

# Splitting Proofs for Interpolation\*

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**Abstract.** We study interpolant extraction from local first-order refutations. We present a new theoretical perspective on interpolation based on clearly separating the condition on logical strength of the formula from the requirement on the common signature. This allows us to highlight the space of all interpolants that can be extracted from a refutation as a space of simple choices on how to split the refutation into two parts. We use this new insight to develop an algorithm for extracting interpolants which are linear in the size of the input refutation and can be further optimized using metrics such as number of non-logical symbols or quantifiers. We implemented the new algorithm in first-order theorem prover VAMPIRE and evaluated it on a large number of examples coming from the first-order proving community. Our experiments give practical evidence that our work improves the state-of-the-art in first-order interpolation.

## 1 Introduction

Starting with the pioneering work of McMillan [15], interpolation became a powerful approach in verification thanks to its use in predicate abstraction and model checking [16,1,19]. To prove program properties over a combination of data structures, such as integers, arrays and pointers, several approaches based on theory-specific reasoning have been proposed, see e.g. [14,5,4]. While powerful, these techniques are limited to quantifier-free fragments of first-order logic. Addressing reasoning in full first-order theories, quantified interpolants are computed in [17,11,3,23] and further optimized with respect to various measures in [9].

In this paper, we address interpolation in full first-order logic and introduce a novel approach to generate interpolants, possibly with quantifiers. Our approach improves and simplifies the aforementioned techniques, in particular [11,9]. In [11,9], the size of computed interpolants is in the worst case quadratic in the size of the proof and the generated interpolants may contain redundant subformulas. Our work addresses these issues and infers interpolants that are linear in the size of the proof and are much simpler than in [11,9]. We proceed as follows. We separate the requirements on a formula being an interpolant into a part restricting the logical strength of an interpolant and a part restricting which symbols are allowed to be used in an interpolant. This way, we first handle formulas, called intermediants, satisfying the requirements on the logical

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strength of interpolants, and only then we restrict the generated space of intermediants to the ones that satisfy the restriction on the interpolants signature.

The work of [11] relies on so-called local proofs (or split proofs) and constructs interpolants by splitting local proofs into (maximal) subproofs. Splitting proofs is determined by the signature of formulas used in the proofs. We observed, however, that there are many ways to split a proof, resulting in interpolants that are different in size and strength. We therefore propose a general framework for splitting proofs and using the boundaries of the resulting sub-proofs to construct the intermediants. The key feature of our work is that the interpolants inferred from our various proof splits are linear in the size of the proof. When constructing interpolants from proof splits, we note that local proofs are exactly the ones that ensure that proof splits yield intermediants that satisfy the requirements of interpolants. Using local proofs and proof splits, we then describe a powerful heuristic and an optimality criterion how to choose the “best” proof split, and hence the resulting interpolant.

**Contributions.** The main contributions of this paper are as follows.

- We present a new algorithm for first-order-interpolation using local proofs in arbitrary sound inference systems. That is, our work can be used in any sound calculus and derives interpolants, possibly with quantifiers, in arbitrary first-order theories.
- Our interpolation algorithm is the first algorithm ensuring that the size of the interpolant is linear in the size of the proof while working with an arbitrary sound logical calculus. This result improves [11] and generalises the work of [17] to any sound inference system.
- We implemented our work in the VAMPIRE theorem prover [12] and evaluated our method on a large number of examples coming from the TPTP library [22]. Our experimental results confirm that our work improves the state-of-the-art in first-order interpolation.

The rest of this paper is structured as follows. The background notation on proofs and interpolation is covered in Section 2. We then show how to construct linear sized interpolants in Section 3 and present optimisations to the procedure in Section 4. We compare to related work in Section 5, describe our experimental results in Section 6, and conclude in Section 7.

## 2 Preliminaries

This section introduces the relevant theoretical notions to our work.

**Formulas.** We deal with standard first-order predicate logic with equality. We allow all standard logical connectives and quantifiers in the language and, in addition, assume that it contains the logical constants  $\top$ ,  $\perp$  for true and false, respectively. Without loss of generality, we restrict ourselves to closed formulas, i.e. we do not allow formulas to contain free variables. The non-logical symbols of a formula  $F$ , denoted by  $\mathcal{N}(F)$ , are all the predicate symbols and function symbols (including constants) occurring in  $F$ . Note that this excludes (quantified) variables and the equality symbol.

An *axiomatisable theory*, or simply a *theory* is any set of formulas. For example, we can use the theory of linear integer arithmetic or the theory of lists. We will from now on restrict ourself to a fixed theory  $\mathcal{T}$  and give all definitions relative to  $\mathcal{T}$ . This

includes that we write  $F_1, \dots, F_n \models F$  (instead of  $F_1, \dots, F_n \models_{\mathcal{T}} F$ ) to denote that every model of  $\mathcal{T}$  which satisfies each  $F_1, \dots, F_n$  also satisfies  $F$ .

**Definition 1.** Let  $F_1, \dots, F_n, F$  be formulas,  $n \geq 0$ . An *inference rule*  $R$  is a tuple  $(F_1, \dots, F_n, F)$ . An *inference system*  $\mathfrak{S}$  is a set of inference rules.

An inference rule  $R = (F_1, \dots, F_n, F)$  is *sound*, if  $F_1, \dots, F_n \models F$ . An inference system  $\mathfrak{S}$  is called *sound*, if it only consists of *sound* inference rules.

From now on, we further restrict ourselves to a fixed inference system  $\mathfrak{S}$  which is sound (relative to  $\mathcal{T}$ ) and give all definitions relative to that system.

**Derivations and proofs.** We model logical proofs as directed hypergraphs in which vertices are associated with formulas and (hyper-)edges with inferences. Because an inference always has exactly one conclusion, we only need hypergraphs where each edge has exactly one end vertex. Moreover, because the order of premises of an inference may be important, we use tuples to model the edges. We will from now on refer to such (hyper-)edges simply as *inferences*.

**Definition 2.** Let  $G$  be a formula and  $\mathcal{F}$  a set of formulas. A *proof of  $G$  from axioms  $\mathcal{F}$*  is a finite acyclic labeled directed hypergraph  $P = (V, E, L)$ , where  $V$  is a set of vertices,  $E$  a set of inferences, and  $L$  is a labelling function mapping each vertex  $v \in V$  to a formula  $L(v)$ . For an inference  $r \in E$  of the form  $(v_1, \dots, v_n, v)$ , where  $n \geq 0$ , we call  $v_1, \dots, v_n$  the *premises of  $r$*  and  $v$  the *conclusion of  $r$* .

Additionally, we require the following:

1. Each vertex  $v \in V$  is a conclusion of exactly one inference  $r \in E$ .
2. There is exactly one vertex  $v_0 \in V$  that is not a premise of any inference  $r \in E$  and  $L(v_0) = G$ .
3. each  $r \in E$  is either (a) an inference of the form  $(v)$  and  $L(v) \in \mathcal{F}$ , or (b) an inference of the form  $(v_1, \dots, v_n, v)$  and  $(L(v_1), \dots, L(v_n), L(v)) \in \mathfrak{S}$ . In the first case, we call  $r$  an *axiom inference*, in the second case,  $r$  is called a *proper inference*.

A *refutation* from axioms  $\mathcal{F}$  is a proof of the formula  $\perp$  from  $\mathcal{F}$ .

Note that in order to support multiple occurrences of the same formula in a proof, one needs to distinguish between vertices and the formulas assigned to them via the labelling function  $L$ . However, because this generality is orthogonal to the ideas we want to present, we will from now on identify each node  $v \in V$  with its formula  $L(v)$  and stop referring to the labelling function explicitly.

In the above definition, condition 1 ensures that any formula of the proof is justified by exactly one inference. Later on we will look at subgraphs of a proof, which are not necessarily proofs themselves and in particular do not satisfy condition 1, since they contain formulas, which are not justified by any inference of the subgraph. We call such a subgraph a *derivation* and call the formulas which are not justified by any inference the *premises of the derivation*. We can see a proof as a derivation having no premises.

**Definition 3.** The definition of a *derivation of  $G$  from axioms  $\mathcal{F}$*  is the same as that of a proof  $P = (V, E, L)$  of  $G$  from  $\mathcal{F}$ , except that condition 1 is generalised to:

1. Each formula  $F \in V$  is a conclusion of *at most one* inference  $r \in E$ .

The *set of premises of a derivation*  $P$ , denoted by  $\text{Prem}(P)$ , consists of all formulas  $F \in V$ , such that there exists no inference  $r \in E$  with conclusion  $F$ .

The definition of a derivation is not natural as it distinguishes between axioms and premises. This distinction is, however, very important for us, as it enables a succinct presentation of the results in Sect. 3.

**Lemma 4 (Soundness).** Let  $P$  be a derivation of  $G$  from axioms  $\mathcal{F}$ . Then we have

$$\mathcal{F} \models (\bigwedge_{F_k \in \text{Prem}(P)} F_k) \rightarrow G.$$

To formalise the idea of a proof traversal in which the inferences are considered one by one from axioms to the final formula  $G$ , we make use of topological orderings.

**Definition 5.** Let  $P = (V, E, L)$  be a derivation. A topological ordering  $<^T$  for  $P$  is a linear ordering  $<^T$  on  $E$  such that for any two inferences  $r_1, r_2 \in E$  if the conclusion of  $r_1$  is a premise of  $r_2$  then  $r_1 <^T r_2$ .

A topological ordering exists for every derivation, because proofs, and thus also derivations, are required to be acyclic.

**Interpolation.** We now recall the notion of a logical interpolant.

**Definition 6.** Let  $A$  and  $B$  be formulas.

1. A non-logical symbol  $s \in \mathcal{N}(A \rightarrow B)$  is called *A-local*, if  $s \in \mathcal{N}(A) \setminus \mathcal{N}(B)$ , *B-local*, if  $s \in \mathcal{N}(B) \setminus \mathcal{N}(A)$ , and *global* otherwise.
2. An *interpolant* for  $A, B$  is a formula  $I$  such that  $\models A \rightarrow I, \models I \rightarrow B$  and all non-logical symbols of  $I$  are global.

Craig's interpolation theorem [6] guarantees the existence of an interpolant for any pair of formulas  $A, B$  for which  $\models A \rightarrow B$ . In the sequel, we assume  $A$  and  $B$  to be fixed and give all definitions relative to  $A$  and  $B$ .

**Refutational theorem proving.** To prove a first-order formula  $F$  in practice, a refutational theorem prover proceeds by negating the input formula, applying a normal form transformation, such as the Conjunctive Normal Form transformation, to the negation, and deriving a contradiction  $\perp$  from the obtained set of formulas  $\mathcal{C}_{\neg F} = \text{CNF}(\neg F)$ . More specifically, in the case of proving the implication  $A \rightarrow B$ , the prover starts with axioms  $\text{CNF}(A \wedge \neg B)$ .

This is relevant for our work, because we rely on refutations as input for our method. However, a complication arises, because the normal form transformations  $\text{CNF}$  typically involves steps like sub-formula naming and Skolemisation [18,20], which 1) introduce new non-logical symbols, 2) in general do not preserve logical equivalence.

To deal with 1) we impose a restriction on  $\text{CNF}$  which dictates that the symbols newly introduced on behalf of  $A$  and  $\neg B$  do not overlap. Formally, we require

$$\mathcal{N}(A) \cap \mathcal{N}(\neg B) = \mathcal{N}(\text{CNF}(A)) \cap \mathcal{N}(\text{CNF}(\neg B)), \quad (1)$$

which is a very natural condition, because the newly introduced symbols are invariably required to be fresh.<sup>3</sup>

<sup>3</sup> This could potentially be violated by an advanced transformation based on formula sharing. In particular, the case would need to involve a common sub-formula of  $A$  and  $\neg B$ .

To deal with 2), let us first recall that steps like sub-formula naming and Skolemisation, although they do not preserve logical equivalence, do preserve satisfiability. While this is sufficient to guarantee soundness of refutational theorem proving, it is not enough for the purposes of interpolation. Fortunately, a stronger property, which is rarely stated explicitly, usually holds for the normal form transformation, namely the *preservation of models over the common symbols*. Formally, we require for every formula  $F$  that

- every model  $\mathcal{M}'$  of  $CNF(F)$  is also a model of  $F$ , and
- every model  $\mathcal{M}$  of  $F$  can be extended to  $\mathcal{M}'$  which is a model of  $CNF(F)$ ,

where extended means that  $\mathcal{M}'$  restricted to  $\mathcal{N}(F)$  equals  $\mathcal{M}$ .

Equipped with a transformation  $CNF$  satisfying the above requirements, the general approach to interpolation from refutations consists of the following steps:

1. Given formulas  $A$  and  $B$ , compute the respective normal forms  $\mathcal{C}_A = CNF(A)$  and  $\mathcal{C}_{\neg B} = CNF(\neg B)$ .
2. Find a refutation  $P$  from axioms  $\mathcal{C}_A \cup \mathcal{C}_{\neg B}$ .
3. Extract from  $P$  a formula  $I$  such that  $\mathcal{C}_A \models I$ ,  $\mathcal{C}_{\neg B}, I \models \perp$ , and all non-logical symbols of  $I$  are global.<sup>4</sup>

**Lemma 7.** The formula  $I$  obtained in the last step is an interpolant for  $A$  and  $B$ .

### 3 Interpolants from refutations

We can separate the properties of an interpolant into two parts, the logical part and the restriction to the global symbols. Instead of considering only interpolants, we now want to look more generally at the formulas, which satisfy the logical part of the properties of interpolants, but not necessarily the restriction to the global symbols. We call such formulas *intermediants*.<sup>5</sup>

**Definition 8.** Let  $A, B$  be two formulas. An *intermediant* for  $A, B$  is a formula  $I$  such that we have both  $\models A \rightarrow I$  and  $\models I \rightarrow B$ .

In the first part of this section, we want to investigate the space of intermediants, which is induced by a given refutation. In the second part, we look at the subspace of those intermediants which also respect the restriction on the global symbols, i.e. the formulas which are interpolants.

#### 3.1 Splitting refutations

Let us now show how to use a refutation of  $A \rightarrow B$  to construct intermediants. Intuitively, we want to split the refutation into two parts and construct a formula which describes the boundaries between the parts.

In the light of the discussion at the end of the previous section, we assume the formulas  $A$  and  $\neg B$  have been transformed to sets of axioms  $\mathcal{C}_A$  and  $\mathcal{C}_{\neg B}$ . It is also natural to extend the notion of an intermediant to axiom sets:

<sup>4</sup> Note that the symbols are global with respect to  $A$  and  $B$  if and only if they are global with respect to  $\mathcal{C}_A$  and  $\mathcal{C}_{\neg B}$  thanks to the requirement (1).

<sup>5</sup> Bonacina and Johansson [2] introduce the notion of a *provisional interpolant* with an analogous definition. However, the intended use of the notion is different. While provisional interpolants are meant to be modified to yield interpolants in a refinement stage, we give conditions under which intermediants are, in fact, interpolants.

**Definition 9.** Let  $\mathcal{C}_A$  and  $\mathcal{C}_{-B}$  be two sets of axioms. An *intermediant* for  $\mathcal{C}_A, \mathcal{C}_{-B}$  is a formula  $I$  such that we have both  $\mathcal{C}_A \models I$  and  $\mathcal{C}_{-B}, I \models \perp$ .

Splitting a proof into two parts for us means mapping each inference to one of the two parts. Formally, we introduce a two element set  $\{\mathcal{A}, \mathcal{B}\}$  to serve as a co-domain of such mapping, where  $\mathcal{A}$  denotes the  $A$ -part and  $\mathcal{B}$  the  $B$ -part. It is natural to map the axioms from  $\mathcal{C}_A$  to  $\mathcal{A}$  and the axioms from  $\mathcal{C}_{-B}$  to  $\mathcal{B}$ , therefore we only consider mappings of this form. All other inferences can be mapped to any part.

**Definition 10.** Let  $P$  be a refutation from axioms  $\mathcal{C}_A \cup \mathcal{C}_{-B}$ . A *splitting function*  $\mathcal{S}$  is a function assigning each inference of  $P$  to either  $\mathcal{A}$  or  $\mathcal{B}$ , such that for each axiom inference  $r = (F)$ , if  $\mathcal{S}(r) = \mathcal{A}$  then  $F \in \mathcal{C}_A$  and  $r$  is called an *A-axiom*, and if  $\mathcal{S}(r) = \mathcal{B}$  then  $F \in \mathcal{C}_{-B}$  and  $r$  is called a *B-axiom*.

A given splitting function  $\mathcal{S}$  splits a proof into several maximal subderivations. We now want to capture this intuitive notion formally. We start with the concept of In-formulas (resp. Out-formulas) of  $P$  and  $\mathcal{S}$ . Intuitively, these are the formulas which occur at the boundary between the subderivations.

**Definition 11.** Let  $P = (V, E, L)$  be a refutation from axioms  $\mathcal{C}_A \cup \mathcal{C}_{-B}$  and let  $\mathcal{S}$  be a splitting function on  $P$ . The *set of in-formulas*, which is denoted  $\text{In}(P, \mathcal{S})$ , consists of those formulas  $F \in V$ , which has the following properties:

- There exists an inference  $r_1 \in E$  with conclusion  $F$  and  $\mathcal{S}(r_1) = \mathcal{B}$ .
- There exists an inference  $r_2 \in E$  with premise  $F$  and  $\mathcal{S}(r_2) = \mathcal{A}$ .

The *set of out-formulas*, denoted  $\text{Out}(P, \mathcal{S})$ , consists of formulas  $F \in V$ , such that

- There exists an inference  $r_1 \in E$  with conclusion  $F$  and  $\mathcal{S}(r_1) = \mathcal{A}$ .
- Either there exists an inference  $r_2 \in E$  with premise  $F$  and  $\mathcal{S}(r_2) = \mathcal{B}$ , or  $F = \perp$ .

Notice that the notions of in- and out-formulas are not entirely symmetrical. The reason for this will become clear later.

We are now able to formally introduce the maximal subderivations.

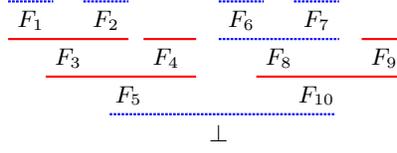
**Definition 12.** Let  $P = (V, E, L)$  be a refutation from axioms  $\mathcal{C}_A \cup \mathcal{C}_{-B}$  and let  $\mathcal{S}$  be a splitting function on  $P$ . Let  $r \in E$  be an inference and let  $\{r_1, \dots, r_l\}$  be the set of those inferences which derive a premise of  $r$  and are mapped by  $\mathcal{S}$  to the same part as  $r$ , i.e.  $\mathcal{S}(r) = \mathcal{S}(r_i)$  for  $i = 1, \dots, l$ . Then we define  $\text{Sub}(r)$  recursively as

$$\text{Sub}(r) = \{r\} \cup \text{Sub}(r_1) \cup \dots \cup \text{Sub}(r_l).$$

Now let  $F \in \text{Out}(P, \mathcal{S})$  (resp.  $F \in \text{In}(P, \mathcal{S})$ ) be a formula and  $r$  be the inference deriving  $F$ . We define the *maximal A-subderivation* (resp. *B-subderivation*) of  $F$ , denoted by  $\text{Sub}(F)$ , as the induced derivation  $(V', \text{Sub}(r), L)$ , where  $V'$  contains every vertex which is either a premise or a conclusion of an inference in  $\text{Sub}(F)$ . We call  $F$  the *conclusion* of  $\text{Sub}(F)$ .

The dependencies of  $F$ , written  $\text{Dep}(F)$ , are defined as the premises of  $\text{Sub}(F)$ .

We can observe that the In-formulas (resp. Out-formulas) are the premises (resp. conclusions) of all maximal  $A$ -subderivations. Dually, the In-formulas (resp. Out-formulas)



**Fig. 1.** Consider the proof above along with the splitting function which is denoted by drawing the inferences assigned to  $\mathcal{A}$  using solid red lines and the inferences assigned to  $\mathcal{B}$  using dashed blue lines. The maximal  $A$ -subderivation of  $F_5$  has premises  $F_1, F_2$  (and conclusion  $F_5$ ), the maximal  $A$ -subderivation of  $F_{10}$  has premise  $F_8$ . The maximal  $B$ -subderivation of  $F_1$  has no premises, the maximal  $B$ -subderivation of  $\perp$  has premises  $F_5$  and  $F_{10}$ . The In-formulas are  $F_1, F_2$  and  $F_8$  and the Out-formulas are  $F_5$  and  $F_{10}$ . The induced simple splitting formula is  $((F_1 \wedge F_2) \rightarrow F_5) \wedge (F_8 \rightarrow F_{10})$ .

are the conclusions (resp. premises) of all maximal  $B$ -subderivations. The use of the introduced concepts is demonstrated in Fig. 3.1.

Note that the  $A$ -subderivations contain all  $A$ -axioms, but no  $B$ -axiom. Therefore the  $A$ -axioms's contribution to the derivation is captured by the  $A$ -subderivations. The key idea of this subsection is that encoding the contribution of the  $A$ -subderivations as a formula therefore yields the intermediant  $I$  we are looking for. The following lemma tells us how to describe the contribution of an  $A$ -subderivation.

**Lemma 13.** Let  $P$  be a refutation from axioms  $\mathcal{C}_A \cup \mathcal{C}_{\neg B}$  and let  $\mathcal{S}$  be a splitting function on  $P$ .

1. Let  $F \in \text{Out}(P, \mathcal{S})$ . Then we have  $\mathcal{C}_A \models (\bigwedge_{F_k \in \text{Dep}(F)} F_k) \rightarrow F$ .
2. Let  $F \in \text{In}(P, \mathcal{S})$ . Then we have  $\mathcal{C}_{\neg B} \models (\bigwedge_{F_k \in \text{Dep}(F)} F_k) \rightarrow F$ .

We therefore arrive at the following definition.

**Definition 14.** Let  $P$  be a refutation from axioms  $\mathcal{C}_A \cup \mathcal{C}_{\neg B}$  and let  $\mathcal{S}$  be a splitting function on  $P$ . The formula

$$I := \bigwedge_{F \in \text{Out}(P, \mathcal{S})} ((\bigwedge_{F_k \in \text{Dep}(F)} F_k) \rightarrow F)$$

is called the *simple splitting formula of  $P$  induced by  $\mathcal{S}$* .

**Theorem 15.** Let  $P$  be a refutation from axioms  $\mathcal{C}_A \cup \mathcal{C}_{\neg B}$  and let  $\mathcal{S}$  be a splitting function on  $P$ . Then the simple splitting formula  $I$  induced by  $\mathcal{S}$  is an intermediant.

**Proof.**

1. For each  $F \in \text{Out}(P, \mathcal{S})$ , we can use Lemma 13.1 to get  $\mathcal{C}_A \models (\bigwedge_{F_k \in \text{Dep}(F)} F_k) \rightarrow F$ . Therefore we have  $\mathcal{C}_A \models \bigwedge_{F \in \text{Out}(P, \mathcal{S})} ((\bigwedge_{F_k \in \text{Dep}(F)} F_k) \rightarrow F)$ .
2. Let  $<^T$  be a topological ordering for  $P$  and let  $F_1, \dots, F_n$  denote the formulas of  $\text{In}(P) \cup \text{Out}(P)$  in the order induced by  $<^T$ . We visit the formulas from  $F_1$  to  $F_n$  and prove by complete induction that  $I, \mathcal{C}_{\neg B} \models F_1, \dots, F_i$ . Since  $F_n = \perp$ , we afterwards are able to conclude  $I, \mathcal{C}_{\neg B} \models \perp$ .

$$\begin{array}{c}
\overline{F_1} \quad \overline{F_2} \\
r_1 \frac{\quad}{F_3} \quad \frac{\quad}{F_5} \quad \frac{\quad}{F_4} r_2 \\
r_4 \frac{\quad}{\perp}
\end{array}$$

**Fig. 2.** Let  $r_1 = (F_1, F_3)$ ,  $r_2 = (F_2, F_4)$ ,  $r_3 = (F_1, F_2, F_5)$ , and  $r_4 = (F_3, F_4, F_5, \perp)$ . Let further  $\mathcal{S}(r_1) = \mathcal{S}(r_2) = \mathcal{S}(r_3) = \mathcal{A}$  and  $\mathcal{S}(r_4) = \mathcal{B}$ . Then Algorithm 1 generates the simple splitting formula  $I = (F_1 \rightarrow F_3) \wedge (F_2 \rightarrow F_4) \wedge ((F_1 \wedge F_2) \rightarrow F_5)$ . There is no intermediant which is both logically equivalent to  $I$  and contains each formula of the given proof at most once.

Inductive step: Let us assume, by the induction hypothesis, that  $I, \mathcal{C}_{\neg B} \models F_1, \dots, F_{i-1}$ .

We make a case distinction on  $\mathcal{S}(r)$ , where  $r$  is the inference which derived  $F_i$ :

- Case  $\mathcal{S}(r) = \mathcal{A}$ : By the definition of  $I$ , we know that  $I \models (\bigwedge_{F_k \in \text{Dep}(F_i)} F_k) \rightarrow F_i$ . Using both the definition of topological orderings and the definition of  $\text{Dep}$  we know that  $\text{Dep}(F_i) \subseteq \{F_1, \dots, F_{i-1}\}$ , so we can combine the previous facts to obtain  $I, \mathcal{C}_{\neg B} \models F_1, \dots, F_i$ .
- Case  $\mathcal{S}(r) = \mathcal{B}$ : We use Lemma 13.2 to conclude  $\mathcal{C}_{\neg B} \models (\bigwedge_{F_k \in \text{Dep}(F_i)} F_k) \rightarrow F_i$ . As in the previous case, we can use  $\text{Dep}(F_i) \subseteq \{F_1, \dots, F_{i-1}\}$  to conclude  $I, \mathcal{C}_{\neg B} \models F_1, \dots, F_i$ .

We summarise the ideas of this subsection in Simple-splitting-formula (Algorithm 1).

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**Algorithm 1** Simple-splitting-formula

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choose a splitting function on  $P$ .  
compute  $\text{Out}(P)$  and  $\text{Dep}(F)$  for all  $F$  using depth first search  
**return**  $I$  as defined in Definition 14

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### 3.2 Intermediants of linear size

Simple-splitting-formula yields an intermediant of size which is in the worst case quadratic in the size of the proof. This may be prohibitively large for large proofs. In this subsection, we describe an algorithm which yields intermediants of size which is linear in the size of the proof. Modifying Algorithm 1 to generate such an intermediant is nontrivial: there are examples, where the simple splitting formula is provably logically stronger than any intermediant which uses every formula of the refutation only once, cf. Fig. 2. We therefore need to modify the algorithm such that it produces an intermediant which is logically weaker but still sufficiently strong to be inconsistent with  $\mathcal{C}_B$ .

The key idea for the new algorithm is contained in the following definition.

**Definition 16.** Let  $P$  be a refutation from axioms  $\mathcal{C}_A \cup \mathcal{C}_{\neg B}$ ,  $\mathcal{S}$  a splitting function on  $P$ , and let  $<^T$  be a topological ordering for  $P$ . Furthermore let  $F_1, \dots, F_n$  denote the formulas of  $\text{In}(P) \cup \text{Out}(P)$  ordered by  $<^T$ . Now let

$$I_i = \begin{cases} \top & \text{if } i = n + 1 \\ F_i \rightarrow I_{i+1} & \text{if } F_i \in \text{In}(P) \\ F_i \wedge I_{i+1} & \text{if } F_i \in \text{Out}(P) \end{cases}$$

---

**Algorithm 2** Linear-splitting-formula

---

choose a splitting function and a topological ordering on  $P$ .  
compute  $\text{In}(P)$  and  $\text{Out}(P)$   
**return**  $I_1$  as defined in Definition 16

---

Then  $I_1$  is called *linear splitting formula of  $P$  induced by  $\mathcal{S}$  and  $<^T$* .

Note that the size of  $I_1$  is linear in the size of  $P$  in Definition 16.

**Theorem 17.** Let  $P$  be a refutation from axioms  $\mathcal{C}_A \cup \mathcal{C}_{-B}$ , let  $\mathcal{S}$  be a splitting function on  $P$  and let  $<^T$  be a topological ordering for  $P$ . Then the linear splitting formula  $I$  induced by  $\mathcal{S}$  and  $<^T$  is an intermediant.

**Proof.** Let

$$I' = \bigwedge_{F_i \in \text{Out}(P)} \left( \left( \bigwedge_{F_k \in \text{In}(P), F_k <^T F_i} F_k \right) \rightarrow F_i \right).$$

First note that  $I'$  is logically equivalent to  $I$ : This can be proved by a simple induction using the two facts that conjunction on the right distributes over implication and that  $A \rightarrow (B \rightarrow C)$  is equivalent to  $(A \wedge B) \rightarrow C$ .

Now we complete the proof by showing that  $I'$  is an intermediant:

1. Using both the definition of topological orderings and the definition of  $\text{Dep}$  we know that  $\text{Dep}(F_i) \subseteq \{F_k \in \text{In}(P) \mid F_k <^T F_i\}$ , so  $I'$  is logically weaker than the simple splitting formula. Therefore  $\mathcal{C}_A \models I'$  follows from Theorem 15.1.
2. We can show  $\mathcal{C}_{-B}, I' \models \perp$  by re-using the proof of Theorem 15.2 with  $<^T$  as the topological ordering and by replacing  $\text{Dep}(P)$  with  $\{F_k \in \text{In}(P) \mid F_k <^T F_i\}$ .

We summarise the presented ideas in Linear-splitting-formula (Algorithm 2) and conclude this subsection by pointing out the following basic lemma, which will become useful later in the paper.

**Lemma 18.** Let  $P = (V, E, L)$  be a refutation from axioms  $\mathcal{C}_A \cup \mathcal{C}_{-B}$  and let  $\mathcal{S}$  be a splitting function on  $P$ . Let further  $I$  be the linear splitting formula induced by  $\mathcal{S}$  and let  $F \in V$  be an arbitrary formula different from  $\perp$ . Then  $F$  occurs in  $I$  if and only if there are two inferences  $r_1, r_2$ , where  $r_1$  derives  $F$ ,  $F$  is a premise of  $r_2$  and  $\mathcal{S}(r_1) \neq \mathcal{S}(r_2)$ .

### 3.3 Interpolants as special intermediants

In the previous subsections, we discussed how to construct intermediants given a splitting function. We now look closer at the question which splitting function to choose. While studying the intermediants induced by different choices of a splitting function is an interesting topic in general, we turn our attention to the problem of choosing a splitting function such that the induced intermediant is an interpolant, i.e. we have the additional requirement that the intermediant contains no local symbols.

Let us recall the notion of local proofs—also called split proofs—introduced by Jhala and McMillan [10]:

**Definition 19 (Local Proof).** A proof  $P = (V, E, L)$  from axioms  $\mathcal{C}_A \cup \mathcal{C}_{-B}$  is *local* if for every inference  $(F_1, \dots, F_n, F) \in E$  we have either:

- $\mathcal{N}(F_1) \cup \dots \cup \mathcal{N}(F_k) \cup \mathcal{N}(F) \subseteq \mathcal{N}(C_A)$  or
- $\mathcal{N}(F_1) \cup \dots \cup \mathcal{N}(F_k) \cup \mathcal{N}(F) \subseteq \mathcal{N}(C_{\neg B})$ .

The definition of local proofs ensures that we can define a splitting function  $\mathcal{S}$  which maps all inferences with  $A$ -local symbols to  $\mathcal{A}$  and those with  $B$ -local symbols to  $\mathcal{B}$ .

**Definition 20.** Let  $P$  be a local proof. A *local splitting function on  $P$*  is a splitting function  $\mathcal{S}$  on  $P$  such that  $\mathcal{S}(r) = \mathcal{A}$  (resp.  $\mathcal{S}(r) = \mathcal{B}$ ) for all inferences  $r$  having as premise or conclusion a formula containing an  $A$ -local (resp. a  $B$ -local) symbol.

The corollary of the following lemma represents the central observation of this subsection: local proofs are exactly the proofs on which we can define a splitting function that induces an interpolant which is an interpolant.

**Lemma 21.** Let  $P = (V, E, L)$  be a refutation from axioms  $C_A \cup C_B$ ,  $\mathcal{S}$  be a local splitting function on  $P$ , and  $I$  the corresponding simple (resp. linear) splitting formula.

- i) Then any formula  $F \in \text{In}(P, \mathcal{S}) \cup \text{Out}(P, \mathcal{S})$  contains neither an  $A$ -local nor a  $B$ -local symbol.
- ii)  $I$  contains neither  $A$ -local nor  $B$ -local symbols.

**Proof.** i) Consider any formula  $F \in \text{Out}(P, \mathcal{S})$ . If  $F = \perp$  then  $F$  trivially contains neither an  $A$ -local nor a  $B$ -local symbol. Otherwise, we know that there exists an inference  $r_1 \in E$  with premise  $F$  and  $\mathcal{S}(r_1) = \mathcal{B}$ . By the locality of  $\mathcal{S}$  we get that  $F$  contains no  $A$ -local symbol. Furthermore, we know that there exists an inference  $r_2 \in E$  with conclusion  $F$  and  $\mathcal{S}(r_2) = \mathcal{A}$ . By the locality of  $\mathcal{S}$  we get that  $F$  contains no  $B$ -local symbol.

Now consider any formula  $F \in \text{In}(P, \mathcal{S})$ . We can use a similar argument to show that  $F$  contains neither an  $A$ -local nor a  $B$ -local symbol.

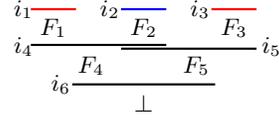
- ii) Follows immediately from i) and the definition of the simple (resp. linear) splitting formula.

**Corollary 22.** Let  $P$  be a local refutation, let  $\mathcal{S}$  be a local splitting function on  $P$  and let  $I$  be either the simple splitting formula or the linear splitting formula. Then  $I$  is an interpolant for  $A, B$ .

## 4 Implementing Local Splitting Functions

By the definition of a local splitting function we know that we need to assign axioms and inferences with local symbols to the corresponding part. All the other inferences—the inferences forming the so called *grey area* [9]—can be assigned freely to either part. Different choices on how to split the grey area result in different  $A$ -subproofs and therefore in different interpolants, which vary, e.g., in size, the number of contained quantifiers and in logical strength.

We want to minimize the interpolant with respect to a given weight function  $w$ , which maps each formula  $F$  to its weight  $w(F)$ . The task we want to solve in this section is, therefore, to be able to come up with a local splitting function which minimises the weight of the resulting interpolant.



**Fig. 3.** Let  $\mathcal{S}(i_1) = \mathcal{S}(i_3) = \mathcal{A}$  and  $\mathcal{S}(i_2) = \mathcal{B}$ . Let further  $w(F_1) = w(F_3) = 2$  and  $w(F_2) = 3$ . For both inferences  $i_4$  and  $i_5$ , the assignment of the inference to  $\mathcal{B}$  is locally optimal, then causes the assignment of  $i_6$  to  $\mathcal{B}$  and finally yields an interpolant of weight 4. Note that  $F_2$  is used as a premise of both  $i_4$  and  $i_5$ , so due to the DAG-structure we would only include it once if we assigned both  $i_4$  and  $i_5$  to  $\mathcal{A}$ . This would then cause the assignment of  $i_6$  to  $\mathcal{A}$  and finally yield a smaller interpolant of weight 3.

We present two different solutions, a heuristical greedy approach and one of expressing the optimal splitting as a minimisation problem. Both solutions are based on the insight from Lemma 18 of Sect. 3: A conclusion  $F$  of an inference  $r_1$  occurs in the linear splitting formula if and only if there is an inference  $r_2$  with  $F$  as a premise such that the splitting function maps  $r_1$  and  $r_2$  to different parts.

#### 4.1 Greedy weighted sum heuristic

Consider an inference  $r$  of the grey area with premises  $C_1, \dots, C_n, D_1, \dots, D_m$  and assume that the inferences deriving  $C_1, \dots, C_n$  are already assigned to  $\mathcal{A}$  and that the inferences deriving  $D_1, \dots, D_m$  are already assigned to  $\mathcal{B}$ . Using Lemma 18, we know that if we assign  $r$  to  $\mathcal{A}$ , then  $D_1, \dots, D_m$  will be added to the interpolant and if we assign  $r$  to  $\mathcal{B}$ , then  $C_1, \dots, C_n$  will be added to the interpolant.

We can therefore use the following greedy strategy to locally minimize the weight of the interpolant: for any inference  $r$  of the grey area, if  $\sum_{k=1}^n w(C_k) > \sum_{k=1}^m w(D_k)$ , map  $r$  to  $\mathcal{A}$ , otherwise to  $\mathcal{B}$ .

This results in Top-down-weighted-sum-heuristic (Algorithm 3):

---

#### Algorithm 3 Top-down-weighted-sum-heuristic

---

```

for each inference  $r$  of  $P$  (top-down) do
  if  $r$  is an  $A$ -axiom or  $r$  contains an  $A$ -local symbol then
    set  $\mathcal{S}(r)$  to  $\mathcal{A}$ 
  else if  $r$  is a  $B$ -axiom or  $r$  contains a  $B$ -local symbol then
    set  $\mathcal{S}(r)$  to  $\mathcal{B}$ 
  else
    if  $\sum_{k=1}^n w(C_k) > \sum_{k=1}^m w(D_k)$  then
      set  $\mathcal{S}(r)$  to  $\mathcal{A}$ 
    else
      set  $\mathcal{S}(r)$  to  $\mathcal{B}$ 
return  $\mathcal{S}$ 

```

---

The two reasons why a locally optimal choice is not a globally optimal choice can be seen in Figures 3 and 4.

$$\begin{array}{c}
i_1 \text{ ---} \\
\frac{F_1}{i_2} \quad \frac{F_3}{i_3} \\
\hline
\perp
\end{array}$$

**Fig. 4.** Let  $\mathcal{S}(i_1) = \mathcal{A}$  and  $\mathcal{S}(i_3) = \mathcal{B}$ . Let further  $w(F_1) < w(F_2)$ . Algorithm 3 would now assign  $i_2$  to  $\mathcal{A}$  and therefore include  $F_2$  in the interpolant. It would be better to assign  $i_2$  to  $\mathcal{B}$  in order to include  $F_1$  in the interpolant instead of  $F_2$ .

## 4.2 Encoding optimal splitting as a minimisation problem

Similar to the idea presented in [9], we can alternatively encode the problem of finding an optimal local splitting function as a minimisation problem and pass it to a pseudo-boolean constraint solver. This yields an optimal assignment, but is computationally more expensive.

The encoding works as follows. We use propositional variables  $x_i$  to denote that inference  $i$  is assigned to  $\mathcal{A}$  and use propositional variables  $L_i$  to denote that the conclusion of  $i$  occurs in the interpolant. We again predict the size of the resulting interpolant using Lemma 18, but this time use the optimisation procedure to make globally optimal choices instead of greedily making locally optimal ones. This leads to algorithm Weighted-sum-optimal (Algorithm 4).

---

### Algorithm 4 Weighted-sum-optimal

---

```

for each inference  $r$  of  $P$  do
  if  $r$  is an  $A$ -axiom or  $r$  contains an  $A$ -local symbol then
    assert  $x_r$ 
  else if  $r$  is a  $B$ -axiom or  $r$  contains a  $B$ -local symbol then
    assert  $\neg x_r$ 
  for each parent inference  $r'$  of  $r$  do
    assert  $(\neg(x_r \leftrightarrow x_{r'})) \rightarrow L_{r'}$ 
  compute model  $M$  which minimises  $\sum_{r \in P} w(\text{concl}(r)) \cdot L_r$ 
  for each inference  $r$  of  $P$  do
    if  $x_r$  evaluates to true in  $M$  then
      set  $\mathcal{S}(r)$  to  $\mathcal{A}$ 
    else
      set  $\mathcal{S}(r)$  to  $\mathcal{B}$ 
  return  $\mathcal{S}$ 

```

---

## 5 Discussion and Related Work

There are two main existing approaches to constructing interpolants from arbitrary local proofs in arbitrary sound first-order proof systems with equality.

First, there is the work from Jhala and McMillan (Theorem 3 of [10]). They present an algorithm which consists of two main phases: A) Extract a propositionally unsatisfiable set of formulas  $F$ , B) obtain a propositional refutation of  $F$  using boolean

constraint propagation and apply McMillan’s interpolation algorithm for propositional logic [15] to the result in order to obtain an interpolant for the original local refutation.

One can easily see that the set  $F$  constructed in phase A consists of both the conjuncts of the splitting formula from Definition 14 and the conjuncts of the simple splitting formula obtained by swapping  $A$  and  $B$  in the proof. In contrast, Algorithm 1 only needs the former conjuncts. Furthermore we know from Corollary 22 that it is sufficient to conjoin all these conjuncts instead of unnecessarily constructing and interpolating from a propositional refutation. Besides conceptually simplifying the algorithm, this also enables the optimisations presented in Sect. 4.

More importantly, in [17], it is claimed that the complexity of the algorithm behind Theorem 3 of [10] is linear in the size of the proof. While phase B of the algorithm is clearly linear, we can see easily from Example 23 below that phase A is worst case quadratic in the size of the proof, making the whole algorithm quadratic, which contrasts to our Algorithm 2, that is linear.

**Example 23.** Consider a split refutation with nodes  $A_1, \dots, A_n, B_1, \dots, B_n, \perp$ ; edges  $(A_i, A_{i+1}), (A_i, B_{i+1}), (B_i, A_{i+1}), (B_i, B_{i+1})$ , for  $1 \leq i < n$  and  $(A_n, \perp), (B_n, \perp)$ ; and labeling  $P(A_i) = A, P(B_i) = B, P(\perp) = \text{arbitrary}$ . Phase A would construct a graph with edges  $(A_i, B_j), (B_i, A_j)$  for all  $0 < i < j \leq n$ , which is quadratic in  $n$ .

The second main approach to constructing interpolants in first-order logic with equality using an arbitrary sound inference system was introduced in [11] and later improved by an optimisation technique in [9]. Let us refer to the interpolation algorithm from [11] as  $\mathcal{SE}$ . In a nutshell,  $\mathcal{SE}$  uses two main concepts:

As a first concept, it constructs the largest subderivations containing only symbols from one of the two partitions (cf. Lemma 8 of [11]). This construction corresponds to a commitment to a specific choice of local splitting function in our framework. In contrast, both Algorithm 1 and Algorithm 2 are parametrized by an arbitrary local splitting function and different choices yield different interpolants.

As the main contribution of [9], the authors extend algorithm  $\mathcal{SE}$  such that it also considers a space of different interpolants and optimise over this space. We can see that the extension simulates different choices of splitting function by merging proof steps. Both the algorithm from [9] and our Algorithm 4 encode the space of candidates and the minimisation objective as a pseudo-boolean constraint problem and then ask an optimising SMT-solver for an optimal solution. While encoding the space of splitting functions is trivial using Algorithm 4, encoding the space of local proofs, which are results from repeated pairwise merging of inferences, is much more involved. More critically, while we can make use of Lemma 18 to predict the size of the resulting interpolant, the approach from [9] uses a notion of so called digest to predict the size of the interpolant computed from the transformed proof. The authors claim that the interpolant is a boolean combination of formulas in the digest (Theorem 3.6, [9]). Unfortunately, this claim is wrong, which can be concluded from the counterexample presented in Fig. 5. Therefore the technique presented in [9] can potentially yield sub-optimal interpolants.

As the second concept, the algorithm  $\mathcal{SE}$  from [11] relies on a recursive construction to compute the interpolant: it computes for each largest subderivation a formula such that the formula of the outermost call yields an interpolant (cf. Lemma 10 of [11]).

$$\begin{array}{c}
\frac{R_1 \quad G_1}{G_3 \quad B_1 \quad G_2} \\
\frac{R_3 \quad \frac{G_3 \quad B_1 \quad G_2}{G_6}}{R_4} \\
\perp
\end{array}$$

**Fig. 5.** Consider the proof above, taken from Example 5.2 in [9]. Let  $R_1, R_3$  and  $R_4$  be formulas containing  $A$ -local symbols,  $B_1$  a formula containing  $B$ -local symbols and let  $G_1, G_2, G_3$  and  $G_6$  be formulas containing no local symbols. Then the digest contains only  $G_6$ , but the algorithm from [11] would construct the interpolant  $G_3 \wedge \neg G_6$ , which also contains  $G_3$ .

We now want to hint at the relation of algorithm  $\mathcal{SE}$  and Algorithm 1. Consider a sub-derivation with premises  $F_1, \dots, F_k$  and conclusion  $F$ . Let further  $I_1, \dots, I_k$  denote the recursively computed formulas. Algorithm  $\mathcal{SE}$  now constructs the following formulas:

- Case A:  $I = ((I_1 \vee F_1) \wedge \dots \wedge (I_k \vee F_k)) \wedge \neg(F_1 \wedge \dots \wedge F_k)$ .
- Case B:  $I = ((I_1 \vee F_1) \wedge \dots \wedge (I_k \vee F_k))$ .

It is not difficult to see that one can reformulate the construction of  $\mathcal{SE}$  as the following one, which we will refer to as  $\mathcal{SE}'$ :

- Case A:  $I = ((I_1 \vee F_1) \wedge \dots \wedge (I_k \vee F_k)) \vee F \wedge ((F_1 \wedge \dots \wedge F_k) \rightarrow F)$ .
- Case B:  $I = (I_1 \wedge \dots \wedge I_k)$ .

Note that although the intermediate formulas of algorithm  $\mathcal{SE}$  and  $\mathcal{SE}'$  are potentially different, the result of the outermost call is the same for  $\mathcal{SE}$  and  $\mathcal{SE}'$ .

We now state a recursive presentation of our Algorithm 1 in order to compare it to  $\mathcal{SE}'$ . The idea is to replace the global view on the refutation, i.e. the iteration over all elements of  $\text{Out}(P, \mathcal{S})$ , by a recursive construction which collects all the formulas describing the boundaries of maximal  $A$ -subderivations.

Let  $P$  be a local proof of a formula  $F$  and let  $r$  be the inference which derives  $F$ . Let further  $\mathcal{S}$  be a local splitting function on  $P$ . We compute a formula using the following recursive construction: Let  $F_1, \dots, F_k$  denote the elements of  $\text{Dep}(F)$  and let  $I_i$  denote the formula computed recursively from  $F_i$ .

- Case  $\mathcal{S}(r) = \mathcal{A}$ :  $I = (I_1 \wedge \dots \wedge I_n) \wedge ((F_1 \wedge \dots \wedge F_n) \rightarrow F)$ .
- Case  $\mathcal{S}(e) = \mathcal{B}$ :  $I = (I_1 \wedge \dots \wedge I_n)$ .

If we now compare algorithm  $\mathcal{SE}'$  and the recursive presentation of Algorithm 1 we see that they are the same with the exception that  $\mathcal{SE}'$  contains redundant sub-formulas. More critically, since we know that Algorithm 1 yields an interpolant of size which is worst-case quadratic in the size of the proof, we know that the same holds for  $\mathcal{SE}'$  and therefore for  $\mathcal{SE}$ , i.e. for the interpolation algorithm of [11]. This represents the most important downside of the approach of [11] and makes it inferior to Algorithm 2.

Finally, interpolation from first-order refutations is also studied in [2] and [13] where the authors present methods for computing interpolants from arbitrary proofs in first-order logic but either without equality or under the assumption that colored function symbols are only constants. While our proof splits are restricted to local proofs, in our approach we handle first-order theories with equality in full generality.

## 6 Experimental Results

We implemented Linear-splitting-formula (Algorithm 2, Sect. 3) in automated theorem prover VAMPIRE [12] and combined it with the two approaches for obtaining a local splitting function: the Top-down-weighted-sum-heuristic (Algorithm 3) and the Weighted-sum-optimal (Algorithm 4). We will from now on refer to the combinations as `LinHeu` and `LinOpt`, respectively. The aim of the experiment is to compare the performance of the new algorithms to algorithm from [11] combined with its optimising improvement from [9], which was already implemented in a previous version of VAMPIRE. We will from now on refer to this latter combination as `SEOpt`.<sup>6</sup>

To compensate for the lack of a representative set of benchmarks explicitly focusing on first-order interpolation, we made use of the first-order problems from the TPTP library [22] (version 6.4.0). We classified each problem using VAMPIRE and split the obtained set of clauses into halves, treating the first half as  $\mathcal{C}_A$  and the second as  $\mathcal{C}_{-B}$ . We attempted to refute each of the obtained problems using VAMPIRE (which was instructed to generate only local proofs as described in [11]) and followed up by one of `LinHeu`, `LinOpt`, or `SEOpt` to compute an interpolant. We imposed a 60s time limit on the proof search in VAMPIRE and a total limit of 100s on each whole run. We ran the experiment on the StarExec compute cluster [21].

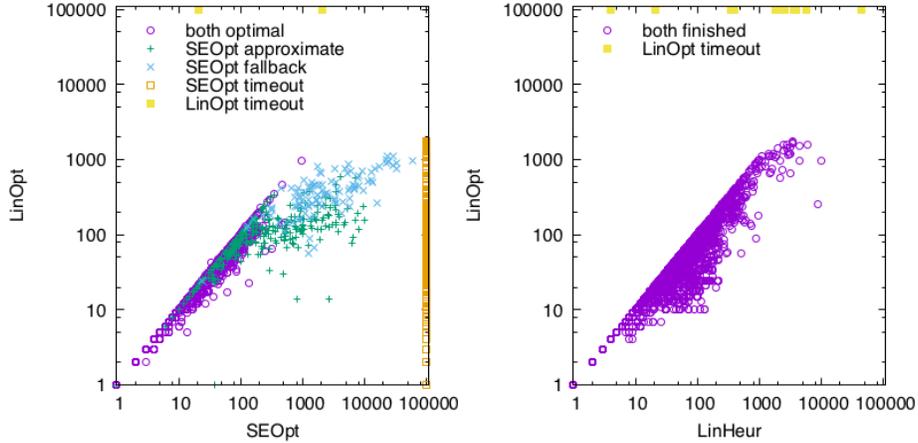
In total, we obtained 7442 local refutations. Out of these `SEOpt` failed to construct an interpolant in 723 cases. In contrast, `LinOpt` failed to construct an interpolant in only 16 cases and `LinHeu` always constructed an interpolant within the time limit. Furthermore, there were 353 cases in which `SEOpt` returned only an approximate result and 108 cases where optimisation failed and the unoptimized version of [11] was used as a fallback instead. The observed higher computational demands of `SEOpt` can be mostly ascribed to the reliance on a different pseudo-boolean solver and different connecting technology.<sup>7</sup> These differences unfortunately exclude the possibility of a meaningful comparison of more detailed timing results. However, we would like to point out that the optimisation problem `SEOpt` constructs is arguably much more complex than the one stemming from Weighted-sum-optimal employed by `LinOpt`.

Fig. 6 (left) contains a scatter plot comparison of the sizes of obtained interpolants for `LinOpt` and `SEOpt`. An artificial large value was substituted whenever a particular algorithm failed to provide an interpolant. This is reflected by the data points on the right and the upper border, respectively. The plot further separates the points to categories based on the optimality guarantee provided by `SEOpt`. We can see that `LinOpt` yields consistently better results. Moreover, the improvement tends to get more pronounced with the growing size of the instances. Finally, even when just focusing on instances where `SEOpt` finished optimising, there are numerous cases where the interpolant from `LinOpt` is several times smaller than that of `SEOpt`. This is because `SEOpt` cannot

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<sup>6</sup> The executables and connecting scripts used in the experiment are available at [http://forsythe.at/static/people/suda/vampire\\_new\\_interpolation.zip](http://forsythe.at/static/people/suda/vampire_new_interpolation.zip).

<sup>7</sup> `SEOpt` uses the SMT solver Yices [8] (version 1.0) and communicates via a file, while `LinOpt` one relies on Z3 [7] (version 4.5) and its API.



**Fig. 6.** Size comparison of interpolant produced by SEOpt and LinOpt (left), and LinHeu and LinOpt (right). Each point corresponds to a single refutation and its position to the sizes of the respective interpolants.

avoid repeating certain formulas from the refutation many times in the interpolant and corresponds to the worst case quadratic complexity discussed in Sect. 5.<sup>8</sup>

Fig. 6 (right) correspondingly compares LinOpt with LinHeu. Although the plot highlights many examples where LinHeu yields a larger interpolant than LinOpt, an optimal interpolant is actually discovered by LinHeu in 79% of the cases and its interpolants are only 11.6% larger on average. Moreover, on the 7429 refutations on which both algorithms finished in time, the accumulated time spent on interpolant extraction by LinHeu was only 8.17s compared to a total of 1901.03s spent by LinOpt. This shows that LinHeu presents a viable alternative to LinOpt when trading the quality of interpolant for computational time becomes desirable.

Given the encouraging results we intend to officially replace SEOpt by LinOpt and LinHeu in VAMPIRE and make it available with the next release of the prover.

## 7 Conclusion

We presented a new technique for constructing interpolants from first-order local refutations. The technique is based on an idea of proof splitting and on a novel non-inductive construction which arguably gives more insight than previous work and yields interpolants of linear size. This leads to a new interpolation algorithm which we implemented in the automated theorem prover VAMPIRE. Finally, we confirmed in an extensive experiment that the algorithm also improves over the state-of-the-art in practice.

<sup>8</sup> An interesting side-effect is an ability of SEOpt to assign two different colors to a formula when considered from the perspective of two different sub-derivations. In rare cases, such formula does not need to appear at all, and the final interpolant may be smaller than what is achievable by LinOpt. An instance of this phenomenon occurred in our experiment on benchmark SYN577-1, which appears in Fig. 6 (left) slightly above the diagonal.

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