

Multi-Criteria Optimization in ASP and its Application to Linux Package Configuration

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Outline

- 1 Introduction
- 2 Package Configuration
- 3 Multi-Criteria Optimization
- 4 Experimental Results
- 5 Discussion

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Motivation

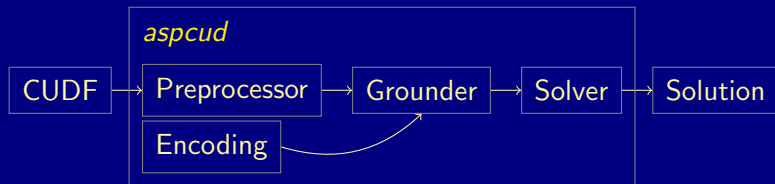
- Maintaining packages in modern Linux distributions is difficult
 - Complex dependencies
 - Large package repositories
 - Ever changing in view of software development
- Challenges for package configuration tools
 - Large problem size
 - Soft (and hard) constraints
 - Multiple optimization criteria
- Contributions of this work
 - Package configuration via Answer Set Programming (ASP)
 - Uniform modeling by encoding plus instances
 - Solving techniques for multi-criteria optimization

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Overview

aspcud Tool for solving package configuration problems



Preprocessor Converts CUDF input to ASP instance

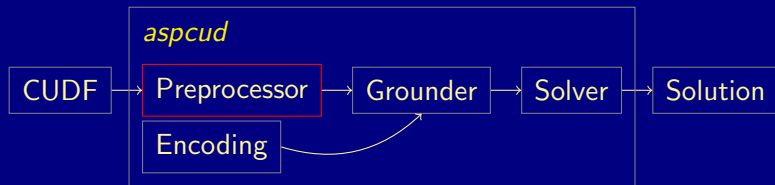
Encoding First-order problem specification

Grounder Instantiates first-order variables

Solver Searches for (optimal) answer sets

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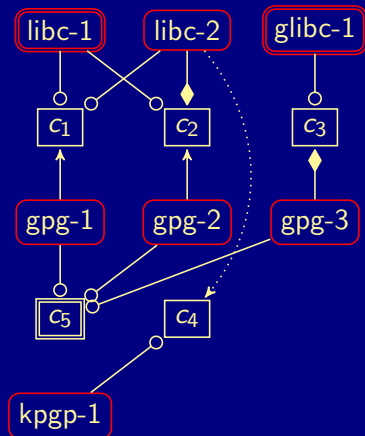
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Instance Format



Installable Packages:

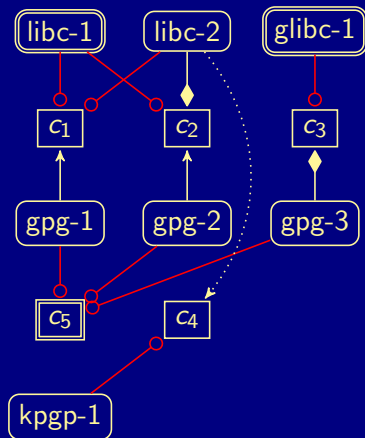
```
package(libc,1).  
package(libc,2).
```

```
package(glibc,1).
```

```
package(gpg,1).  
package(gpg,2).  
package(gpg,3).
```

```
package(kpgp,1).
```


Instance Format



Clauses:

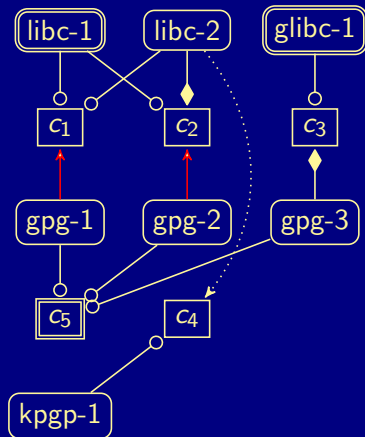
```
satisfies(libc,1,c1).  
satisfies(libc,1,c2).  
satisfies(libc,2,c1).
```

```
satisfies(glibc,1,c3).
```

```
satisfies(gpg,1,c5).  
satisfies(gpg,2,c5).  
satisfies(gpg,3,c5).
```

```
satisfies(kpgp,1,c4).
```

Instance Format

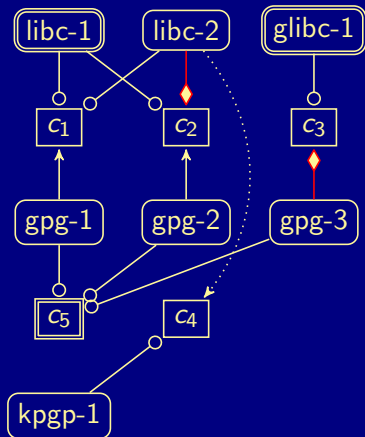


Package Dependencies:

`depends (gpg, 1, c1) .`

`depends (gpg, 2, c2) .`

Instance Format

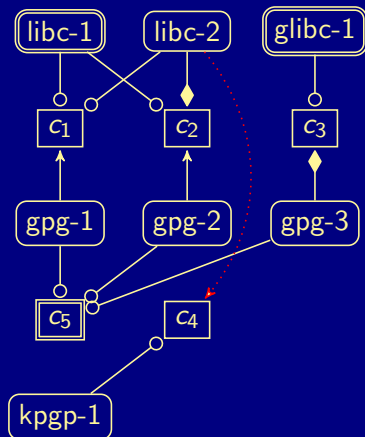


Package Conflicts:

```
conflicts(libc,2,c2).
```

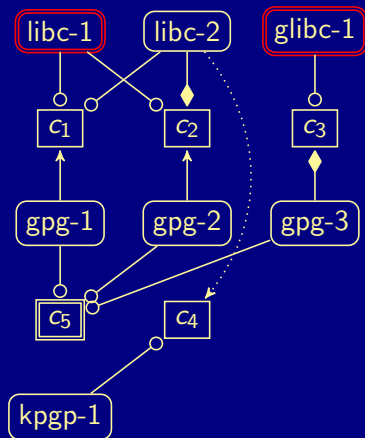
```
conflicts(gpg,3,c3).
```

Instance Format



Package Recommendations:
`recommends(libc,2,c4).`

Instance Format

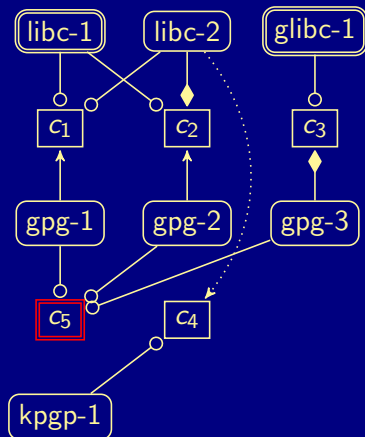


Installed Packages:

```
installed(libc,1).
```

```
installed(glibc,1).
```

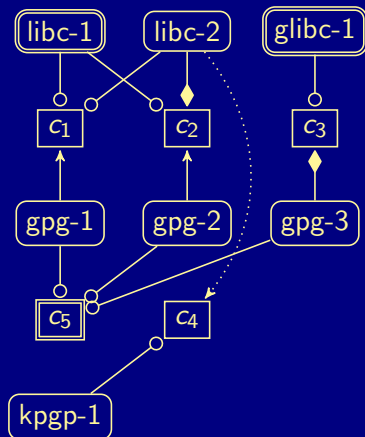
Instance Format



Requests:

`requested(c5).`

Instance Format



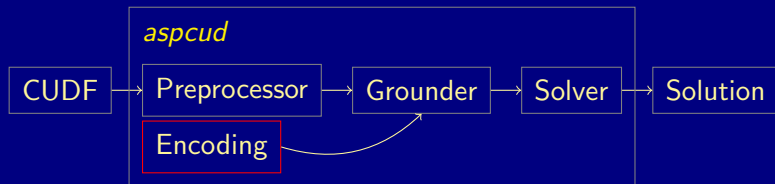
Optimization Criteria:

```
utility(delete,-1).
```

```
utility(change,-2).
```

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Hard Constraints

```
% choose packages to install
{ install(N,V) } :- package(N,V).

% derive required clauses
exclude(C) :- install(N,V), conflicts(N,V,C).
include(C) :- install(N,V), depends(N,V,C).
% derive satisfied clauses
satisfy(C) :- install(N,V), satisfies(N,V,C).

% assert required clauses to be (un)satisfied
:- exclude(C),      satisfy(C).
:- include(C), not satisfy(C).
:- request(C), not satisfy(C).
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Soft Constraints

```
% auxiliary definitions
install(N)    :- install(N,V).
installed(N)  :- installed(N,V).

% derive optimization criteria violations
violate(newpkg,N) :-
    utility(newpkg,L), install(N), not installed(N).
violate(delete,N) :-
    utility(delete,L), installed(N), not install(N).
% similar for other criteria
...

% impose soft constraints
#minimize[ violate(U,T) = 1 @ -L : utility(U,L) : L < 0 ].
#maximize[ violate(U,T) = 1 @  L : utility(U,L) : L > 0 ].
```

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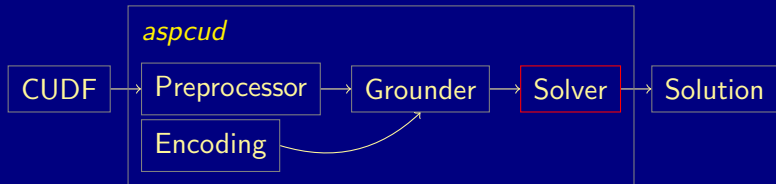
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Optimization Algorithm



- Package configuration problems are often under-constrained
- Lexicographical optimization algorithm enumerates too much

Alternative Approach

- Optimize criteria in the order of significance
- Decrease upper bounds (costs) w.r.t. witnesses
- Proceed to next criterion upon unsatisfiability

Design Goals

- Incorporate into conflict-driven solving
- Keep as much learned information as possible
- Build upon standard features like assumptions

```

1 Model  $\leftarrow \perp$ 
2 foreach Criterion do
3   Lower  $\leftarrow 0$ 
4   Upper  $\leftarrow \text{eval}(\text{Criterion}, \text{Model})$ 
5   while Lower < Upper do
6     add((Criterion  $\cup \langle \sim \text{Aux} = -\infty \rangle$ ) < Upper)
7     M  $\leftarrow \text{solve}(\{\text{Aux}\})$ 
8     if M  $\neq \perp$  then
9       Model  $\leftarrow$  M
10      Upper  $\leftarrow \text{eval}(\text{Criterion}, \text{Model})$ 
11      simplify( $\{\text{Aux}\}$ )
12    else
13      if Model =  $\perp$  then return  $\perp$ 
14      Lower  $\leftarrow$  Upper
15      simplify( $\{\sim \text{Aux}\}$ )
16 return M

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Setup

- Benchmarks
 - 117 instances from the 3rd MISC-live run
 - Optimization criteria
 - paranoid, trendy
 - user1 (-notuptodate, -removed, -changed)
 - user2 (-changed, -removed, -unsat_recommends, -new)
 - user3 (-changed, -notuptodate, -removed, -new)
- Optimization algorithms
 - $clasp_0$: lexicographical optimization
 - $clasp_1$: hierarchical optimization
 - $clasp_2$: hierarchical optimization with exponential steps
- Optimization heuristics
 - $clasp_i^0$: no optimization-specific heuristic
 - $clasp_i^1$: falsify literals to minimize upon branching
 - $clasp_i^2$: falsify literals to minimize until conflict
 - $clasp_i^3$: combines $clasp_i^1$ and $clasp_i^2$
- Search restarts
 - $clasp_i^j$ -r: perform restart after each model
(mandatory with $clasp_i^2$ and $clasp_i^3$)
- Scoring like in MISC-live run

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	paranoid		trendy		user1		user2		user3	
Solver	S	T/O	S	T/O	S	T/O	S	T/O	S	T/O
<i>clasp</i> ₀ ⁰ -r	431	2,287/6	1730	23,829/ 80	935	14,349/35	525	5,097/12	1031	14,184/37
<i>clasp</i> ₀ ⁰	416	2,294/6	2375	29,781/105	1727	21,897/73	1224	14,697/45	671	11,178/21
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<i>cudf2msu</i>	610	3,051/8	669	5,318 / 8	1270	8,709/18	548	3,238 / 7	504	4,750/ 9
<i>cudf2pbo</i>	465	2,727 /7	1082	21,302/ 68	520	6,168/13	462	3,575/ 7	537	3,487 / 8
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<i>clasp</i> ₂ ² -r	445	2,402/6	706	16,551/ 50	528	5,788/13	419	2,700/ 5	436	5,519/13
<i>clasp</i> ₂ ³ -r	434	2,345/6	748	16,982/ 51	518	5,850/14	415	2,559/ 5	457	5,360/13
<i>cudf2msu</i>	610	3,051/8	669	5,318 / 8	1270	8,709/18	548	3,236 / 7	504	4,750/ 9
<i>cudf2pbo</i>	465	2,727 /7	1082	21,302/ 68	520	6,168/13	462	3,575/ 7	537	3,487 / 8
<i>p2cudf</i>	463	2,920/8	696	19,105/ 60	516	3,947 / 7	573	6,927/16	577	8,063/21

	paranoid		trendy		user1		user2		user3	
Solver	S	T/O	S	T/O	S	T/O	S	T/O	S	T/O
<i>clasp</i> ₀ ⁰ -r	431	2,287/6	1730	23,829/ 80	935	14,349/35	525	5,097/12	1031	14,184/37
<i>clasp</i> ₀ ⁰	416	2,294/6	2375	29,781/105	1727	21,897/73	1224	14,697/45	671	11,178/21
<i>clasp</i> ₀ ¹ -r	410	2,210/6	1560	22,660/ 73	898	13,466/30	502	4,654/ 9	980	13,682/35
<i>clasp</i> ₀ ¹	410	2,326/6	2079	26,471/ 92	1723	21,525/72	922	10,767/31	658	10,675/23
<i>clasp</i> ₀ ² -r	427	2,135/6	712	16,867/ 51	527	5,891/11	426	2,981/ 5	587	7,628/20
<i>clasp</i> ₀ ³ -r	429	2,134 /6	740	17,079/ 52	507	5,863/12	425	3,044/ 6	576	7,769/21
<i>clasp</i> ₁ ⁰ -r	425	2,428/6	579	16,713/ 50	550	5,819/14	434	3,000/ 6	710	8,958/25
<i>clasp</i> ₁ ⁰	417	2,418/6	549	16,544/ 50	475	5,318/12	411	2,538/ 5	502	6,279/16
<i>clasp</i> ₁ ¹ -r	429	2,405/6	622	17,304/ 50	518	5,908/13	438	2,976/ 6	676	8,938/23
<i>clasp</i> ₁ ¹	427	2,372/6	613	16,946/ 49	490	5,478/12	416	2,562/ 5	496	6,144/16
<i>clasp</i> ₁ ² -r	427	2,352/6	571	16,646/ 50	518	5,358/13	418	2,582/ 5	471	6,356/16
<i>clasp</i> ₁ ³ -r	429	2,346/6	547	16,386 / 50	499	5,306 /12	413	2,498/ 5	497	6,255/16
<i>clasp</i> ₂ ⁰ -r	425	2,392/6	806	16,598/ 50	523	5,583/13	421	2,677/ 6	479	5,548/12
<i>clasp</i> ₂ ⁰	417	2,364/7	748	17,132/ 50	487	5,823/14	422	2,583/ 5	482	5,592/15
<i>clasp</i> ₂ ¹ -r	416	2,378/6	752	17,269/ 52	492	5,663/12	414	2,409 / 5	451	5,349 /11
<i>clasp</i> ₂ ¹	425	2,365/6	864	17,128/ 51	517	6,151/15	412	2,681/ 5	463	5,972/14
<i>clasp</i> ₂ ² -r	445	2,402/6	706	16,551/ 50	528	5,788/13	419	2,700/ 5	436	5,519/13
<i>clasp</i> ₂ ³ -r	434	2,345/6	748	16,982/ 51	518	5,850/14	415	2,559/ 5	457	5,360/13
<i>cudf2msu</i>	610	3,051/8	669	5,318 / 8	1270	8,709/18	548	3,238 / 7	504	4,750/ 9
<i>cudf2pbo</i>	465	2,727 /7	1082	21,302/ 68	520	6,168/13	462	3,575/ 7	537	3,487 / 8
<i>p2cudf</i>	463	2,920/8	696	19,105/ 60	516	3,947 / 7	573	6,927/16	577	8,063/21

	paranoid		trendy		user1		user2		user3	
Solver	S	T/O	S	T/O	S	T/O	S	T/O	S	T/O
<i>clasp</i> ₀ ⁰ -r	431	2,287/6	1730	23,829/ 80	935	14,349/35	525	5,097/12	1031	14,184/37
<i>clasp</i> ₀ ⁰	416	2,294/6	2375	29,781/105	1727	21,897/73	1224	14,697/45	671	11,178/21
<i>clasp</i> ₀ ¹ -r	410	2,210/6	1560	22,660/ 73	898	13,466/30	502	4,654/ 9	980	13,682/35
<i>clasp</i> ₀ ¹	410	2,326/6	2079	26,471/ 92	1723	21,525/72	922	10,767/31	658	10,675/23
<i>clasp</i> ₀ ² -r	427	2,135/6	712	16,867/ 51	527	5,891/11	426	2,981/ 5	587	7,628/20
<i>clasp</i> ₀ ³ -r	429	2,134 /6	740	17,079/ 52	507	5,863/12	425	3,044/ 6	576	7,769/21
<i>clasp</i> ₁ ⁰ -r	425	2,428/6	579	16,713/ 50	550	5,819/14	434	3,000/ 6	710	8,958/25
<i>clasp</i> ₁ ⁰	417	2,418/6	549	16,544/ 50	475	5,318/12	411	2,538/ 5	502	6,279/16
<i>clasp</i> ₁ ¹ -r	429	2,405/6	622	17,304/ 50	518	5,908/13	438	2,976/ 6	676	8,938/23
<i>clasp</i> ₁ ¹	427	2,372/6	613	16,946/ 49	490	5,478/12	416	2,562/ 5	496	6,144/16
<i>clasp</i> ₁ ² -r	427	2,352/6	571	16,646/ 50	518	5,358/13	418	2,582/ 5	471	6,356/16
<i>clasp</i> ₁ ³ -r	429	2,346/6	547	16,386 / 50	499	5,306 /12	413	2,498/ 5	497	6,255/16
<i>clasp</i> ₂ ⁰ -r	425	2,392/6	806	16,598/ 50	523	5,583/13	421	2,677/ 6	479	5,548/12
<i>clasp</i> ₂ ⁰	417	2,364/7	748	17,132/ 50	487	5,823/14	422	2,583/ 5	482	5,592/15
<i>clasp</i> ₂ ¹ -r	416	2,378/6	752	17,269/ 52	492	5,663/12	414	2,409 / 5	451	5,349 /11
<i>clasp</i> ₂ ¹	425	2,365/6	864	17,128/ 51	517	6,151/15	412	2,681/ 5	463	5,972/14
<i>clasp</i> ₂ ² -r	445	2,402/6	706	16,551/ 50	528	5,788/13	419	2,700/ 5	436	5,519/13
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Outline

- 1 Introduction
- 2 Package Configuration
- 3 Multi-Criteria Optimization
- 4 Experimental Results
- 5 Discussion

Discussion

- Multi-criteria optimization algorithm
 - optimizing criteria in the order of significance
 - keeping learned information whenever possible
 - retracting invalid constraints using assumptions
 - avoiding solver relaunches after unsatisfiability proofs
- Optimization-oriented heuristics to guide search for optima
- Techniques used in package configuration tool *aspcud*
- Future work: combination with lower bound refinement