorgan rules, is mapped recursively over the new sequent

This is possible since classic linear logic is symmetric and negation is involutive.

The purpose of this process is to cut away a great deal of possible rules applicable to the sequent, sacrificing some of the structure of the sequent. In fact the number of rules we need to implement after normalization is more than halved, since we now need just the right rules, without the ones for negation and linear implication.

As a second transformation, to each formula we assign its why-not height, as defined in Definition 1. Why-not height is used during proof search to decide which branch to do first and which exponential to decide first. This is a technique borrowed from APLL and its obvious purpose is to try first the branches with less exponentials: in the case the first probably simpler branch fails, we do not have to try the other. After this transformation formulae are attribute trees, with at each node the why not height of the subformula.

As a third and final transformation, each formula gets annotated. This means we associate a variable to each formula in the sequent as in Definition 2. Given a sequent Δ we obtain

$$\hat{\Delta} = \{ af(\phi, x) \mid x \text{ new}, \phi \in \Delta \}$$

To be clear, a variable is only assigned to the "top-level" formula, and subformulae are left unchanged.

In the implementation the concept of variable is split in two: the name of the variable – represented by a Prolog atom, and the value of the variable – represented by a Prolog variable. This is needed since, after checking the constraints, the SAT-solver unifies the variable to its value if it finds a satisfaiable solution, so the purpose of the atom is to associate the variable value to its name if the final proof. The process of annotation is implemented by the predicate annotate/3

```
1 %! annotate (+[Formulae], -[AFs], -[Constraints]) is det.

2 annotate (Fs, Afs, Cns) :-

3 maplist ([F

4 , af (F, X, Var)

5 , v(Var) =:= 1

6 ]>>(gensym(x, X)), Fs, Afs, Cns).
```

which is a simple map over the sequent. The constraints returned state that each formula must have its variable to one, that is to say each formulae must be used. These are the constraints the proof search starts with.

3.2 Constraint propagation

Some care is to be given to explaining how the constraints propagate. In fact, in constrast to Figure 2.3 the implementation does not have explicit propagation of the solution to the constraints. This is because Prolog's unification implicitly propagates a solution from one branch to another.

3.3 Helper predicates

We now define some helper predicates to work with the constraints. What we defined as Δ ¬used in Definition 3 corresponds to the predicate set_to_zero/2

```
\%! set_to_zero(+[AFs], -[Constr]) is det. set_to_zero(Fs, Cns):- maplist([af(_, _, E), v(E) =:= 0]>> true, Fs, Cns).
```

The other helper predicate implements the split function defined in Definition 5.

It is again a map over the list of formulae, that generates the new formulae and the constraints accordingly. Three new Prolog variables are introduced: X, Y and Z. X is the new variable, the annotated formulae refer to the variable Y and Z, and constraints are added so that

$$y = x \wedge e$$
$$y = \overline{x} \wedge e$$

Compare this to the original definition of Definition 5, one can see that the two are basically identical other than the fact that here the name of the variable (the atom) and its value are treated separately.

3.4 Focusing

When explaining the code we will use some common names for variables, these are

- A is a set of unrestricted atoms;
- U is a set of unrestricted formulae;
- F, F1, ..., are formulae, and Fs and D are a lists of them;
- S is the queue of currently usable unrestricted formulae;
- In is a list of constraints.

3.4.1 Asynchronous and focusing phase

During the asynchronous phase we have a list of formulae which are being worked on and a list of formulae which are put to the side. With the former being called Fss and the latter D. At each step we analyze the first element of the list Fs, and we keep scomposing the members of the list untill we can't anymore. This process can be seen for example in the predicate for &

Here we can see both the choice being made based on the why-not height of the two subformulae, and how the with is scomposed. Compare this with the & rule in Figure 2.3. The cut at line 2 represents the main concept of the asynchronous phase: if an asynchronous connective is encountered the only thing we ought to do is to scompose it.

If a formula cannot be further be broken apart – i.e. it is either an atom, a negated atom, or it has a toplevel synchronous connective – then it is put to the side in D. This can be seen in the rule to_delta which implements the rule $R \uparrow$

This process goes as long as Fs has still formulae inside.

When Fs is empty the phase switches, and the focusing process begins: we choose a formula – called decide – from either D or U and we scompose it untill either an asynchronous connective is left or a negated atom. This is represented by the rules D_1 and D_2 that will be discussed further ahead in Section 3.4.3.

3.4.2 Identity rules

This process of alternating asynchronous and synchronous phases in classic focusing goes on untill we have a only positive literal (in our case a negated atom) in Fs and the corresponding negative literal (in our case just an atom) in either U or D. When this happens the axioms – rules I_1 or I_2 – are applied to close the branch. In our case when we are focusing and we have a positive literal in Fs, we check if the corresponding negative literal exists in D. If this is true, then the variables of all the other formulae in D are set to zero using the predicate $set_to_zero/2$ defined in Section 3.3, and the constraints are checked. This is encoded in the clause

A slightly different process happens if instead a correspondence is found in A instead of D. Here A is a special set containing just unrestricted atoms. This is a slight modification to APLL approach based on the fact that once negative literals are put in a sequent they can never leave it, and is due to the fact that since U may be sorted many times, we try to keep the number of formulae in it small.

3.4.3 Decide rules

For the decide rules, particularly for D_2 , we use a slight modification to APLL algorithm defined in Section 1.2.1. Like APLL we keep a queue of ordered unrestricted formulae which can be refilled only a certain number of times per-branch. This can be seen in the definition of the rule decide_2

```
refill(U, M, S, M1),
early_stop(A, U, D, S, M1, In, _).
```

In particular, if the queue S is refilled, we do not directly call async/8, but instead call another predicate: early_stop/7 (line 11).

```
\begin{array}{l} early\_stop\left(\_,\_,\_,\_,\left[\right],\_,\_,\_\right) := \\ false. \\ early\_stop\left(A,\ U,\ D,\ [H|T],\ M,\ In\,,\_\right) := \\ gensym\left(x,\ X\right), \\ focus\left(A,\ U,\ D,\ af\left(H,\ X,\ E\right),\ T,\ M,\ [v(E) =:= 1|In]\,,\_\right). \\ early\_stop\left(A,\ U,\ D,\ \left[\_|T\right],\ M,\ In\,,\_\right) := \\ early\_stop\left(A,\ U,\ D,\ T,\ M,\ In\,,\_\right). \end{array}
```

This is due the simple fact that if the branch was not provable and we instead called directly async/8 at line 11, we would try to refill the branch M times. What early stop does is fail if the queue has just been refilled and it turns out the branch was not provable.

All the rules D_1 , I_1 are put before the unrestricted counterparts, so that they are tried first.

3.5 Building the tree

In the listings above we omitted one parameter of the predicates, which purpose is to build the proof tree. At each call of async and focus one node of the proof tree is built. This contains the context of the call. For example in

we can see clearly the structure of the node: a label, the context and an – optionally empty – list of subtrees. A leaf is just a node with an empty list of subtrees.

This tree can be used in the end to reconstruct the actual proof tree, by visiting it and – for each formula – querying whether its variable is set of one, and cancelling it otherwise. A more sophisticated algorithm may even cancel out unwanted unrestricted formulae, that otherwise remain lingering in the sequent.

Chapter 4

Testing

!!! VERY MUCH STILL WIP !!!

4.1 Infrastructure

4.1.1 Reproducibility

The prover's tests and benchmarks are made using a jupyter notebook. To ensure reproducibility we use Nix, which is a build system based on reproducile and declarative recipes, called *derivations*. A docker image is also given, which in turn calls nix without the need to install it on the system. Furthermore for the project infrastructure we used nix flakes, which are an experimental feature of nix that makes the process more hermetic and pure.

4.1.2 Prefix format

Since for benchmarking we will interface with a lot of different provers, each with its own syntax for expressions, the need for a common format which was easy to parse and translate arose. For this purpose we define use a prefix format for linear logic formulae inspired by the format used by [4] for implicational formulae

symbol		
*AB		
AB		
+AB		
&AB		
@AB		
^A		
?A		
! A		

Furthermore each single character not representing an operator is considered as a variable name. Longer names can be specified by enclosing them in single apices as in 'varname'. As an example we give the translation of DeMorgan for the tensor:

$$\operatorname{trans}((a \otimes b)^{\perp} \multimap a^{\perp} \, \mathfrak{A} \, b^{\perp}) = @\hat{a}b|\hat{a}b$$

4.1.3 File formats

As a standard format to store the tests we use json because of its vast adoption by most programming languages. A test suite is defined as a list of test cases

```
TestCase ::= {
   "id": <Number>,
   "premises": [ <PrefixFormula>+ ],
   "conclusions": [ <PrefixFormula> ]
}
where
```

id

is a number with the sole purpose of tracing back the test case from the output;

premises

is a list of premises as prefix formulae

conclusions

a list of conclusions as prefix formulae

Other arbitrary fields may be present, for example we will use the following optional fields:

$_{ m thm}$

whether this test case is a tautology or not, may be null;

*, &, +, ...

the number of times a specific appears in the test case;

notes

human readable text about the test case, for example its infix representation;

size

an indicative number of the size of the formula;

atoms

the upper bound on the number of atoms.

4.1.4 Formula generation

One of the sources of formulae we'll use in Section 4.2 is APLL's random formula generator. The version we'll use is a slight modification of it where:

• the output is in the json format described in Section 4.1.3

- one can choose to generate normalized formulae or not;
- one can choose which connectives appear in the generated formula.

A noteworthy detail is how the parameters size and atoms mentioned in Section 4.1.3 are defined, since these are directly related to how the formulae are generated:

- when one specifies a number of atoms atoms, the generator initializes an array containing atoms atoms, their negations, and the constants ⊥, ⊤,.... During the generation of the formula this array is randomly accessed, extracting a random element when needed. This means that atoms represents an upper bound to the number of different atoms that may appear in the formula, not their exact number.
- when a formula is generated at each step, based on a threshold, it is chosen whether to generate a unary or binary connective:
 - if a unary connective is chosen, the process continues with a size of size –1:
 - if a binary connective is chosen, the program chooses a random value between 0 and size, and it generates the two branches of the formula, with size respectively k and size -k.

4.2 Benchmarking

We'll mainly use three sources for formulae:

- llprover's tests
- LLTP, especially the translations of Kleene's intuistic formulae
- randomly generated formulae made by the genrator described in 4.1.4

llprovers tests are composed mainly by simple linear logic tautologies, e.g. the DeMorgan rules, for this reason these tests are used more as a simple and fast suite to catch the most obvious bugs between iterations of the prover.

We now show the results of running the provers on two datasets: KLE-cbn and KLE-cbv, respectively the call-by-name and call-by-value translations of Kleene's theorems. These translations introduce a high number of exponentials, and this causes, other than timeouts because of the added complexity, some failures. These failures are not due to bugs, but instead happen because the prover has reached its bound for that formula. The benchmarks are done using a timeout of 60 s and a bound of 3.

Kleene CBN								
prover	timeouts	failures	successes	success rate	avg. time (succ.)	avg. time (tot.)		
APLL	0	16	72	≈ 0.80	$0.037\mathrm{s}$	$0.055\mathrm{s}$		
llprover	20	6	62	≈ 0.70	$1.709\mathrm{s}$	$3.253\mathrm{s}$		
sat-ll	4	18	66	≈ 0.75	$0.130\mathrm{s}$	$0.185\mathrm{s}$		
Kleene CBV								
prover	timeouts	failures	successes	success rate	avg. time (succ.)	avg. time (tot.)		
APLL	0	17	71	≈ 0.80	$0.035\mathrm{s}$	$0.326\mathrm{s}$		
llprover	20	6	62	≈ 0.70	$0.981\mathrm{s}$	$2.179\mathrm{s}$		
sat-ll	5	15	68	≈ 0.77	$0.443\mathrm{s}$	$0.496\mathrm{s}$		

There are mainly two downsides with using random formulae:

- most cases do not hold any structure,
- ???

Using random formulae we can clearly see that our prover outperforms APLL (and llprover) with formulae rich in multiplicatives.

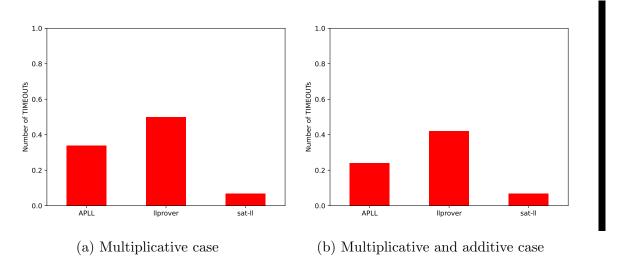


Figure 4.1: Percentage of number of timeouts out of a hundred formulae

We can see that in the multiplicative and additive case the difference begin to level. The additive case is not that significant as the formulae remain manageable and no major differences can be seen.

As soon as exponentials come into play the differences level out. In full linear logic our prover performs slightly worse than APLL.

All these test were done generating suits of 100 tests. Figure 4.1a's tests were normalized formulae of size 100, and atoms 50, with just the connectives \otimes and \Im . Figure 4.1b's tests were normalized formulae formulae of size 500, and atoms 250,

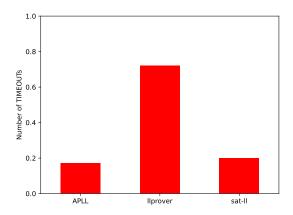


Figure 4.2: Full linear logic

with the connectives \otimes , \Re , & and \oplus . Finally 4.2's tests were normalized formulae of size 500, and atoms 250, with all the connectives.

When looking at the results of full linear logic, it is important to note that unlike the tests with multiplicatives and additives, some of the results may be early failures because of the bound. Since liprover uses incremental search, its times are often of the slowest. On the other hand our prover is consistently slightly slower that APLL, this difference is negligible and due to the fact that APLL is compiled, whereas our prover is interpreted.

Bibliography

- [1] Jean-Yves Girard. Linear logic. Theoretical Computer Science, 1987.
- [2] James Harland and David Pym. Resource-distribution via boolean constraints. 2018.
- [3] Chuck Liang and Dale Miller. Focusing and polarization in linear, intuitionistic, and classical logics. *Theoretical Computer Science*, 2009.
- [4] Paul Tarau and Valeira de Paiva. Deriving theorems in implicational linear logic. 2020.