



Correction to: Incremental maintenance of three-way regions with variations of objects and values in hybrid incomplete decision systems

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Correction to: Applied Intelligence (2022)

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The original version of this article unfortunately contained three mistakes due to the transformation document being edited through different formulas. The presentation of Definition 5, Theorem 3 and Table 3 were incorrect. The correct versions are given below:

Definition 5 Given a HIDS, for $\forall x_i, x_j \in U$ and $\exists \in C^I$, suppose that $f(x_i, a) = [l_i^a, h_i^a]$ and $f(x_j, a) = [l_j^a, h_j^a]$, the distance metric $d_a(x_i, x_j)$ of objects x_i, x_j pertaining to a can be depicted as follows:

$$d_a(x_i, x_j) = \begin{cases} 1 - \frac{|f(x_i, a) \cap f(x_j, a)|}{|f(x_i, a) \cup f(x_j, a)|} & a \in C^I, f(x_i, a) \neq' \wedge f(x_j, a) \neq' \\ 0 & a \in C^I, f(x_i, a) = ' \vee f(x_j, a) = ' \end{cases}$$

where $|f(x_i, a) \cap f(x_j, a)| = \left\{ \min\{h_i^a, h_j^a\} - \max\{l_i^a, l_j^a\} \right\}$
 $\min\{h_i^a, h_j^a\} - \max\{l_i^a, l_j^a\} > 0$
 $\min\{h_i^a, h_j^a\} - \max\{l_i^a, l_j^a\} < 0$, and $|f(x_i, a) \cup f(x_j, a)| = \max\{h_i^a, h_j^a\} - \min\{l_i^a, l_j^a\}$.

Theorem 3 Let δ -HIDS^(t) be a δ -HIDS at t , and let $\mathbf{M}_D^{(t)} = [d_{ir}^{(t)}]_{n \times s}$ and $\mathbf{M}_A^{(t)} = [m_{ij}^{(t)}]_{n \times n}$ be the decision matrix and the neighborhood relation matrix in regard to $A \subseteq C$ at t , respectively. When modifying the object set ΔV , adding the object set $\Delta^+ U$ at $t+1$ and deleting the object set $\Delta^- U$, let the decision matrix be $\mathbf{M}_D^{(t+1)} = [d_{ir}^{(t+1)}]_{(n+n'-p) \times s}$ and the neighborhood relation matrix be $\mathbf{M}_A^{(t+1)} = [m_{ij}^{(t+1)}]_{(n+n'-p) \times (n+n'-p)}$. The following properties hold:

(1) for $p + 1 \leq i \leq n$, $\omega_{ir}^{+1} = \sum_{j=p+1}^n \left(m_{ij}^{(t)} \oplus m_{ij}^{(t+1)} \right) \cdot d_{jr}^{(t+1)} \cdot m_{ij}^{(t+1)}$;

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Table 3 Comparison of the time complexities of static and incremental algorithms at $t+1$

Steps	MSTW	MITW-OV
Compute decision matrix	$O(n+n'-p)$	$O(n')$
Compute relation matrix	$O(m \times (n+n'-p)^2)$	$O((n-p)^2 \times (+s) + (n')^2 \times m)$
Compute intermediate matrix	$O(s \times (n+n'-p)^2)$	$O(s \times (n^2 + n' \times (n-p) + (n')^2))$
Compute column vector	$O(s \times (n+n'-p))$	$O(n+n'-p)$
Compute basic matrix	$O(s \times (n+n'-p))$	$O(s \times (n+n'-p))$
Compute three-way regions	$O(s \times (n+n'-p))$	$O(s \times (n+n'-p))$
Total	$O((m+s) \times (n+n'-p)^2 + (s+1) \times (n+n'-p))$	$O(n' + (n-p)^2 \times (m+s) + (n')^2 \times m + s \times (p^2 + n' \times (n-p) + (n')^2) + (s+1) \times (n+n'-p))$

- (2) for $p + 1 \leq i \leq n$, $\omega_{ir}^{-1} = \sum_{k=1}^n \left(\left(m_{ij}^{(t)} \oplus m_{ij}^{(t+1)} \right) \cdot d_{jr}^{(t+1)} \cdot m_{ij}^{(t)} \right)$;
- (3) for $p + 1 \leq i \leq n$, $\omega_{ir}^{-2} = \sum_{j=1}^p \left(m_{ij}^{(t+1)} \cdot d_{jr}^{(t+1)} \right)$;
- (4) for $p + 1 \leq i \leq n$, $\omega_{ir}^{+2} = \sum_{j=n+1}^{n+n'} \left(m_{ij}^{(t+1)} \cdot d_{jr}^{(t+1)} \right)$;
- (5) for $n + 1 \leq i \leq n + n'$, $\omega_{ir}^{+3} = \sum_{j=p+1}^{n+n'} \left(m_{ij}^{(t+1)} \cdot d_{jr}^{(t+1)} \right)$.

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The original article has been corrected.