# Fuzzy Information Retrieval Based on Weighted Power-Mean Averaging Operators

Won-Sin Hong\*, Shyi-Ming Chen\*, and Shi-Jay Chen\*\*

\*Department of Computer Science and Information Engineering, National Taiwan University of Science and Technology, Taipei, Taiwan, R. O. C.

University of Science and Technology, Taiper, Taiwan, K. O. C.

\*\*Department of Information Management, Ching-Yun University, Jung-Li, Taiwan R. O. C.

### Abstract

In recent years, some researchers used averaging operators (i.e., Infinite-One operators, Waller-Kraft operators, P-Norm operators and GMA operators) to deal with AND and OR operations of users' queries for fuzzy information retrieval, but they still have some drawbacks, e.g., sometimes query results do not coincide with the intuition of the human being. In this paper, we present new averaging operators, called weighted power-mean averaging (WPMA) operators, based on the weighted power-mean for dealing with fuzzy information retrieval to overcome the drawbacks of the existing methods. The proposed WPMA operators are more flexible and more intelligent than the existing averaging operators to deal with users' fuzzy queries for fuzzy information retrieval.

## 1 Introduction

In [1], [2], [3], [4] and [5], the T-operators (i.e., T-norms and T-cornorms) are used to deal with fuzzy information retrieval. Although T-operators support the ranking facility, they still have some drawbacks, e.g., sometimes query results do not coincide with the intuition of the human being. In [6], Lee et al. pointed out that three averaging operators (i.e. Waller-Kraft operators [7], P-Norm operators [8] and Infinite-One operators [9]) have been proposed to achieve high retrieval effectiveness for fuzzy information retrieval, where these three averaging operators can avoid the drawbacks of T-operators. However, in [10], Chen et al. pointed out that these three averaging operators still have some drawbacks, i.e., it is subjective and hard to determine appropriate values for the parameters of these averaging operators, respectively. Thus, in [10], Chen et al. presented new averaging operators based on the geometric mean, called the geometric-mean averaging (GMA) operators, to overcome the drawbacks of the Waller-Kraft operators, the P-Norm operators and the Infinite-One operators. However, the GMA operators still have some drawbacks, i.e., in some specific situations, the retrieval results do not coincide with the intuition of the human being. Thus, it is necessary to develop new averaging operators to overcome the drawbacks of the existing averaging operators for dealing with fuzzy information retrieval.

In this paper, we present new averaging operators, called weighted power-mean averaging (WPMA) operators, based on the concept of the weighted power-mean [8] to deal with fuzzy information retrieval. We also prove that the proposed WPMA operators are "positively compensatory" operators. In [3] and [6], Kim et al. pointed out that operators which have the "positively compensatory" property could provide high retrieval effectiveness. The proposed WPMA operators are more flexible and more intelligent than the average operators presented in [7], [9], [10] and [11] for dealing with fuzzy information retrieval.

# 2 Preliminaries

In [8], the definition of the weighted power means is defined as follows:

**Definition 2.1:** Assume that  $\underline{a}$  and  $\underline{w}$  are two positive n-tuples and  $r \in R$ , then the *rth* power mean  $M_n^{[r]}(a, w)$  of  $\underline{a}$  with weight  $\underline{w}$  is define as follows:

$$M_{n}^{[r]}(\underline{a}, \underline{w}) = \left(\frac{1}{W_{n}} \sum_{i=1}^{n} a_{i}^{r} w_{i}\right)^{\frac{1}{r}}, \qquad (1)$$

where  $W_n = w_1 + w_2 + \cdots + w_n$ .

Since  $M_n^{[1]} = A_n$  and  $M_n^{[-1]} = H_n$ , in [8], Bullen et al. defined  $M_n^{[0]} = G_n$  and proved that the definition of the weighted power mean is reasonable. Thus, the weighted power means form a natural extension of elementary means. Furthermore, when  $r \to \infty$ ,  $M_n^{[r]}(\underline{a}, \underline{w}) = \max \underline{a}$ ; when  $r \to -\infty$ ,  $M_n^{[r]}(\underline{a}, \underline{w}) = \min \underline{a}$ .

In [3], Kim et al. pointed out that an information retrieval system based on the conventional fuzzy set model is defined by a quadruple <T, Q, D, F>, where

- (1) T is a set of index terms,  $T = \{t_1, t_2, \dots, t_m\}$ , where these index terms are used for representing queries and documents.
- (2) Q is a set of queries, where query  $q \in Q$  is a Boolean expression composed of index terms  $t_i$ ,

 $1 \le j \le m$ , and the logical operators "AND", "OR"

and "NOT".

(3) D is a set of documents,  $D = \{d_1, d_2, \dots, d_n\}$ , where each document  $d_i \in D$  is represented by  $((t_1, e_{i1}), (t_2, e_{i2}), \dots, (t_m, e_{im})), e_{ij}$  denotes the degree of strength of term  $t_j$  in document  $d_i$ ,  $e_{ii} \in [0, 1], 1 \le i \le n$ , and  $1 \le j \le m$ .

(4) 
$$F$$
 is an evaluation function,

$$F: \mathbf{D} \times \mathbf{Q} \to [0, 1], \tag{2}$$

which assigns a real value in the closed interval [0, 1] to each pair (d, q). It is a similarity measure between document *d* and query *q*.

From [12], we can see that the weight  $e_{ij}$  of term  $t_j$  in document  $d_i$  is determined either subjectively by domain experts or objectively by some algorithmic procedures, where  $1 \le i \le n$  and  $1 \le j \le m$ . One way for determining the degree of strength  $e_{ij}$  of term  $t_j$  in document  $d_i$  objectively is to consider the frequency of occurrence of index term  $t_j$  in document  $d_i$ .

# 3 Fuzzy Information Retrieval Based on the Weighted Power-Mean Averaging Operators

In this section, we present new averaging operators, called the Weighted Power-Mean Averaging (WPMA) operators, for fuzzy information retrieval, shown as follows:

$$F(d_i, \mathbf{q}_{AND}) = F(d_i, \mathbf{t}_1 \text{ AND } \mathbf{t}_2 \text{ AND } \cdots \text{ AND } \mathbf{t}_m)$$
$$= \left[\frac{1}{m^2} \sum_{k=1}^m (2m - 2k + 1) e^{*}_{ik} r^k\right]^{\frac{1}{r}}, \qquad (3)$$
$$F(d_i, \mathbf{q}_{OR}) = F(d_i, \mathbf{t}_1 \text{ OR } \mathbf{t}_2 \text{ OR } \cdots \text{ OR } \mathbf{t}_m)$$

$$= 1 - \left[\frac{1}{m^2} \sum_{k=1}^{m} (2k-1)(1-e^*_{ik})^r\right]^{\frac{1}{r}}, \qquad (4)$$

where  $r \in \{0.0001, 0.5\}$ ,  $e_{ij}$  denotes the degree of strength of term  $t_j$  in document  $d_i$ ,  $e^*_{ik}$  denotes the *kth* smallest value of  $e_{ij}$ ; the weight of the term in document  $d_i$  which is associated with the *kth* smallest value of  $e_{ij}$  in the AND query  $q_{AND}$  is 2m - 2k + 1; the weight of the term in document  $d_i$ which is associated with the *kth* smallest value of  $e_{ij}$ in the OR query  $q_{OR}$  is 2k - 1,  $1 \le i \le n$ ,  $1 \le j \le m$ ,  $1 \le k \le m$ ,  $F(d_i, q_{AND}) \in [0,1]$  and  $F(d_i, q_{OR}) \in [0,1]$ .

In the following, we use two cases to discuss how the proposed WPMA operators are controlled by a parameter *r*. Assume that there are four documents  $d_1, d_2, d_3$  and  $d_4$ , and assume that there are two queries

$$q_{1} \text{ and } q_{2}, \text{ shown as follows} :$$

$$d_{1} = \{(t_{1}, 0), (t_{2}, 0)\},$$

$$d_{2} = \{(t_{1}, 0), (t_{2}, 1)\},$$

$$d_{3} = \{(t_{1}, 1), (t_{2}, 0)\},$$

$$d_{4} = \{(t_{1}, 1), (t_{2}, 1)\},$$

$$q_{1} = t_{1} \text{ AND } t_{2},$$

$$q_{2} = t_{1} \text{ OR } t_{2}.$$

**Case 1**: If the parameter r = 0.0001, then the degree of satisfaction  $F(d_3, q_1)$  of the document  $d_3$  with respect to the query  $q_1$  and the degree of satisfaction  $F(d_2, q_2)$  of the document  $d_2$  with respect to the query  $q_2$  can be evaluated, shown as follows:

$$F(d_3, q_1) = \left[\frac{1}{4} \left(3 \times 0^{0.0001} + 1 \times 1^{0.0001}\right)\right]^{10000} = 0,$$
  

$$F(d_2, q_2) = 1 - \left[\frac{1}{4} \left(1 \times (1-0)^{0.0001} + 3 \times (1-1)^{0.0001}\right)\right]^{10000} = 1.$$
  
In the same way, we can calculate the values of  $F(d_1, q_1), F(d_2, q_1), F(d_4, q_1), F(d_1, q_2),$ 

 $F(d_3, q_2)$  and  $F(d_4, q_2)$ , respectively, as shown in Table 1. From Table 1, we can see that when r = 0.0001, the proposed WPMA operators become the traditional Boolean operators. The operator graph [10] of the proposed WPMA operators is shown in Fig. 3.

Table 1. Query Result of the Proposed WPMA Operators (when r = 0.0001)

Documents	Terms		Queries	
	$t_1$	$t_2$	$q_1 = t_1 \text{ AND t}_2$	$q_2 = t_1 \operatorname{OR} \mathbf{t}_2$
$d_1$	0	0	0	0
$d_2$	0	1	0	1
$d_3$	1	0	0	1
$d_4$	1	1	1	1

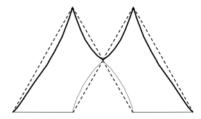


Fig. 3. The operator graphs of the proposed WPMA operators (when r = 0.0001).

**Case 2:** If the parameter r = 0.5, then the degree of satisfaction  $F(d_3, q_1)$  of the document  $d_3$  with respect to the query  $q_1$  and the degree of satisfaction  $F(d_2, q_2)$  of the document  $d_2$  with respect to the query  $q_2$  can be evaluated, shown as follows:

$$F(d_3, q_1) = \left[\frac{1}{4} \left(3 \times 0^{0.5} + 1 \times 1^{0.5}\right)\right]^2 = 0.063,$$
  

$$F(d_2, q_2) = 1 - \left[\frac{1}{4} \left(1 \times (1 - 0)^{0.5} + 3 \times (1 - 1)^{0.5}\right)\right]^2 = 0.938.$$
  
In the same way, we can calculate the values of  $F(d_1, q_1), F(d_2, q_1), F(d_3, q_2), F(d_4, q_1), F(d_1, q_2), F(d_3, q_3)$ 

 $F(d_3, q_2)$  and  $F(d_4, q_2)$ , respectively, as shown in Table 2. From Table 2, we can see that when r = 0.5, the proposed WPMA operators are compatible with the extended Boolean operators [11]. The operator graph of the proposed WPMA operators is shown in Fig. 4.

Table 2. Query result of the Proposed WPMA Operators (when r = 0.5)

Documents	Terms		Queries	
	$t_1$	$t_2$	$q_1 = t_1 \text{ AND t}_2$	$q_2 = t_1 \operatorname{OR} \mathbf{t}_2$
$d_1$	0	0	0	0
$d_2$	0	1	0.063	0.938
$d_3$	1	0	0.063	0.938
$d_4$	1	1	1	1

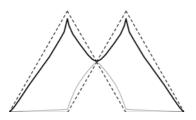


Fig. 4. The operator graph of the proposed WPMA operators (r = 0.5).

In [6], Lee pointed out that an operator which has the "*positively compensatory*" property could provide higher retrieval effectiveness. The "*positively compensatory*" operators are functions of the form  $p:[0,1]\times[0,1] \rightarrow [0,1]$ . They must satisfy the following two properties:

(1) p(x, x) = x; i.e., *p* is an idempotent function.

(2) Min(x, y) < p(x, y) < Max(x, y), where  $x \neq y$ .

In the following, we assume that  $0 \le x \le 1$ ,  $0 \le y \le 1$ , and prove that the proposed WPMA operators satisfy the following two properties.

**Property 3.1:**  $F(d, t_1 \text{ AND } t_2)$  and  $F(d, t_1 \text{ OR } t_2)$ are idempotent, where document  $d = \{(t_1, x), (t_2, x)\}, t_1$ and  $t_2$  are terms, and  $0 \le x \le 1$ .

**Proof:** Based on formula (3), we can see that

$$F(d, t_1 \text{ AND } t_2) = \left[\frac{1}{4}(3 \times x^r + 1 \times x^r)\right]^{\frac{1}{r}}$$
$$= \left[x^r\right]^{\frac{1}{r}}$$
$$= x.$$

Based on formula (4), we can see that

$$F(d, t_1 \text{ OR } t_2) = 1 - \left[\frac{1}{4}(1 \times (1 - x)^r + 3 \times (1 - x)^r)\right]^{\frac{1}{r}}$$
  
=  $1 - \left[(1 - x)^r\right]^{\frac{1}{r}}$   
=  $1 - (1 - x)$   
=  $x$ .

Thus, the proposed WPMA operators are idempotent.

Q.E.D.

**Property 3.2:** Assume that there is a document  $d = \{(t_1, x), (t_2, y)\}$ , where  $t_1$  and  $t_2$  are terms, and  $0 \le x \le 1$  and  $0 \le y \le 1$ . Then,

 $\begin{array}{ll} \textit{Min}(x, y) &< F(d, t_1 \text{ AND } t_2) &< F(d, t_1 \text{ OR } t_2) &< \\ \textit{Max}(x, y) \text{ , where } x \neq y \text{ .} \end{array}$ 

Proof:

(i) If x > y, then we can see that Min(x, y) = y and Max(x, y) = x. Furthermore, based on formula (3), we can see that

$$F(d, t_1 \operatorname{AND} t_2) = \left[\frac{1}{4}(3 \times y^r + 1 \times x^r)\right]^{\frac{1}{r}}$$
$$\Longrightarrow \left[\frac{1}{4}(3 \times y^r + 1 \times y^r)\right]^{\frac{1}{r}} < \left[\frac{1}{4}(3 \times y^r + 1 \times x^r)\right]^{\frac{1}{r}} < \left[\frac{1}{4}(3 \times x^r + 1 \times x^r)\right]^{\frac{1}{r}}$$
$$\Longrightarrow F(d, t_2 \operatorname{AND} t_2) < F(d, t_1 \operatorname{AND} t_2) < F(d, t_1 \operatorname{AND} t_1)$$

$$(by \text{ formula } 3)$$

$$v \leq E(d \ t, \text{ AND } t_{-}) \leq r$$

$$(by \text{ Property } 3 \ 1)$$

$$\implies Min(x, y) < F(d, t_1 AND t_2) < Max(x, y).$$

Based on formula on (4), we can see that

$$F(d, t_{1} \operatorname{OR} t_{2}) = 1 - \left[\frac{1}{4}(1 \times (1 - y)^{r} + 3 \times (1 - x)^{r})\right]^{\frac{1}{r}}$$

$$\implies 1 - \left[\frac{1}{4}(1 \times (1 - y)^{r} + 3 \times (1 - y)^{r})\right]^{\frac{1}{r}} < 1 - \left[\frac{1}{4}(1 \times (1 - y)^{r} + 3 \times (1 - x)^{r})\right]^{\frac{1}{r}} < 1 - \left[\frac{1}{4}(1 \times (1 - x)^{r} + 3 \times (1 - x)^{r})\right]^{\frac{1}{r}}$$

$$\implies F(d, t_{2} \operatorname{OR} t_{2}) < F(d, t_{1} \operatorname{OR} t_{2}) < F(d, t_{1} \operatorname{OR} t_{1}),$$

$$(by \text{ formula (4)})$$

$$\implies y < F(d, t_{1} \operatorname{OR} t_{2}) < x,$$

$$(by \text{ Property 3.1})$$

$$\implies Min(x, y) < F(d, t_{1} \operatorname{OR} t_{2}) < Max(x, y).$$

Then, we prove " $F(d, t_1 \text{ AND } t_2) < F(d, t_1 \text{ OR } t_2)$ " based on formulas (3) and (4), shown as follows:

$$F(d, t_1 \text{ AND } t_2) = \left(\frac{y^r + y^r + y^r + x^r}{4}\right)^{\frac{1}{r}} <$$

$$\begin{split} & \left(\frac{y+y+y+x}{4}\right) < \left(\frac{y+x+x+x}{4}\right) = \\ & 1 - \left(\frac{(1-y) + (1-x) + (1-x) + (1-x)}{4}\right) < \\ & 1 - \left(\frac{(1-y)^r + (1-x)^r + (1-x)^r + (1-x)^r}{4}\right)^{\frac{1}{r}} \\ & = F(d, \ t_1 \operatorname{OR} t_2). \end{split}$$

From the above discussions, we can see that  $Min(x, y) < F(d, t_1 \text{ AND } t_2) < F(d, t_1 \text{ OR } t_2) < Max(x, y)$ . (ii) In the same way, if x < y, we can get

 $F(d, t_1 \operatorname{AND} t_1) < F(d, t_1 \operatorname{AND} t_2) < F(d, t_2 \operatorname{AND} t_2),$ 

 $x < F(d, t_1 \text{ AND } t_2) < y,$   $F(d, t_1 \text{ OR } t_1) < F(d, t_1 \text{ OR } t_2) < F(d, t_2 \text{ OR } t_2),$   $F(d, t_1 \text{ OR } t_2) < F(d, t_2 \text{ OR } t_2),$   $F(d, t_1 \text{ OR } t_2) < y,$   $F(d, t_1 \text{ OR } t_2) < y,$  F(d

and  $F(d, t_1 \text{ AND } t_2) < F(d, t_1 \text{ OR } t_2)$ .

(by formulas (3) and (4))

Thus,  $Min(x, y) < F(d, t_1 \text{ AND } t_2) < F(d, t_1 \text{ OR } t_2) < Max(x, y)$ , where  $x \neq y$ . Q.E.D.

According to the above properties, we can see that the proposed WPMA operators have the "*positively compensatory*" property. Moreover, they have neither the "single operand dependent" property nor the "negatively compensatory" property. Therefore, they can overcome the drawback of the T-operators.

# 4 Conclusions

In this paper, we have presented the weighted power-mean averaging (WPMA) operators for fuzzy information retrieval. The proposed WPMA operators are more flexible and more intelligent than the averaging operators presented in [7], [9], [10] and [11] to deal with users' fuzzy queries for fuzzy information retrieval due to the fact that the proposed WPMA operators have the following advantages:

- (1) The proposed WPMA operators can overcome the drawbacks of the existing averaging operators for fuzzy information retrieval.
- (2) The retrieval results of the proposed WPMA operators are much closer to the intuition of the human being than the existing averaging operators.
- (3) We can easily determine appropriate values for the parameters of the proposed WPMA averaging operators. If we use the proposed WPMA operators to deal with traditional Boolean query processing for fuzzy information retrieval, then we can set the parameter r = 0.0001; if we use the proposed WPMA operators to deal with extended Boolean query processing for fuzzy information retrieval, then we can set the parameter r = 0.5.

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