Body-Decoupled Grounding via Solving: A Novel Approach on the ASP Bottleneck

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Abstract

Answer-Set Programming (ASP) has seen tremendous progress over the last two decades and is nowadays successfully applied in many real-world domains. However, for certain problems, the well-known ASP grounding bottleneck still causes severe problems. This becomes virulent when grounding of rules, where the variables have to be replaced by constants, leads to a ground program that is too huge to be processed by the ASP solver. In this work, we tackle this problem by a novel method that decouples non-ground atoms in rules in order to delegate the evaluation of rule bodies to the solving process. Our procedure translates a nonground normal program into a ground disjunctive program that is exponential only in the maximum predicate arity, and thus polynomial if this arity is bounded by a constant. We demonstrate the feasibility of this new method experimentally by comparing it to standard ASP technology in terms of grounding size, grounding time and total runtime.

1 Introduction

Answer set programming (ASP) [Brewka et al., 2011; Gebser et al., 2019; Janhunen and Niemelä, 2016] is a modeling and solving framework that can be seen as an extension of propositional satisfiability (SAT), where knowledge is expressed by means of rules comprising a (logic) program, whose solutions are sets of atoms, called answer sets, that obey every rule. Its rich first-order like language made ASP an appealing tool for modeling industrial applications (see e.g., [Falkner et al., 2018]). Efficient systems are readily available ([Gebser et al., 2019; Calimeri et al., 2019]) and mainly build on a ground-and-solve technique, replacing variables by constants and feeding the resulting ground program into an ASP solver.

However, for certain types of problems, this ground-and-solve approach leads to the well-known ASP *grounding bot-tleneck* [Cuteri *et al.*, 2020; Tsamoura *et al.*, 2020], i.e., the instantiation of rules yields a program that is exponentially larger and thus too huge to be processed by the solver efficiently. We recall that the complexity for ground programs is relatively mild: consistency for disjunctive programs is located at the second level of the polynomial hierarchy,

cf. [Eiter and Gottlob, 1995]; normal programs or tight programs yield NP-complete fragments [Bidoít and Froidevaux, 1991; Marek and Truszczyński, 1991]. Variables in rules increase the expressiveness and reflect the potentially huge costs of grounding: even for normal non-ground programs the complexity jumps up to NEXPTIME completeness [Dantsin et al., 2001]. While the standard grounding via rule instantation leads to an exponential blow up even for programs with bounded predicate arities, there are results that indicate that the costs of grounding can be decreased when assuming such a setting. In particular, Eiter et al. [2007] have shown that the complexity of consistency for non-ground normal programs is Σ_2^P -complete. This indicates that assuming the predicate arity as fixed, there exists a polynomial translation to ground disjunctive programs, and thus an alternative grounding procedure that delegates certain efforts to the solving process.

Contributions. We provide an alternative grounding approach that utilizes the aforementioned complexity result, thereby decoupling rule bodies during grounding, which is rectified during solving. Our contributions are as follows.

- 1. We present a novel reduction from tight (non-ground) logic programs to disjunctive programs that encodes grounding via search, in order to identify unsatisfiable ground rules and unjustified (unfounded) atoms. In contrast to traditional grounding, our reduction allows us to *decouple* predicates occurring in the body of a rule, which might be particularly useful for larger bodies with a very dense structure.
- 2. We extend this approach to normal, non-ground programs, where for ensuring justifiability we additionally encode the idea of orderings (level mappings) in our reduction.
- 3. Finally, we present a prototype¹ that allows to translate critical parts of the program using our reduction, thereby empowering the grounding process by decoupling body predicates. Preliminary experiments indicate that this approach can lead to significant speed-ups, where the size of standard groundings would be excessively large.

Related Work. Efficient grounding is an active field of research and requires powerful rule initialization procedures trying to evade the intractable evaluation of non-ground programs. The literature distinguishes many approaches, rang-

¹Our system including supplemental material of this work is publicly available at https://github.com/viktorbesin/newground.

ing from traditional instantiation tactics [Gebser et al., 2019; Alviano et al., 2019; Kaminski and Schaub, 2021] over fruitful estimations [Hippen and Lierler, 2021] and lazy grounding [Bomanson et al., 2019; Weinzierl et al., 2020]. Besides, there are further attempts to avoid the grounding bottleneck, like constraint-programming or ASP modulo theory extensions, e.g., [Banbara et al., 2017; Janhunen et al., 2017; Cabalar et al., 2020], and methods based on graph invariants like treewidth [Bichler et al., 2020; Calimeri et al., 2019; Bliem et al., 2020; Mitchell, 2019]. Similar to our work, [Eiter et al., 2010] draws on the complexity of bounded arities to design space-efficient ASP evaluation methods, but their method is more in the spirit of meta-programming than on an alternative grounding procedure. Instead, we focus on decoupling body predicates and a translation to ground ASP.

2 Preliminaries

We use mathematical vectors $X = \langle x_1, \ldots, x_m \rangle$, $Y = \langle y_1, \ldots, y_n \rangle$ in the usual way; we *combine vectors* by $\langle X, Y \rangle := \langle x_1, \ldots, x_m, y_1, \ldots, y_n \rangle$ and test whether x_1 is *contained in* X by $x_1 \in X$. Without loss of generality, we assume that elements of vectors are given in any fixed total order; for a given set S, we *construct* its unique vector by $\langle S \rangle$.

Ground ASP. Let ℓ , m, n be non-negative integers such that $\ell \leq m \leq n$; a_1, \ldots, a_n be distinct propositional atoms. A (disjunctive) program P is a set of (disjunctive) rules of form

 $a_1\vee\ldots\vee a_\ell\leftarrow a_{\ell+1},\ldots,a_m, \neg a_{m+1},\ldots, \neg a_n.$ For a rule r, we let $H_r:=\{a_1,\ldots,a_\ell\},\ B_r^+:=\{a_{\ell+1},\ldots,a_m\},$ and $B_r^-:=\{a_{m+1},\ldots,a_n\}.$ We denote the sets of atoms occurring in a rule r or in a program P by $\operatorname{at}(r):=H_r\cup B_r^+\cup B_r^-$ and $\operatorname{at}(\mathsf{P}):=\bigcup_{r\in\mathsf{P}}\operatorname{at}(r).$ A rule r is normal if $|H_r|\leq 1$ and a program P is normal if all its rules are normal. The dependency graph \mathcal{D}_P is the directed graph defined on the set $\bigcup_{r\in\mathsf{P}}H_r\cup B_r^+$ of atoms, where for every rule $r\in\mathsf{P}$ two atoms $a\in B_r^+$ and $b\in H_r$ are joined by an edge (a,b). A program P is tight if \mathcal{D}_P has no directed cycle [Fages, 1994].

An interpretation I is a set of atoms. I satisfies a rule rif $(H_r \cup B_r^-) \cap I \neq \emptyset$ or $B_r^+ \setminus I \neq \emptyset$. I is a model of P if it satisfies all rules of P. The Gelfond-Lifschitz (GL) reduct of P under I is the program P^I obtained from P by first removing all rules r with $B_r^- \cap I \neq \emptyset$ and then removing all $\neg z$ where $z \in B_r^-$ from the remaining rules r [Gelfond and Lifschitz, 1991]. I is an answer set of a program P if I is a minimal model (w.r.t. \subseteq) of P^I. The problem of deciding whether an ASP program has an answer set is called *consistency*, which is $\Sigma_2^{\bar{p}}$ -complete [Eiter and Gottlob, 1995]. If the input is restricted to normal programs, the complexity drops to NP-complete [Bidoít and Froidevaux, 1991; Marek and Truszczyński, 1991]. The following characterization of answer sets is often applied for normal programs [Lin and Zhao, 2003; Janhunen, 2006]. Let I be a model of a normal program P and φ be a function (ordering) $\varphi: I \to$ $\{0,\ldots,|I|-1\}$ over I. We say a rule $r\in P$ is suitable for justifying $a \in I$ if (i) $a \in H_r$, (ii) $B_r^+ \subseteq I$, (iii) $I \cap B_r^- = \emptyset$, as well as (iv) $I \cap (H_r \setminus \{a\}]) = \emptyset$. An atom $a \in I$ is founded if there is a rule $r \in P$ justifying a, which is the case if r is suitable for justifying a and $\varphi(b) < \varphi(a)$ for every $b \in B_r^+$, and unfounded otherwise. Then, I is an answer set of P if

(i) I is a model of P, and (ii) I is founded, i.e., every $a \in I$ is founded. For tight programs, the ordering φ is not needed.

Non-ground ASP. Let $p_1, \dots p_n$ be predicates, where each takes *arity* $|p_i|$ many variables for $1 \le i \le n$. A *(non-ground)* program Π is a set of *(non-ground)* rules of the form

$$p_1(X_1) \lor \ldots \lor p_{\ell}(X_{\ell}) \leftarrow p_{\ell+1}(X_{\ell+1}), \ldots, p_m(X_m), (1)$$

 $\neg p_{m+1}(X_{m+1}), \ldots, \neg p_n(X_n),$

where for every variable vector X_i we have $|X_i| = |p_i|$, and whenever $x \in \langle X_1, \dots, X_\ell, X_{m+1}, \dots, X_n \rangle$, then $x \in \langle X_{\ell+1}, \dots, X_m \rangle$ (safeness). For a non-ground rule r, we let $H_r := \{p_1(X_1), \dots, p_\ell(X_\ell)\}, \ B_r^+ := \{p_{\ell+1}(X_{\ell+1}), \dots, p_m(X_m)\}, \ B_r^- := \{p_{m+1}(X_{m+1}), \dots, p_n(X_n)\},$ and $\mathrm{var}(r) := \{x \in X \mid p(X) \in H_r \cup B_r^+ \cup B_r^-\}.$ We use heads(\$\Pi\$) := \{p(X) \in H_r \| r \in \Pi\$}, hpreds(\$\Pi\$):= \{p \ p(X) \in \text{heads}(\Pi\$)}. Without loss of generality, we assume that variables are unique per rule, i.e., for every two rules $r, r' \in \Pi$, we have $\mathrm{var}(r) \cap \mathrm{var}(r') = \emptyset$. Attributes disjunctive, normal, and tight naturally carry over to nonground rules (programs). The rule size corresponds to $||r|| := |B_r^+| + |B_r^-| + |H_r|$ and program size $||\Pi|| := \sum_{r \in \Pi} ||r||$. In order to ground \$\Pi\$, we require a given set \$\mathcal{F}\$ of facts, i.e.,

In order to ground Π , we require a given set \mathcal{F} of facts, i.e., atoms of the form p(D) with p being a predicate of Π and D being a vector over domain values of size |D| = |p|. We say that D is part of the domain of Π , defined by $\mathrm{dom}(\Pi) := \{d \in D \mid p(D) \in \mathcal{F}\}$. We refer to the domain vectors over $\mathrm{dom}(\Pi)$ for a variable vector X of size |X| by $\mathrm{dom}(X)$. Let D be a domain vector over variable vector X and vector Y contain only variables of X. We refer to the domain vector of D restricted to Y by D_Y . The grounding $\mathcal{G}(\Pi)$ consists of \mathcal{F} and ground rules obtained by replacing each rule T of Form (1) for every domain vector $D \in \mathrm{dom}(\langle \mathrm{var}(T) \rangle)$ by

$$p_1(D_{X_1}) \lor \ldots \lor p_{\ell}(D_{X_{\ell}}) \leftarrow p_{\ell+1}(D_{X_{\ell+1}}), \ldots, p_m(D_{X_m}), \\ \neg p_{m+1}(D_{X_{m+1}}), \ldots, \neg p_n(D_{X_n}).$$

Example 1. Consider the non-ground program $\Pi := \{r\}$ with $r = a(X,Y) \leftarrow b(X), c(Y,Z)$ and $\mathcal{F} := \{b(1), c(1,2).\}$. Observe that $\operatorname{dom}(X) = \{1\}$ and $\operatorname{dom}(Y) = \operatorname{dom}(Z) = \{1,2\}$. The grounding $\mathsf{P} = \mathcal{G}(\Pi)$ of Π consists of:

$$\{a(1,1) \leftarrow b(1), c(1,1). \ a(1,1) \leftarrow b(1), c(1,2). \ a(1,2) \leftarrow b(1), c(2,1). \ a(1,2) \leftarrow b(1), c(2,2). \}$$

The only answer set of P *is* $\{b(1), c(1, 2), a(1, 1)\}$.

3 Body-Decoupled Grounding via Search

In this section, we introduce our concept of *body-decoupled* grounding, whose idea is to reduce the grounding size by decoupling dependencies between different predicates of rule bodies. We briefly motivate the potential of this idea.

Example 2. Assume the following non-ground program Π' that decides in (13) for each edge (e) of a given graph, whether to pick it (p) or not (\bar{p}) . Then, in (14) it is ensured that the choice of edges does not form triangles.

$$p(A,B) \vee \bar{p}(A,B) \leftarrow e(A,B)$$
 (13)

$$\leftarrow p(X,Y), p(Y,Z), p(X,Z), X \neq Y, Y \neq Z, X \neq Z. \tag{14}$$

The typical grounding effort of (14) is in $\mathcal{O}(|\text{dom}(\Pi')|^3)$. Our approach grounds body predicates of (14) individually, yielding linear bounds in the size of facts, i.e., $\mathcal{O}(|\text{dom}(\Pi')|^2)$.

```
Guess Answer Set Candidates
h(D) \vee h(D) \leftarrow
                                                                                                              for every h(X) \in \text{heads}(\Pi), D \in \text{dom}(X)
                                                                                                                                                                                                                                                                          (2)
Ensure Satisfiability
                  \operatorname{sat}_x(d) \leftarrow
                                                                                                              for every r \in \Pi, x \in var(r)
                                                                                                                                                                                                                                                                          (3)
d \in dom(x)
                                                                                                              for every r \in \Pi, p(X) \in B_r^+, D \in \text{dom}(X), X = \langle x_1, \dots, x_\ell \rangle
\operatorname{sat}_r \leftarrow \operatorname{sat}_{x_1}(D_{\langle x_1 \rangle}), \dots, \operatorname{sat}_{x_\ell}(D_{\langle x_\ell \rangle}), \neg p(D)
                                                                                                                                                                                                                                                                          (4)
                                                                                                              for every r \in \Pi, p(X) \in H_r \cup B_r^-, D \in \text{dom}(X), X = \langle x_1, \dots, x_\ell \rangle
                                                                                                                                                                                                                                                                          (5)
\operatorname{sat}_r \leftarrow \operatorname{sat}_{x_1}(D_{\langle x_1 \rangle}), \dots, \operatorname{sat}_{x_\ell}(D_{\langle x_\ell \rangle}), p(D)
                                                                                                              where \Pi = \{r_1, \ldots, r_n\}
                                                                                                                                                                                                                                                                          (6)
\operatorname{sat} \leftarrow \operatorname{sat}_{r_1}, \dots, \operatorname{sat}_{r_n}
                                                                                                              for every r \in \Pi, x \in var(r), d \in dom(x)
\operatorname{sat}_x(d) \leftarrow \operatorname{sat}
                                                                                                                                                                                                                                                                          (7)
 \leftarrow \neg sat
                                                                                                                                                                                                                                                                          (8)
Prevent Unfoundedness
                  \operatorname{uf}_y(\langle D, d \rangle) \leftarrow h(D)
                                                                                                             for every r \in \Pi, h(X) \in H_r, D \in \text{dom}(X), y \in \text{var}(r), y \notin X
                                                                                                                                                                                                                                                                          (9)
d \in dom(y)
\operatorname{uf}_r(D_X) \leftarrow \operatorname{uf}_{y_1}(D_{\langle X,y_1\rangle}), \dots, \operatorname{uf}_{y_\ell}(D_{\langle X,y_\ell\rangle}), \neg p(D_Y) \quad \text{for every } r \in \Pi, h(X) \in H_r, p(Y) \in B_r^+, D \in \operatorname{dom}(\langle X,Y\rangle), Y = \langle y_1, \dots, y_\ell \rangle \quad (10)
\operatorname{uf}_r(D_X) \leftarrow \operatorname{uf}_{y_1}(D_{\langle X,y_1\rangle}), \dots, \operatorname{uf}_{y_\ell}(D_{\langle X,y_\ell\rangle}), p(D_Y) \quad \text{ for every } r \in \Pi, h(X) \in H_r, p(Y) \in B_r^-, D \in \operatorname{dom}(\langle X,Y\rangle), Y = \langle y_1, \dots, y_\ell \rangle \quad (11)
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Figure 1: Body-decoupled grounding procedure \mathcal{R} for a given tight non-ground program Π , which creates a disjunctive *ground* program.

Body-Decoupled Grounding for Tight ASP. First, we present our approach for cycle-free programs without disjunction. To this end, we assume a given non-ground, tight program Π and a set \mathcal{F} of facts. For each predicate p(X) in heads (Π) , we use every instantiation of p(X) and its negation $\bar{p}(X)$ over dom (Π) , resulting in atoms AtPred := $\{p(D), \bar{p}(D) \mid p(X) \in \text{heads}(\Pi), D \in \text{dom}(X)\}$; the atoms over $\bar{p}(X)$ are for technical reasons only, and are not needed when emplyoing e.g., choice rules [Simons $et\ al.$, 2002].

 $\leftarrow \operatorname{uf}_{r_1}(D), \ldots, \operatorname{uf}_{r_m}(D)$

In addition to AtPred and in accordance with the semantics of ASP, we require to ensure (i) satisfiability and (ii) foundedness. For (i) computing models of rules, we require atoms AtSat := {sat, sat_r, sat_x(d) | $r \in \Pi, x \in$ $var(r), d \in dom(x)$, where sat (sat_r) indicates satisfiability (of non-ground rule r), respectively. of the form $sat_x(d)$ indicates that for checking satisfiability, we assign variable x of non-ground rule r to domain value $d \in dom(x)$. For (ii) ensuring foundedness, we use variables AtUf := $\{ \operatorname{uf}_r(D_X), \operatorname{uf}_y(D_{\langle X,y \rangle}) \mid r \in \Pi, D \in \Pi \}$ $\operatorname{dom}(\langle \operatorname{var}(r) \rangle), h(X) \in H_r, y \in \operatorname{var}(r), y \notin X\}.$ Intuitively, $\operatorname{uf}_r(D)$ indicates that r fails to justify h(D) for head predicate $h(X) \in H_r$ and domain vector $D \in dom(h)$, where $\mathrm{uf}_y(D_{\langle X,y\rangle})$ refers to the assigned domain value d that variable y gets in this foundedness check for h(D). Overall, the number $|AtPred \cup AtSat \cup AtUf|$ of auxiliary atoms is in $\mathcal{O}(\|\Pi\| \cdot |\text{dom}(\Pi)|^{a+1})$, where a is the largest predicate arity.

Next, we are in position to explain our reduction for tight programs. The overall idea consists of three parts and is part of the reduction \mathcal{R} , which transforms the non-ground, tight program Π into a ground, disjunctive program $\mathcal{R}(\Pi)$ consisting of \mathcal{F} and the rules given in Figure 1. Rules (2) take care of guessing answer set candidates, then Rules (3)–(8) ensure (i) satisfiability, and Rules (9)–(12) model (ii) foundedness.

Interestingly, the only disjunction that is indeed crucial and cannot be modeled via choice rules, is the disjunction part of Rules (3) for (i) satisfiability, which is responsible for guessing and saturating assignments of variables to domain values. The construction is such that whenever for a non-ground rule $r \in \Pi$, there is an assignment of variables to domain val-

ues such that the resulting ground rule is satisfied, Rules (4) or (5) yield sat_r . If such an atom sat_r can be derived for all non-ground rules of $r \in \Pi$, we follow sat by Rules (6), which is mandatory (cf. Rules (8)). Then, Rules (7) apply *saturation*, which causes the assignment of all domain values to every variable. Assuming that a grounding of a rule $r \in \Pi$ was not satisfied, then there would exist a \subseteq -smaller model of the reduct, invalidating the answer set candidate. Intuitively, the construction takes care that there is an answer set of $\mathcal{R}(\Pi)$, only if the grounding of Π admits an answer set.

for every $h(X) \in \text{heads}(\Pi), D \in \text{dom}(X), \{r_1, \dots, r_m\} = \{r \in \Pi \mid h(X) \in H_r\}$ (12)

Rules (9) are for (ii) preventing unfoundedness, ensuring that for each head atom h(D) contained in a model of $\mathcal{R}(\Pi)$, variables get assigned domain values for proving foundedness. Unjustifiability of a rule $r \in \Pi$ for atom h(D) is derived by Rules (10) and (11), which is prevented by Rules (12).

Appendix A shows reduction $\mathcal{R}(\Pi)$ for Π from Example 1.

Theorem 1 (Correctness). Let Π be any tight, non-ground program. Then, the grounding procedure \mathcal{R} on Π is correct, i.e., the answer sets of $\mathcal{R}(\Pi)$ restricted to $\operatorname{at}(\mathcal{G}(\Pi))$ match the answer sets of $\mathcal{G}(\Pi)$. Precisely, for every answer set M' of $\mathcal{R}(\Pi)$ there is an answer set $M' \cap \operatorname{at}(\mathcal{G}(\Pi))$ of $\mathcal{G}(\Pi)$.

 $\begin{array}{l} \textit{Proof}\left(\textit{Sketch}\right). \;\; \Leftarrow : \; \text{Assume that} \; M \; \text{is an answer set of} \; \mathcal{G}(\Pi), \\ \text{but there is no} \; M' \supsetneq M \; \text{with} \; M' \cap \text{at}(\mathcal{G}(\Pi)) = M \cap \text{at}(\mathcal{G}(\Pi)) \\ \text{that is an answer set of} \; \mathcal{R}(\Pi). \; \; \text{This results in either (i)} \; M \\ \text{being not a model of} \; \mathcal{G}(\Pi), \; \text{or (ii)} \; M \; \text{being unfounded.} \;\; \Rightarrow : \\ \text{Let} \; M' \; \text{be an answer set of} \; \mathcal{R}(\Pi) \; \text{and assume that} \; M \; := \\ M' \cap \text{at}(\mathcal{G}(\Pi)) \; \text{is not an answer set of} \; \mathcal{G}(\Pi). \; \text{Then, either (i)} \\ M' \; \text{is not a model of} \; \mathcal{R}(\Pi), \; \text{or (ii)} \; M' \; \text{is unfounded.} \end{array} \ \Box$

Notably, one can split the program and partially apply \mathcal{R} .

Corollary 1 (Partial Reducibility). Given a tight, non-ground program Π and a partition of Π into programs Π_1, Π_2 with hpreds $(\Pi_1) \cap \text{hpreds}(\Pi_2) = \emptyset$. Then, the answer sets of $\mathcal{R}(\Pi_1) \cup \mathcal{G}(\Pi_2)$ restricted to $\text{at}(\mathcal{G}(\Pi))$ match those of $\mathcal{G}(\Pi)$. Interestingly, if the predicate arities a are fixed, the grounding procedure \mathcal{R} works in polynomial time, since our technique does not suffer from large rules (or large rule bodies).

Theorem 2 (Polynomial Runtime and Grounding Size). Let Π be any tight, non-ground program, where every pred-

Improved Foundedness, replacing (10)–(12) of $\mathcal R$

$$\begin{aligned} & \text{uf}_{\text{rch}(Y,X)}(D_{\langle \text{rch}(Y,X)\rangle}) \leftarrow \text{uf}_{y_1}(D_{\langle X,y_1\rangle}), \dots, \text{uf}_{y_\ell}(D_{\langle X,y_\ell\rangle}), \neg p(D_Y) & \text{for every } r \in \Pi, h(X) \in H_r, p(Y) \in B_r^+, D \in \text{dom}(\langle X,Y\rangle), \\ & Y = \langle y_1, \dots, y_\ell \rangle & \text{(15)} \end{aligned} \\ & \text{uf}_{\text{rch}(Y,X)}(D_{\langle \text{rch}(Y,X)\rangle}) \leftarrow \text{uf}_{y_1}(D_{\langle X,y_1\rangle}), \dots, \text{uf}_{y_\ell}(D_{\langle X,y_\ell\rangle}), p(D_Y) & \text{for every } r \in \Pi, h(X) \in H_r, p(Y) \in B_r^-, D \in \text{dom}(\langle X,Y\rangle), \\ & Y = \langle y_1, \dots, y_\ell \rangle & \text{(16)} \end{aligned} \\ & \leftarrow h(D), [\bigvee_{t \in h(Y,X)}(D_{\langle \text{rch}(Y,X)\rangle})], \dots, [\bigvee_{t \in h(Y,X)}(D_{\langle \text{rch}(Y,X)\rangle})] & \text{for every } r \in \Pi, h(X) \in H_r, p(Y) \in B_r^-, D \in \text{dom}(\langle X,Y\rangle), \\ & Y = \langle y_1, \dots, y_\ell \rangle & \text{(16)} \end{aligned} \\ & \leftarrow h(D), [\bigvee_{t \in h(Y,X)}(D_{\langle \text{rch}(Y,X)\rangle})], \dots, [\bigvee_{t \in h(Y,X)}(D_{\langle \text{rch}(Y,X)\rangle})] & \text{for every } r \in \Pi, h(X) \in H_r, p(Y) \in B_r^-, D \in \text{dom}(\langle X,Y\rangle), \\ & Y = \langle y_1, \dots, y_\ell \rangle & \text{(16)} \end{aligned}$$

Figure 2: More involved formalization of checking for unfounded atoms based on Observation 1, yielding alternative \mathcal{R}' of (cf. \mathcal{R} of Figure 1).

Additional Rules for Foundedness of Normal Programs

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[p(D) \prec p'(D')] \lor [p'(D') \prec p(D)] \leftarrow \qquad \text{for every } p(X), p'(X') \in \text{heads}(\Pi), D \in \text{dom}(X), D' \in \text{dom}(X'), p(D) \neq p'(D') \quad (18)
\leftarrow [p_1(D_1) \prec p_2(D_2)], [p_2(D_2) \prec p_3(D_3)], \qquad \text{for every } p_1(X_1), p_2(X_2), p_3(X_3) \in \text{heads}(\Pi), D_1 \in \text{dom}(X_1), D_2 \in \text{dom}(X_2),
[p_3(D_3) \prec p_1(D_1)] \qquad \qquad D_3 \in \text{dom}(X_3), p_1(D_1) \neq p_2(D_2), p_1(D_1) \neq p_3(D_3), p_2(D_2) \neq p_3(D_3) \quad (19)
\text{uf}_r(D_X) \leftarrow \text{uf}_{y_1}(D_{\langle X, y_1 \rangle}), \neg [p(D_Y) \prec h(D_X)] \qquad \text{for every } r \in \Pi, h(X) \in H_r, p(Y) \in B_r^+, D \in \text{dom}(\langle X, Y \rangle), Y = \langle y_1, \dots, y_\ell \rangle \quad (20)
\text{uf}_{y_\ell}(D_{\langle X, y_\ell \rangle}), \neg [p(D_Y) \prec h(D_X)] \qquad \text{for every } p(X), p'(X') \in \text{heads}(\Pi), D \in \text{dom}(X), D' \in \text{dom}(X'), p(D) \neq p'(D') \quad (18)
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Figure 3: For normal, non-ground programs, we require additional rules for ensuring foundedness, yielding \mathcal{R}'' (cf. \mathcal{R} of Figure 1).

icate has arity at most a. Then, the grounding procedure \mathcal{R} on Π is polynomial, i.e., runs in time $\mathcal{O}(\|\Pi\| \cdot |\mathrm{dom}(\Pi)|^{2 \cdot a})$. We do not expect a significant runtime improvement in the worst case. Further, the expressiveness increase from normal to disjunctive programs is inevitable (already for fixed arity).

Proposition 1 (Disjunctive Programs Inevitable). Let Π be any tight non-ground program, where every predicate has arity at most a. Then, unless $NP = \Sigma_2^P$, there cannot be a polynomial grounding procedure \mathcal{R}' , where $\mathcal{R}'(\Pi)$ is normal. Proof (Idea). While consistency for normal ground programs is in NP, Σ_2^P -hardness for disjunctive (head-cycle-free) nonground programs, cf. [Eiter et al., 2007, Lem. 6], can be lifted to tight non-ground programs, by observing tightness after converting disjunctive rules into normal ones via shifting. \square

Utilizing Variable Independencies. Having established the basic grounding procedure \mathcal{R} , we provide an improvement based on reachable paths. For a non-ground program Π , we define the *variable graph* \mathcal{V}_{Π} , whose vertices are the variables $\operatorname{var}(r)$ of rules $r \in \Pi$, with an edge between two $x, y \in \operatorname{var}(r)$ whenever there is a predicate p(X) in r with $x, y \in X$. Let X, Y be variable vectors. We refer by $\operatorname{rch}(Y, X)$ to those vertices in X that are *reachable* by some $y \in Y$ in \mathcal{V}_{Π} .

Observation 1 (Variable-Justification Independency). Given a non-ground program Π , any rule $r \in \Pi$ with $h(X) \in H_r$ and $p(Y) \in B_r^+$. Further, let I be a set of atoms over $\mathcal{G}(\Pi)$. Then, if r does not justify $h(D_X) \in I$ for $D \in \operatorname{dom}(\langle X, Y \rangle)$ due to $p(D_Y) \notin I$, we have that r fails to justify any $h(D') \in I$ with $D' \in \operatorname{dom}(X)$ and $D'_{\langle \operatorname{rch}(Y,X) \rangle} = D_{\langle \operatorname{rch}(Y,X) \rangle}$ as well.

Example 3. Recall Π of Example 1. Let $I = \{b(2)\}$ and assume an arbitrary atom a(2,1). The rule $a(X,Y) \leftarrow b(X)$, c(Y,Z) does not justify $a(2,1) \in I$ due to $b(1) \notin I$. Hence, this rule fails to justify any atom a(2,y) with $y \in dom(Y)$.

Based on the observation above, we provide an alternative of our reduction \mathcal{R} , which is preferred in case of more independent variables according to Observation 1. Instead of AtUf, we use atoms AtUfI := $\{\operatorname{uf}_{\operatorname{rch}(Y,X)}(D_X),\operatorname{uf}_y(D_{\langle X,y\rangle})|r\in\Pi,D\in\operatorname{dom}(\langle\operatorname{var}(r)\rangle),h(X)\in H_r,p(Y)\in B_r^+\cup B_r^-,y\in Y,y\notin X\}$. The updated reduction \mathcal{R}' is given in Figure 2, where Rules (15) and (16)

replace Rules (10) and (11) with the only difference that a different head predicate is used with a potentially smaller domain vector. Further, Rules (12) are replaced by Rules (17), which contain disjunctions in their bodies, as in the well-known weight rules [Gebser *et al.*, 2011]. Alternatively, one can do the cross product among the sets of disjuncts of (17).

Body-Decoupled Grounding for Normal ASP. The idea of our reduction \mathcal{R} can be also lifted to normal, non-ground programs. To this end, one can rely on orderings (level mappings) as defined in the preliminaries. To simplify the presentation, we only show a variant that uses a quadratic number of auxiliary atoms for comparison, instead of encoding orderings. We therefore use for every two distinct ground atoms p(D), p'(D) of Π , an additional auxiliary predicate $[p(D) \prec p'(D')]$ responsible for storing precedence in the order of derivation. Then, given \mathcal{R} of Figure 1, for normal programs we just need to add those rules of Figure 3, resulting in \mathcal{R}'' . Intuitively, Rules (18) determine precedence among different atoms and Rules (19) take care of transitivity of " \prec ". In addition to (15) and (16), Rules (20) add a case of unfoundedness, if precedence is not suitable for justifying.

Analogously to Theorem 1, one can show correctness of \mathcal{R}'' for normal, non-ground programs, including when applied to program parts as in Corollary 1. To this end, we capture predicate dependencies as follows. The dependency graph \mathcal{D}_{Π} has as vertices the predicates $hpreds(\Pi)$ with a directed edge from p to q whenever there is a rule $r \in \Pi$ with $p(X) \in B_r^+$ and $q(Y) \in H_r$. Then, a set $C \subseteq hpreds(\Pi)$ of predicates is a strongly-connected component (SCC) if C is a \subseteq -largest set such that for every two distinct predicates p,q in C there is a directed path from p to q in \mathcal{D}_{Π} . Then, SCCs yield sufficient conditions for partially applying \mathcal{R}'' .

Theorem 3. Given a normal, non-ground program Π , a partition Π_1, Π_2 with $\operatorname{hpreds}(\Pi_1) \cap \operatorname{hpreds}(\Pi_2) = \emptyset$ s.t. for any $SCCs\ C_1$ of $\mathcal{D}_{\Pi_1},\ C_2$ of $\mathcal{D}_{\Pi_2}\colon C_1 \cap C_2 = \emptyset$. The answer sets of $\mathcal{R}''(\Pi_1) \cup \mathcal{G}(\Pi_2)$ restricted to $\operatorname{at}(\mathcal{G}(\Pi))$ match those of $\mathcal{G}(\Pi)$.

4 Implemented Prototype & Experiments

We implemented a software tool, called newground, realizing body-decoupled grounding via search as described above. The system newground is written in Python3 and

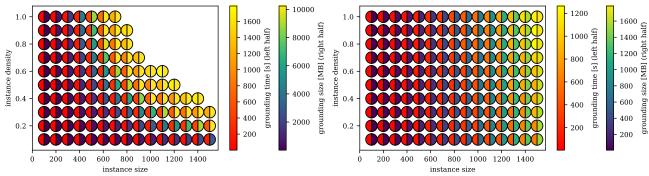


Figure 4: Grounding profile of S1 for gringo (left) and newground (right). The x-axis refers to the instance size; the y-axis indicates density. Circles mark instances grounded < 1800s; the left (right) half depicts grounding time (size), respectively. Mind the scales (10000 vs. 1600).

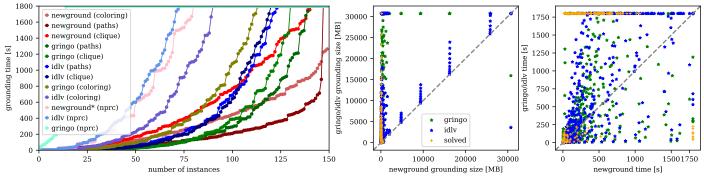


Figure 5: (Left): Cactus plot of grounding time over Scenarios S1–S4 for newground (red color tones), gringo (green tones), and idlv (blue tones). The x-axis shows the number of instances; the y-axis is the runtime in seconds, sorted in ascending order for each solver individually. The legend is sorted from best to worst. (Middle): Scatter plot of grounding size over Scenarios S1–S4 of newground (x-axis) compared to both gringo (blue) and idlv (green) on the y-axis. Those instances that could be solved are highlighted in orange. (Right): Scatter plot of grounding time over Scenarios S1–S4 of newground (x-axis) compared to gringo and idlv (y-axis); grounding and solving time in orange.

uses, among others, the API and of clingo 5.5 and its ability to efficiently parse logic programs via syntax trees. As Corollary 1 suggests, we opted for an implementation enabling partial reducability, allowing users to select program parts that shall be reduced and others that are (traditionally) grounded. Besides Figure 1, we utilize variable independencies, as highlighted in Figure 2. Internally, we optimize by pre-compiling usual compare-operators to achieve more compact programs. Notably, newground is *not restricted to decision problems*, which enables, e.g., counting and (quantitative) reasoning.

In order to evaluate newground, we design a series of benchmarks. Clearly, we cannot beat highly optimized grounders in all imaginable scenarios. Instead, we discuss potential use cases, where body-decoupled grounding is preferrable, since this approach can be incorporated into every grounder. We consider these (directed) graph *scenarios*:

- S1 (coloring): Compute *edge colorings* over three colors s.t. no incoming nor outgoing edge has the same color.
- S2 (paths): Find *reachable paths* between source and destination, where each node admits only one outgoing edge.
- S3 (clique): Obtain directed subgraphs containing *cliques* (fully connected subgraphs of size at least three).
- S4 (nprc): Compute non-partition-removal colorings, whose encoding is taken from [Weinzierl *et al.*, 2020].
- S5 (stable marriage): Obtain so-called *stable marriages*; encoding taken from the ASP competition 2014.

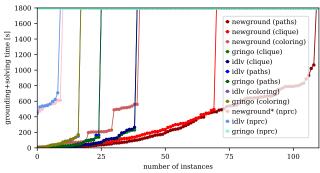
S1 aims at providing a basic coloring problem. The decision of S2 is in P, but it can be extended; counting such paths is hard (#P-complete problem [Valiant, 1979]). Deciding S3 is also polynomial-time computable, but ready to be used as part of more expressive problems. Finally, we consider known ASP benchmarks like S4 (nprc) and S5 (stable marriage).

Experimental Hypotheses. We study Hypotheses H1–H4.

- H1: In contrast to traditional grounding, newground suffers less from increased *instance density* and *instance size*.
- H2: Body-decoupled grounding can massively reduce *grounding sizes* and *grounding times* of large instances.
- H3: Body-decoupled grounding can *improve overall* (+solv-ing) performance on crafted and application instances.
- H4: The idea of body-decoupled grounding, where suitable, efficiently interoperates with other approaches.

Benchmark Instances. For answering the hypotheses, we use crafted (random) and applicable ASP competition instances. Note, that *competition instances are not designed* to run into grounding bottlenecks. Therefore, we randomly generated instances for S1–S4, from 100 to 1500 vertices, with an edge probability (density) from 0.1 to 1.0. These are particularly useful to answer H1–H3. For S5, we took competition instances (plus crafted ones), which aid in analyzing H2–H4.

Compared Grounders. We study the performance of these *exact (full) grounders*, neither considering ASP modulo the-



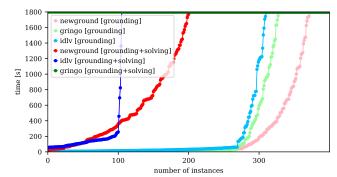


Figure 6: (Left): Corresponding cactus plot of overall (grounding and solving) time over Scenarios S1-S4 (cf. grounding performance of Figure 5 (Left)). (Right): Cactus plot of grounding time as well as grounding and solving performance for Scenario S5 (stable marriage).

ory nor lazy grounding, which are just different problems.

- gringo version 5.5.1; idlv version 1.1.6
- newground: Body-decoupled grounding is applied on certain (manually fixed, see Appendix C) non-ground rules of the respective programs that potentially cause grounding overhead. The remaining part is grounded with gringo.
- newground*: newground with idlv instead of gringo.

Benchmark Measures. We measure the *grounding sizes*, *grounding times*, and solving capabilities of groundings. For full intercomparability, grounding size measures the size of the corresponding text output without I/O operations. The reason is that newground can be interleaved with other program parts, which fails to work with smodels or aspif formats out-of-the-box. Anyway, relative orders of magnitude of measured grounding sizes unambiguously stand. For comparing *overall performance*, we use solver clingo 5.5.1 with "-q--stats=2", i.e., we compute one answer set; for newground we add "--project" to ensure answer sets are over the same atoms. We *limit* main memory (RAM) to 16GB and restrict runs to 1800s overall runtime (grounding *and solving*). For clarity, plots use *cut-off* grounding sizes of 10GB or 30GB.

Results: Grounding Scalability Study. For systematically studying the grounding of S1–S5, we use crafted instances and create for each solver a grounding profile. Figure 4 depicts the grounding profiles of S1 for gringo (left) and newground (right), showing grounding times and sizes, depending on the instance size (x-axis) and density (y-axis). Interestingly, for newground grounding times and sizes for a fixed instance size (column) of Figure 4 (right) are quite similar. This is in contrast to Figure 4 (left), suggesting that compared to gringo, newground is not that sensitive to instance density. Further, the rows of Figure 4 (right) are also similar. Overall, we confirm H1 (see Appendix C for more data).

Results: Grounding Performance. In Figure 5 (left), we show a cactus plot of grounding times over S1–S4 for all grounders. Interestingly, newground clearly outperforms here; gringo performs second-best, except for S4 (nprc), where treewidth-based grounding seems promising. The grounding sizes over all instances for S1–S4 are given in the scatter plot Figure 5 (middle), which shows that newground massively reduces sizes (almost all dots above diagonal). Interestingly, newground *solves* instances, where gringo and idly output groundings beyond 30GB. Figure 5 (right) shows

a scatter plot revealing that newground improves grounding times (blue/green dots above diagonal). Among those below the diagonal, most dots are below 250s (y-axis), suggesting a portfolio that uses gringo or idlv for 250s and switches to newground if unsuccessful. The orange dots represent overall runtimes among solved instances, showing that no instance below the diagonal (where newground falls behind) can be solved in more than 250s. Overall, we confirm H2.

Results: Overall Performance. For a deeper study of overall (solving) performance, we refer to the cactus plot of Figure 6 (left), showing runtimes for S1–S4. While newground performs best, we still see a clear difference between solving and grounding performance (cf. Figure 5). One could believe that analyzing combined grounding and solving is unfair for gringo and idly, since newground grounds faster. However, Figure 5 (left) shows that for, e.g., S2 and S3, there are many instances grounded by gringo within 200s. Interestingly, Figure 6 (left) reveals, that only a small amount of those can then actually be solved by clingo within the remaining 1600s. This is also visible in Figure 6 (right), which depicts a cactus plot for S5 over grounding as well as overall runtime. We expect potential to optimize clingo's internal handling of large programs (big data) for better utilizing newground's grounding performance. However, we confirm H3; and H4 due to the integration of newground with gringo and idlv.

5 Conclusion, Discussion & Future Work

This work introduces a technique on body-decoupled grounding, where grounding-intense ASP rules are treated by means of a novel reduction. The reduction translates tight (normal) non-ground rules into disjunctive ground rules, thereby being exponential only in the maximum predicate arity. Our empirical evaluation shows an advantage: Compared to state-of-the-art exact grounders, *body-decoupled* grounding applied on crucial program parts massively reduces grounding size.

We plan on integrating our approach into (intelligent) grounders, aiming for heuristics that automatically estimate program parts where newground likely pays off. Further, our approach could be extended to disjunctive non-ground programs, which unfortunately is Σ_3^P -complete for bounded arity [Eiter *et al.*, 2007]. Nevertheless, we expect "almost tightness" notions, e.g., [Fandinno and Hecher, 2021; Hecher, 2022], to aid in reducing heavy rules for disjunctive atoms. Finally, we see potential in solving on big data.

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A Additional Examples

 $a(1,1) \vee \bar{a}(1,1)$. $a(1,2) \vee \bar{a}(1,2)$.

Example 4. Consider the non-ground program Π and facts $\mathcal{F} = \{b(1).\ c(1,2).\}$ from Example 1. Grounding the program as described above and shown in Figure 1, results in the ground program $P = \mathcal{R}(\Pi)$ of Figure 7. The answer sets of P restricted to symbols a,b,c of Π yield $\{a(1,1),b(1),c(1,2)\}$. The rules for the foundedness can be decreased via the approach from Figure 2, here shown in Figure 8 for this example. As defined in Section 3, Rules (10) are replaced with Rules (15), and Rules (17) replace Rules (12), which, apart from the number of rules, also reduces predicates arities.

$$\begin{array}{lll} \operatorname{sat}_X(1).\ \operatorname{sat}_Y(1) \vee \operatorname{sat}_Y(2).\ \operatorname{sat}_Z(1) \vee \operatorname{sat}_Z(2). & (3) \\ \operatorname{sat}_r \leftarrow \operatorname{sat}_X(1), \operatorname{sat}_Y(1), a(1,1). & (4) \\ \operatorname{sat}_r \leftarrow \operatorname{sat}_X(1), \operatorname{sat}_Y(2), a(1,2).\ \operatorname{sat}_r \leftarrow \operatorname{sat}_X(1), \neg b(1). \\ \operatorname{sat}_r \leftarrow \operatorname{sat}_Y(1), \operatorname{sat}_Z(1), \neg c(1,1).\ \operatorname{sat}_r \leftarrow \operatorname{sat}_Y(1), \operatorname{sat}_Z(1), \neg c(1,2). \\ \operatorname{sat}_r \leftarrow \operatorname{sat}_Y(2), \operatorname{sat}_Z(1), \neg c(2,1).\ \operatorname{sat}_r \leftarrow \operatorname{sat}_Y(2), \operatorname{sat}_Z(1), \neg c(2,2). \\ \operatorname{sat} \leftarrow \operatorname{sat}_r. & (6) \\ \operatorname{sat}_X(1) \leftarrow \operatorname{sat}.\ \operatorname{sat}_Y(1) \leftarrow \operatorname{sat}.\ \operatorname{sat}_Y(2) \leftarrow \operatorname{sat}. & (7) \\ \operatorname{sat}_Z(1) \leftarrow \operatorname{sat}.\ \operatorname{sat}_Z(2) \leftarrow \operatorname{sat}. & (8) \\ \operatorname{uf}_Z(1,1,1) \vee \operatorname{uf}_Z(1,1,2) \leftarrow a(1,1). & (9) \\ \operatorname{uf}_Z(1,2,1) \vee \operatorname{uf}_Z(1,2,2) \leftarrow a(1,2). \\ \operatorname{uf}_r(1,1) \leftarrow \neg b(1).\ \operatorname{uf}_r(1,2) \leftarrow \neg b(1). & (10) \\ \operatorname{uf}_r(1,1) \leftarrow \neg c(1,1), \operatorname{uf}_Z(1,1,1).\ \operatorname{uf}_r(1,1) \leftarrow \neg c(1,2), \operatorname{uf}_Z(1,2,2). \\ \operatorname{uf}_r(1,2) \leftarrow \neg c(2,1), \operatorname{uf}_Z(1,2,1).\ \operatorname{uf}_r(1,2) \leftarrow \neg c(2,2), \operatorname{uf}_Z(1,2,2). \\ \leftarrow a(1,1), \operatorname{uf}_r(1,1). \leftarrow a(1,2), \operatorname{uf}_r(1,2). & (12) \\ \operatorname{Figure 7:} \mathcal{R}(\Pi) \text{ with } \Pi \text{ from Example 1, guided by } \mathcal{R} \text{ of Figure 1.} \end{array}$$

$$\begin{split} & \text{uf}_{\{X\}}(1) \leftarrow \neg b(1). \\ & \text{uf}_{\{Y\}}(1) \leftarrow \neg c(1,1), \text{uf}_{Z}(1,1). \text{ uf}_{\{Y\}}(1) \leftarrow \neg c(1,2), \text{uf}_{Z}(1,2). \\ & \text{uf}_{\{Y\}}(2) \leftarrow \neg c(2,1), \text{uf}_{Z}(2,1). \text{ uf}_{\{Y\}}(2) \leftarrow \neg c(2,2), \text{uf}_{Z}(2,2). \\ & \leftarrow a(1,1), [\text{uf}_{\{X\}}(1) \vee \text{uf}_{\{Y\}}(1)]. \\ & \leftarrow a(1,2), [\text{uf}_{\{X\}}(1) \vee \text{uf}_{\{Y\}}(2)]. \end{split} \tag{15}$$

Figure 8: Optimization for the reduction of Π from Example 4.

Example 5. Consider the non-ground, normal program $\Pi_2 := \Pi \cup \{r_2\}$ with $r_2 = c(X,Y) \leftarrow a(X,Y)$, and program Π and facts $\mathcal{F} = \{b(1).\ c(1,2).\}$ from Example 1. While the grounding procedure \mathcal{R} can be used identical for normal programs, the additional rules, shown in Figure 9, ensure orderings as described above. For storing the order of derivation the auxiliary predicate p is used. Notice that the the number of rules were kept minimal by using only the relevant variables per rule, i.e. X,Y,Z for r and X,Y for r_2 , and taking the domain of each variable into consideration. The answer sets of $P = \mathcal{R}''(\Pi_2)$ restricted to symbols a,b,c of Π_2 yield $\{a(1,1),b(1),c(1,1),c(1,2)\}$, as expected.

$$\begin{split} & \mathbf{p}_{ac}(1,1) \vee \mathbf{p}_{ca}(1,1,1). \ \mathbf{p}_{ac}(1,1) \vee \mathbf{p}_{ca}(1,1,2). \\ & \mathbf{p}_{ac}(1,2) \vee \mathbf{p}_{ca}(1,2,1). \ \mathbf{p}_{ac}(1,2) \vee \mathbf{p}_{ca}(1,2,2). \\ & \mathbf{uf}_{r}(1,1) \leftarrow \neg \mathbf{p}_{ca}(1,1,1). \mathbf{uf}_{Z}(1,1,1). \\ & \mathbf{uf}_{r}(1,1) \leftarrow \neg \mathbf{p}_{ca}(1,1,2). \mathbf{uf}_{Z}(1,1,2). \\ & \mathbf{uf}_{r}(1,2) \leftarrow \neg \mathbf{p}_{ca}(1,2,1). \mathbf{uf}_{Z}(1,2,1). \\ & \mathbf{uf}_{r}(1,2) \leftarrow \neg \mathbf{p}_{ca}(1,2,2). \mathbf{uf}_{Z}(1,2,2). \\ & \mathbf{uf}_{r2}(1,1) \leftarrow \neg \mathbf{p}_{ac}(1,1). \ \mathbf{uf}_{r2}(1,2) \leftarrow \neg \mathbf{p}_{ac}(1,2). \end{split}$$

Figure 9: Additional rules for the reduction of program Π_2 from Example 5, guided by \mathcal{R}'' of Figure 3.

B Omitted Proofs

(2)

Theorem 2 (Polynomial Runtime and Grounding Size). Let Π be any tight, non-ground program, where every predicate has arity at most a. Then, the grounding procedure \mathcal{R} on Π is polynomial, i.e., runs in time $\mathcal{O}(\|\Pi\| \cdot |\text{dom}(\Pi)|^{2 \cdot a})$.

Proof (Sketch). The reduction \mathcal{R} constructs $\mathcal{O}(\|\Pi\| \cdot |\text{dom}(\Pi)|^a)$ many Rules (2) of constant size, $\mathcal{O}(\|\Pi\| \cdot a)$ many Rules (3) of size $|\text{dom}(\Pi)|$, as well as $\mathcal{O}(\|\Pi\| \cdot |\text{dom}(\Pi)|^a)$ many Rules (4) and (5) of size $\mathcal{O}(a)$. Then, there is one Rule (6) of size $\mathcal{O}(|\Pi|)$, $\mathcal{O}(\|\Pi\| \cdot |\text{dom}(\Pi)|)$ many Rules (7) and one Rule (8). Finally, there are $\mathcal{O}(\|\Pi\| \cdot |\text{dom}(\Pi)|^a \cdot a)$ many Rules (9) of size $\mathcal{O}(|\text{dom}(\Pi)| \cdot a)$, and we require $\mathcal{O}(\|\Pi\| \cdot |\text{dom}(\Pi)|^{2 \cdot a})$ effort for Rules (10) and (11), as well as $\mathcal{O}(\|\Pi\| \cdot |\text{dom}(\Pi)|)$ effort for Rules (12). □

Theorem 1 (Correctness). Let Π be any tight, non-ground program. Then, the grounding procedure \mathcal{R} on Π is correct, i.e., the answer sets of $\mathcal{R}(\Pi)$ restricted to $\operatorname{at}(\mathcal{G}(\Pi))$ match with the answer sets of $\mathcal{G}(\Pi)$. Precisely, for every answer set M' of $\mathcal{R}(\Pi)$ there is exactly one answer set $M' \cap \operatorname{at}(\mathcal{G}(\Pi))$ of $\mathcal{G}(\Pi)$.

Proof(Sketch). \Leftarrow : Let M be an answer set of $\mathcal{G}(\Pi)$ and assume towards a contradiction that there is no extension $M' \supset M$ of M with $M \cap \operatorname{at}(\mathcal{G}(\Pi)) = M' \cap$ $\operatorname{at}(\mathcal{G}(\Pi))$ such that M' is an answer set of $\mathcal{R}(\Pi)$. First, we construct $N := \{\bar{h}(D) \mid h(X) \in \text{heads}(\Pi), D \in \text{heads}(\Pi)\}$ $dom(X), h(D) \notin M$, which collects those head atoms that are not in M. Further, we gather satisfied rules by S := $\{\operatorname{sat}, \operatorname{sat}_r, \operatorname{sat}_x(d) \mid r \in \Pi, x \in \operatorname{var}(r), d \in \operatorname{dom}(x)\}.$ Then, we set appropriate domain values for the founded atoms $F := \{ \operatorname{uf}_{y_1}(\langle D, D'_{y_1} \rangle), \dots, \operatorname{uf}_{y_\ell}(\langle D, D'_{y_\ell} \rangle) \mid r \in \mathcal{F} \}$ $\Pi, h(X) \in H_r, \{y \in var(r) \mid y \notin X\} = \{y_1, \dots, y_\ell\}, D' \in$ $\operatorname{dom}(\langle \operatorname{var}(r) \rangle), h(D) \in M, D_X = D_X, p(Z) \in B_r^+ \cup D_X$ $B_r^-, (p(Z) \in B_r^+) \text{ iff } (p(D'_Z) \in M)\}.$ we set the domain values for unfounded atoms U := $\{\operatorname{uf}_r(D),\operatorname{uf}_u(\langle D,d(y)\rangle)\mid r\in\Pi,h(X)\in H_r,h(D)\in$ $M, y \in \text{var}(r), y \notin X$, there is no $\text{uf}_y(\langle D, d' \rangle) \in F$, where d(y) yields any arbitrary, fixed domain value in dom(y). Then, we let $M' := M \cup N \cup F \cup U$. Since by assumption M' is not an answer set, either (i) one of the rules is not satisfied, or (ii) M' is not minimal.

We proceed by case distinction. Case (i): Some rule $r \in \mathcal{R}(\Pi)$ is not satisfied by M'. It is easy to see that all heads of Rules (2)–(9) are satisfied by construction of M'. Rules (9) are satisfied by construction of F and G. The construction

of U also ensures that Rules (10) and (11) are satisfied. Observe that Rules (12) are satisfied, since M is an answer set of $\mathcal{G}(\Pi)$ and therefore every atom of M is founded, as constructed by F. This concludes Case (i), since every rule of $\mathcal{R}(\Pi)$ is satisfied.

Case (ii): Model M' of $\mathcal{R}(\Pi)$ is not minimal, i.e., there exists a model $M'' \subseteq M'$ with M'' being a model of $\mathcal{R}(\Pi)^{M'}$. Note that the difference between M' and M'' cannot be due to Rule (2), since then M' and M'' would be incomparable. If sat $\notin M''$, then by Rules (6), there is a rule $r_i \in$ Π with sat_r, \notin M", which by Rules (4) and (5) implies that $\mathcal{G}(r_i)$ is not satisfied by M''. This, however, contradicts the assumption that M is an answer set of $\mathcal{G}(\Pi)$. Consequently, by Rules (7), the difference between M' and M''cannot be due to any sat_x predicate. Further, M' and M''cannot differ in any predicate of the form uf_y either, since by Rules (9), for every $h(D) \in M'$, there is precisely one uf_u($\langle D, d \rangle$) $\in M'$. Similarly, Rules (10)–(12) yield deterministic consequences, depending only on the choice of Rules (2) and (9). As a result, we conclude that M' = M''cannot differ.

 \Rightarrow : Let M' be an answer set of $\mathcal{R}(\Pi)$ and assume towards a contradiction that $M:=M'\cap\operatorname{at}(\mathcal{G}(\Pi))$ is not an answer set of $\mathcal{G}(\Pi)$. Then, either (i) M is not a model of $\mathcal{G}(\Pi)$, or (ii) there is an $h(D)\in M$ that is unfounded by M. We proceed by case distinction.

Case (i): M is not a model of $\mathcal{G}(\Pi)$. Assume towards a contradiction that there is a rule $r_i \in \Pi$ with M not satisfying $\mathcal{G}(\{r_i\})$. Then, there is a subset $M'' \subsetneq M'$ with $\operatorname{sat}_{r_i} \notin M''$, $\operatorname{sat} \notin M''$ and sat_x predicate set accordingly for $x \in \operatorname{var}(r_i)$, that is a model of $\mathcal{G}(\Pi)$, which contradicts that M' is an answer set (\subseteq -minimal model) of $\mathcal{R}(\Pi)$.

Case (ii): There exists an $h(D) \in M$ that is unfounded by M. Then, no matter how the predicates uf_x are assigned in M', since $h(D) \in M$ is not founded, there is no rule $r_i \in \Pi$ that can be instantiated such that the body is satisfied and thereby justifying h(D). Consequently, for every such rule r_i , by Rules (10) and (11), there is at least one atom over predicates uf_{r_i} in M'. Finally, Rules (12) are not satisfied (for h(D)), which contradicts that M' is an answer set of $\mathcal{R}(\Pi)$.

Corollary 1 (Partial Reducibility). Given a tight, non-ground program Π and a partition of Π into programs Π_1, Π_2 with $\operatorname{hpreds}(\Pi_1) \cap \operatorname{hpreds}(\Pi_2) = \emptyset$. Then, the answer sets of $\mathcal{R}(\Pi_1) \cup \mathcal{G}(\Pi_2)$ restricted to $\operatorname{at}(\mathcal{G}(\Pi))$ match those of $\mathcal{G}(\Pi)$.

Proof (Idea). The result is a direct consequence of the proof construction of Theorem 1, which can be extended. Satisfiability checking of $\mathcal{R}(\Pi_1) \cup \mathcal{G}(\Pi_2)$ works similarly, since satisfiability of rules in $\mathcal{G}(\Pi_1)$ are ensured by Rules (3)–(8), and satisfiability of rules $\mathcal{G}(\Pi_2)$ is treated directly.

Since $\operatorname{hpreds}(\Pi_1) \cap \operatorname{hpreds}(\Pi_2) = \emptyset$, predicates in $\operatorname{hpreds}(\Pi_1)$, $\operatorname{hpreds}(\Pi_2)$ can only appear in *rule bodies* of Π_2 , Π_1 , respectively. Consequently, for a given model of $\mathcal{G}(\Pi)$, the foundedness of atoms over predicates $\operatorname{hpreds}(\Pi_1)$ is decided by $\mathcal{R}(\Pi_1)$, namely Rules (9)–(12), whereas foundedness of atoms over $\operatorname{hpreds}(\Pi_2)$ is checked by $\mathcal{G}(\Pi_2)$.

Theorem 3 (Partial Reducibility/Normal ASP). Given a normal, non-ground program Π and a partition of Π into programs Π_1, Π_2 with $\operatorname{hpreds}(\Pi_1) \cap \operatorname{hpreds}(\Pi_2) = \emptyset$, where for every SCC C_1 of \mathcal{D}_{Π_1} and SCC C_2 of \mathcal{D}_{Π_2} , we have $C_1 \cap C_2 = \emptyset$. Then, the answer sets of $\mathcal{R}''(\Pi_1) \cup \mathcal{G}(\Pi_2)$ restricted to $\operatorname{at}(\mathcal{G}(\Pi))$ match those of $\mathcal{G}(\Pi)$.

Proof(Sketch). \Leftarrow : Let M be an answer set of $\mathcal{G}(\Pi)$ and assume towards a contradiction that there is no extension $M' \supseteq M$ of M with $M \cap \operatorname{at}(\mathcal{G}(\Pi)) = M' \cap \operatorname{at}(\mathcal{G}(\Pi))$ such that M' is an answer set of $\mathcal{R}''(\Pi_1) \cup \mathcal{G}(\Pi_2)$. Since M is an answer set of $\mathcal{G}(\Pi)$, there is an order $\varphi: M \to \{0,\ldots,|M|-1\}$ that is used for showing foundedness of M. Similar to the proof of Theorem 1, we collect those head atoms not in M by $N := \{\bar{h}(D) \mid h(X) \in \operatorname{heads}(\Pi_1), D \in$ $dom(X), h(D) \notin M$. Then, we gather satisfied rules by $S := \{ \operatorname{sat}, \operatorname{sat}_r, \operatorname{sat}_x(d) \mid r \in \Pi_1, x \in \operatorname{var}(r), d \in \Pi_1 \}$ dom(x). We set appropriate domain values for the founded atoms $F := \{\widehat{\operatorname{uf}}_{y_1}(\langle D, D'_{y_1} \rangle), \dots, \operatorname{uf}_{y_\ell}(\langle D, D'_{y_\ell} \rangle)\}$ $r \in \Pi_1, h(X) \in H_r, \{y \in var(r) \mid y \notin X\}$ $\{y_1, \dots, y_\ell\}, D' \in \text{dom}(\langle \text{var}(r) \rangle), h(D) \in M, D_X = D'_X, p(Z) \in B^+_r \cup B^-_r, (p(Z) \in B^+_r) \text{ iff } (p(D'_Z) \in M), [p(Z) \in B^-_r] \text{ or } [\varphi(p(D'_Z)) < \varphi(h(D))] \}.$ Then, we set domain values for unfounded atoms U $\{\operatorname{uf}_r(D),\operatorname{uf}_y(\langle D,d(y)\rangle)\mid r\in\Pi_1,h(X)\in H_r,h(D)\in$ $M, y \in \text{var}(r), y \notin X$, there is no $\text{uf}_y(\langle D, d' \rangle) \in F$ }, where d(y) yields any arbitrary, fixed domain value in dom(y).

We encode ordering φ by setting $O := \{[a \prec b] \mid \{a,b\} \subseteq M, \varphi(a) < \varphi(b)\}$ and adding those over atoms not only in M by $P := \{[p(D) \prec q(D')] \mid \{p(X),q(Y)\} \subseteq \operatorname{heads}(\Pi_1),p(D) \in \operatorname{dom}(X),q(D') \in \operatorname{dom}(Y),p(D)\notin M \text{ or }q(D')\notin M,q(D')\succ p(D)\}, \text{ where }\succ \text{ is any arbitrary total ordering over atoms at}(\mathcal{G}(\Pi_1)).$ Finally, we let $M' := M \cup N \cup F \cup U \cup O \cup P$. Since by assumption M' is not an answer set, either (i) one of the rules is not satisfied, or (ii) M' is not minimal.

We proceed by case distinction. Case (i): Some rule $r \in \mathcal{R}(\Pi_1) \cup \mathcal{G}(\Pi_2)$ is not satisfied by M'. Obviously $\mathcal{G}(\Pi_2)$ is satisfied since M is an answer set of $\mathcal{G}(\Pi)$. Further, it is easy to see that all heads of Rules (2)–(9) are satisfied by construction of M'. Rules (9) are satisfied by construction of F and F and F and F are satisfied. Observe that Rules (12) are satisfied, since F is an answer set of F and therefore every atom of F is founded, as constructed by F. Further, by construction of F and F and F and F are satisfied. This concludes Case (i), since every rule of F and F are satisfied.

Case (ii): Model M' of $\mathcal{R}(\Pi_1) \cup \mathcal{G}(\Pi_2)$ is not minimal, i.e., there exists a model $M'' \subseteq M'$ with M'' being a model of $(\mathcal{R}(\Pi_1) \cup \mathcal{G}(\Pi_2))^{M'}$. Note that the difference between M' and M'' cannot be due to Rule (2), since then M' and M'' would be incomparable. If sat $\notin M''$, then by Rules (6), there is a rule $r_i \in \Pi$ with $\operatorname{sat}_{r_i} \notin M''$, which by Rules (4) and (5) implies that $\mathcal{G}(r_i)$ is not satisfied by M''. This, however, contradicts the assumption that M is an answer set of $\mathcal{G}(\Pi)$. Consequently, by Rules (7), the difference between M' and M'' cannot be due to any sat_x predicate. Further, M' and M'' cannot differ in any pred-

icate of the form uf_y either, since by Rules (9), for every $h(D) \in M'$, there is precisely one $\mathrm{uf}_y(\langle D,d\rangle) \in M'$. Similarly, Rules (10)–(12) yield deterministic consequences, depending only on the choice of Rules (2) and (9). Rules (18) cannot be responsible for the difference between M' and M'', since then M' and M'' would be incomparable (every answer set has to contain precisely one of these head atoms per grounding of Rules (19)). Finally, Rules (20) also yield deterministic consequences, given the choice of Rules (9) and (20). As a result, we conclude that M' and M'' can only differ due to Π_2 . Then, however, $M'' \cap \mathrm{at}(\mathcal{G}(\Pi))$ is also a model of $\mathcal{G}(\Pi_2)^M$. Further, $M'' \cap \mathrm{at}(\mathcal{G}(\Pi))$ is a model of $\mathcal{G}(\Pi)^M$ as well, since M'' is model of $(\mathcal{R}(\Pi_1) \cup \mathcal{G}(\Pi_2))^{M'}$, which by heads $(\Pi_1) \cap \mathrm{heads}(\Pi_2) = \emptyset$, could only be prevented by atoms over body predicates of Π_1 . Then, however, M is not an answer set since $M'' \cap \mathrm{at}(\mathcal{G}(\Pi)) \subsetneq M$.

 \Rightarrow : Let M' be an answer set of $\mathcal{R}(\Pi_1) \cup \mathcal{G}(\Pi_2)$ and assume towards a contradiction that $M := M' \cap \operatorname{at}(\mathcal{G}(\Pi))$ is not an answer set of $\mathcal{G}(\Pi)$. Then, either (i) M is not a model of $\mathcal{G}(\Pi)$, or (ii) there is an $h(D) \in M$ that is unfounded by M. We proceed by case distinction.

Case (i): M is not a model of $\mathcal{R}(\Pi_1) \cup \mathcal{G}(\Pi_2)$. Assume towards a contradiction that there is a rule $r_i \in \Pi$ with M not satisfying $\mathcal{G}(\{r_i\})$. Then, $r_i \in \Pi_1$ since M is by definition a model of $\mathcal{G}(\Pi_2)$. Then, there is a subset $M'' \subsetneq M'$ with $\operatorname{sat}_{r_i} \notin M''$, $\operatorname{sat} \notin M''$ and sat_x predicate set accordingly for $x \in \operatorname{var}(r_i)$, that is a model of $\mathcal{G}(\Pi)$, which contradicts that M' is an answer set (\subseteq -minimal model) of $\mathcal{R}(\Pi_1) \cup \mathcal{G}(\Pi_2)$.

Case (ii): There exists an $h(D) \in M$ that is unfounded by $\mathcal{R}(\Pi_1) \cup \mathcal{G}(\Pi_2)$. If $h \in \operatorname{heads}(\Pi_2)$, then by construction of M, $h(D) \in M'$ is also unfounded by every rule in Π_2 (as well as by every rule in Π_1 since $\operatorname{heads}(\Pi_1) \cap \operatorname{heads}(\Pi_2) = \emptyset$ and there is no SCC overlap that could cause foundedness), contradicting that M' is an answer set. Otherwise, irrelevant how the predicates uf_x or predicates of the form $\cdot \prec \cdot$ are assigned in M', since $h(D) \in M$ is not founded, there is no rule $r_i \in \Pi_1$ that can be grounded such that the resulting body is satisfied and the ground rule justifies h(D). Consequently, for every such rule r_i , by Rules (10), (11), and (20) there is at least one atom over predicates uf_{r_i} in M'. Finally, Rules (12) are not satisfied (for h(D)), which contradicts that M' is an answer set of $\mathcal{R}(\Pi_1) \cup \mathcal{G}(\Pi_2)$.

C Additional Experimental Data and Plots

Benchmark Platform. All our solvers ran on a cluster consisting of 12 nodes. Each node of the cluster is equipped with two Intel Xeon E5-2650 CPUs, where each of these 12 physical cores runs at 2.2 GHz clock speed and has access to 256 GB shared RAM. Results are gathered on Ubuntu 16.04.1 LTS powered on kernel 4.4.0-139 with hyperthreading disabled using version 3.7.6 of Python3.

Partial Reductions for newground. Grounder newground applies body-decoupled grounding for those rules that hold a crucial role in terms of grounding size (grounding bottleneck). Body-decoupled grounding is applied for

S1: on the rules prohibiting identical edge colors

S2: on the rule restricting to 1 chosen outgoing edge per node

S3: on the rule checking for a clique

S4: on a rule that ensures non-partition (over reachability)

S5: on the rules preventing polygamy

For detailed encodings and the rules subject to reduction, we refer to the supplemental material, which will be made publicly available after submission. There, the respective rules below "#program rules." are the ones subject to reduction and newground grounds all other rules via gringo or idly (newground*). For the vanilla solvers gringo and idly, the lines containing "#program rules." are removed before those solvers are invoked.

Additional Data and Plots. Table 1 presents detailed benchmark results over Scenarios S1–S5. Figure 10 shows the grounding profile of idlv for Scenario S1. Figures 11, 12, and 13 show the grounding profiles of gringo and newground for Scenarios S2, S3, and S4, respectively. Figure 14 provides scatter plots of grounding size and grounding time for S5, similar to the scatter plots of Figure 5.

	grounded instances					grounded+solved instances			
solver (scenario)	Σ	grounding time[h]	grounding[GB]	$\max(grounding[MB])$	$\max(\texttt{gringo}\ size[MB])$	\sum	time[h]	fastest	unique
newground (S1:coloring)	150	14.5	87.0	1785.69	102400.0	40 /150	57.27	32	23
gringo (S1:coloring)	114	29.44	4346.11	34612.96	34612.96	17/150	66.78	5	0
idlv (S1:coloring)	91	39.37	6165.15	13070.43	13070.43	17/150	66.78	3	0
newground (S2:paths)	147	8.6	315.17	368.1	102400.0	109 /150	30.55	101	84
gringo (S2:paths)	140	15.54	1783.91	34431.63	34431.63	25/150	62.8	8	0
idlv (S2:paths)	124	24.61	2998.41	19011.75	19011.75	25/150	62.82	0	0
newground (S3:clique)	141	25.02	931.96	708.82	102400.0	70 /150	43.12	52	38
gringo (S3:clique)	131	18.73	2397.33	27465.99	27465.99	39/150	56.24	16	0
idlv (S3:clique)	120	24.84	3280.83	17836.62	17836.62	39/150	56.29	9	0
newground * (S4:nprc)	80	46.21	7953.04	38732.34	102400.0	10 /150	71.51	6	1
gringo (S4:nprc)	15	69.27	13187.82	57268.14	57268.14	0/150	75.0	0	0
idlv (S4:nprc)	73	47.96	8505.09	43224.31	102400.0	9/150	71.91	4	0
newground (S5,competition)	50	0.1	1.12	32.49	84.26	49 /50	1.47	45	1
gringo (S5,competition)	50	0.05	2.88	84.26	84.26	0/50	25.0	0	0
idlv (S5,competition)	50	0.14	3.21	94.19	84.26	48/50	2.56	4	0
newground (S5,crafted)	322	29.76	3319.58	23596.12	102400.0	152 /350	125.94	100	95
gringo (S5,crafted)	278	47.27	8096.1	45380.83	45380.83	0/350	175.0	0	0
idlv (S5,crafted)	261	54.39	9302.68	21822.21	21593.23	57/350	149.28	52	0
newground	890	124.18	12607.87	38732.34	102400.0	430 /1000	329.86	336	242
gringo	728	180.3	30314.16	57268.14	57268.14	81/1000	460.82	29	0
idlv	719	191.31	30255.38	43224.31	102400.0	195/1000	409.63	72	0

Table 1: Detailed results over Scenarios S1–S5, where best performance values are given in bold-face. The block "grounded instances" shows the number of grounded instances (\sum), the total ground time in hours ("grounding time[h]", timeouts count as 1800s), the total grounding size in GB ("ground[GB]", non-groundable instances count as 100GB), the largest grounding size encountered ("max(grounding[MB])"), and the largest groundable instance given in the corresponding gringo grounding size ("max(gringo size[MB])"). Then, the block "grounded+solved instances" refers to overall performance values, where " \sum " shows the number of solved instances (/ the number of total instances), "time[h]" gives the overall solving time in hours (timeouts count as 1800s), "fastest" refers to the number of instances solved first, and "unique" gives the number of uniquely solved instances.

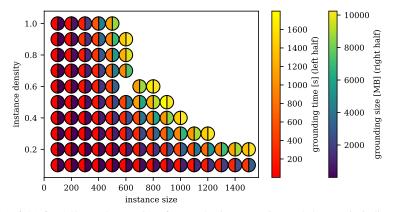


Figure 10: Grounding profile of S1 for idlv. The x-axis refers to the instance size and the y-axis indicates its density. Circles mark groundable instances (< 1800s), whose left half represents the grounding time and the right color depicts the grounding size, cf. Figure 4.

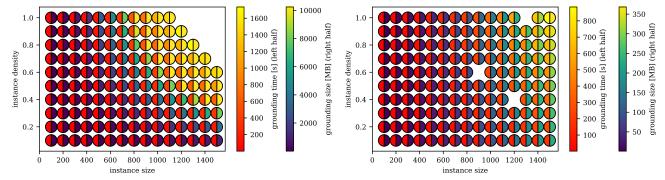


Figure 11: (Left): Grounding profile for gringo of S2, similar to Figure 4. (Right): Grounding profile for newground of S2. Compared to gringo (and idlv), newground grounds larger and denser instances faster, with a grounding size reduction of up to $\frac{1}{50}$.

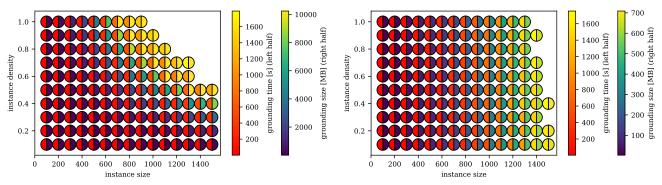


Figure 12: (Left): Grounding profile for gringo of S3. (Right): Grounding profile for newground of S3.

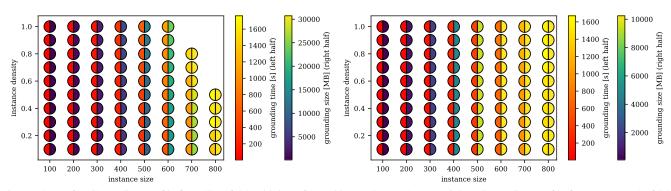


Figure 13: (Left): Grounding profile for idlv of S4, which performed better than gringo. (Right): Grounding profile for newground of S4. Both plots depict the same grounding cut-off size of 30GB, but the solvers scale differently.

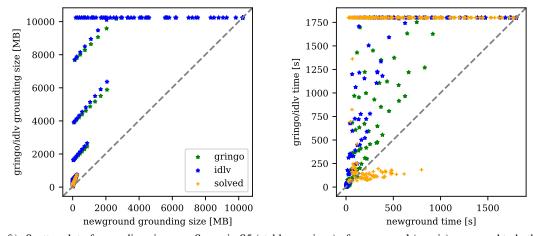


Figure 14: (Left): Scatter plot of grounding size over Scenario S5 (stable marriage) of newground (x-axis) compared to both gringo (blue) and idlv (green) on the y-axis. Those instances that could be solved are highlighted in orange. (Right): Scatter plot of grounding time over Scenario S5 of newground (x-axis) compared to gringo and idlv (y-axis); grounding and solving time in orange.