



MARIO ALVIANO
DeMaCS@UNICAL

Deep Reasoning in AI with ASP

Joint effort with Francesco Ricca



<https://asp-chef.alviano.net/s/dimacs-to-truth-table>

<https://asp-chef.alviano.net/s/tutorials/essai2024/1>

<https://asp-chef.alviano.net/s/tutorials/essai2024/2>

Suggested Book

Potassco Guide

<https://github.com/potassco/guide/releases/download/v2.2.0/guide.pdf>

L2

<https://asp-chef.alviano.net/s/asp-playground>

<https://asp-chef.alviano.net/s/grounding-playground>

<https://asp-chef.alviano.net/s/tutorials/basic-usage>

<https://asp-chef.alviano.net/s/tutorials/aquarium>

L3

<https://asp-chef.alviano.net/s/tutorials/billy-the-kid>

<https://asp-chef.alviano.net/s/tutorials/fortress>

<https://asp-chef.alviano.net/s/tutorials/essai2024/3>

L4

Structured Declarative Language (SDL)

Explainability Issues

ASP and LLMs interaction

L5

Structured Declarative Language

<https://asp-chef.alviano.net/s/SDL>

GitHub Repository

<https://github.com/dodaro/SDL>

Declarative Programming

Expresses the **logic** of a computation
without describing its **control flow**

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Answer Set Programming
is (at its core) declarative

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But is ASP a
good speaker?

Relational Algebra vs Structured Query Language (SQL)

π_{id}

$\sigma_{username = "root" \text{ AND } password = "toor"} user$

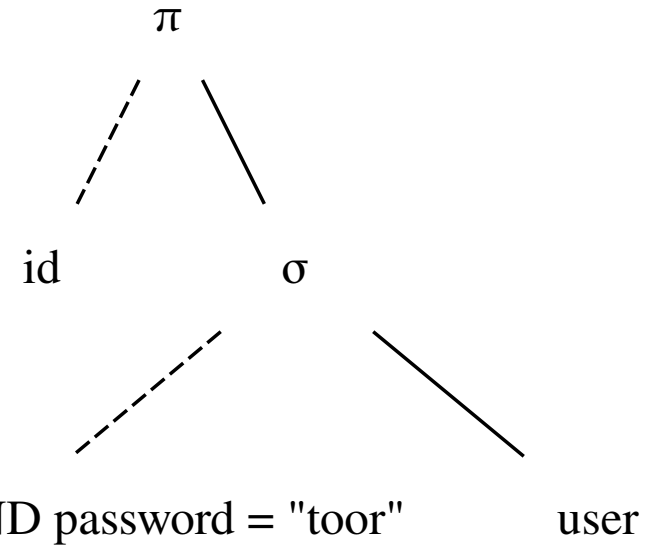
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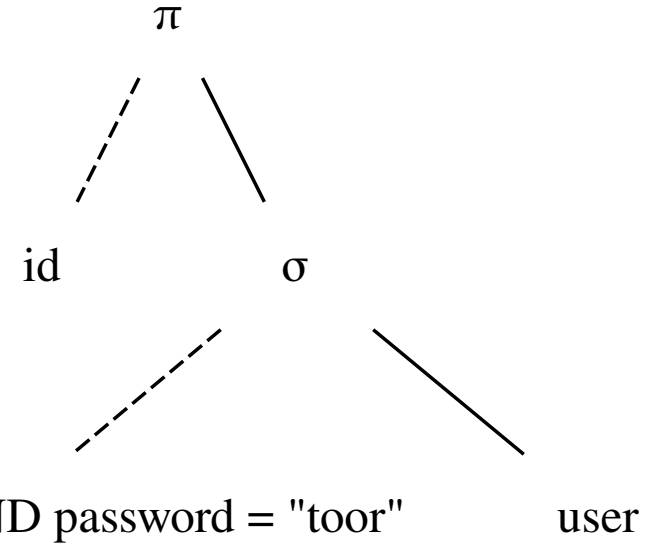
OK for the engine (to optimize)



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OK in papers (for compactness)
OK for the engine (to optimize)



But programmers
prefer to express
themselves this way

```
SELECT id  
FROM user  
WHERE username = 'root' AND password = 'toor'
```

ASP for Papers vs ASP for Programmers

$\{assign(X, C) : color(C)\} = 1 \leftarrow node(X).$

$\perp \leftarrow edge(X, Y), assign(X, C), assign(Y, C).$

$[1@1, C] \leftarrow assign(_, C).$

ASP for Papers vs ASP for Programmers

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```
% guess one color for each node
```

```
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```
% adjacent nodes must have different colors
```

```
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```
% minimize the number of used colors
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But... they are essentially
the same!

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The SAME... but for Different Purposes!

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**But a long-standing codebase needs
maintenance, readability, and robustness**

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But a long-standing codebase needs
maintenance, readability, and robustness

Three downsides of long-standing ASP codebases

- 1) Propagating Changes
- 2) Use of Object Variables
- 3) Lack of Semantic Annotations

ASP Example

```
%      cab( cab_id, driver)
% customer(cust_id,  name, title)
%  assign(cust_id, cab_id)

% assign one cab to every customer
{assign(C,C') : cab(C',D)} = 1 :- customer(C,N,T).

% don't assign more than one customer to each cab
:- cab(C,D), #count{C' : assign(C',C)} > 1.
```

ASP Example

If driver is moved in a different predicate,
both rules HAVE TO be changed!

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but the driver affect the rules!

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The rules don't use the driver...
but the driver affect the rules!

Object variables are often shorts and used inconsistently (e.g., C and C')

BTW... is assign(cust, cab) or assign(cab, cust)?
There is no semantic annotation within the syntax!

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Like SQL... but mapping to ASP!

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Abstraction and Simplification

Problems are expressed in a language
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No order of attributes,
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References to records are tracked to ease debugging, and to avoid comparing apples and oranges

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SDL maps to semantically annotated atoms

```
assign(  
  customer(CUST_ID, NAME, TITLE),  
  cab(CAB_ID, DRIVER)  
)  
vs  
assign(C,C')
```

SDL Example

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% cab( cab_id, driver)
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```
record Cab:          id: int, driver: str;
record Customer:   id: int, name: str, title: str;
record Assign:     customer: Customer, cab: Cab;

guess from Customer exactly 1
  Assign from Cab
    where Assign.customer == Customer and Assign.cab == Cab;
deny from Cab having
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SDL Example

A Cab has attributes
id (an integer) and
driver (a string)

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Just two little observations...

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record Cab:      id: int, driver: str;  
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record Cab:      id: int, driver: str;  
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record Assign:   customer: Customer, cab: Cab;  
guess from Customer exactly 1 —————  
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    where Assign.customer == Customer and Assign.cab == Cab;  
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Doesn't use the driver... and
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Attributes accessed by name, not by position!

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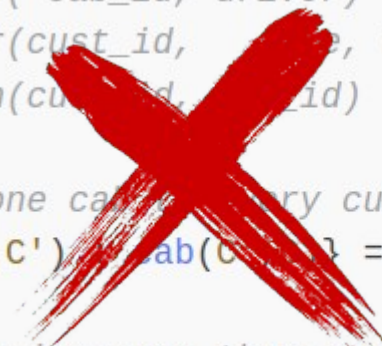
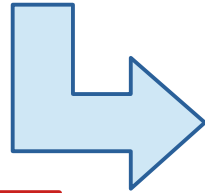
```

% cab( cab_id, driver)
% customer(cust_id, name, title)
% assign(customer_id, cab_id)

% assign one cab to every customer
{assign(C,C') & cab(C,D)} = 1 :- customer(C,N,T).

% don't assign more than one customer to each cab
:- cab(C,D), #count{C' : assign(C',C)} > 1.

```

```

{assign(
  customer(ID,NAME,TITLE),
  cab(ID,DRIVER)
) : cab(ID,DRIVER)} = 1 :-
  customer(ID,NAME,TITLE).

:- cab(ID,DRIVER), #count{
  customer(ID,NAME,TITLE) :
  assign(
    customer(ID,NAME,TITLE),
    cab(ID,DRIVER)
  )
} > 1.

```

Structure Instruction

```
record RecordName: Attributes;
```

List of **name: type** pairs

int, str or a record name

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Acyclicity of structure instructions is
required and checked

Query Instruction

```
show RecordNames;
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List of record names

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Query Instruction

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show RecordNames;
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List of record names

We may include more expressive query instructions in the future, but for now... KISS!

By default, we only show SAT/UNSAT...

to avoid any ambiguity

```
malvi@pandora:~ [ven giu 07 15:01]
$ echo "a(foo). b(bar). #show a/1. " | clingo --outf=1 -V0
ANSWER
a(foo).
malvi@pandora:~ [ven giu 07 15:01]
$ echo "a(foo). b(bar). #show a(X) : a(X)." | clingo --outf=1 -V0
ANSWER
a(foo). b(bar). a(foo).
```

Model Instructions (several)

Refer records by defining aliases (optional);
prepend **not** to refer the complement

RecordName **as** *alias*

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Attributes are accessed with dot operations,
as in OOP languages

```
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alias.attribute_name.sub_attribute
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alias.attribute_name  
alias.attribute_name.sub_attribute
```

Values (of attributes, or constants)
can be combined in expressions

Model Instructions: Definitions

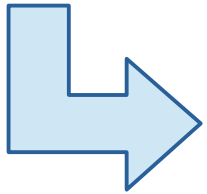
```
record Node: id: int;  
record Edge: first: Node, second: Node;
```

```
define Edge as self from Edge as other  
where self.first == other.second and self.second == other.first;
```

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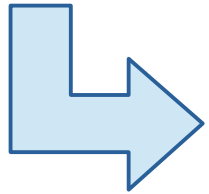


```
edge(node(SelfFirst), node(SelfSecond)) :-  
  edge(node(OtherFirst), node(OtherSecond)),  
  node(SelfFirst) == node(OtherSecond),  
  node(SelfSecond) == node(OtherFirst).
```

Model Instructions: Definitions

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record Node: id: int;  
record Edge: first: Node, second: Node;
```

```
define Edge as self from Edge as other  
where self.first == other.second and self.second == other.first;
```



```
edge(node(SelfFirst), node(SelfSecond)) :-  
  edge(node(OtherFirst), node(OtherSecond)),  
  node(SelfFirst) == node(OtherSecond),  
  node(SelfSecond) == node(OtherFirst).
```

Essentially, the same of

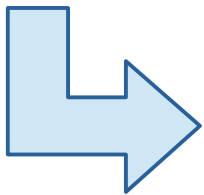
```
edge(OtherSecond, OtherFirst) :-  
  edge(OtherFirst, OtherSecond).
```

```
record Node: id: int;  
record Edge: first: Node, second: Node;  
record In: node: Node;  
record Size: value: int;
```

```
define Size having count {In.node from In} == Size.value;
```

```
record Node: id: int;  
record Edge: first: Node, second: Node;  
record In: node: Node;  
record Size: value: int;
```

```
define Size having count {In.node from In} == Size.value;
```



```
size(Value) :- #count{node(Id) : in(node(Id))} = Value.
```

Model Instructions: Guesses

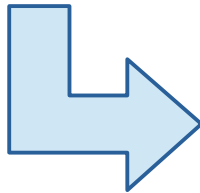
```
record Node: id: int;  
record Edge: first: Node, second: Node;  
record In: node: Node;  
record Size: value: int;
```

```
guess from Node at most 1  
  In where Node == In.node;
```

Model Instructions: Guesses

```
record Node: id: int;  
record Edge: first: Node, second: Node;  
record In: node: Node;  
record Size: value: int;
```

```
guess from Node at most 1  
In where Node == In.node;
```



```
0 <= {  
  in(node(Id')) :  
    node(Id) == node(Id')  
} <= 1 :- node(Id).
```

Model Instructions: Denies

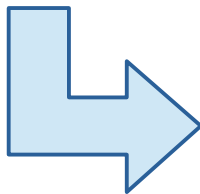
```
record Node: id: int;  
record Edge: first: Node, second: Node;  
record In: node: Node;  
record Size: value: int;
```

```
deny from In as in1, In as in2, not Edge  
where in1.node == Edge.first and  
        in2.node == Edge.second and  
        in1.node < in2.node;
```


Model Instructions: Denies

```
record Node: id: int;  
record Edge: first: Node, second: Node;  
record In: node: Node;  
record Size: value: int;
```

```
deny from In as in1, In as in2, not Edge  
where in1.node == Edge.first and  
      in2.node == Edge.second and  
      in1.node < in2.node;
```



```
:- in(node(Id)), in(node(Id')), not edge(node(F), node(S)),  
   node(Id) == node(F),  
   node(Id') == node(S),  
   node(Id) < node(Id').
```

Model Instructions: Denies with Penalty

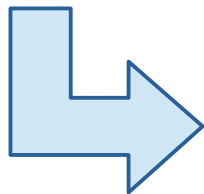
```
record Node: id: int;  
record Edge: first: Node, second: Node;  
record In: node: Node;  
record Size: value: int;
```

```
deny from Node, not In  
where In.node == Node  
or pay 1 at 1;
```

Model Instructions: Denies with Penalty

```
record Node: id: int;  
record Edge: first: Node, second: Node;  
record In: node: Node;  
record Size: value: int;
```

```
deny from Node, not In  
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```
:- node(Id), not in(node(Id')),  
node(Id') == node(Id).  
[1@1, Id, Id']
```

And if you need “assembly”...

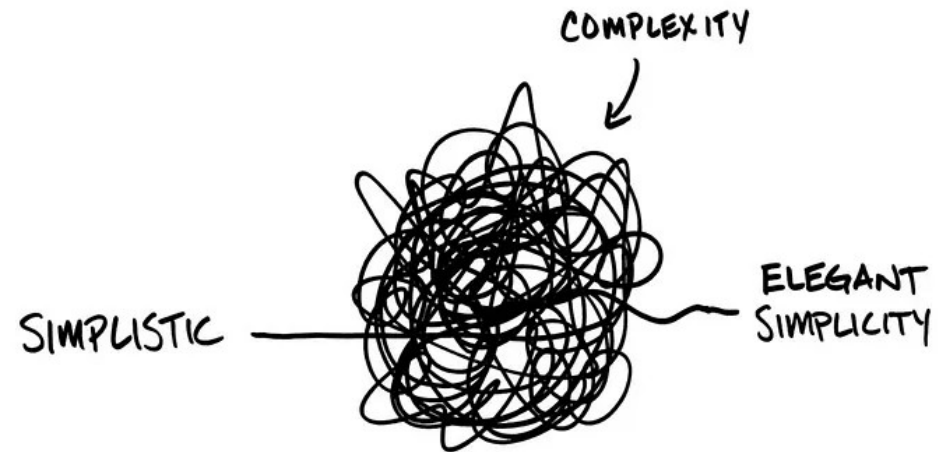
```
@asp_block $  
  to_be | not_to_be :-  
    plain, hold, asp, code.  
  
  possible.  
  :- suggested.  
$
```

We don't do any syntactic or semantic check in

@asp_block \$... \$

Summing UP

ASP syntax is not suitable for long-standing codebases
(it doesn't even look like a programming language)



BEHAVIOR GAP

Image Credits to Hill Investment Group

Summing UP

ASP syntax is not suitable for long-standing codebases
(it doesn't even look like a programming language)

SDL is a programming language of higher level,
enforcing a proper structure of records, and
introducing dependencies only if explicitly needed



BEHAVIOR GAP

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Summing UP

ASP syntax is not suitable for long-standing codebases
(it doesn't even look like a programming language)

SDL is a programming language of higher level,
enforcing a proper structure of records, and
introducing dependencies only if explicitly needed

On the long run, programmers may code
in SDL without thinking at all to ASP
(as SQL programmers never rarely think
to relational algebra)



BEHAVIOR GAP

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Explainability Issues

<https://asp-chef.alviano.net/s/ucoexplain/>

Outline

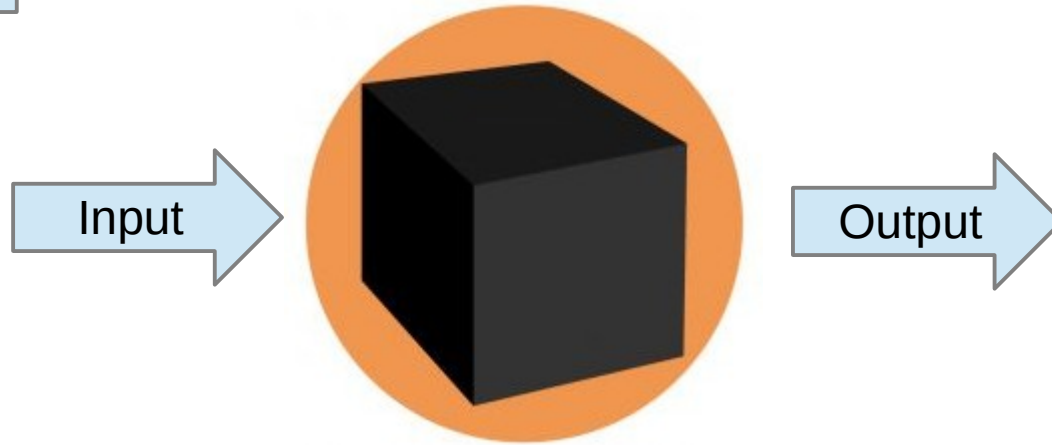
- Who has an issue with explainability
- What you are asked to believe in ASP
- How to ask for less faith
- ASP powered approach
- Interactive representation

Explainability: A ML issue?



**Black-box ML algorithms for AI:
The link between input and output is unclear
(i.e., it cannot be explained)**

Explainability: A ML issue?



**Black-box ML algorithms for AI:
The link between input and output is unclear
(i.e., it cannot be explained)**

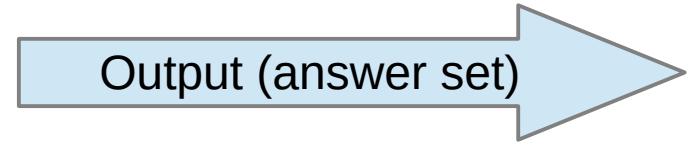
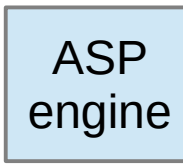
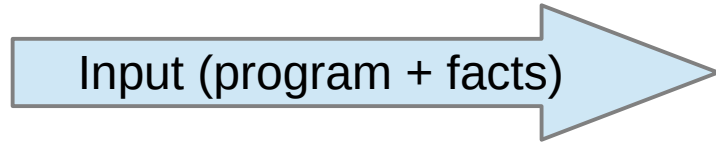


White-box ML algorithms for AI

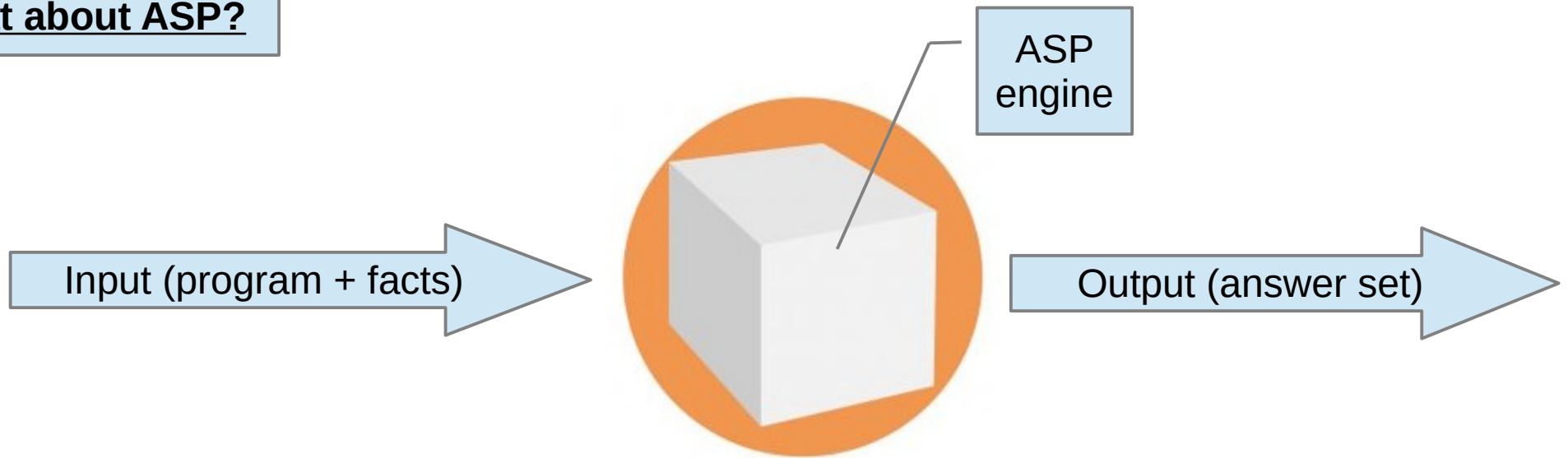


The link between input and output is understandable by domain experts
(e.g., a collection of features that influenced the result)

What about ASP?

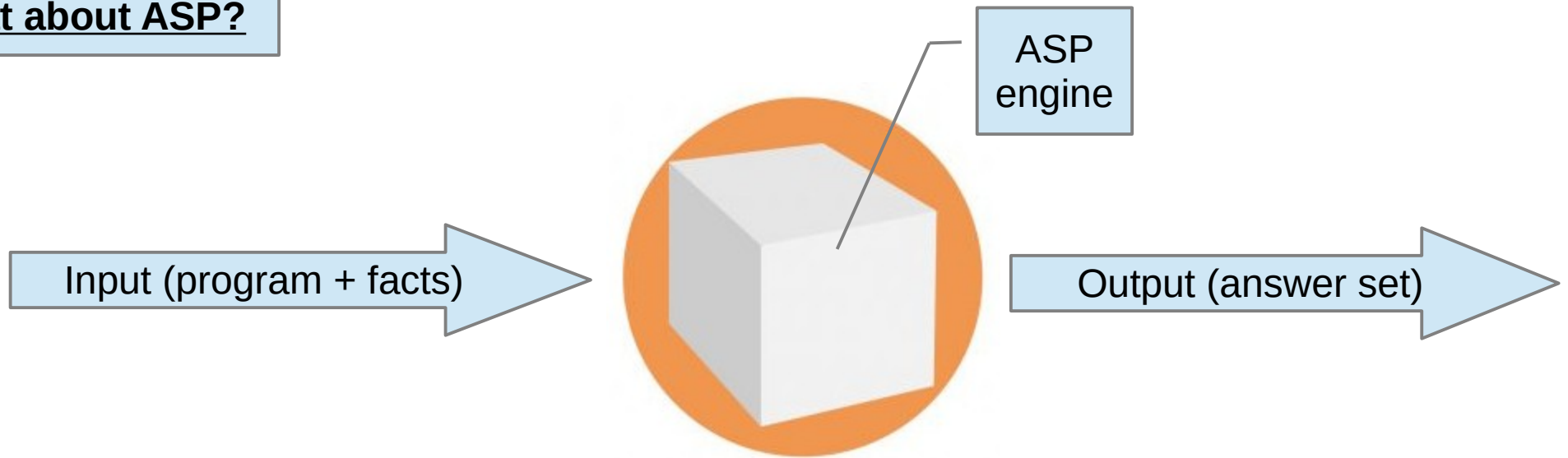


What about ASP?



**The algorithm is known, the semantics is intuitive.
The answer set can be checked in polynomial time
(in the normal case, wrt. the ground program).**

What about ASP?



The algorithm is known, the semantics is intuitive.
The answer set can be checked in polynomial time
(in the normal case, wrt. the ground program).

So... is everything perfect here?



Recipe



#1. Search Models

Height Decode ECHO USE CONSTRAINTS

```
1 node(X) :- link(X,Y).  
2 node(Y) :- link(X,Y).
```

of models RAISE ERROR

#2. Search Models



Height Decode ECHO USE CONSTRAINTS

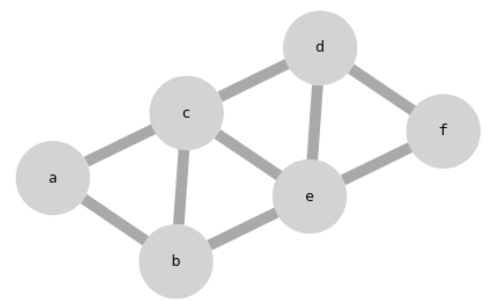
```
1 g(node(X), label(X)) :- node(X).  
2 g(link(X,Y)) :- link(X,Y).  
3 g(defaults, undirected).
```

of models RAISE ERROR

#3. Graph



Height Predicate ECHO



Input

```
1 link(a,b).  
2 link(a,c).  
3 link(b,c).  
4 link(b,e).  
5 link(c,e).  
6 link(c,d).  
7 link(d,e).  
8 link(d,f).  
9 link(e,f).
```

<https://asp-chef.alviano.net>

Recipe



Input

<https://asp-chef.alviano.net>

#1. Search Models

Height 60 Decode __base64__ ECHO USE CONSTRAINTS

```
1 node(X) :- link(X,Y).
2 node(Y) :- link(X,Y).
```

of models 1 RAISE ERROR

```
1 link(a,b).
2 link(a,c).
3 link(b,c).
4 link(b,e).
5 link(c,e).
6 link(c,d).
7 link(d,e).
8 link(d,f).
9 link(e,f).
```

#2. Search Models

Height 80 Decode __base64__ ECHO USE CONSTRAINTS

```
1 g(node(X), label(X)) :- node(X).
2 g(link(X,Y)) :- link(X,Y).
3 g(defaults, undirected).
```

of models 1 RAISE ERROR

#4. Search Models

Height 100 Decode __base64__ ECHO USE CONSTRAINTS

```
1 assign(X, red) :- node(X), not assign(X, green), not assign(X, blue).
2 assign(X, green) :- node(X), not assign(X, red), not assign(X, blue).
3 assign(X, blue) :- node(X), not assign(X, red), not assign(X, green).
4 :- link(X,Y), assign(X,C), assign(Y,C).
```

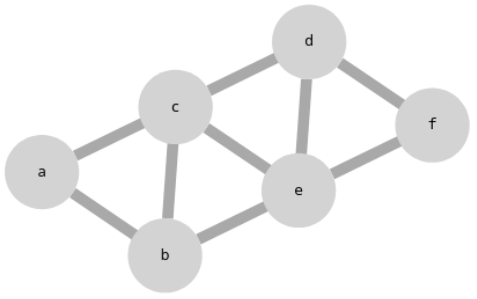
of models 1 RAISE ERROR

Input

```
1 link(a,b).
2 link(a,c).
3 link(b,c).
4 link(b,e).
5 link(c,e).
6 link(c,d).
7 link(d,e).
8 link(d,f).
9 link(e,f).
```

#3. Graph

Height 500 Predicate g



#5. Search Models

Height 80 Decode __base64__ ECHO USE CONSTRAINTS

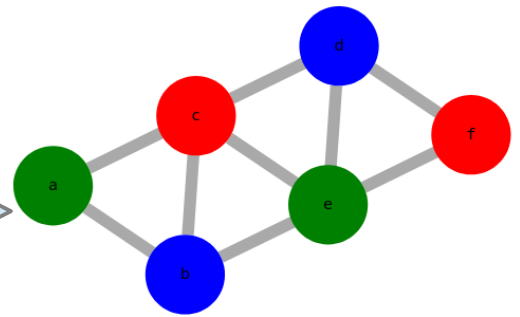
```
1 g(node(X), label(X), color(C)) :- node(X), assign(X,C).
2 g(link(X,Y)) :- link(X,Y).
3 g(defaults, undirected).
```

of models 1 RAISE ERROR

3-coloring of a graph obtained with ASP (Chef)

#6. Graph

Height 500 Predicate g ECHO



Output

```
1 assign(b,blue).
2 assign(c,red).
3 assign(e,green).
4 assign(d,blue).
5 assign(f,red).
6 assign(a,green).
```

Recipe



Input

<https://asp-chef.alviano.net>

```

1 link(a,b).
2 link(a,c).
3 link(b,c).
4 link(b,e).
5 link(c,e).
6 link(c,d).
7 link(d,e).
8 link(d,f).
9 link(e,f).

```

#1. Search Models

Height 60 Decode __base64__ ECHO USE CONSTRAINTS

```

1 node(X) :- link(X,Y).
2 node(Y) :- link(X,Y).

```

of models 1 RAISE ERROR

#2. Search Models

Height 80 Decode __base64__ ECHO USE CONSTRAINTS

```

1 g(node(X), label(X)) :- node(X).
2 g(link(X,Y)) :- link(X,Y).
3 g(defaults, undirected).

```

of models 1 RAISE ERROR

#4. Search Models

Height 100 Decode __base64__ ECHO USE CONSTRAINTS

```

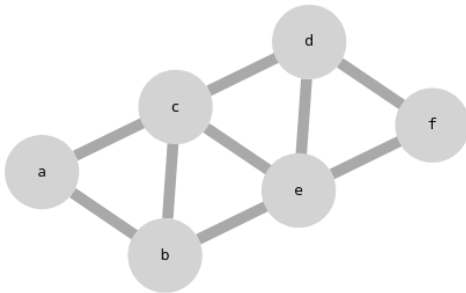
1 assign(X, red) :- node(X), not assign(X, green), not assign(X, blue).
2 assign(X, green) :- node(X), not assign(X, red), not assign(X, blue).
3 assign(X, blue) :- node(X), not assign(X, red), not assign(X, green).
4 :- link(X,Y), assign(X,C), assign(Y,C).

```

of models 1 RAISE ERROR

#3. Graph

Height 500 Predicate g



#5. Search Models

Height 80 Decode __base64__ ECHO USE CONSTRAINTS

```

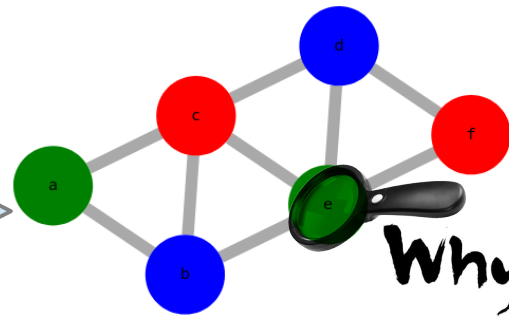
1 g(node(X), label(X), color(C)) :- node(X), assign(X,C).
2 g(link(X,Y)) :- link(X,Y).
3 g(defaults, undirected).

```

of models 1 RAISE ERROR

#6. Graph

Height 500 Predicate g ECHO



Input

```

1 link(a,b).
2 link(a,c).
3 link(b,c).
4 link(b,e).
5 link(c,e).
6 link(c,d).
7 link(d,e).
8 link(d,f).
9 link(e,f).

```

Output

```

1 assign(b,blue).
2 assign(c,red).
3 assign(e,green).
4 assign(d,blue).
5 assign(f,red).
6 assign(a,green).

```

3-coloring of a graph obtained with ASP (Chef)

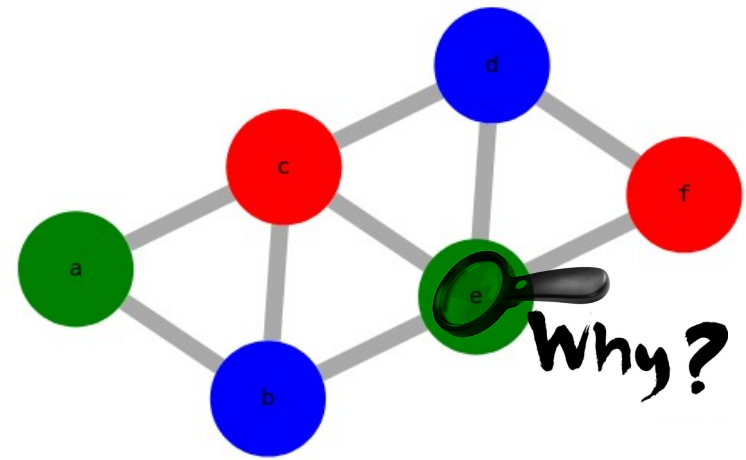
Why?

Input

```
1 link(a,b).
2 link(a,c).
3 link(b,c).
4 link(b,e).
5 link(c,e).
6 link(c,d).
7 link(d,e).
8 link(d,f).
9 link(e,f).
```

Output

```
1 assign(b,blue).
2 assign(c,red).
3 assign(e,green).
4 assign(d,blue).
5 assign(f,red).
6 assign(a,green).
```



```
10
11 node(X) :- link(X,Y).
12 node(Y) :- link(X,Y).
13
14 assign(X, red) :- node(X), not assign(X, green), not assign(X, blue).
15 assign(X, green) :- node(X), not assign(X, red), not assign(X, blue).
16 assign(X, blue) :- node(X), not assign(X, red), not assign(X, green).
17 :- link(X,Y), assign(X,C), assign(Y,C).
```

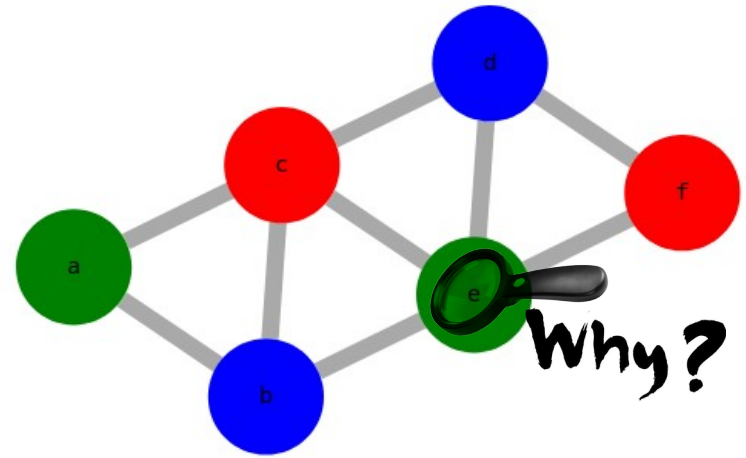
**An answer set is a minimal model of its program reduct (a Datalog program).
Datalog models can be computed by support inference.
The computation gives a DAG. Extract a proof tree for the query atom.**

The program reduct is obtained from the ground version of the program in input by fixing the interpretation of negative literals

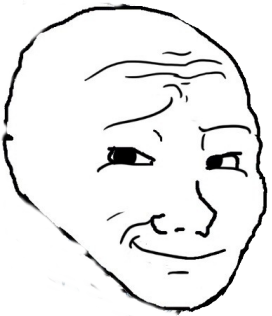


In the program reduct...

```
3 → link(e,f).  
2 → node(e) :- link(e,f).  
  
% assign(e, red) :- node(e), not assign(e, green), not assign(e, blue).  
assign(e, red) :- node(e), #false, #true.  
  
1 → % assign(e, green) :- node(e), not assign(e, red), not assign(e, blue).  
assign(e, green) :- node(e), #true, #true.  
  
% assign(e, blue) :- node(e), not assign(e, red), not assign(e, green).  
assign(e, blue) :- node(e), #true, #false.
```

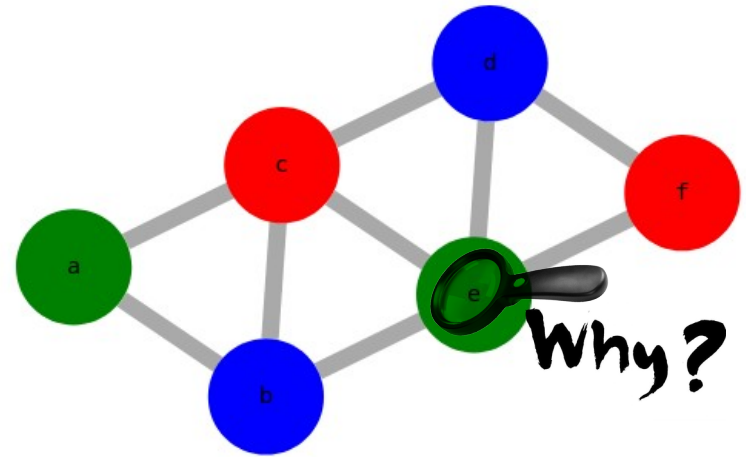


- 1) assign(e, green) is supported by node(e)
- 2) node(e) is supported by link(e,f)
- 3) link(e,f) is a fact



In the program reduct...

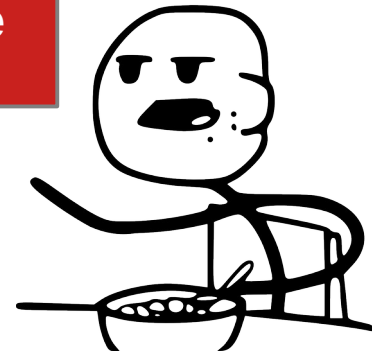
```
3 → link(e,f).  
2 → node(e) :- link(e,f).  
  
% assign(e, red) :- node(e), not assign(e, green), not assign(e, blue).  
assign(e, red) :- node(e), #false, #true.  
  
1 → % assign(e, green) :- node(e), not assign(e, red), not assign(e, blue).  
assign(e, green) :- node(e), #true, #true.  
  
% assign(e, blue) :- node(e), not assign(e, red), not assign(e, green).  
assign(e, blue) :- node(e), #true, #false.
```



- 1) assign(e, green) is supported by node(e)
- 2) node(e) is supported by link(e,f)
- 3) link(e,f) is a fact

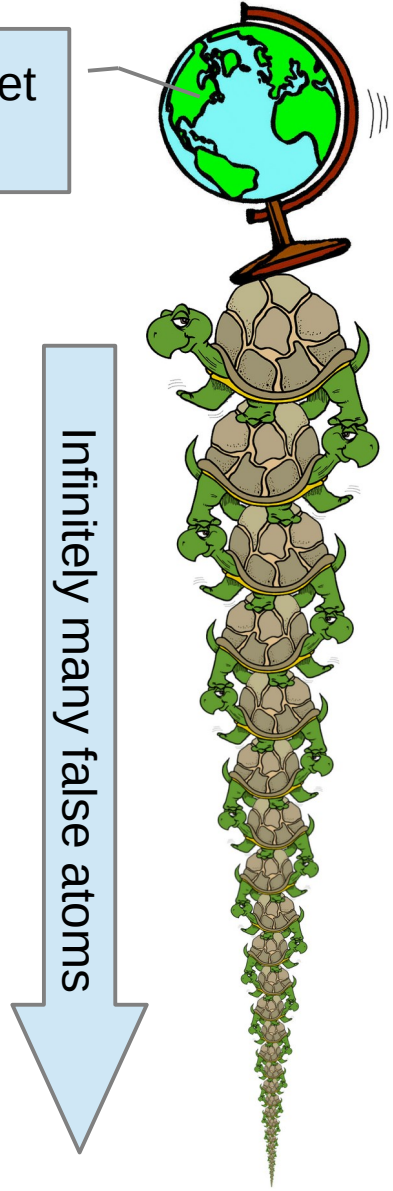
Why
#true, #true
?

Because ALL false atoms are assumed false...





The answer set
(true atoms)

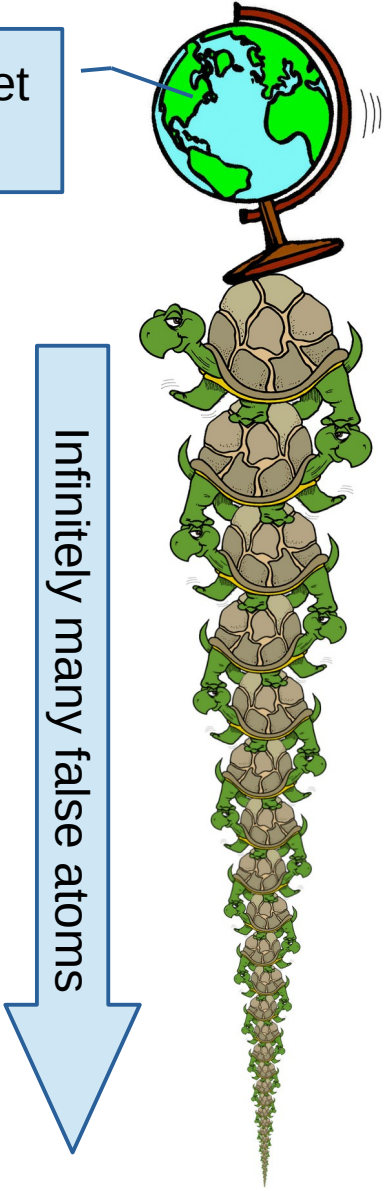




Fixing here means (blindly) trusting!

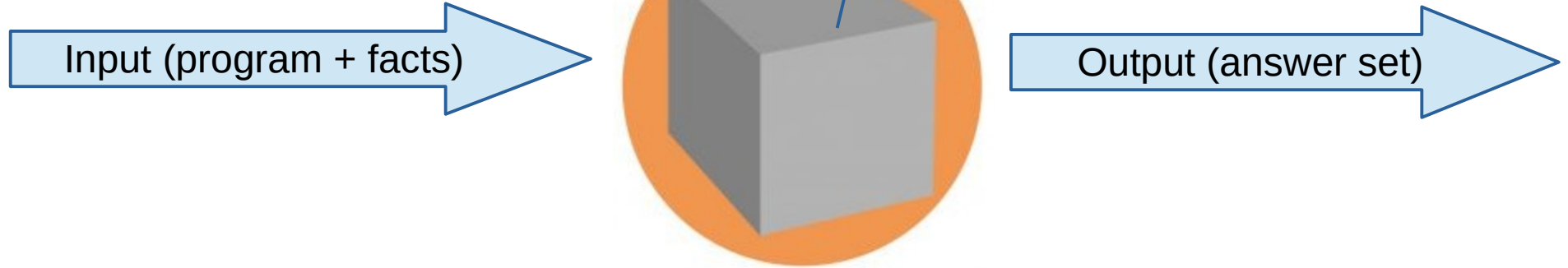
How big must our faith be?

The answer set
(true atoms)



The truth about ASP

ASP engine
(by itself)



Input (program + facts)

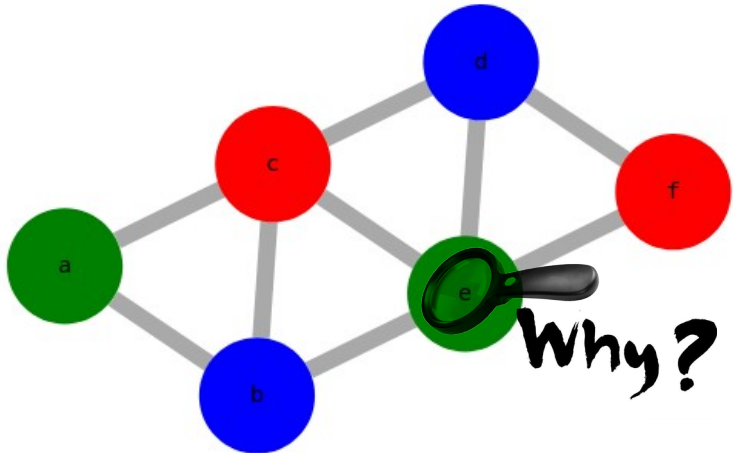
Output (answer set)

Let's say it is a gray-box!

**You can check the answer set in polynomial time,
but who want to do this?**

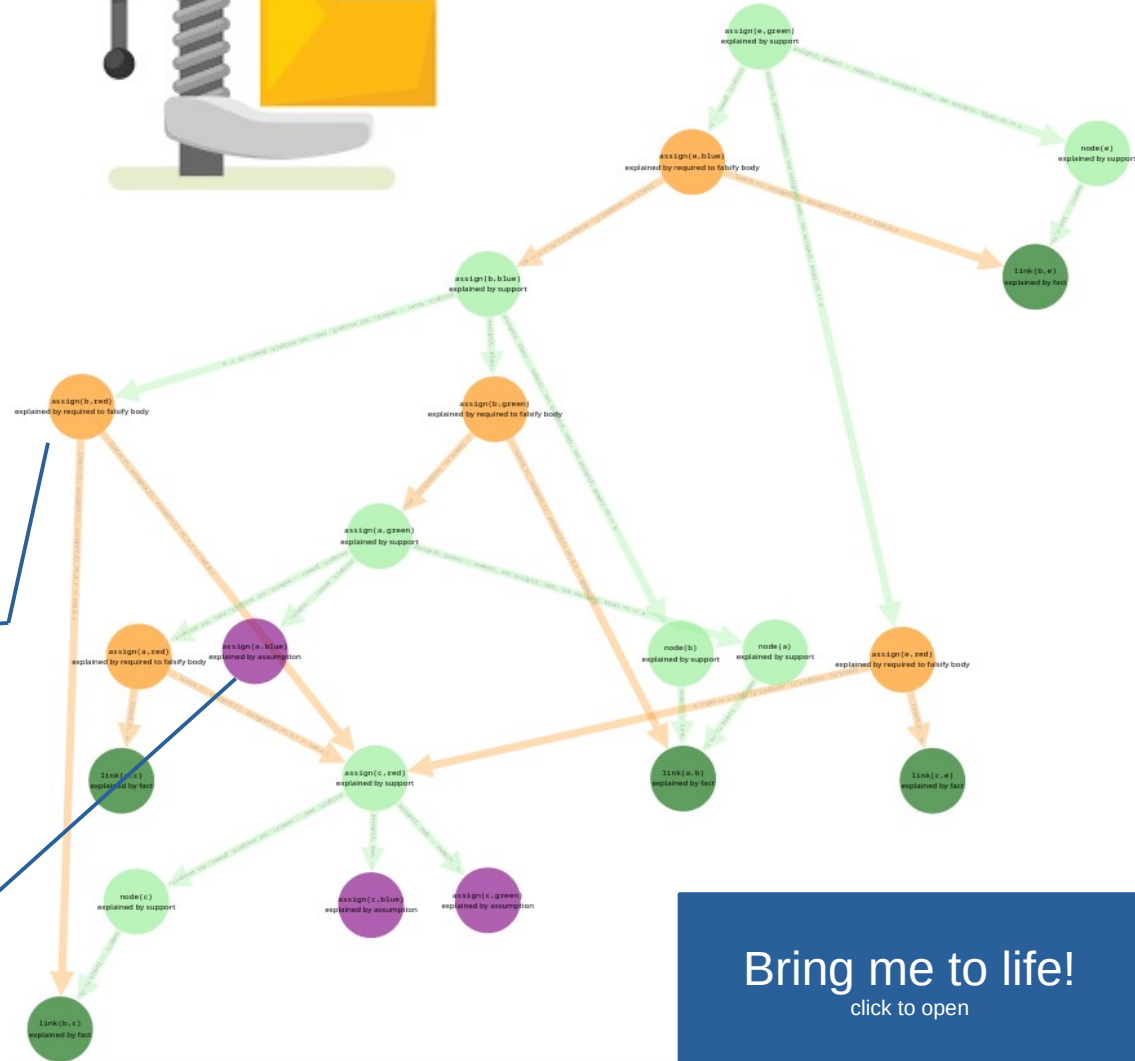
**Anyhow you must trust all false atoms,
a huge act of faith!**

Let's try to make the assumption smaller!



Rules with false heads (e.g., constraints) can infer falsity of some atoms!

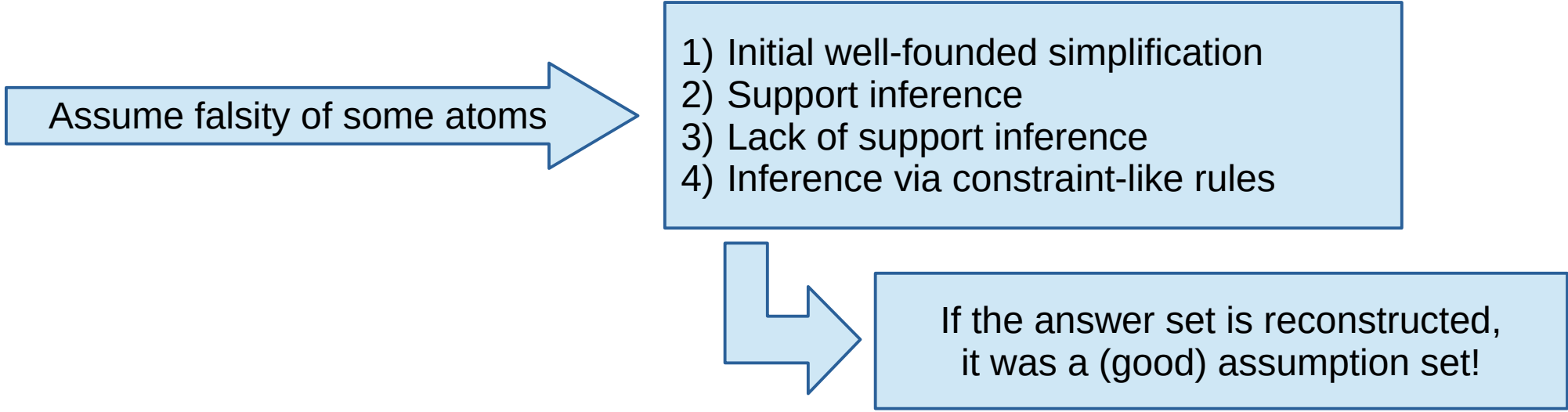
Some other atoms are simply assumed false



Bring me to life!
click to open

How? Step 1: Minimal Assumption Set!

Assume falsity of some atoms

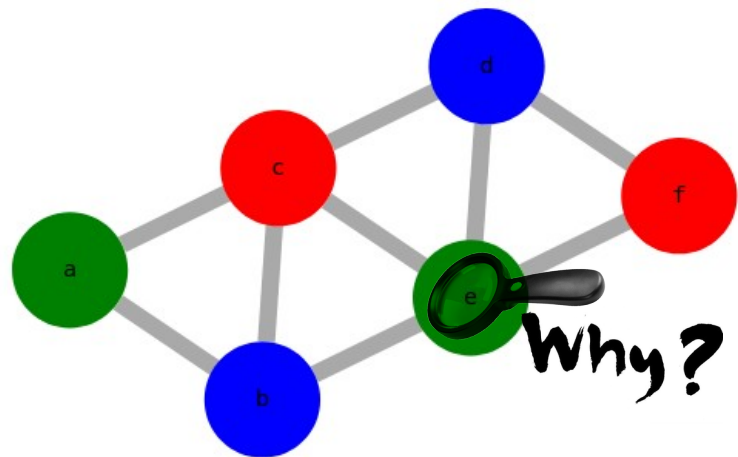


- 1) Initial well-founded simplification
- 2) Support inference
- 3) Lack of support inference
- 4) Inference via constraint-like rules

If the answer set is reconstructed,
it was a (good) assumption set!

Prefer subset-minimal assumption sets not containing the query atom

Example of Minimal Assumption Set



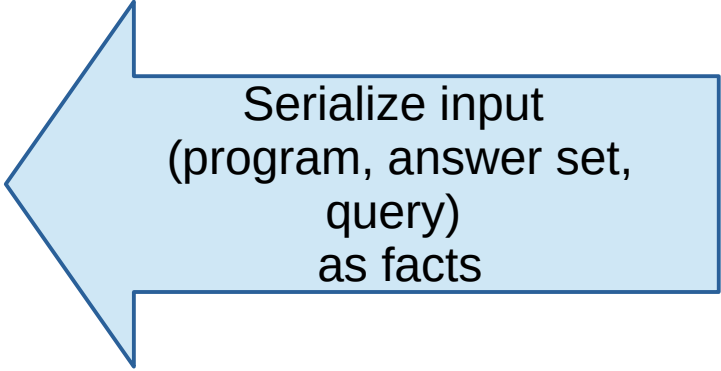
```
assign(a, blue)
assign(c, blue)
assign(c,
green)
```

VS



```
assign(b,blue).
assign(c,red).
assign(e,green).
assign(d,blue).
assign(f,red).
assign(a,green).
node(b).
node(c).
node(e).
node(d).
node(f).
node(a).
link(a,b).
link(a,c).
link(b,c).
link(b,e).
link(c,e).
link(c,d).
link(d,e).
link(d,f).
link(e,f).
```

MAS is a combinatorial optimization problem



Serialize input
(program, answer set,
query)
as facts

MAS is a combinatorial optimization problem

```
1 {assume_false(Atom)} :- false(Atom).
2 :- false(Atom), assume_false(Atom), not explain(Atom). [1@1, Atom]
3 :- false(Atom), assume_false(Atom), explain(Atom). [1@2, Atom]

4 has_explanation(Atom) :- explained_by(Atom,_).
5 :- atom(X), #count{Reason: explained_by(Atom,Reason)} != 1.

6 explained_by(Atom, assumption) :- assume_false(Atom).

7 {explained_by(Atom, (support, Rule))} :- head(Rule,Atom), true(Atom);
8   true(BAtom) : pos_body(Rule,BAtom);
9   has_explanation(BAtom) : pos_body(Rule,BAtom);
10  false(BAtom) : neg_body(Rule,BAtom);
11  has_explanation(BAtom) : neg_body(Rule,BAtom).

12 {explained_by(Atom, lack_of_support)} :- false(Atom);
13   false_body(Rule) : head(Rule,Atom).
14 false_body(Rule) :- rule(Rule);
15   pos_body(Rule,BAtom), false(BAtom), has_explanation(BAtom).
16 false_body(Rule) :- rule(Rule);
17   neg_body(Rule,BAtom), true(BAtom), has_explanation(BAtom).

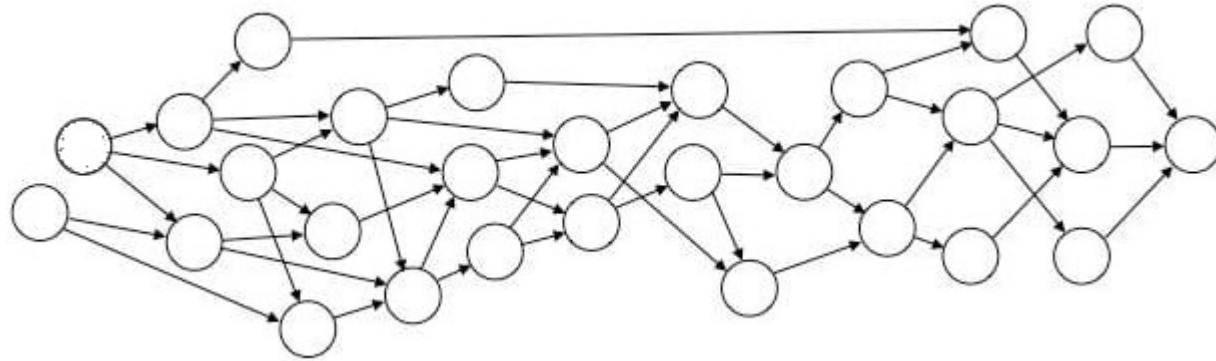
18 {explained_by(Atom, (required_to_falsify_body, Rule))} :- false(Atom);
19   pos_body(Rule,Atom), false_head(Rule);
20   true(BAtom) : pos_body(Rule,BAtom), BAtom != Atom;
21   has_explanation(BAtom) : pos_body(Rule,BAtom), BAtom != Atom;
22   false(BAtom) : neg_body(Rule,BAtom);
23   has_explanation(BAtom) : neg_body(Rule,BAtom).
24 false_head(Rule) :- rule(Rule); false(HAtom) : head(Rule,HAtom);
25   has_explanation(HAtom) : head(Rule,HAtom).
```

Serialize input
(program, answer set,
query)
as facts

Each optimal answer set is a MAS
(and comes with a derivation)



How? Step 2: Directed Acyclic Graph (DAG)



Program, answer set, and query serialized as facts...
derivation serialized (and indexed) as explained_by/3

The DAG can be materialized by a Datalog program

```
1 link(Atom, BAtom) :- explained_by(_, Atom, (support, Rule));
2   pos_body(Rule, BAtom).
3 link(Atom, BAtom) :- explained_by(_, Atom, (support, Rule));
4   neg_body(Rule, BAtom).

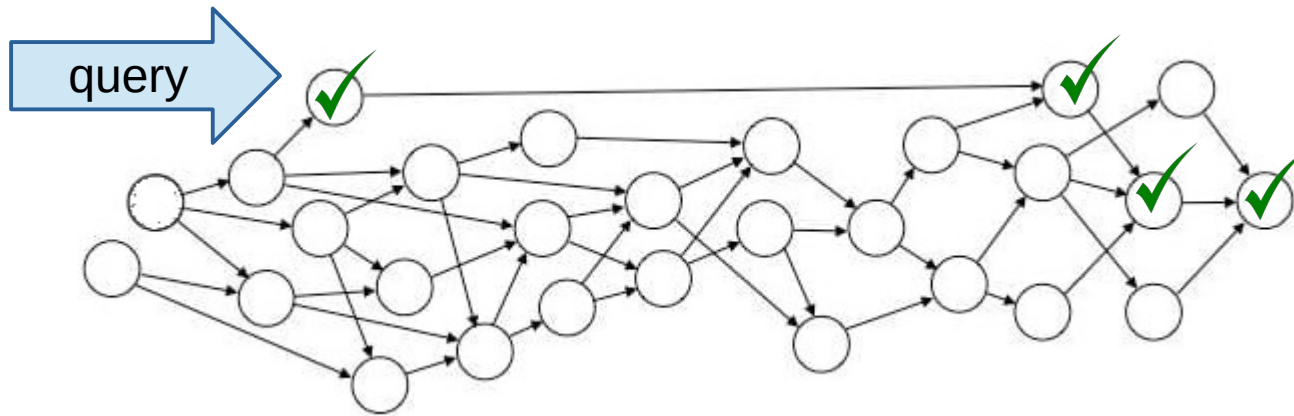
5 {link(Atom, A) : pos_body(Rule,A), false(A), explained_by(I,A,_), I < Index;
6 link(Atom, A) : neg_body(Rule,A), true (A), explained_by(I,A,_), I < Index}
   = 1 :- explained_by(_, Atom, lack_of_support); head(Rule, Atom).

7 link(Atom,A) :- explained_by(_, Atom, (required_to_falsify_body, Rule));
8   head(Rule,A).
9 link(At,A) :- explained_by(_,At,(required_to_falsify_body, Rule));
10  pos_body(Rule,A), A != At.
11 link(Atom,A) :- explained_by(_,Atom,(required_to_falsify_body, Rule));
12  neg_body(Rule,A).
```



The DAG can be materialized by a Datalog program

How? Step 3: Extract the relevant subDAG



Induced subgraph on nodes reachable from the query

Mix of ASP and igraph for better performances
(we don't really compute the full DAG)

How? Step 4: Interactive Representation

XASP NAVIGATOR

Search...

.*

Aa

↑

↓

↶

↷

🔍

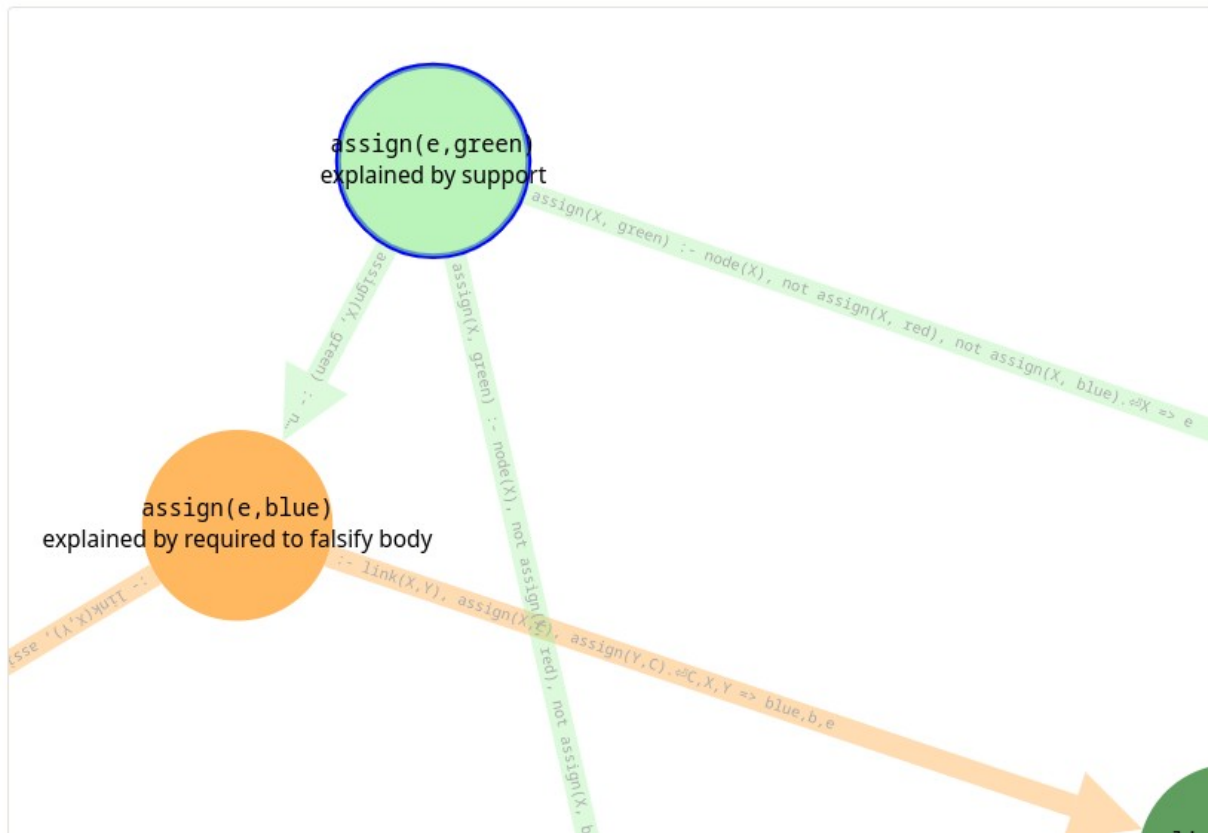
PANEL 37%

1.9x AXIS

⏪

⏩

☰



Selected node (press S to lock)

```
assign(e, green)
```

Explained by support

```
assign(X, green) :- node(X), not assign(X, red), not  
assign(X, blue).  
X => e
```

due to

```
assign(e, blue)  
assign(e, red)  
node(e)
```

Rules

```
node(Y) :- link(X, Y).  
X, Y => a, b
```

```
node(X) :- link(X, Y).  
X, Y => a, b
```

```
node(Y) :- link(X, Y).  
X, Y => b, c
```

```
node(Y) :- link(X, Y).  
X, Y => b, e
```

```
assign(X, red) :- node(X), not assign(X, green), not  
assign(X, blue)
```

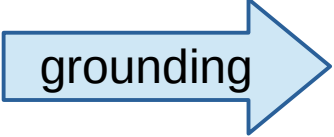
Everything is in the URL!

click to open

Assessment: Marcello had a program...

420 rules
651 facts

grounding



4261 ground rules
4468 ground atoms

Assessment: Marcello had a program...

420 rules
651 facts

grounding

4261 ground rules
4468 ground atoms

Number of answer sets
- expected: 1
- actual: 2



FFFFFFF
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Assessment: Marcello had a program...

420 rules
651 facts

grounding

4261 ground rules
4468 ground atoms

Number of answer sets

- expected: 1

- actual: 2

The two answer sets differ in one atom

- DAG computed in ~15 seconds
- 87 links, 45 internal nodes
- 20 leaves (1 assumption)
- 30 symbolic rules, 11 facts
- 48 ground rules, 65 ground atoms



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Bring me to life!

click to open

Summing up

- Shrink the set of atoms assumed false
 - as much as possible
 - answer set still reconstructible (with simple inferences)
 - computation powered by ASP
- Compute a DAG with ASP (and igragh)
 - open it in xASP Navigator
 - share it by sending a link
 - search in the graph (symbolic rules, substitutions, atoms)



ASP and LLMs interaction

Large Language Models

Computational models achieving **general-purpose language generation** and other natural language processing tasks.

Large Language Models

Computational models achieving **general-purpose language generation** and other natural language processing tasks.

We have General AI here!



Large Language Models

Computational models achieving **general-purpose language generation** and other natural language processing tasks.

We have General AI here!



False! It is good at NLP

Logical reasoning is still our subject

Answer Set Programming

KRR language, good for combinatorial search and optimization

KRR language, good for combinatorial search and optimization

```
% guess one color for each node  
{assign(X,C) : color(C)} = 1 :- node(X).  
  
% adjacent nodes must have different colors  
:- edge(X,Y), assign(X,C), assign(Y,C).  
  
% minimize the number of used colors  
:- assign(_,C). [1@1, C]
```

Answer Set Programming

KRR language, good for combinatorial search and optimization

```
% guess one color for each node  
{assign(X,C) : color(C)} = 1 :- node(X).  
  
% adjacent nodes must have different colors  
:- edge(X,Y), assign(X,C), assign(Y,C).  
  
% minimize the number of used colors  
:- assign(_,C). [1@1, C]
```

Possibly (read it certainly)
there is a (huge) barrier to entry

Computer Programming



Computer Programming



People chatting and telling around they are programming

ChatGPT

Input

Andrew is free from 11 am to 3 pm, Joanne is free from noon to 2 pm and then 3:30 pm to 5 pm. Hannah is available at noon for half an hour, and then 4 pm to 6 pm. What are some options for start times for a 30 minute meeting for Andrew, Hannah, and Joanne?

Output

Here is a list of their availability;

- Andrew: 11 am to 3 pm
- Joanne: noon to 2 pm, and 3:30 pm to 5 pm
- Hannah: noon to 12:30 pm, and 4 pm to 6 pm

Based on their availability, there is a 30-minute window where all three of them are available, which is from 4 pm to 4:30 pm. So, the meeting can be scheduled at 4 pm.

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As we are not a church... **let's take the best of both worlds!**

Idea Overview

Start with natural language

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Obtain a relational
representation (facts)

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Combine facts with a KB
(an ASP program)

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Map the answer set to
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Map the answer set to
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Users will not see ASP
(not that before we were showing them ASP)

LLMs themselves

A simple PROMPT asking for facts

ChatGPT ▾



Encode as Datalog facts the following sentences:
Tonight I want to go to eat some pizza with Marco and Alessio. Marco really like the pizza with onions as toppings.



Sure! Here are the sentences encoded as Datalog facts:

datalog

Copia codice

```
% Facts about the desire to go eat pizza  
wants_to_eat_pizza(tonight, I).
```

```
% Facts about the people involved  
going_with(I, Marco, tonight).  
going_with(I, Alessio, tonight).
```

```
% Facts about pizza preferences  
likes_pizza_with_topping(Marco, onions).
```

Explanation of the facts:

- `wants_to_eat_pizza(tonight, I).` expresses that "I" want to eat pizza tonight.
- `going_with(I, Marco, tonight).` and `going_with(I, Alessio, tonight).` express that "I" will go with Marco and Alessio tonight.
- `likes_pizza_with_topping(Marco, onions).` expresses that Marco likes pizza with onions as toppings.



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We need **more control** on these facts!

ChatGPT



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LLMs themselves

A simple PROMPT asking for facts

We need **more control** on these facts!

And we don't need this garbage

ChatGPT



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Prompt Engineering

Chat with the LLM...
give instructions
on the format and
task you want to achieve

ChatGPT ▾



You are a Natural Language to Datalog translator. To translate your input to Datalog, you will be asked a sequence of questions. The answers are inside the user input provided with [USER_INPUT]input[/USER_INPUT] and the format is provided with [ANSWER_FORMAT]predicate(terms).[/ANSWER_FORMAT]. Predicate is a lowercase string (possibly including underscores). Terms is a comma-separated list of either double quoted strings or integers. Be sure to control the number of terms in each answer! An answer MUST NOT be answered if it is not present in the user input. Remember these instructions and don't say anything!



Understood. Please provide the input for translation.

Prompt Engineering

Chat with the LLM...
give instructions
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We can start with general
instructions for our task

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Understood. Please provide the input for translation.



Here is some context that you MUST analyze and remember.

The user provides a request to obtain catering suggestions. The user can mention a day, other persons, and their cuisine preferences.

Remember this context and don't say anything!

Something specific of the application we are working on



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Understood. Please provide the user input for translation.

And we can start extracting facts!



[USER_INPUT]Tonight I want to go to eat some pizza with Marco and Alessio. Marco really like the pizza with onions as toppings.

[/USER_INPUT]

List all the persons mentioned including me if indirectly included.

[ANSWER_FORMAT]person("who").

[/ANSWER_FORMAT]



person("I").
person("Marco").
person("Alessio").



[USER_INPUT]Tonight I want to go to eat some pizza with Marco and Alessio. Marco really like the pizza with onions as toppings.

[/USER_INPUT]

For each person, list any restaurant preferences.

[ANSWER_FORMAT]cuisine_preferences("who", "country").[/ANSWER_FORMAT]



cuisine_preferences("Marco", "pizza").

cuisine_preferences("Alessio", "pizza").

cuisine_preferences("I", "pizza").



[USER_INPUT]Tonight I want to go to eat some pizza with Marco and Alessio. Marco really like the pizza with onions as toppings.

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cuisine_preferences("Marco", "pizza").

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cuisine_preferences("I", "pizza").

[USER_INPUT]Tonight I want to go to eat some pizza with Marco and Alessio. Marco really like the pizza with onions as toppings.

[/USER_INPUT]

For each person, list what they want to eat.

[ANSWER_FORMAT]want_food("who", "what").

[/ANSWER_FORMAT]



want_food("I", "pizza").

want_food("Marco", "pizza with onions").

want_food("Alessio", "pizza").



Reasoning with ASP

```
person("I").  
person("Marco").  
person("Alessio").  
  
cuisine_preferences("Marco", "pizza").  
cuisine_preferences("Alessio", "pizza").  
cuisine_preferences("I", "pizza").  
  
want_food("I", "pizza").  
want_food("Marco", "pizza with onions").  
want_food("Alessio", "pizza").
```

Extracted facts

Reasoning with ASP

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person("Alessio").

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want_food("Marco", "pizza with onions").
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```

Extracted facts

Combine facts
with a KB

```
can_go_together(X,Y,Z) :-
    person(X), person(Y), X < Y,
    want_food(X,Z), want_food(Y,Z).

can_go_together(X,Y,Z) :-
    person(X), person(Y), X < Y,
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#show can_go_together/3.
```


Reasoning with ASP

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person("I").
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#show can_go_together/3.
```

Obtain an
answer set

```
can_go_together("Alessio", "Marco", "pizza").
can_go_together("I", "Marco", "pizza").
can_go_together("Alessio", "I", "pizza").
```

Back to Natural Language

You are now a Datalog to Natural Language translator.

You will be given relational facts and mapping instructions.

Relational facts are given in the form
[FACTS]atoms[/FACTS].

Remember these instructions and don't say anything!

General instructions
on the task



Understood. Please provide the relational facts and mapping instructions.

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Something specific of the
application we are working
on

Here is some context that you **MUST** analyze and remember.
Explain the facts in a human readable way, as a paragraph.
Remember this context and don't say anything!



Understood. Please provide the relational facts for translation.

Provide facts and mapping instructions

```
[FACTS]can_go_together("me", "marco", "pizza").  
can_go_together("me", "alessio", "pizza").  
can_go_together("marco", "alessio", "pizza"). [/FACTS]  
Each fact matching can_go_together("person 1",  
"person 2", "cuisine preference") must be  
interpreted as follows: Say that "person 1" can go  
with "person 2" to eat "cuisine preference".
```

Collect answers



I can go with Marco to eat pizza. I can go with Alessio to eat pizza. Marco can go with Alessio to eat pizza.

Ask to summarize the collected responses



Summarize the following responses:
I can go with Marco to eat pizza. I can go with Alessio to eat pizza. Marco can go with Alessio to eat pizza.



The responses indicate that the speaker (presumably "me") can accompany Marco and Alessio to eat pizza, and Marco can also accompany Alessio to eat pizza.

This is the output shown to the user

Let's Automate IT!

YAML (yet another markup language)

Like JSON... but more readable

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Like JSON... but more readable

```
name: Lorenzo
degrees:
- Bachelor
short bio: |
  I'm Lorenzo...
  I'm a student at UNICAL...
```

Here is an object with properties
name, **degrees** and **short bio**

degrees is a list (of values)

short bio is a string (block notation)

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YAML (yet another markup language)

Like JSON... but more readable

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YAML is used in many applications,
among them Kubernetes

We use YAML to define an application

preprocessing:

- _: The user provides a request to obtain catering suggestions.
The user can mention a day, other persons, and their cuisine preferences.
- person("who"): List all the persons mentioned including me if indirectly included.
- cuisine_preferences("who", "country"): For each person, list any restaurant preferences.
- want_food("who", "what"): For each person, list what they want to eat.

knowledge base: |

```
can_go_together(X,Y,Z) :-  
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```

```
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```

postprocessing:

- _: Explain the facts in a human readable way, as a paragraph.
- can_go_together("person 1", "person 2", "cuisine preference"): |
Say that "person 1" can go with "person 2" to eat "cuisine preference".

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```
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```

postprocessing:

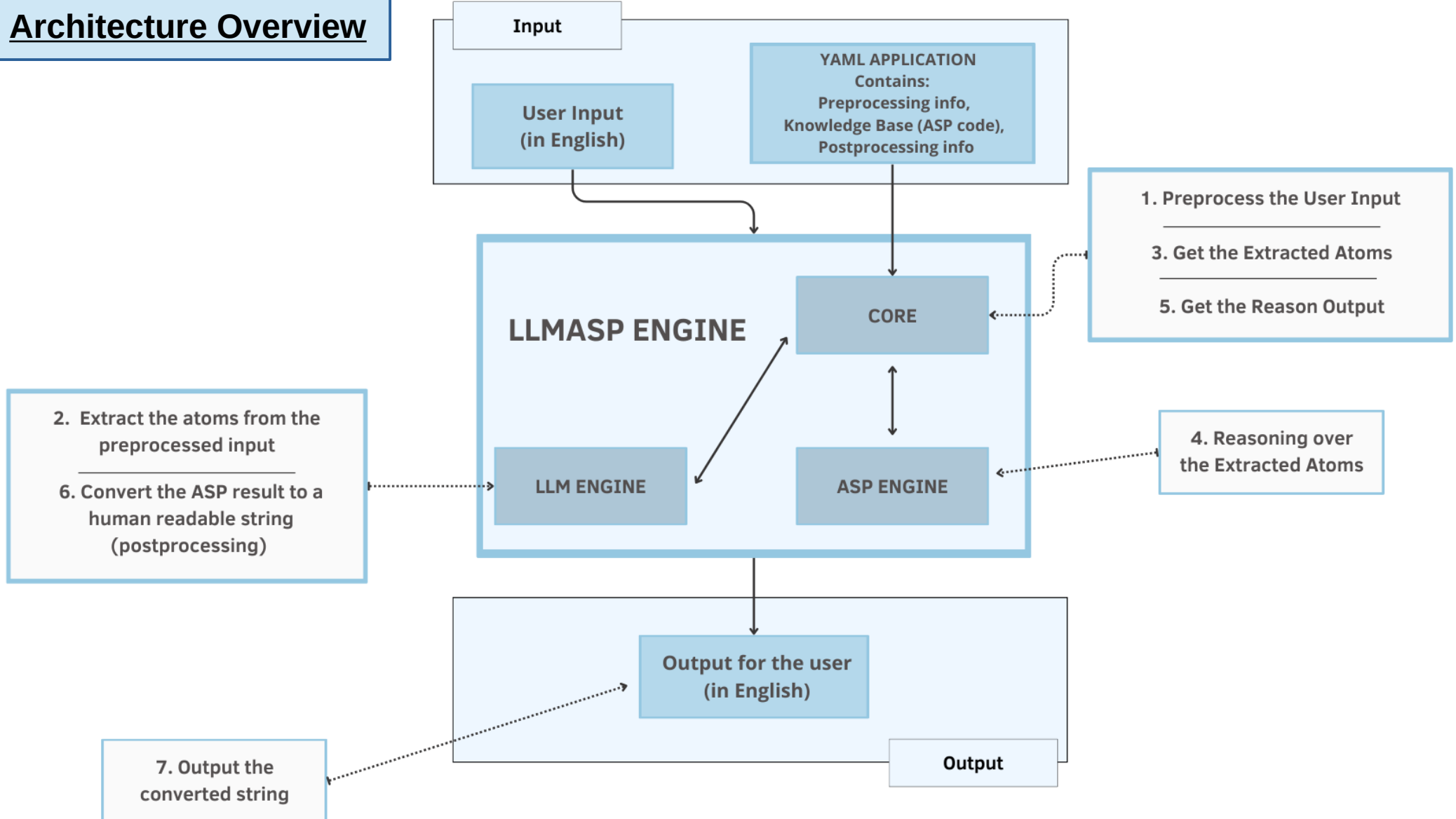
- _: Explain the facts in a human readable way, as a paragraph.
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Say that "person 1" can go with "person 2" to eat "cuisine preference".

Preprocessing includes the context and the instructions on the atoms to extract from the input

KB is simply an ASP program

Postprocessing includes the context and the mapping instructions

Architecture Overview



Summing UP

LLMASP combines
NLP capabilities of LLMs
with the deep reasoning of ASP

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We automated the interaction
with LLM and ASP engines

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Still a lot to do!

Why those prompts and not something different?

We have to build a dataset.

We have to try different prompts and measure their quality.

We can start with the preprocessing and then evaluate the postprocessing by extracting facts from the produced output.

Questions

