

Approximate Reasoning

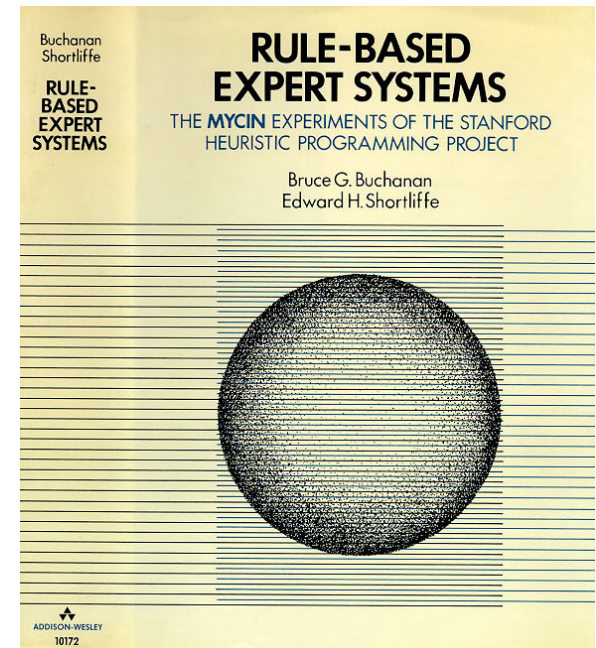
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Certainty Factors

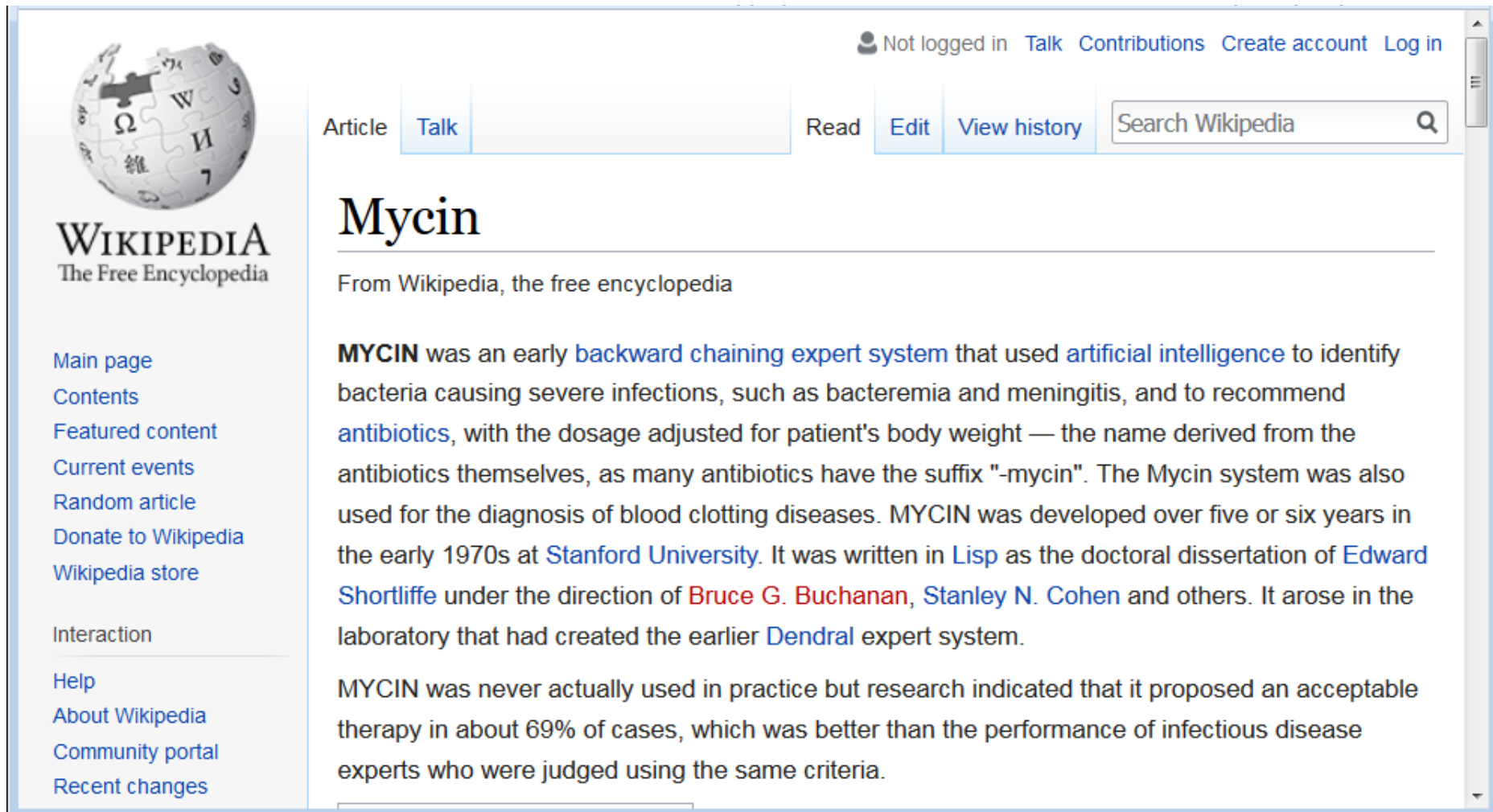
MYCIN

Introduction

- Certainty Factors is a well-known a quasi-probabilistic model that is founded on the classic probability theory.
- It is one of the pioneer techniques for approximate reasoning.
- It was defined in the MYCIN expert system.



Introduction



The screenshot shows the Wikipedia article for "Mycin". The page layout includes a sidebar on the left with navigation links, a top navigation bar with user status and actions, and a main content area with the article title and text. The article text describes MYCIN as an early backward chaining expert system for identifying bacteria and recommending antibiotics.

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Mycin

From Wikipedia, the free encyclopedia

MYCIN was an early [backward chaining expert system](#) that used [artificial intelligence](#) to identify bacteria causing severe infections, such as bacteremia and meningitis, and to recommend [antibiotics](#), with the dosage adjusted for patient's body weight — the name derived from the antibiotics themselves, as many antibiotics have the suffix "-mycin". The Mycin system was also used for the diagnosis of blood clotting diseases. MYCIN was developed over five or six years in the early 1970s at [Stanford University](#). It was written in [Lisp](#) as the doctoral dissertation of [Edward Shortliffe](#) under the direction of [Bruce G. Buchanan](#), [Stanley N. Cohen](#) and others. It arose in the laboratory that had created the earlier [Dendral](#) expert system.

MYCIN was never actually used in practice but research indicated that it proposed an acceptable therapy in about 69% of cases, which was better than the performance of infectious disease experts who were judged using the same criteria.

Motivation

- The Bayesian model based on probabilities makes some assumptions that sometimes are not appropriate for the experts.
- For example,

$$p(c|e) = x \Rightarrow p(\neg c|e) = 1 - x$$

- The probability of the conclusion and its negation gives a sum of 1. But some experts argue that a decrease in the confidence about one conclusion c , does not increase the confidence in its contrary.

Certainty Factors

- MYCIN expert system introduced the CF model (1970s).
- It is based on calculating two measures:
 - Measure of Belief: MB
 - Measure of Disbelief: MD

$$e \xrightarrow{MB(c,e), MD(c,e)} c$$

Certainty Factors

- Measure of Belief: MB

$$MB(c, e) = \begin{cases} 1 & \text{if } p(c) = 1 \\ \frac{\max(p(c | e), p(c)) - p(c)}{1 - p(c)} & \text{if } p(c) \neq 1 \end{cases}$$

- Measure of Disbelief: MD

$$MD(c, e) = \begin{cases} 1 & \text{if } p(c) = 0 \\ \frac{p(c) - \min(p(c | e), p(c))}{p(c)} & \text{if } p(c) \neq 0 \end{cases}$$

Certainty Factors. Properties of MB and MD

1. Both values are in $[0,1]$
2. If $MB(c, e) > 0 \Rightarrow MD(c, e) = 0$. In this case the evidences increase the belief about the conclusion c .
3. If $MD(c, e) > 0 \Rightarrow MB(c, e) = 0$. In this case the evidences decrease the confidence about c .

Therefore, a certain evidence cannot increase and decrease the confidence at the same time.

Certainty Factors

- Both measures can be combined into CF as:

$$CF(c, e) = \frac{MB(c, e) - MD(c, e)}{1 - \min(MB(c, e), MD(c, e))}$$

IF <evidence e > THEN <conclusion c > $\{CF_{rule}\}$

where CF represents belief in conclusion c
given that evidence e has occurred.

- Properties:
 1. CF is in $[-1, 1]$
 2. If $CF(c, e) > 0$ the evidence increases the belief about c
 3. If $CF(c, e) < 0$ the evidence decreases the belief about c

Certainty Factors

- The certainty factor assigned by a rule is propagated through the rule. This involves establishing the net certainty of the rule consequent when the evidence in the rule antecedent is uncertain:

$$CF(c) = CF(e) \times CF(c, e)$$

For example:

IF sky is clear
THEN the forecast is sunny $\{CF_{rule} 0.8\}$

and the current certainty factor of *sky is clear* is 0.5, then

$$CF(c) = 0.5 \times 0.8 = 0.4$$

This result can be interpreted as "It may be sunny".

Certainty Factors. Combination rules.

- For **conjunctive rules** such as:

IF <evidence e1>
AND <evidence e2> AND ...
THEN <conclusion c> {CF}

the certainty of conjunction of evidences is:

$$CF(e_1 \cap e_2 \cap \dots \cap e_n) = \min [CF(e_1), CF(e_2), \dots, CF(e_n)]$$

For example:

IF sky is clear AND the forecast is sunny
THEN the action is 'wear sunglasses' {CF=0.8}

and the certainty of **sky is clear is 0.9** and the certainty of the forecast of **sunny is 0.7**, then

$$CF(c, e_1 \cap e_2) = \min [0.9, 0.7] \times 0.8 = 0.7 \times 0.8 = 0.56$$

Certainty Factors

- For **disjunctive rules** such as:

IF <evidence e1>
OR <evidence e2> OR ...
THEN <conclusion c> {CF}

the certainty of disjunction of the evidences is:

$$CF(e_1 \cup e_2 \cup \dots \cup e_n) = \max [CF(e_1), CF(e_2), \dots, CF(e_n)]$$

For example:

IF sky is overcast OR the forecast is rain
THEN the action is 'take an umbrella' {CF=0.9}

and the certainty of **sky is overcast** is **0.8** and the certainty of the forecast of **rain** is **0.5**, then

$$CF(c, e_1 \cup e_2) = \max [0.8, 0.5] \times 0.9 = 0.8 \times 0.9 = 0.72$$

Certainty Factors. Combination rules.

- When the same consequent is obtained as a result of the execution of two or more rules, the individual certainty factors of these rules must be merged to give a combined certainty factor for a hypothesis.

Rule 1: IF A is X
 THEN C {CF=0.8}

Rule 2: IF B is Y
 THEN C {CF=0.6}

- What certainty should be assigned C if both Rule 1 and Rule 2 are fired?

Certainty Factors. Parallel combination rule.

To calculate a combined certainty factor from two others we must consider three cases, depending on the sign of the two CF values that are combined.

$$CF(c) = CF1(c) \text{ combined } CF2(c) = \begin{cases} CF1 + CF2(1 - CF1) & CF1 > 0 \text{ \& } CF2 > 0 \\ CF1 + CF2(1 + CF1) & CF1 < 0 \text{ \& } CF2 < 0 \\ \frac{CF1 + CF2}{1 - \min(|CF1|, |CF2|)} & \text{otherwise} \end{cases}$$

This combination rule is commutative and associative
It must be applied only to two inputs (in pairs)

Bayesian reasoning vs Certainty Factors

- The certainty factors theory provides a practical alternative to classic Bayes theorem in probability.
- The heuristic manner of combining certainty factors is different from the manner in which they would be combined if they were probabilities.
- The certainty theory is not “mathematically pure” but does mimic the thinking process of a human expert.

Probabilistic reasoning vs Certainty Factors

- The probabilistic method is likely to be the most appropriate if reliable statistical data exists.
- In the absence of any of the specified conditions, the probabilistic approach might be too arbitrary and even biased to produce meaningful results.
- Although the certainty factors approach lacks the mathematical correctness of the probability theory, it outperforms probabilistic reasoning in real application areas, such as medical diagnostics.

Additional Readings

- **Artificial Intelligence. A new synthesis.** Nils. J. Nilsson, Morgan-Kauffman, 1998 (part III). (004.8 Nil at URV in Spanish)
- **Artificial Intelligence.** Elaine Rich & Kevin Knight. Ed. McGraw Hill, 1991. Chapter 8 (004.8 Ric at URV)
- See paper in Moodle from:

<https://www.microsoft.com/en-us/research/wp-content/uploads/2016/11/The-Certainty-Factor-Model.pdf>