

# “At Least $k$ out of $n$ ” under Fuzzy Uncertainty: Efficient Algorithm for General “And”-Operations

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## 1. “At least $k$ out of $n$ ” is practically useful

- In medicine, often a diagnosis is based on the condition that a certain number of symptoms are present.
- Usually, it is at least  $k$  symptoms out of  $n$  possible symptoms, for some  $k$ .
- Let us give two examples.
- “A patient is classified as high risk for septic shock if they show at least four out of the following signs:
  - hypotension, tachycardia, fever, leukocytosis,
  - altered mental status, elevated serum lactate levels.”

## 2. “At least $k$ out of $n$ ” is practically useful (cont-d)

- “A patient is classified as low risk for septic shock if they show at most two out of the following signs:
  - hypotension, tachycardia, fever, leukocytosis,
  - altered mental status, elevated serum lactate levels.”
- This means that when there are at least three signs, the risk is medium or high.

### 3. Need to take fuzzy-type uncertainty into account

- In principle, one can use “yes”-“no” (crisp) definitions of the symptoms.
- For example, we can say that a body temperature of 38 C or higher is fever, while anything smaller than 38 is not a fever.
- However, from the commonsense viewpoint, it does not make sense to call 38.0 a fever and 37.9 not a fever.
- The difference between the two values is close to the accuracy of the thermometer.

#### 4. Need to take fuzzy-type uncertainty into account (cont-d)

- From this viewpoint, it makes more sense to talk about to what degree the patient has a fever:
  - temperature much smaller than 38 means that this is definitely not a fever,
  - temperature much larger than 38 means that this is definitely a fever, but
  - temperature close to 38 mean that we have a fever to some degree.
- The technique to deal with such degrees is what is usually called *fuzzy logic*.
- In these terms, what we need is to take into account fuzzy uncertainty.

## 5. How to estimate to what extent is the at-least- $k$ -out-of- $n$ condition satisfied: a natural idea

- When each statement is either true or false, it is straightforward to decide when we have at least  $k$  out of  $n$  symptoms.
- However:
  - when we have symptoms satisfied to certain degrees  $m_1, \dots, m_n$ ,
  - the at-least- $k$ -out-of- $n$  condition is only satisfied to some degree.
- How can we compute this degree?
- In fuzzy techniques, a usual way to assign a degree to a complex statement is:
  - to represent this statement in terms of the basic logical connectives
    - “and”, “or”, and “not” – and then
  - to use fuzzy analogs of these connectives.

## 6. How to estimate to what extent is the at-least- $k$ -out-of- $n$ condition satisfied: a natural idea (cont-d)

- These analogs include:
  - “and”-operations  $f_{\&}(a, b)$  (also known as t-norms),
  - “or”-operations  $f_{\vee}(a, b)$  (also known as t-conorms), and
  - negation operations  $f_{\neg}(a)$ .
- For example, “at least  $k$  out of  $n$ ” means that the set of symptoms can be any set  $S$  with at least  $k$  elements.
- For each of these sets, the above approach leads to the formula

$$m(S) = f_{\&}(m_{i_1}, \dots, n_{i_\ell}, f_{\neg}(m_{j_1}), \dots, f_{\neg}(m_{j_p})).$$

- Here  $i_1, \dots, i_\ell$  are all elements of the set  $S$ , while  $j_1, \dots, j_p$  are all symptoms that do not belong to the set  $S$ .

## 7. How to estimate to what extent is the at-least- $k$ -out-of- $n$ condition satisfied: a natural idea (cont-d)

- Then, the degree  $m$  to which the condition is satisfied:
  - can be computed as  $f_{\vee}(m(S_1), m(S_2), \dots)$ ,
  - where  $S_1, S_2$ , etc. are all the subsets of the set  $\{1, 2, \dots, n\}$  that have at least  $k$  elements.

## 8. Limitations of the above natural idea

- In the above natural-idea approach:
  - to find the desired degree  $m$ ,
  - we need to compute  $m(S)$  for a large number of sets – close to  $2^n$ .
- For large  $n$ , this number becomes astronomical.
- Because of this limitation, it is desirable to come up with an efficient way to define and compute the desired degree.

## 9. What is known and remaining problem

- We want our degree to reflect the practice of the corresponding discipline – e.g., of the corresponding branch of medicine.
- In general, reasoning in different branch of knowledge is best described by different “and”-operations.
- This was first discovered when the first expert systems appeared.
- Historically the first expert system was the system MYCIN that was focused on a certain class of blood diseases.
- Designers of this system spent a lot of time and efforts:
  - searching for the “and”-operation
  - that provides the most accurate description of the reasoning of the medical doctors dealing with these diseases.
- At first, they were under the impression that they found universal laws of human reasoning.

## 10. What is known and remaining problem (cont-d)

- However:
  - when they tried to apply the same “and”-operation to a different application area – geophysics,
  - they found out that their degrees for “and”-statements were very different from what geophysicists’,
  - and thus, that a different “and”-operation is needed for geophysical applications.
- This difference in “and”-operations make perfect sense.
- In medicine, one needs to be very cautious, and to prescribe some cure only if we are reasonable sure that it will help.
- Medical mistakes can be deadly.

## 11. What is known and remaining problem (cont-d)

- In contrast, in mining applications of geophysics:
  - if a company waits too long for a perfect conformation that there is oil in a field,
  - it may lose to competitors.
- Too much caution can ruin a company.
- An efficient algorithm for estimating the desired degree  $m$  is known for the case when the “and”-operation is  $f_{\&}(a, b) = \max(a + b - 1, 0)$ .
- We want to take care of all possible applications where the “at least  $k$  out of  $n$ ” idea is used.
- So, we need to extend the above-mentioned feasible algorithm for a *specific* “and”-operation to the case of *general* “and”-operations.
- In this talk, we solve this problem by providing an efficient algorithm for general “and”-operations.

## 12. How can we describe a general “and”-operation: a brief reminder

- It is known that any continuous “and”-operation can be approximated:
  - with any given accuracy  $\varepsilon > 0$ ,
  - by a *strictly Archimedean* “and”-operation, i.e., operation of the type  $f_{\&}(a, b) = f^{-1}(f(a) \cdot f(b))$ .
- Here,  $f(a)$  is a continuous strictly increasing function  $f(a)$ .
- And  $f^{-1}(a)$  means the inverse function, i.e., the function for which  $f^{-1}(b) = a$  if and only if  $f(a) = b$ .
- Based on finitely many experiments with real experts, we can only determine the “and”-operation with some accuracy.

### 13. How can we describe a general “and”-operation: a brief reminder (cont-d)

- This means that we can have a strictly Archimedean “and”-operation that is in perfect accordance with the experimental data.
- Thus, to describe actual expert reasoning, we can always use strictly Archimedean “and”-operations.

## 14. Our first idea

- The use of strictly Archimedean “and”-operation means, in effect, that:
  - if we re-scale degrees, i.e., replace each degree  $m_i$  with the degree  $m'_i = f(m_i)$ ,
  - then we can simply use multiplication as “and”, and
  - at the end of computations, we need to transform the resulting degree  $m'$  back to the original scale by taking  $m = f^{-1}(m')$ .
- So this is our idea of how to estimate the desired degree  $m$  under a general strictly Archimedean “and”-operation:
  - first, we transform the original degrees  $m_i$  into re-scaled ones  $m'_i = f(m_i)$ ,
  - then, we use the values  $m'_i$  to perform multiplication-based estimation of the degree  $m$ , and
  - after that, compute the desired estimate  $m$  as  $f^{-1}(m)$ .

## 15. Our second idea: how to perform multiplication-based computations

- We have reduced the problem for a general “and”-operation to its specific case, when the “and”-operation is simply multiplication.
- So, to solve the general problem, it is sufficient to solve it for the case of the product “and”-operation  $f_{\&}(a, b) = a \cdot b$ .
- For this operation, the “and”-formula is similar to the probabilistic case, where:
  - for two independent events,
  - the probability that both will occur is equal to the product  $a \cdot b$  of the probabilities  $a$  and  $b$  of each of the two events.
- Probability theory exists for many centuries, many algorithms have been designed for it.

## 16. Our second idea: how to perform multiplication-based computations (cont-d)

- So, for the multiplication “and”-operation, a natural way to provide an estimate is:
  - to view the values  $m'_i$  as probabilities and
  - to estimate the probability that at least  $k$  out of  $n$  symptoms are satisfied.
- To use this idea, we need to come up with the efficient algorithm for computing this probability.
- Of course, it is important to emphasize that the use of probabilities here is simply a mathematical trick.
- It does not mean that expert-produced degrees  $m_i$  somehow became probabilities.

## 17. Numerical example

- Let us illustrate our approach on a simple numerical example, in which  $f_{\&}(a, b) = a \cdot b$ . In this case,  $f(m) = m$ .
- Let us consider the case when  $n = 4$ ,  $k = 3$ ,  $m_1 = m_2 = 0.5$ , and  $m_3 = m_4 = 0.6$ .
- In this case, to find the desired degree to which at least 3 out of 4 symptoms are satisfied, we need to consider the following cases:
- We need to consider cases when exactly 3 symptoms are present:

$$m(\{1, 2, 3, 4\} - \{1\}) = m(\{2, 3, 4\}) = (1 - 0.5) \cdot 0.5 \cdot 0.6 \cdot 0.6 = \\ 0.5 \cdot 0.5 \cdot 0.6 \cdot 0.6 = 0.09;$$

$$m(\{1, 2, 3, 4\} - \{2\}) = m(\{1, 3, 4\}) = 0.5 \cdot (1 - 0.5) \cdot 0.6 \cdot 0.6 = \\ 0.5 \cdot 0.5 \cdot 0.6 \cdot 0.6 = 0.09;$$

$$m(\{1, 2, 3, 4\} - \{3\}) = m(\{1, 2, 4\}) = 0.5 \cdot 0.5 \cdot (1 - 0.6) \cdot 0.6 = \\ 0.5 \cdot 0.5 \cdot 0.4 \cdot 0.6 = 0.06.$$

## 18. Numerical example (cont-d)

- One more such case:

$$\begin{aligned}m(\{1, 2, 3, 4\} - \{4\}) &= m(\{1, 2, 3\}) = 0.5 \cdot 0.5 \cdot 0.6 \cdot (1 - 0.6) = \\ &0.5 \cdot 0.5 \cdot 0.6 \cdot 0.4 = 0.06;\end{aligned}$$

- We also need the case when all four symptoms are present:

$$m(\{1, 2, 3, 4\}) = 0.5 \cdot 0.5 \cdot 0.6 \cdot 0.6 = 0.09.$$

- Then, the desired degree  $m'$  that we have at least 3 symptoms is  $m' = m(\{2, 3, 4\}) + m(\{1, 3, 4\}) + m(\{1, 2, 3\}) + m(\{1, 2, 3\}) + m(\{1, 2, 3, 4\}) = 0.09 + 0.09 + 0.06 + 0.06 + 0.09 = 0.39.$

## 19. Comment

- For small  $n$ , we could simply enumerate all possible sets of desired size.
- However, for larger  $n$ , as we have mentioned earlier:
  - this will be very computationally intensive – and
  - even, for large  $n$ , infeasible.
- So, we need to come up with a more efficient algorithm.

## 20. How to effectively compute the desired probability: towards an algorithm

- The probability  $m'$  that at least  $k$  out of  $n$  events occur can be described as  $m' = 1 - p_0 - p_1 - p_2 - \dots - p_{k-1}$ .
- Here  $p_j$  is the probability that exactly  $j$  events happened independently.
- So, to compute  $m'$ , it is sufficient to be able to estimate the values  $p_j$ .
- One can easily check that we have

$$p_j = \sum_{S:|S|=j} \left( \prod_{i \in S} m'_i \cdot \prod_{i \notin S} (1 - m'_i) \right).$$

- Here  $|S|$  denotes the number of elements in the set  $S$ .

## 21. How to effectively compute the desired probability: towards an algorithm (cont-d)

- One can check that such sums appear as coefficients at  $z^j$  when we compute the product  $P(z)$  of all  $n$  terms  $(1 - m'_i) + z \cdot m'_i$ :

$$P(z) \stackrel{\text{def}}{=} \prod_{i=1}^n ((1 - m'_i) + z \cdot m'_i) = \sum_j p_j \cdot z^j.$$

- Indeed, let us use distributivity and represent the product of the sums as the sum of all possible products of terms  $m'_i$  and  $1 - m'_i$ .
- We can see that the only such products that lead to a coefficient at  $z^j$  are the products that have:
  - exactly  $j$  terms  $m'_i$  and,
  - correspondingly,  $n - j$  terms of the type  $1 - m'_i$ .
- By using the above formula, we can compute the values of the polynomial  $P(z)$  for different inputs  $z$ .

## 22. How to effectively compute the desired probability: towards an algorithm (cont-d)

- So, to find the values  $p_j$ , we need to find the coefficients of a polynomial based on its values.
- This is a known computational problem with a known solution.
- Namely, we compute  $P(z)$  for the values

$$z_k = \exp\left(i \cdot \frac{2\pi \cdot k}{n}\right), k = 0, \dots, n-1, \text{ where } i \stackrel{\text{def}}{=} \sqrt{-1}.$$

- The values  $P(z_k)$  form the Fourier transform of the sequence  $(p_0, \dots, p_{n-1})$ .
- Thus, to compute the values  $p_j$ , we can apply the inverse Fourier transform to the sequence  $(P(z_0), \dots, P(z_{n-1}))$ .

## 23. How to effectively compute the desired probability: towards an algorithm (cont-d)

- There is a known efficient algorithm for performing the inverse Fourier transform.
- It is known as inverse Fast Fourier Transform, or iFFT, for short.
- Thus, we arrive at the following algorithm for solving our problem.

## 24. Resulting algorithm

- We know the degrees  $m_1, \dots, m_n$  of each of  $n$  symptoms.
- We also know the function  $f(x)$  for which:
  - the “and”-operation  $f^{-1}(f(a) \cdot f(b))$
  - best describes the reasoning of people from this particular application area.
- We want to estimate the degree to which, based on this information, at least  $k$  symptoms are present.
- So first, we compute the values  $m'_i = f(m_i)$  for  $i = 1, \dots, n$ .
- Then, for  $k = 0, \dots, n - 1$ , we compute the values  $P(z_k)$  as described earlier.
- After that, we apply the inverse Fast Fourier Transform algorithm to the resulting sequence of values  $(P(z_0), \dots, P(z_{n-1}))$ .
- This produces the values  $p_0, \dots, p_{n-1}$ .

## 25. Resulting algorithm (cont-d)

- Then, we compute the value  $m'$  by using the above formula.
- Finally, we compute the desired value  $m = f^{-1}(m')$ .

## 26. What is the computational complexity of this algorithm

- Computing each of  $n$  terms  $(1 - m'_i) + z \cdot m'_i$  requires one multiplication, one subtraction, and one addition.
- So, overall we need 3 computational steps, if we count each arithmetic operation as one step.
- To compute each product  $P(z_k)$ , we need:
  - to compute  $n$  such terms – that will take  $3n$  steps,
  - and then to compute their product – which requires  $n$  more steps.
- So, computing each value  $P(z_k)$  takes  $4n$  computational steps.
- In our algorithm, we need  $n$  values  $P(z_k)$  corresponding to  $k = 0, \dots, n - 1$ .
- Thus, the overall computation of all these values takes time  $n \cdot 4n = 4n^2$ , which is  $O(n^2)$ .
- Inverse Fast Fourier Transform takes  $O(n \cdot \log(n))$  steps.

## 27. What is the computational complexity of this algorithm (cont-d)

- Applying the formula for  $m'$  takes  $k \leq n$  steps, i.e.,  $O(n)$ .
- Thus, overall, we need

$$O(n^2) + O(n \cdot \log(n)) + O(n) = O(n^2) \text{ computational steps.}$$

- So, our algorithm requires quadratic time, which is very feasible – and definitely much faster than computing  $2^n$  terms.

## 28. Clinical testing

- We are currently testing our technique on numerical values coming from actual clinical practice.
- Our preliminary results show that the results of applying our approach are in good accordance with the opinion of medical doctors.
- In particular:
  - preliminary results of applying our approach to the first example
  - of checking whether there is a high risk of septic shock
  - are described in a to-appear medical paper.

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