

Specifying Weak Constraints in Processable English

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Overview

- Processable English (PENGASP)
- PENGASP System and Architecture
- A Simple Specification in PENGASP
- Strong and Weak Constraints in Answer Set Programming
- Specifying Weak Constraints in PENGASP

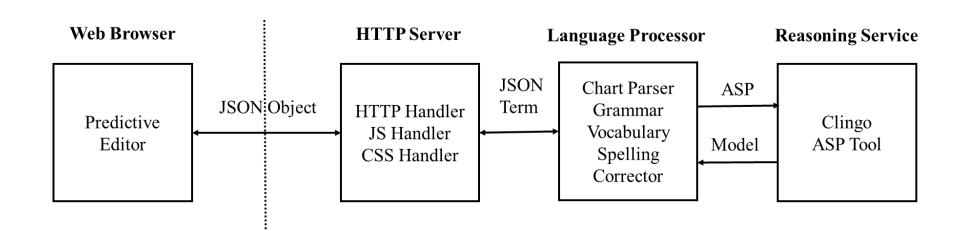
Processable English (PENGASP)

- Processable English (PENG^{ASP}) is a controlled natural language that serves as a high-level specification language for Answer Set Programs (ASP).
- ASP is a form of declarative programming.
- ASP has its roots in: knowledge representation, logic programming, deductive databases, and constraint solving.
- The language processor of the PENG^{ASP} system translates a specification in controlled natural language into an executable ASP program and vice versa.

The PENGASP System

- A predicative text editor is used to guide the writing of a specification in controlled natural language.
- The specification is parsed incrementally during the writing process with the help of a chart parser.
- All natural language processing tasks occur in <u>parallel</u>:
 - anaphoric expressions are resolved,
 - -look-ahead information is generated, and
 - an executable ASP is generated.
- New content words can be defined on the fly.

PENG^{ASP} Architecture



PENG^{ASP} at Work

PENG Editor × +			-		×
← → C (i) 130.56.243.30:8085/peng/			☆ 🤤) R	:
PENG^{ASP} File - Options - Help - Contact -					
CNL Input					
Enter sentence here		Submit			
Processed Input:		,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,			
Lookahead Information ▼ Anaphoric Expressions ▼					
CNL Text					
 Every student who works is successful. 					
 Every student who studies at Macquarie University works o It is not the encoded to the encoded who is encoded in Information 	-				
 It is not the case that a student who is enrolled in Informatic Tom is a student. 	r rechnology parties.				
5. Tom studies at Macquarie University and is enrolled in Infor	nation Technology.				
6. Bob is a student.					
7. Bob studies at Macquarie University and does not work.					
8. Who is successful?					

Answer Set Program

C Answer Set Program

```
Background Theory: OFF
```

```
prop(A,successful) :- class(A,student), pred(A,work).
pred(B,work) ; pred(B,party) :- class(B,student), pred(B,C,study_at), named(C,macquarie_university).
:- class(D,student), prop(D,E,enrolled_in), named(E,information_technology), pred(D,party).
named(1,tom).
class(1,student).
pred(1,2,study_at).
named(2,macquarie_university).
prop(1,3,enrolled_in).
named(3,information_technology).
named(4,bob).
class(4,student).
pred(4,2,study_at).
-pred(4,work).
answer(named(F,G)) :- named(F,G), prop(F,successful).
```

Answer Sets

C Answer Sets

```
clingo version 5.3.0
Reading from asp.lp
Solving...
Answer: 1
-pred(4,work) class(1,student) class(4,student) named(1,tom) named(2,macquarie_university)
named(3,information_technology) named(4,bob) pred(1,2,study_at) pred(4,2,study_at) prop(1,
3,enrolled_in) pred(1,work) pred(4,party) prop(1,successful) answer(named(1,tom))
SATISFIABLE
Models
             : 1
Calls
             : 1
             : 0.001s (Solving: 0.00s 1st Model: 0.00s Unsat: 0.00s)
Time
CPU Time
             : 0.000s
```

Strong and Weak Constraints

- ASP supports strong constraints and weak constraints.
- Strong constraints weed out answer sets.
- Weak constraints should be satisfied if possible, they rank answer sets.
- Strong constraint:
 - :- Literal.
- Weak constraint:
 - :~ Literal. [Weight@Level, Term]

Weak Constraints

- Weak constraints can be <u>weighted</u> according to their importance.
- In the presence of weights, best answer sets minimize the sum of the weights of the violated weak constraints.
- Weak constraints can also be prioritized.
- The violation of the constraints of the highest priority level is minimised first, followed by the lower priority levels in descending order.

Weak Constraints

• The weak constraint:

:~ pred(I, N, W). [W@L, I]

can be expressed alternatively as a minimize statement:
#minimize { W@L, I : pred(I, N, W) }

• Minimize statements can be interpreted as maximize statements with inverse weights:

#maximize { -W@L, I : pred(I, N, W) }.

Scenario in Full Natural Language

We want to choose one among five available accommodations. These accommodations are identified via their names (Amora, Grace, Metro, Posh and Adina), and each accommodation owns a certain number of stars and costs a certain amount of money.

Of course, the more stars an accommodation has, the more it costs per night. Additionally, we know that the motel Posh is located on a main street, which is why we expect its rooms to be noisy.

Avoiding noise is very important to us, minimizing the cost per star is less important, and maximizing the star rating is least important to us.

Reconstructed Scenario in PENGASP

- 1. Choose either the accommodation Amora or Grace or Metro or Posh or Adina.
- 2. The hotel Amora owns five stars and costs 170 dollars.
- 3. The hotel Grace owns four stars and costs 140 dollars.
- 4. The hotel Metro owns three stars and costs 90 dollars.
- 5. The guesthouse Adina owns two stars and costs 60 dollars.
- 6. The motel Posh that is located on the main street owns three stars and costs 75 dollars.

Reconstructed Scenario in PENGASP

- 7. If an accommodation is located on a main street then the accommodation is noisy.
- 8. If an accommodation costs N dollars and owns M stars then N / M is the cost per star of the accommodation.
- 9. If an accommodation owns N stars then N is the star rating of the accommodation.

Reconstructed Scenario in PENGASP

- 10. Minimizing that an accommodation is noisy has the high priority H.
- 11. Minimizing the cost per star of an accommodation has the medium priority M.
- 12. Maximizing the star rating of an accommodation has the low priority L.
- 13. The high priority is 3.
- 14. The medium priority is 2.
- 15. The low priority is 1.

Choose either the accommodation Amora or Grace or Metro or Posh or Adina.

1 { class(X, accommodation) :
 named(X, (amora ; grace ; metro ; posh ; adina)) } 1.

Sentences 2-5: 4

The hotel Metro owns three stars and costs 90 dollars.

```
class(7, hotel).
named(7, metro).
pred(7, 8, own).
data_prop(8, pos_int(3), cardinal).
class(8, star).
pred(7, 9, cost).
data_prop(9, pos_int(90), cardinal).
class(9, dollar).
```

The motel Posh that is located on the main street owns three stars and costs 75 dollars.

```
class(13, motel).
named(13, posh).
prop(13, 14, located_on).
class(14, main_street).
pred(13, 8, own).
pred(13, 15, cost).
data_prop(15, pos_int(75), cardinal).
class(15, dollar).
```

Sentence 6: Relative Clause

The motel Posh that is located on the main street owns three stars and costs 75 dollars.

```
class(13, motel).
named(13, posh).
prop(13, 14, located_on).
class(14, main_street).
pred(13, 8, own).
pred(13, 15, cost).
data_prop(15, pos_int(75), cardinal).
class(15, dollar).
```

Sentence 6: Anaphoric Reference

The motel Posh that is located on the main street owns three stars and costs 75 dollars.

```
class(13, motel).
named(13, posh).
prop(13, 14, located_on).
class(14, main_street).
pred(13, 8, own).
pred(13, 15, cost).
data_prop(15, pos_int(75), cardinal).
class(15, dollar).
```

If an accommodation is located on a main street then the hotel is noisy.

```
prop(X, noisy) :-
    class(X, accommodation),
    prop(X, Y, located_on),
    class(Y, main_street).
```

If an accommodation costs N dollars and owns M stars then N / M is the cost per star of the accommodation.

```
rel(N/M, X, cost_per_star) :-
    class(X, accommodation),
    pred(X, Y, cost),
    data_prop(Y, pos_int(N), cardinal),
    class(Y, dollar),
    pred(X, Z, own),
    data_prop(Z, pos_int(M), cardinal),
    class(Z, star).
```

If an accommodation owns N stars then N is the star rating of the accommodation.

```
rel(N, X, star_rating) :-
   class(X, accommodation),
   pred(X, Y, own),
   data_prop(Y, pos_int(N), cardinal),
   class(Y, star).
```

Minimizing that an accommodation is noisy has the high priority H.

```
#minimize { 10H,
        X : class(X, accommodation),
        prop(X, noisy),
        prop(P, high),
        class(P, priority),
        data prop(P, pos int(H), nominal) }.
```

Minimizing the cost per star of an accommodation has the medium priority M.

```
#minimize { X@M,
    Y : rel(X, Y, cost_per_star),
        class(Y, accommodation),
        prop(P, medium),
        class(P, priority),
        data prop(P, pos int(M), nominal) }.
```

Maximizing the star rating of an accommodation has the low priority L.

```
#maximize { X@L,
    Y : rel(X, Y, star_rating),
        class(Y, accommodation),
        prop(P, low),
        class(P, priority),
        data prop(P, pos int(L), nominal) }.
```

Sentences 13-15: 13

```
The high priority is 3.
prop(16, high).
class(16, priority).
data_prop(16, pos_int(3), nominal).
```

:~ prop(13, 1	noisy), class(13,	accommodation).	[1@3,	13]
:~ class(1,	accommodation).	[34@2, 1]		
:~ class(4,	accommodation).	[35@2, 4]		
:~ class(7,	accommodation).	[30@2, 7]		
:~ class(10,	accommodation).	[30@2, 10]		
:~ class(13,	accommodation).	[25@2, 13]		
:~ class(1,	accommodation).	[-5@1, 1]		
:~ class(4,	accommodation).	[-4@1, 4]		
:~ class(7,	accommodation).	[-3@1, 7]		
:~ class(10,	accommodation).	[-201, 10]		
:~ class(13,	accommodation).	[-3@1, 13]		
				20

:~ prop(13, noisy), cla	ass(13, accom	modation).	[1@3,	13]
:~ class(1, accommoda	tion). [34@2,	1]		
:~ class(4, accommoda	tion). [35@2,	4]		
:~ class(7, accommoda	tion). [30@2,	7]		
:~ class(10, accommoda	tion). [30@2,	10]		
:~ class(13, accommoda	tion). [25@2,	13]		
:~ class(1, accommoda	tion). [-5@1,	1]		
:~ class(4, accommoda	tion). [-4@1,	4]		
:~ class(7, accommoda	tion). [-3@1,	7]		
:~ class(10, accommoda	tion). [-2@1,	10]		
:~ class(13, accommoda	tion). [-3@1,	13]		
				20

:~]	prop(13, 1	noisy),	class(13,	accom	modation).	[1@3,	13]
:~ (class(1,	accommo	dation).	[34@2,	1]		
:~ (class(4,	accommo	dation).	[35@2,	4]		
:~	class(7,	accommo	dation).	[30@2,	7]		
:~	class(10,	accommo	dation).	[30@2,	10]		
:~	class(13,	accommo	dation).	[25@2,	13]		
:~ (class(1,	accommo	dation).	[-5@1,	1]		
:~ (class(4,	accommo	dation).	[-4@1,	4]		
:~	class(7,	accommo	dation).	[-3@1,	7]		
:~	class(10,	accommo	dation).	[-2@1,	10]		
:~	class(13,	accommo	dation).	[-3@1,	13]		
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:~ prop(13, r	noisy), class(13,	accommodation).	[103,	13]
:~ class(1,	accommodation).	[34@2, 1]		
:~ class(4,	accommodation).	[35@2, 4]		
:~ class(7,	accommodation).	[30@2, 7]		
:~ class(10,	accommodation).	[30@2, 10]		
:~ class(13,	accommodation).	[25@2, 13]		
:~ class(1,	accommodation).	[-5@1, 1]		
:~ class(4,	accommodation).	[-4@1, 4]		
:~ class(7,	accommodation).	[-3@1, 7]		
:~ class(10,	accommodation).	[-2@1, 10]		
:~ class(13,	accommodation).	[-3@1, 13]		
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Solution

- % named(7, metro) ... class(7, hotel) ...
- % class(7, accommodation)
- % Optimization: 0 30 -3
- **% OPTIMUM FOUND**
- % Time : 0.004s

Take-Home Messages

- PENG^{ASP} is human-understandable and machine-processable controlled natural language.
- PENG^{ASP} can bridge the gap between English and a formal languages like ASP.
- PENG^{ASP} can be used to specify not only strong constraints but also weak constraints in controlled language.
- Weak constraints rank answer sets.
- Future research: PENG^{ASP} for smart contracts, for dynamic domains and for conceptual modelling.