Knowledge Technology CLIPS PROJECT - JOBEX

The Problem

- Using CLIPS 6.3, I have implemented an intelligent system that acts as a career advisor. For every user it produces a set of jobs that are best suited for him/her, based on the answers he/she will give to a questionnaire;
- Every answer generates 1 or more facts. Every fact has an associated numeric value between -1 and 1 called certainty factor (CF);

```
(deftemplate a-fact
    (slot name)
    (slot cf (default 0)))
```

Through a repeated application of modus ponens the system infers the best set of jobs, also with an associated certainty factor (CF).

Certainty Theory

- An uncertainty methodology has been implemented to account for the inherent unreliability of the data, since it is generated from answers given by a human, thus affected by human factors such as level of attention, degree of honesty, lack of knowledge etc.
- Certainty Theory is a system that works with judgmental measures of belief and associates to every fact a certainty factor (CF).

Certainty Factor

► Certainty Factor (CF) = Belief (0 <= CF <= 1) – Disbelief (-1 <= CF <= 0)

Meaning	Value
Definitely not	-1
Almost certainly not	-0.8
Probably not	-0.6
Maybe not	-0.4
Unknown	-0.2 to 0.2
Maybe	0.4
Probably	0.6
Almost certainly	0.8
Definitely	1

Certainty Factor Propagation #1

- A certainty factor can be associated to a fact axiomatically;
- ▶ Basic rule of derivation: IF E THEN H CF(Rule) CF(H, E) = CF(E)*CF(Rule);
- ► Conjuction: IF (E1 \cap E2 \cap ...) THEN H CF(Rule) CF(H, E1 \cap E2 \cap ...) = min{CF(Ei)}*CF(Rule);
- **Disjunction**: IF (E1 U E2 U ...) THEN H CF(Rule) CF(H, E1 U E2 U ...) = $max\{CF(Ei)\}*CF(Rule);$
- Negation: CF(!E) = -CF(E);
- Mix: IF E1 AND E2 OR E3 AND E4 THEN H CF(H) = Max={min(E1, E2), min(E3, E4)}* CF(Rule);

Certainty Factor Propagation #2

- When more rules conclude the same hypothesis, there is a way to combine the values of belief/disbelief;
- Both > 0: CF(CF1, CF2) = CF1 + CF2*(1 CF1);
- One < 0: CF(CF1, CF2) = (CF1 + CF2)/(1-min{|CF1|, |CF2|});</p>
- ▶ Both < 0: CF(CF1, CF2) = CF1 + CF2*(1 + CF1).

Certainty Factor Propagation CLIPS #1

► In CLIPS IF (E1 U E2 U ...) THEN H CF(Rule) - CF(H, E1 U E2 U ...) = max{CF(Ei)}*CF(Rule) is implemented as follows:

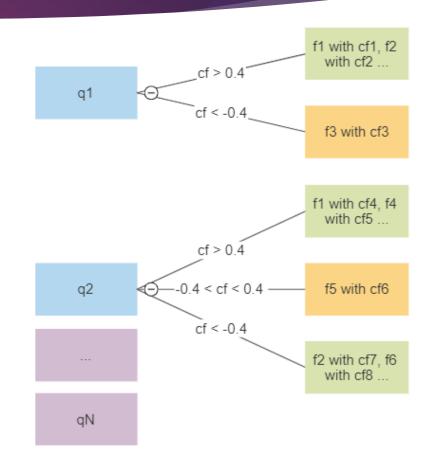
Certainty Factor Propagation CLIPS #2

In CLIPS CF(CF1, CF2) = (CF1 + CF2)/(1-min{|CF1|, |CF2|}) is implemented as follows:

```
(defrule combine-certainties-3 (declare (salience 100)(auto-focus TRUE))
  ?fact1 <- (a-fact (name ?id) (cf ?cf1))
  ?fact2 <- (a-fact (name ?id) (cf ?cf2))
  (test (neq ?fact1 ?fact2))
  (test (> ?cf1 0))
  (test (> ?cf2 0))
  =>
    (retract ?fact1)
  (modify ?fact2 (cf (/ (+ ?cf1 ?cf2) (- 1 (min (abs ?cf1) (abs ?cf2)))))))
```

Information Flow #1

- qi = questionnaire questions;
- ▶ The answer to a qiis:-1 < CF < 1;
- Depending on the value of the answer, a set of facts are asserted. The certainty factor propagates from the answer to the facts according to the rules of propagation;
- ▶ Different qis can infer the same fact, with different or equal certainty factors. The certainty factor of this fact is resolved into one according the rules of propagation.



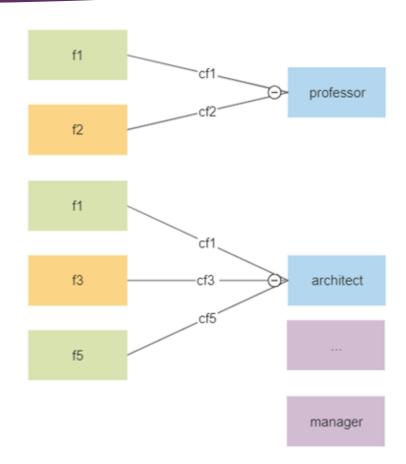
Information Flow #1 in CLIPS

Information Flow #2

- In the end after enough evidence has been collected is possible to infer a set of best jobs;
- Each job also has a CF associated to it. For example:

Mathematician with cf 0.6496

Construction Worker with cf 0.899397152038656



Information Flow #2 in CLIPS

How to run the program

- ▶ Download and install CLIPS from http://www.clipsrules.net/;
- ► Loadjobex.clp, reset and run;

THANK YOU!