Expert Systems a.k.a. Business Rules Engines, *another kind of AI*

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Let's begin with... some history



Expert Systems, the AI precursor & AI savior of the 80'

"Expert systems were formally introduced around 1965."

"... such as diagnosing infectious diseases (Mycin) and identifying unknown organic molecules (Dendral)."

"In the early 1980s, **AI research was revived by the commercial success of expert systems**. By 1985, the market for AI had reached over a billion dollars."

https://en.wikipedia.org/wiki/Artificial_intelligence



A large variety of applications

- Finance / Banks
 - Morgan Stanley, Fidelity Investments, Casden Banque Populaire, BNP Paribas, Banque de France (AMF)
- Insurance
 - Shenandoah Life Insurance, Farmers, Manulife, Intact
- Public Sector
 - Hydro Quebec, State of Nevada, Teranet
- HealthCare
 - Partners HealthCare System, Optum



Fine, but what is... an expert system?



Are Expert Systems…

- Systems that reproduce the reasoning of (human) experts?
 - Actually, Experts are mostly relaying on past experiences… (Case-based reasoning)
- Systems that (can) replace (human) experts?
 - Systems that can compute complex decisions
 - Not based on hardcoded sequences of formulas/functions
 - But on "Unknown" combinations of individual/local reasoning



The key element of the ES: Inference



Inference…

- Starting from a set of initial piece of knowledge
- Then, applying successive rules to this knowledge base to produce additional knowledge
- Two types of inference (based on logical "rules"):
 - Forward Chaining (the most common)
 - To deduce (all possible) new knowledge from existing "facts"
 - Backward Chaining
 - To prove hypothesis or to answer specific questions



"Logical" Rules…

- Logical relations between facts
- Can be expressed by p(f) => p'(f)
 - If/when conditions then actions (!)



To be (a vehicle) or not to be?

- if car(x) then vehicle(x)
- if car(x) then hasWheels(x)
- if plane(x) then vehicle(x)
- if vehicle(x) then canMoveByItself(x)

car(Corolla)

- 1.plane(Corolla)? canMoveByItself(Corrola)?
 - backward or forward chaining?
- 2.=> vehicle(Corolla), canMoveByItself(Corolla), hasWheels(Corrola)

Forward Chaining: how to implement it?



The *RETE* algorithm

- The RETE algorithm was designed by Charles L. Forgy of Carnegie Mellon University between 1974 and 1982
- Was developed to *efficiently* apply many patterns (i.e. the conditions of the rules) to many objects
- Became the root of the **Rules Engines**
- (The RETE algorithm is no more the most efficient one, but is still usually the most flexible)



The pattern matching of the RETE algorithm

```
A( a1=3 a2=10 )
                                                           discrimination
                                                           tree
B( b1=2 b2=4 b3=10 )
                                                                                                                class=
                                                                        class=
                                                                                           class=
B( b1=2 b2=7 b3=10 )
                                                                                                     tests
C(c1=4)
                                                                         a1=3
                                                                                            b1=2
                                                           alpha
                                                                                                                  C
                                                                                       b1=2 b2=7 b3=10
                                                                                                                 c1=4
                                                           nodes
                                                                      a1=3 a2=10
                                                                                                         tokens
rulefilter {
                                                                                       b1=2 b2=4 b3=10
    when {
                                                                                a2=b3
        A (a1==3; ?x:a2);
                                                                                                                           tests
        B (b1==2; ?y:b2; b3==?x);
                                                                                                                       giving rise
                                                                                b1=2 b2=7 b3=10
                                                                                                                      to join nodes
                                                                       a1=3 a2=10
        C (c1 = ?y);
                                                                       A B b1=2 b2=4 b3=10
                                                                                                                 b2=c
    then {
                                                                                                    A
a1=3 a2=10 B
b1=2 b2=4 b3=10
                                                                                                                            С
         System.out.println("filter");
                                                                                        tuples
                                                                                                                           c1=4
}
                                                                                                                  rules
```



The key advantages of the RETE algorithm

- No need for a *full* reevaluation of the RETE network when:
 - Objects / facts are added
 - objects / facts are removed
 - objects / facts are updated
- Some rules can be partially combined
 - Only when they start with the *same conditions*, in the *same order*
- (The most discriminant conditions should be declared first)



An iterative process… to explore all derivative Knowledge Agenda



















An iterative process… until no more eligible rules



Rule Engines based on the RETE algorithm



Rule Engines based on the RETE algorithm

- CLIPS (created at NASA in 1985)
 - Written in C but with a Lisp like syntax
- JESS
 - An alternative written in Java, again using a Lisp like syntax

```
Jess> (bind ?x 1)

1

Jess> (if (> ?x 100) then

(printout t "X is big" crlf)

else

(printout t "X is small" crlf))

X is small

...
```



Experta, the inspired by CLIPS Python library

```
from random import choice
from experta import *

class Light(Fact):
    """Info about the traffic light."""
    pass

class RobotCrossStreet(KnowledgeEngine):
    @Rule(Light(color='green'))
        def green_light(self):
            print("Walk")
    @Rule(Light(color='red'))
        def red_light(self):
            print("Don't walk")
    @Rule(AS.light << Light(color=L('yellow') | L('blinking-yellow')))
        def cautious(self, light):
            print("Be cautious because light is", light["color"])</pre>
```

Experta, the inspired by CLIPS Python library

```
>>> engine = RobotCrossStreet()
>>> engine.reset()
>>> engine.declare(
    Light(color=choice(['green', 'yellow', 'blinking-yellow', 'red'])))
>>> engine.run()
Be cautious because light is blinking-yellow
```

Great fit when you have to call simple rules from your existing Python code



Rule Engine: is it enough?



(simple) Rule Engines do lack…

- A language to easily express rules (by SME)
- Ways to logically regroup rules
- Rules represented by Decision Tables
- A collaborative platform (through a common Web interface)
- Integrated rules governance, packaging & deployment
- Robust, very fast & distributed engines

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Business Rule Management Systems, beyond basic Rule Engines



Business Rule Management Systems





ODM and Drools, two examples of BRMS

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Expert Systems a.k.a. Business Rules Engines,

Conclusion



(Business) Rule Engines…

- Can be a *very powerful* way to implement *certain* types of AI, when:
 - Knowledge is not made of data, but of individual "rules"
 - Explanations (& some reproducibility) are required
- Developing "Expert Systems" requires:
 - Dedicated tools (Rule Engine or BRMS)
 - Specific mindset (not procedural programming, not ML···)