Maximal rough neighborhoods with a medical application

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Abstract
In this paper, we focus on the main concepts of rough set theory induced from the idea of neighborhoods. First, we put forward new types of maximal neighborhoods (briefly, $\mathcal{M}_{\sigma}$-neighborhoods) and explore master properties. We also reveal their relationships with foregoing neighborhoods and specify the sufficient conditions to obtain some equivalences. Then, we apply $\mathcal{M}_{\sigma}$-neighborhoods to define $\mathcal{M}_{\sigma}$-lower and $\mathcal{M}_{\sigma}$-upper approximations and elucidate which one of Pawlak’s properties are preserved (evaporated) by these approximations. Moreover, we research $\mathcal{A}_{\mathcal{M}_{\sigma}}$-accuracy measures and prove that they keep the monotonic property under any arbitrary relation. We provide some comparisons that illustrate the best approximations and accuracy measures are obtained when $\sigma = \langle i \rangle$. To show the importance of $\mathcal{M}_{\sigma}$-neighborhoods, we present a medical application of them in classifying individuals of a specific facility in terms of their infection with COVID-19. Finally, we scrutinize the strengths and limitations of the followed technique in this manuscript compared with the previous ones.

Keywords $\mathcal{N}_{\sigma}$-neighborhood · $\mathcal{M}_{\sigma}$-neighborhood · Lower and upper approximations · Accuracy measure · Rough set

1 Introduction

Pawlak (1982, 1991) familiarized the idea of rough sets theory as a mathematical tool to address incomplete information systems and uncertainty. The essential notions in this theory such as upper and lower approximations and accuracy measures are firstly characterized using an equivalence relation. In past few years, many novel models have been espoused, by changing or relaxing some underlying conditions.

After the superb start of studying rough sets using right and left neighborhoods (Yao 1996, 1998), some rough set models have been established such as minimal left neighborhoods (Allam et al. 2006) and minimal right neighborhoods (Allam et al. 2005). Then, intersection and union neighborhoods, intersection minimal and union minimal neighborhoods were introduced in (Abd El-Monsef et al. 2014). Generating right neighborhoods from finite class of arbitrary relations was investigated in (Abu-Donia 2008). Mareay (2016) studied four types of neighborhoods using the equality relation between $\mathcal{N}_{\sigma}$-neighborhoods. Sun et al. (2019) defined the dual idea of neighborhood systems under the name of remote neighborhood systems. Recently, Al-shami et al. (2021b) have applied the intersection operation between $\mathcal{N}_{\sigma}$-neighborhoods to define $\mathcal{E}_{\sigma}$-neighborhoods. Also, novel types of neighborhoods called $\mathcal{C}_{\sigma}$-neighborhoods (Al-shami 2021a) and $\mathcal{S}_{\sigma}$-neighborhoods (Al-shami and Ciucci 2022) have been initiated using the inclusion relation between $\mathcal{N}_{\sigma}$-neighborhoods.

Following this line of study, we display the concept of $\mathcal{M}_{\sigma}$-neighborhood systems. The motivations are twofold. From a theoretical standpoint, the lower and upper approximations satisfy most of the properties of the standard Pawlak model, and the accuracy measures preserve the monotonic property under any arbitrary relation as illustrated in Proposition 25 and Corollary 26. Whereas, this property is losing or keeping under strict conditions in some foregoing methods such as those introduced by Mareay (2016) and Al-shami (2021a). From an application prospective, $\mathcal{M}_{\sigma}$-neighborhood systems can be used in concrete situations as we explain in Sect. 5. Also, $\mathcal{M}_{\sigma}$- and $\mathcal{M}_{\langle \sigma \rangle}$-neighborhoods improve the approximations and accuracy measures more than $\mathcal{M}_{\sigma}$-neighborhoods given by Dai et al. (2018).

The arrangement of the rest of this manuscript is as follows. Sect. 2 reviews the literature needed to understand the article’s results. In Sect. 3, we embraced a new class...
of neighborhoods system called $\mathcal{M}_\sigma$-neighborhoods, where $\sigma \in \{l, r, \langle l \rangle, i, u, \langle i \rangle, \langle u \rangle\}$. In Sect. 4, we mainly aim to introduce different types of approximations and accuracy measures. We compare between them and elucidate that the best case given by (i). In Sect. 5, we propose an algorithm to classify the individuals who work in a specific facility with respect to corona-virus COVID-19 infection. Finally, discussions and conclusions are provided in Sects. 6 and 7, respectively.

## 2 Preliminaries

In this section, we provide some definitions and results of different types of neighborhoods to make this manuscript self-contained and also to rationalize the need to introduce the concept of $\mathcal{M}_\sigma$-neighborhoods.

**Definition 1** (see, Abd El-Monsef et al. 2014) A binary relation $\theta$ on $B$ is a subset of $B \times B$. We write $a \theta b$ if $(a, b) \in \theta$. $\theta$ is called reflexive (resp., symmetric, transitive) if $a \theta a$ for each $a \in B$ (resp., $a \theta b \iff b \theta a$, $a \theta b$ whenever $a \theta b$ and $b \theta c$). If $\theta$ is reflexive, symmetric and transitive, then it called an equivalence relation. Also, we call $\theta$ a comparable relation if $a \theta b$ or $b \theta a$ for each $a, b \in B$.

**Definition 2** (Pawlak 1982, 1991) Let $\theta$ be an equivalence relation on $B$. We associate every $W \subseteq B$ with two subsets:

$\theta(W) = \bigcup \{ E \in B/\theta : E \subseteq W \}$ (It is called the lower approximation of $W$).

$\overline{\theta}(W) = \bigcup \{ E \in B/\theta : E \cap W \neq \emptyset \}$ (It is called the upper approximation of $W$).

The following result lists the core properties of Pawlak’s rough set model. In the previous generalized rough set models, the validity of these properties was examined. Of course, some of them are evaporated in these models; however, keeping as many as possible of these properties is considered as an advantage of the model.

**Proposition 1** (Pawlak 1982, 1991) Let $V, W$ be two subsets of $B$. The following properties are satisfied if $\theta$ is an equivalence relation on $B$.

| (L1) | $\theta(V) \subseteq V$ |
| (U1) | $\overline{\theta}(V) \subseteq \overline{\theta}(V)$ |
| (L2) | $\theta(\emptyset) = \emptyset$ |
| (U2) | $\overline{\theta}(\emptyset) = \emptyset$ |
| (L3) | $\theta(B) = B$ |
| (U3) | $\overline{\theta}(B) = B$ |
| (L4) | If $V \subseteq W$, then $\theta(V) \subseteq \theta(W)$ |
| (U4) | If $V \subseteq W$, then $\overline{\theta}(V) \subseteq \overline{\theta}(W)$ |

We can also numerically describe rough sets by using the following two measures.

**Definition 3** (Pawlak 1982, 1991) Let $\theta$ be an equivalence relation on $B$. The $A$-accuracy and $R$-roughness measures of a nonempty subset $W \subseteq B$ with respect to an equivalence relation $\theta$ are respectively defined by

$\mathcal{A}(W) = \frac{| \theta(W) |}{| \overline{\theta}(W) |}$.

$\mathcal{R}(W) = 1 - \mathcal{A}(W)$.

The equivalence relations are a strict condition and cannot be satisfied in many circumstances. So that, the classical model has been extended using weaker relations than the equivalence.

**Definition 4** (Abd El-Monsef et al. 2014; Abo-Tabl 2011; Allam et al. 2005, 2006; Yao 1996, 1998) The $\mathcal{N}_\sigma$-neighborhoods of an $a \in B$ (denoted by $\mathcal{N}_\sigma(a)$) are defined under an arbitrary relation $\theta$ on $B$, where $\sigma \in \{r, l, \langle r \rangle, \langle l \rangle, i, u, \langle i \rangle, \langle u \rangle\}$, as follows.

(i) $\mathcal{N}_l(a) = \{ b \in B : a \theta b \}$.  

(ii) $\mathcal{N}_r(a) = \{ b \in B : b \theta a \}$.

$\mathcal{N}_\langle r \rangle(a) = \left\{ \bigcup_{a \in \mathcal{N}_\perp(\beta)} \mathcal{N}_\langle r \rangle(\beta) : \exists \mathcal{N}_\langle r \rangle(\beta) \text{ containing } a \right\}$

$\mathcal{N}_\langle l \rangle(a) = \left\{ \bigcup_{a \in \mathcal{N}_\perp(\beta)} \mathcal{N}_\langle l \rangle(\beta) : \exists \mathcal{N}_\langle l \rangle(\beta) \text{ containing } a \right\}$

Henceforth, we consider $\sigma \in \{r, l, \langle r \rangle, \langle l \rangle, i, u, \langle i \rangle, \langle u \rangle\}$, unless stated otherwise.
Definition 5 (Abd El-Monsef et al. 2014) Let $\theta$ be an arbitrary relation on $B$ and $\zeta_\theta$ be a map from $B$ to $2^B$ which associated each $a \in B$ with its $\sigma$-neighborhood in $2^B$. We call the triple $(B, \theta, \zeta_\theta)$ a $\sigma$-neighborhood space (briefly, $\sigma$-NS).

Definition 6 (Al-shami et al. 2021b) The $\mathcal{E}$-neighborhoods of an $a \in B$ (briefly, $\mathcal{E}_\sigma(a)$) are defined for each $\sigma$ under an arbitrary relation $\theta$ on $B$ as follows.

(i) $\mathcal{E}_\sigma(a) = \{b \in B : N^\sigma_{\mathcal{E}}(b) \cap N^\sigma_{\mathcal{E}}(a) \neq \emptyset\}$.
(ii) $\mathcal{E}_\sigma(a) = \{b \in B : N^\sigma_{\mathcal{E}}(b) \cap N^\sigma_{\mathcal{E}}(a) \neq \emptyset\}$.
(iii) $\mathcal{E}_\sigma(a) = \mathcal{E}_\sigma(a) \cap \mathcal{E}_\sigma(a)$. (iv) $\mathcal{E}_\sigma(a) = \mathcal{E}_\sigma(a) \cup \mathcal{E}_\sigma(a)$. (v) $\mathcal{E}_\sigma(a) = \{b \in B : \mathcal{E}_\sigma(b) \cap \mathcal{E}_\sigma(a) \neq \emptyset\}$. (vi) $\mathcal{E}_\sigma(a) = \{b \in B : N^\sigma_{\mathcal{E}}(b) \cap N^\sigma_{\mathcal{E}}(a) \neq \emptyset\}$. (vii) $\mathcal{E}_\sigma(a) = \mathcal{E}_\sigma(a) \cap \mathcal{E}_\sigma(a)$. (viii) $\mathcal{E}_\sigma(a) = \mathcal{E}_\sigma(a) \cup \mathcal{E}_\sigma(a)$.}

Definition 7 (Al-shami 2021a) The $\mathcal{C}$-neighborhoods of an $a \in B$ (briefly, $\mathcal{C}_\sigma(a)$) are defined for each $\sigma$ under an arbitrary relation $\theta$ on $B$ as follows.

(i) $\mathcal{C}_\sigma(a) = \{b \in B : N^\sigma_{\mathcal{C}}(b) \subseteq N^\sigma_{\mathcal{C}}(a)\}$.
(ii) $\mathcal{C}_\sigma(a) = \{b \in B : N^\sigma_{\mathcal{C}}(b) \subseteq N^\sigma_{\mathcal{C}}(a)\}$.
(iii) $\mathcal{C}_\sigma(a) = \mathcal{C}_\sigma(a) \cap \mathcal{C}_\sigma(a)$.
(iv) $\mathcal{C}_\sigma(a) = \mathcal{C}_\sigma(a) \cup \mathcal{C}_\sigma(a)$.
(v) $\mathcal{C}_\sigma(a) = \{b \in B : N^\sigma_{\mathcal{C}}(b) \subseteq N^\sigma_{\mathcal{C}}(a)\}$.
(vi) $\mathcal{C}_\sigma(a) = \{b \in B : N^\sigma_{\mathcal{C}}(b) \subseteq N^\sigma_{\mathcal{C}}(a)\}$.
(vii) $\mathcal{C}_\sigma(a) = \mathcal{C}_\sigma(a) \cap \mathcal{C}_\sigma(a)$.
(viii) $\mathcal{C}_\sigma(a) = \mathcal{C}_\sigma(a) \cup \mathcal{C}_\sigma(a)$.

The foregoing types of neighborhoods were applied to define new types of lower and upper approximations and accuracy (roughness) measures. Comparisons between them in terms of improving the approximations and increasing the accuracy measures were done in various papers.

Definition 8 (Abd El-Monsef et al. 2014; Abo-Tabl 2011; Allam et al. 2005, 2006; Al-shami 2021a; Al-shami et al. 2021b; Yao 1996, 1998) Let $\theta$ be an arbitrary relation on $B$ and $\mathcal{K} \in \{\mathcal{N}, \mathcal{E}, \mathcal{C}\}$. We define lower and upper approximations of each $W \subseteq B$ with respect to the types of neighborhoods as follows.

$\mathcal{H}_{\mathcal{K},\sigma}(W) = \{a \in B : K_{\sigma}(a) \subseteq W\}$, and $\mathcal{H}^{\prime\sigma}(W) = \{a \in B : K_{\sigma}(a) \cap W = \emptyset\}$. $\mathcal{A}_{\mathcal{K},\sigma}(W) = |\mathcal{H}_{\mathcal{K},\sigma}(W) \cap W|$, and $\mathcal{A}^{\prime\sigma}(W) = |\mathcal{H}^{\prime\sigma}(W)|$.

Definition 9 (Abd El-Monsef et al. 2014; Allam et al. 2005, 2006; Al-shami 2021a; Al-shami et al. 2021b; Yao 1996, 1998) Let $\theta$ be an arbitrary relation on $B$. The $\mathcal{A}_{\mathcal{K},\sigma}$-accuracy and $\mathcal{R}_{\mathcal{K},\sigma}$-roughness measures of a nonempty subset $W$ of $B$ with respect to $\theta$ are respectively defined by

$\mathcal{A}_{\mathcal{K},\sigma}(W) = |\mathcal{H}_{\mathcal{K},\sigma}(W) \cap W|$, and $\mathcal{A}^{\prime\sigma}(W) = |\mathcal{H}^{\prime\sigma}(W)|$.

$\mathcal{R}_{\mathcal{K},\sigma}(W) = 1 - \mathcal{A}_{\mathcal{K},\sigma}(W)$.

3 Maximal neighborhoods system

In this section, we introduce the concept of maximal neighborhoods of an element with respect to any binary relation. We explore their main properties and determine the conditions under which some of them are identical. We compare between them as well as compare them with the previous ones.

Definition 10 (see, (Dai et al. 2018)) Let $\mathcal{K} \in \{\mathcal{N}, \mathcal{E}, \mathcal{C}\}$ and consider $\theta_1$ and $\theta_2$ are two relations on $B$ such that $\theta_1 \subseteq \theta_2$. We say that the approximations induced from $\mathcal{K}$-neighborhoods have the property of monotonicity-accuracy (resp., monotonicity-roughness) if $\mathcal{A}_{\mathcal{K},\sigma}(W) \geq \mathcal{A}_{\mathcal{K},\sigma}(W)$ (resp., $\mathcal{R}_{\mathcal{K},\sigma}(W) \leq \mathcal{R}_{\mathcal{K},\sigma}(W)$).

Definition 11 (Dai et al. 2018) The maximal right neighborhood of an $a \in B$ (briefly, $\mathcal{M}_\sigma(a)$) is defined under an arbitrary relation $\theta$ on $B$ as follows.

$\mathcal{M}_\sigma(a) = \bigcup_{a \in \mathcal{N}_{\mathcal{E}}(b)} \mathcal{N}_{\mathcal{E}}(b)$.
Proposition 3 Let \((B, \theta, \zeta_\sigma)\) be a \(\sigma\)-NS. Then \(\beta \in \mathcal{M}_\sigma(\alpha)\) iff \(\alpha \in \mathcal{M}_\sigma(\beta)\) for each \(\sigma \in \{r, l, i\}\).

**Proof** Let \(\beta \in \mathcal{M}_\sigma(\alpha)\). Then there exists \(x \in B\) such that \(\beta \in N_\sigma(x)\), where \(\alpha \in N_\sigma(x)\) as well. This automatically means that \(\alpha \in \mathcal{M}_\sigma(\beta)\).

Similarly, one can prove the proposition when \(\sigma = l\). \(\square\)

Corollary 4 Let \((B, \theta, \zeta_\sigma)\) be a \(\sigma\)-NS. Then \(\beta \in \mathcal{M}_\sigma(\alpha)\) iff \(\alpha \in \mathcal{M}_\sigma(\beta)\) for each \(\sigma \in \{i, u\}\).

Proposition 5 Let \((B, \theta, \zeta_\sigma)\) be a \(\sigma\)-NS such that \(\theta\) is symmetric and transitive. If \(\alpha, \beta \in N_\sigma(x)\), then \(\mathcal{M}_\sigma(\alpha) = \mathcal{M}_\sigma(\beta)\) for \(\sigma \in \{r, l, i, u\}\).

**Proof** Consider \(\sigma = r\) and let \(\alpha, \beta \in N_\sigma(x)\). Then \(x \theta \alpha\) and \(x \theta \beta\). Since \(\theta\) is symmetric and transitive, we obtain \(\alpha \theta \beta\) and \(\beta \theta \alpha\). Now, suppose that \(x \in \mathcal{M}_\sigma(\alpha)\). Then there exists \(y \in B\) such that \(y \in N_\sigma(y)\), where \(x \in N_\sigma(y)\) as well. This implies that \(x \theta \zeta\). It follows from that \(x \theta \alpha\) and \(\theta\) is transitive that \(x \theta \beta\). Therefore, \(N_\sigma(y)\) means that \(x \in \mathcal{M}_\sigma(y)\). Thus, \(\mathcal{M}_\sigma(\alpha) \subseteq \mathcal{M}_\sigma(\beta)\). Similarly, we prove that \(\mathcal{M}_\sigma(\beta) \subseteq \mathcal{M}_\sigma(\alpha)\).

Hence, the coveted result is obtained.

The other cases are proved in a similar way. \(\square\)

Now, we show the relationships between the different types of \(\mathcal{M}_\sigma\)-neighborhoods as well as we determine their relationships with some neighborhoods already introduced in the literature.

Proposition 6 Let \((B, \theta, \zeta_\sigma)\) be a \(\sigma\)-NS and \(\alpha \in B\). Then

(i) \(\mathcal{M}_\sigma(\alpha) \subseteq \mathcal{M}_\sigma(\alpha) \cup \mathcal{M}_\sigma(\alpha) \subseteq \mathcal{M}_\sigma(\alpha)\).

(ii) \(\mathcal{M}_\sigma(\alpha) \subseteq \mathcal{M}_\sigma(\alpha) \cup \mathcal{M}_\sigma(\alpha) \subseteq \mathcal{M}_\sigma(\alpha)\).

(iii) \(\mathcal{M}_\sigma(\alpha) \subseteq \mathcal{M}_\sigma(\alpha)\) for all \(\sigma \in \{r, l, i, u\}\).

(iv) If \(\theta\) is symmetric and transitive, then \(\mathcal{M}_\sigma(\alpha) = \mathcal{M}_\sigma(\alpha)\) for all \(\sigma \in \{r, l, i, u\}\).

**Proof** The proofs of (i) and (ii) come from the fact that \(N_\sigma(\alpha) \subseteq N_\sigma(\alpha) \cup N_\sigma(\alpha) \subseteq N_\sigma(\alpha)\).

To prove (iii), if \(\mathcal{M}_\sigma(\alpha) = \emptyset\), then the proof is trivial. So, suppose that \(\emptyset \in \mathcal{M}_\sigma(\alpha)\). Then there exists \(x \in B\) such that \(\alpha, \beta \in \mathcal{M}_\sigma(x)\). It follows from Proposition 3 that \(x \in \mathcal{M}_\sigma(\alpha)\).

According to Proposition 2, \(\alpha \in \mathcal{M}_\sigma(\alpha)\). This implies that \(\emptyset \in \mathcal{M}_\sigma(\alpha)\). This proves that \(\mathcal{M}_\sigma(\alpha) \subseteq \mathcal{M}_\sigma(\alpha)\). Following a similar manner, one can prove the other cases of \(\sigma\).

To prove (iv), if \(\mathcal{M}_\sigma(\alpha) = \emptyset\), then the proof is trivial. So, suppose that \(\beta \in \mathcal{M}_\sigma(\alpha)\). Then there exists \(x \in B\) such that \(\alpha, \beta \in N_\sigma(x)\). Since \(\theta\) is symmetric and transitive, it follows from Proposition 5 that
\[ M_\sigma(\alpha) = M_\theta(\beta) \]  

Suppose that \( \beta \notin M_\theta(\alpha) \). Then there exists \( y \in B \) such that \( \alpha \in M_\theta(y) \) and \( \beta \notin M_\theta(y) \). According to Proposition 5, \( y \in M_\sigma(\alpha) \). It follows from equality (1) that \( y \in M_\theta(\beta) \) as well. This means that \( \beta \in M_\sigma(y) \). But this contradicts assumption. Therefore, \( \beta \in M_\theta(\alpha) \). Thus, \( M_\sigma(\alpha) \subseteq M_\theta(\alpha) \). The side \( M_\sigma(\alpha) \subseteq M_\theta(\alpha) \) is proved in (iii). Hence, the proof is complete. \( \square \)

The following corollary gives one of the unique characterizations of \( M_\sigma \)-neighborhoods which is the determination of the smallest and largest one of neighborhoods from all \( \sigma \). The previous type of neighborhoods do not have this characterization; their determination is limited on disjoint two sets \( \{r, l, i, u\} \) and \( \{(r), (l), (i), (u)\} \).

**Corollary 7** Let \((B, \theta, \zeta_\sigma)\) be a \( \sigma \)-NS and \( \alpha \in B \). Then \( M_{(r)}(\alpha) \subseteq M_\sigma(\alpha) \subseteq M_{(i)}(\alpha) \) for all \( \sigma \in \{r, l, i, u\} \).

The converses of (i), (ii) and (iv) in the Proposition 6 fail as illustrated in Example 1. To show the converse of (iii) need not be true, we provide the next example.

**Example 2** Consider \( \theta = \{(\delta, \delta), (\alpha, \beta), (\beta, \alpha), (\delta, \beta), (\beta, \delta)\} \) is a binary relation on \( B = \{\alpha, \beta, \delta\} \). Then we calculate the maximal neighborhoods of each element of \( B \) in the Table 2. Since \( \theta \) is symmetric, it is sufficient to compare between them when \( \sigma = r \).

**Proposition 8** Let \((B, \theta, \zeta_\sigma)\) be a \( \sigma \)-NS and \( \alpha \in B \). Then

(i) If \( \theta \) is symmetric, then \( M_\sigma(\alpha) = M_\delta(\alpha) = M_\mu(\alpha) \) and \( M_{(r)}(\alpha) = M_{(l)}(\alpha) = M_{(i)}(\alpha) = M_{(u)}(\alpha) \).

(ii) If \( \theta \) is an equivalence, then all kinds of maximal neighborhoods are equal.

**Proof**

(i) : Since \( \theta \) is symmetric, then \( N_\sigma(\alpha) = N_\delta(\alpha) \). So that, \( M_\sigma(\alpha) = M_\delta(\alpha) = M_\mu(\alpha) = M_\mu(\alpha) \). Thus, \( M_{(r)}(\alpha) = M_{(l)}(\alpha) = M_{(i)}(\alpha) = M_{(u)}(\alpha) \).

(ii) : If \( \theta \) is an equivalence, then all kinds of \( N_\sigma \)-neighborhoods of any element are equal. Hence, all kinds of \( M_\sigma \)-neighborhoods of any element are also equal. \( \square \)

In the next results, we scrutinize the interrelations between \( M_\sigma \)-neighborhoods and some of the previous ones.

**Proposition 9** Let \((B, \theta, \zeta_\sigma)\) be a \( \sigma \)-NS and \( \alpha \in B \). Then

(i) If \( \theta \) is reflexive, then \( N_\sigma(\alpha) \subseteq M_\sigma(\alpha) \) and \( \zeta_\sigma(\alpha) \subseteq M_\sigma(\alpha) \) for each \( \sigma \).

(ii) \( E_\sigma(\alpha) \subseteq M_\sigma(\alpha) \) and \( E_\sigma(\alpha) \subseteq M_\sigma(\alpha) \) for each \( \alpha \in B \).

**Proof**

(i) : It is clear that \( N_\sigma(\alpha) \subseteq M_\sigma(\alpha) \) under a reflexivity condition. Let \( \beta \in \zeta_\sigma(\alpha) \). Then \( N_\sigma(\beta) \subseteq N_\sigma(\alpha) \). Now, \( \beta \in N_\sigma(\beta) \cap N_\sigma(\alpha) \) and \( \alpha \in N_\sigma(\alpha) \) because \( \theta \) is reflexive. This means that \( \beta \in M_\sigma(\alpha) \). Thus, \( \zeta_\sigma(\alpha) \subseteq M_\sigma(\alpha) \), as required.

(ii) : Let \( \beta \in E_\sigma(\alpha) \). Then there exists \( y \in B \) such that \( y \in N_\sigma(\beta) \cap N_\sigma(\alpha) \). Therefore, \( \alpha \) and \( \beta \) are members of \( N_\sigma(\alpha) \). Hence, \( E_\sigma(\alpha) \subseteq M_\sigma(\alpha) \). In a similar manner, one can prove that \( E_\sigma(\alpha) \subseteq M_\sigma(\alpha) \).

**Corollary 10** Let \((B, \theta, \zeta_\sigma)\) be a \( \sigma \)-NS and \( \alpha \in B \). Then

(i) \( E_\sigma(\alpha) \subseteq M_\sigma(\alpha) \) for each \( \sigma \in \{i, u\} \)

(ii) If \( \theta \) is symmetry, then \( E_\sigma(\alpha) = M_\sigma(\alpha) \) for each \( \sigma \in \{r, l, i, u\} \).

**Proof**

(i) : Obvious.

(ii) : It follows from the above proposition that \( E_\sigma(\alpha) \subseteq M_\sigma(\alpha) \). Conversely, let \( \beta \in M_\sigma(\alpha) \). Then there exists \( y \in B \) such that \( \alpha \in N_\sigma(y) \) and \( \beta \in N_\sigma(y) \). Since \( \theta \) is symmetric, \( \gamma \in N_\sigma(\beta) \cap N_\sigma(\alpha) \). Therefore, \( \beta \in E_\sigma(\alpha) \). Thus, \( M_\sigma(\alpha) \subseteq E_\sigma(\alpha) \). Hence, we obtain the desired result.

**Proposition 11** Let \((B, \theta, \zeta_\sigma)\) be a \( \sigma \)-NS. Then the equality

\[ N_\sigma(\alpha) = E_\sigma(\alpha) = \zeta_\sigma(\alpha) = M_\sigma(\alpha) \]

holds for each \( \sigma \) provided that \( \theta \) is an equivalence relation.
Proof The proof comes from the fact that every $N_{\sigma}$-neighbourhood forms a partition of $B$ under an equivalence relation.

Remark 1 If we want to compute $M_{\sigma}(a)$ from two different $\sigma$-NSs $(B, \theta_1, \zeta_\sigma)$ and $(B, \theta_2, \zeta_\sigma)$ we write $M_{\sigma}(a)$ and $M_{\sigma}(a)$.

Proposition 12 Let $(B, \theta_1, \zeta_\sigma)$ and $(B, \theta_2, \zeta_\sigma)$ be two $\sigma$-NSs such that $\theta_1 \subseteq \theta_2$. Then $M_{1\sigma}(a) \subseteq M_{2\sigma}(a)$ for each $a \in B$ and $\sigma \in \{r, l, i, u\}$.

Proof Let $\sigma = u$. Suppose that $\beta \in M_{1\sigma}(a)$. Then $\beta \in M_{1\sigma}(a)$ or $\beta \in M_{1\sigma}(a)$. Say, $\beta \in M_{1\sigma}(a)$. Then there exists $y \in B$ such that $\alpha, \beta \in N_{1\sigma}(y)$. Since $\theta_1 \subseteq \theta_2$, $N_{1\sigma}(y) \subseteq N_{2\sigma}(y)$. This implies that $\beta \in M_{2\sigma}(a)$. Hence, $M_{1\sigma}(a) \subseteq M_{2\sigma}(a)$, as required. Following similar technique, one can prove the other cases. □

Proposition 13 Let $(B, \theta, \zeta_\sigma)$ be a $\sigma$-NS. Then.

(i) If $x \in B$ such that $x N_{\sigma}$ for each $a \in B$, then $M_{\sigma}(a) = B$ for each $a \in B$ and each $\sigma \in \{r, u, (r), (u)\}$. Moreover, $M_{\sigma}(a) = M_{\sigma}(a)$ and $M_{(r)}(a) = M_{(r)}(a)$ for each $a \in B$.

(ii) If $\theta$ is comparable, then $M_{\sigma}(a) = B$ or $B \setminus \{a\}$ for each $a \in B$.

Proof (i): Since $x N_{\sigma}$ for each $a \in B$, $N_{\sigma}(x) = B$. Therefore, $M_{\sigma}(a) = B$. Now, $a \in M_{\sigma}(a)$ which implies that $M_{\sigma}(a) = B$. This automatically leads to that this result holds for each $\sigma \in \{u, (r), (u)\}$. Also, it is clear that $M_{\sigma}(a) \subseteq B = M_{\sigma}(a)$ and $M_{(r)}(a) \subseteq B = M_{(r)}(a)$. Hence, $M_{\sigma}(a) = M_{\sigma}(a)$ and $M_{(r)}(a) = M_{(r)}(a)$ for each $a \in B$.

Proposition 14 Let $(B, \theta, \zeta_\sigma)$ be a $\sigma$-NS. If $x \in B$ such that $a N_{\sigma}$ for each $a \in B$, then $M_{\sigma}(a) = B$ for each $a \in B$ and each $\sigma \in \{l, u, (l), (u)\}$. Moreover, $M_{\sigma}(a) = M_{\sigma}(a)$ and $M_{(l)}(a) = M_{(l)}(a)$ for each $a \in B$.

Proof (i): Since $x N_{\sigma}$ for each $a \in B$, $N_{\sigma}(x) = B$. Therefore, $M_{\sigma}(a) = B$. Now, $a \in M_{\sigma}(a)$ which implies that $M_{\sigma}(a) = B$. This automatically leads to that this result holds for each $\sigma \in \{u, (l), (u)\}$. Also, it is clear that $M_{\sigma}(a) \subseteq B = M_{\sigma}(a)$ and $M_{(l)}(a) \subseteq B = M_{(l)}(a)$. Hence, $M_{\sigma}(a) = M_{\sigma}(a)$ and $M_{(l)}(a) = M_{(l)}(a)$ for each $a \in B$.

4 Novel types of rough set models based on $M_{\sigma}$-neighborhoods

In this section, we introduce two new approximations called $M_{\sigma}$-lower and $M_{\sigma}$-upper approximations which we utilize to define new regions and accuracy measures of a set. We elucidate that $M_{\sigma}$-neighborhood produces the best approximations and highest accuracy measures. Illustrative examples are provided.

Definition 13 Let $(B, \theta, \zeta_\sigma)$ be a $\sigma$-NS. We associate a set $W$ with two approximations ($H_{M_\sigma}(W), H_{M_\sigma}(W)$) defined as follows.

- $H_{M_\sigma}(W) = \{a \in B : M_{\sigma}(a) \subseteq W\}$
- $H_{M_\sigma}(W) = \{a \in B : M_{\sigma}(a) \cap W \neq \emptyset\}$

Definition 14 The $M_{\sigma}$-boundary, $M_{\sigma}$-positive, and $M_{\sigma}$-negative regions of a subset $W$ of a $\sigma$-NS $(B, \theta, \zeta_\sigma)$ are respectively given by

- $B_{M_\sigma}(W) = H_{M_\sigma}(W) \setminus H_{M_\sigma}(W)$
- $P_{M_\sigma}(W) = H_{M_\sigma}(W)$
- $N_{M_\sigma}(W) = B \setminus H_{M_\sigma}(W)$

Definition 15 The $M_{\sigma}$-accuracy and $M_{\sigma}$-roughness measures of a subset $W \neq \emptyset$ of a $\sigma$-NS $(B, \theta, \zeta_\sigma)$ are respectively given by

- $A_{M_\sigma}(W) = \frac{|H_{M_\sigma}(W) \cap W|}{|H_{M_\sigma}(W) \cup W|}$
- $R_{M_\sigma}(W) = 1 - A_{M_\sigma}(W)$

We offer the next example to explain how the approximations, boundary regions, and accuracy measures defined above are computed for all $\sigma$.

Example 3 In Example 1, consider $W = \{a, \delta\}$ as a subset of a $\sigma$-NS $(B, \theta, \zeta_\sigma)$. We have the following computations.

(i) If $\sigma \in \{r, (r)\}$, then $H_{M_\sigma}(W) = \{a\}$, $H_{M_\sigma}(W) = \{\beta, \delta\}, B_{M_\sigma} = \{\beta, \delta\}$ and $A_{M_\sigma}(W) = \frac{1}{3}$.

(ii) If $\sigma \in \{l, (l)\}$, then $H_{M_\sigma}(W) = \{\delta, \gamma\}$, $H_{M_\sigma}(W) = \{a, \beta, \delta\}, B_{M_\sigma} = \{a, \beta\}$ and $A_{M_\sigma}(W) = \frac{1}{3}$.

(iii) If $\sigma \in \{i, (i)\}$, then $H_{M_\sigma}(W) = \{a, \delta, \gamma\}$, $H_{M_\sigma}(W) = \{\delta\}, B_{M_\sigma} = \emptyset$ and $A_{M_\sigma}(W) = 1$.

(iv) If $\sigma \in \{u, (u)\}$, then $H_{M_\sigma}(W) = \emptyset$, $H_{M_\sigma}(W) = \{a, \beta, \delta\}$, $B_{M_\sigma} = \{a, \beta, \delta\}$ and $A_{M_\sigma}(W) = 0$. 

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In the following four results, we examine which one of Pawlak's properties are preserved by $M_\sigma$-lower and $M^\sigma$-upper approximations.

**Theorem 15** Let $(B, \theta, \zeta_\sigma)$ be a $\sigma$-NS and $V, W \subseteq B$. The next statements hold true.

(i) $\emptyset \subseteq \mathcal{H}_{M_\sigma}(\emptyset)$.
(ii) $\mathcal{H}_{M_\sigma}(B) = B$.
(iii) If $V \subseteq W$, then $\mathcal{H}_{M_\sigma}(V) \subseteq \mathcal{H}_{M_\sigma}(W)$.
(iv) $\mathcal{H}_{M_\sigma}(V \cap W) = \mathcal{H}_{M_\sigma}(V) \cap \mathcal{H}_{M_\sigma}(W)$.
(v) $\mathcal{H}_{M_\sigma}(W^c) = (\mathcal{H}_{M_\sigma}(W))^c$.

**Proof**

(i) Obvious.
(ii) It follows from the fact that $M_\sigma(a) \subseteq B$ for each $a \in B$.
(iii) If $V \subseteq W$, then $\mathcal{H}_{M_\sigma}(V) = \{a \in B : M_\sigma(a) \subseteq V \} \subseteq \{a \in B : M_\sigma(a) \subseteq W \} = \mathcal{H}_{M_\sigma}(W)$.
(iv) It follows from (iii) that $\mathcal{H}_{M_\sigma}(V \cap W) \subseteq \mathcal{H}_{M_\sigma}(V) \cap \mathcal{H}_{M_\sigma}(W)$. Conversely, let $a \in \mathcal{H}_{M_\sigma}(V) \cap \mathcal{H}_{M_\sigma}(W)$. Then $a \in \mathcal{H}_{M_\sigma}(V)$ and $a \in \mathcal{H}_{M_\sigma}(W)$ which means that $M_\sigma(a) \subseteq V$ and $M_\sigma(a) \subseteq W$. Therefore, $M_\sigma(a) \subseteq V \cap W$. Thus, $a \in \mathcal{H}_{M_\sigma}(V \cap W)$.
(v) $\alpha \in \mathcal{H}_{M_\sigma}(W^c) \iff M_\sigma(\alpha) \subseteq W^c \iff M_\sigma(\alpha) \cap W = \emptyset \iff \alpha \notin M_\sigma(\alpha) \iff \alpha \in (\mathcal{H}_{M_\sigma}(W))^c$. \hfill \Box

**Corollary 16** Let $(B, \theta, \zeta_\sigma)$ be a $\sigma$-NS. Then $\mathcal{H}_{M_\sigma}(V) \cup \mathcal{H}_{M_\sigma}(W) \subseteq \mathcal{H}_{M_\sigma}(V \cup W)$ for any $V, W \subseteq B$.

**Theorem 17** Let $(B, \theta, \zeta_\sigma)$ be a $\sigma$-NS and $V, W \subseteq B$. The next statements hold true.

(i) $\mathcal{H}_{M^\sigma}(\emptyset) = \emptyset$.
(ii) $\mathcal{H}_{M^\sigma}(B) \subseteq B$.
(iii) If $V \subseteq W$, then $\mathcal{H}_{M^\sigma}(V) \subseteq \mathcal{H}_{M^\sigma}(W)$.
(iv) $\mathcal{H}_{M^\sigma}(V \cup W) = \mathcal{H}_{M^\sigma}(V) \cup \mathcal{H}_{M^\sigma}(W)$.
(v) $\mathcal{H}_{M^\sigma}(W^c) = (\mathcal{H}_{M^\sigma}(W))^c$.

**Proof** The proof is similar to that of Theorem 15. \hfill \Box

**Corollary 18** Let $(B, \theta, \zeta_\sigma)$ be a $\sigma$-NS. Then $\mathcal{H}_{M^\sigma}(V \cap W) \subseteq \mathcal{H}_{M^\sigma}(V) \cap \mathcal{H}_{M^\sigma}(W)$ for any $V, W \subseteq B$.

The converses of items (i) and (iii) of Theorem 15, items (ii) and (iii) of Theorem 17, Corollaries 16 and 18 fail as the next example shows.

**Example 4** Let a $\sigma$-NS $(B, \theta, \zeta_\sigma)$ as given in Example 1. We have the following computations.

(i) $\mathcal{H}_{M_\sigma}(\emptyset) = \{\alpha\}$, $\mathcal{H}_{M_\sigma}(\emptyset) = \{\gamma\}$ and $\mathcal{H}_{M_\sigma}(\emptyset) = \{\alpha, \gamma\}$. Then $\mathcal{H}_{M_\sigma}(\emptyset) \not\subseteq \emptyset$ in general.
(ii) $\mathcal{H}_{M^\sigma}(B) = \{\beta, \delta, \gamma\}$, $\mathcal{H}_{M^\sigma}(B) = \{\alpha, \beta, \delta\}$ and $\mathcal{H}_{M^\sigma}(B) = \{\beta, \delta\}$. Then $B \not\subseteq \mathcal{H}_{M_\sigma}(B)$ in general.
(iii) Let $V = \{\beta\}$ and $W = \{\delta\}$. Then $\mathcal{H}_{M_\sigma}(V \cup W) = \{\alpha, \beta, \delta\} \supset \mathcal{H}_{M_\sigma}(V) \cup \mathcal{H}_{M_\sigma}(W) = \{\alpha\}$. Also, $\mathcal{H}_{M^\sigma}(V \cap W) = \emptyset \subset \mathcal{H}_{M_\sigma}(V) \cap \mathcal{H}_{M_\sigma}(W) = \{\beta, \delta\}$.
(iv) Let $V = \{\alpha, \delta\}$ and $W = \{\delta, \gamma\}$. Then $\mathcal{H}_{M_\sigma}(V) = \mathcal{H}_{M_\sigma}(W) = \{\alpha, \delta, \gamma\}$. Also, $\mathcal{H}_{M^\sigma}(V) = \{\beta, \delta\}$ and $\mathcal{H}_{M^\sigma}(W) = \{\beta, \delta, \gamma\}$. But neither $V \not\subseteq W$ nor $W \not\subseteq V$.

Note that some properties of Pawlak (which presented in Definition 1) may lose for the approximations $M_\sigma$ and $M^\sigma$. Some of these properties are partially losing such as $L_2$, $U_3$, whereas some of them are completely losing such as $L_1$, $U_1$, $L_8$, and $U_8$. To elucidate this matter, consider $V = \{\alpha, \beta, \gamma\}$ and $W = \{\delta, \gamma\}$ as subsets of a $\sigma$-NS $(B, \theta, \zeta_\sigma)$ given in Example 1. Then $\mathcal{H}_{M_\sigma}(V) \neq \{\alpha, \gamma\}$ and $\mathcal{H}_{M^\sigma}(W) \neq \{\beta, \delta, \gamma\}$, but $\mathcal{H}_{M_\sigma}(\mathcal{H}_{M_\sigma}(V)) = \{\gamma\}$ and $\mathcal{H}_{M^\sigma}(\mathcal{H}_{M^\sigma}(W)) = B$.

**Proposition 19** For any relation $\sigma$, we have $0 \leq A_{M_\sigma}(W) \leq 1$ for any subset $W \neq \emptyset$ of $B$.

**Proof** Let $W$ be a nonempty subset of $B$. Then $\mathcal{H}_{M_\sigma}(W) \cup W \neq \emptyset$ for each $\sigma$. It is clear that $\emptyset \subseteq \mathcal{H}_{M_\sigma}(W) \subseteq \mathcal{H}_{M_\sigma}(W) \cup W$. This automatically means that $0 \leq |\mathcal{H}_{M_\sigma}(W)| \leq |\mathcal{H}_{M_\sigma}(W) \cup W|$. Therefore, $0 \leq \frac{|\mathcal{H}_{M_\sigma}(W)|}{|\mathcal{H}_{M_\sigma}(W)\cup W|} \leq 1$. Hence, $A_{M_\sigma}(W) \in [0, 1]$.

**Proposition 20** $\mathcal{A}_{M_\sigma}(B) = 1$ for each $\sigma$-NS $(B, \theta, \zeta_\sigma)$.

**Proof** Straightforward. \hfill \Box

We devoted the rest of this section to comparing between the given approximations and accuracy measures.

**Proposition 21** Let $(B, \theta, \zeta_\sigma)$ be a $\sigma$-NS and $W \subseteq B$. Then

(i) $\mathcal{H}_{M_\sigma}(W) \subseteq \mathcal{H}_{M_\sigma}(W) \subseteq \mathcal{H}_{M_\sigma}(W)$.
(ii) $\mathcal{H}_{M^\sigma}(W) \subseteq \mathcal{H}_{M^\sigma}(W) \subseteq \mathcal{H}_{M^\sigma}(W)$.
(iii) $\mathcal{H}_{M_\sigma}(W) \subseteq \mathcal{H}_{M_\sigma}(W) \subseteq \mathcal{H}_{M_\sigma}(W)$.
(iv) $\mathcal{H}_{M^\sigma}(W) \subseteq \mathcal{H}_{M^\sigma}(W) \subseteq \mathcal{H}_{M^\sigma}(W)$.
(v) $\mathcal{H}_{M_\sigma}(W) \subseteq \mathcal{H}_{M_\sigma}(W) \subseteq \mathcal{H}_{M_\sigma}(W)$.
(vi) $\mathcal{H}_{M_\sigma}(W) \subseteq \mathcal{H}_{M_\sigma}(W) \subseteq \mathcal{H}_{M^\sigma}(W)$.
(vii) $\mathcal{H}_{M^\sigma}(W) \subseteq \mathcal{H}_{M^\sigma}(W) \subseteq \mathcal{H}_{M^\sigma}(W)$.
(viii) $\mathcal{H}_{M^\sigma}(W) \subseteq \mathcal{H}_{M^\sigma}(W) \subseteq \mathcal{H}_{M^\sigma}(W)$.
**Proof** To prove (i), let $\alpha \in H_{Ma}(W)$. Then $M_{s}(\alpha) \subseteq W$. As we showed in Proposition 6 that $M_{s}(\alpha) \subseteq M_{a}(\alpha)$, so $\alpha \in H_{Ma}(W)$. Thus, $H_{Ma}(W) \subseteq H_{Mr}(W)$. Similarly, we prove that $H_{Mr}(W) \subseteq H_{Mr}(W)$.

To prove (ii), let $\alpha \in H_{Mr}(W)$. Then $M_{r}(\alpha) \cap W \neq \emptyset$. As we showed in Proposition 6 that $M_{r}(\alpha) \subseteq M_{r}(\alpha)$, so $\alpha \in H_{Mr}(W)$. Thus, $H_{Mr}(W) \subseteq H_{Mr}(W)$. Similarly, we prove that $H_{Mr}(W) \subseteq H_{Mr}(W)$.

Following similar arguments, one can prove the other cases.

**Corollary 22** Let $(B, \theta, \zeta_{e})$ be a $\sigma$-NS and $W$ be a nonempty subset of $B$. Then

(i) $A_{Ms}(W) \leq A_{Mr}(W) \leq A_{Mr}(W)$.

(ii) $A_{Ms}(W) \leq A_{Mr}(W) \leq A_{Mr}(W)$.

(iii) $A_{Mr}(W) \leq A_{Mr}(W) \leq A_{Mr}(W)$.

(iv) $A_{Mr}(W) \leq A_{Mr}(W) \leq A_{Mr}(W)$.

**Proof** We only prove (i) and one may prove the other cases following similar technique.

Since $H_{Ma}(W) \subseteq H_{Ma}(W) \subseteq H_{Mr}(W)$,

$$|H_{Ma}(W) \cap W| \leq |H_{Ma}(W) \cap W| \leq |H_{Mr}(W) \cap W| \quad (2)$$

Since $H_{Mr}(W) \subseteq H_{Mr}(W) \subseteq H_{Mr}(W)$,

$$|H_{Mr}(W) \cup W| \leq |H_{Mr}(W) \cup W| \leq |H_{Mr}(W) \cup W|. \quad (3)$$

By assumption, $W$ is nonempty; therefore, $|H_{Mr}(W) \cup W| > 0$ for all $\sigma$. Thus,

$$\frac{1}{|H_{Mr}(W) \cup W|} \leq \frac{1}{|H_{Mr}(W) \cup W|} \leq \frac{1}{|H_{Mr}(W) \cup W|}$$

By (2) and (3) we find

$$\frac{|H_{Ma}(W) \cap W|}{|H_{Mr}(W) \cap W|} \leq \frac{|H_{Ma}(W) \cap W|}{|H_{Mr}(W) \cap W|} \leq \frac{|H_{Ma}(W) \cap W|}{|H_{Mr}(W) \cap W|}$$

Hence, we obtain the desired result.

The proofs of the next two results follow from item (iii) of Proposition 6 and Corollary 7.

**Proposition 23** Let $(B, \theta, \zeta_{e})$ be a $\sigma$-NS and $W \subseteq B$. Then

(i) $H_{Ma}(W) \subseteq H_{Mr(\sigma)}(W)$, where $\sigma \in \{r, l, i, u\}$.

(ii) $H_{Ms}(W) \subseteq H_{Mr(\sigma)}(W)$, where $\sigma \in \{r, l, i, u\}$.

(iii) $H_{Ms}(W) \subseteq H_{Ma}(W) \subseteq H_{Mr(\sigma)}(W)$, $w h e e r e$ $\sigma \in \{r, l, i, \langle r \rangle, \langle i \rangle, \langle u \rangle\}$.

(iv) $H_{Ms}(W) \subseteq H_{Ma}(W) \subseteq H_{Mr(\sigma)}(W)$, $w h e e r e$ $\sigma \in \{r, l, i, \langle r \rangle, \langle i \rangle, \langle u \rangle\}$.

**Corollary 24** Let $(B, \theta, \zeta_{e})$ be a $\sigma$-NS and $W$ be a nonempty subset of $B$. Then

(i) $A_{Ms}(W) \leq A_{Mr(\sigma)}(W)$, $\sigma \in \{r, l, i, u\}$.

(ii) $A_{Ms}(W) \leq A_{Mr}(W) \leq A_{Mr(\sigma)}(W)$, $w h e e r e$ $\sigma \in \{r, l, i, \langle r \rangle, \langle i \rangle, \langle u \rangle\}$.

To confirm the obtained results and to illustrate that the converses of the above Proposition 21, Corollary 22, Proposition 23 and Corollary 24 fail we provide the following two examples.

**Example 5** Let a $\sigma$-NS $(B, \theta, \zeta_{e})$ as given in Example 2. Note that in this example all $M_{\sigma}$-neighborhoods are identical and all $M_{(\sigma)}$-neighborhoods are identical, where $\sigma \in \{u, r, l, i\}$. Therefore, the calculations in Table 3 show that the approximations and accuracy values are obtained from $\langle \sigma \rangle$ are better than their counterparts obtained from $\sigma$.

**Example 6** Let a $\sigma$-NS $(B, \theta, \zeta_{e})$ as given in Example 1. Note that in this example $M_{\sigma}$-neighborhood and $M_{(\sigma)}$-neighborhood are identical for all $\sigma \in \{u, r, l, i\}$. Therefore, in Table 4, we suffice by comparing the cases of $\sigma \in \{u, r, l, i\}$.

Also, the calculations in Table 4 show that the best approximations and accuracy values are obtained in cases of $\sigma \in \{i, \langle i \rangle\}$. This means that our method is better than the method given by Dai et al. (2018).

### Table 3 The approximations and accuracy measures in cases of $\sigma$ and $\langle \sigma \rangle$

|            | $H_{Ma}(W)$ | $H_{Mr}(W)$ | $A_{Ms}(W)$ | $H_{Mr(\sigma)}(W)$ | $H_{Mr}(W)$ | $A_{Mr(\sigma)}(W)$ | $A_{Mr}(W)$ |
|------------|-------------|-------------|-------------|----------------------|-------------|----------------------|-------------|
| $\{\alpha\}$ | $\emptyset$  | $\{\alpha, \delta\}$ | $0$         | $\emptyset$          | $\emptyset$  | $\emptyset$          | $\emptyset$  |
| $\{\beta\}$ | $\emptyset$  | $\{\beta, \delta\}$ | $0$         | $\emptyset$          | $\emptyset$  | $\emptyset$          | $\emptyset$  |
| $\{\delta\}$ | $\emptyset$  | $B$         | $0$         | $\{\delta\}$         | $B$          | $\frac{1}{2}$        | $\frac{1}{2}$ |
| $\{\alpha, \beta\}$ | $\emptyset$  | $B$         | $0$         | $\emptyset$          | $\{\alpha, \beta\}$ | $B$          | $\frac{1}{2}$        |
| $\{\alpha, \delta\}$ | $\emptyset$  | $B$         | $\frac{1}{3}$ | $\{\alpha, \delta\}$ | $B$          | $\frac{2}{3}$        | $\frac{2}{3}$ |
| $\{\beta, \delta\}$ | $\emptyset$  | $B$         | $\frac{1}{3}$ | $\{\beta, \delta\}$ | $B$          | $\frac{2}{3}$        | $\frac{2}{3}$ |
| $B$         | $B$         | $1$         | $B$         | $B$                  | $B$          | $1$                  | $1$          |
We close this section by showing that $\mathcal{M}_\sigma$-accuracy and $\mathcal{M}_\sigma$-roughness measures have the monotonicity property under any arbitrary relation.

**Proposition 25** Let $(B, \theta_1, \xi_\sigma)$ and $(B, \theta_2, \xi_\sigma)$ be two $\sigma$-NSs such that $\theta_1 \subseteq \theta_2$. Then $A_{2M_\sigma}(W) \leq A_{1M_\sigma}(W)$ for any subset $W \neq \emptyset$ and $\sigma \in \{ r, l, i, u \}$.

**Proof** It is easy to prove the trivial case when $\mathcal{H}_{2M_\sigma}(W) = \emptyset$ or $\mathcal{H}_{1M_\sigma}(W) = \emptyset$. So, suppose that $a \in \mathcal{H}_{2M_\sigma}(W)$ and $b \in \mathcal{H}_{1M_\sigma}(W)$, where $W \neq \emptyset$ and $\sigma \in \{ r, l, i, u \}$. Then $\mathcal{M}_{2\sigma}(a) \subseteq W$ and $\mathcal{M}_{1\sigma}(b) \cap W \neq \emptyset$. It follows from Proposition 12 that $\mathcal{M}_{2\sigma}(a) \subseteq W$ and $\mathcal{M}_{2\sigma}(b) \cap W \neq \emptyset$. Therefore, $a \in \mathcal{H}_{1M_\sigma}(W)$ and $b \in \mathcal{H}_{1M_\sigma}(W)$. Thus, $\mathcal{H}_{2M_\sigma}(W) \subseteq \mathcal{H}_{1M_\sigma}(W)$ and $\mathcal{H}_{1M_\sigma}(W) \subseteq \mathcal{H}_{2M_\sigma}(W)$. This implies that $\mathcal{H}_{2M_\sigma}(W) \cap W \subseteq \mathcal{H}_{1M_\sigma}(W) \cap W$ and $1 \leq \frac{1}{|\mathcal{H}_{2M_\sigma}(W)\cap W|} \leq \frac{1}{|\mathcal{H}_{1M_\sigma}(W)\cap W|}$. Hence, $1 \leq \frac{|\mathcal{H}_{1M_\sigma}(W)\cap W|}{|\mathcal{H}_{2M_\sigma}(W)\cap W|} \leq \frac{|\mathcal{H}_{2M_\sigma}(W)\cap W|}{|\mathcal{H}_{1M_\sigma}(W)\cap W|}$ which means that $A_{2M_\sigma}(W) \leq A_{1M_\sigma}(W)$.

**Corollary 26** Let $(B, \theta_1, \xi_\sigma)$ and $(B, \theta_2, \xi_\sigma)$ be two $\sigma$-NSs such that $\theta_1 \subseteq \theta_2$. Then we have the following results for any subset $W \neq \emptyset$ and $\sigma \in \{ r, l, i, u \}$.

(i) $\mathcal{R}_{1M_\sigma}(W) \subseteq \mathcal{R}_{2M_\sigma}(W)$
(ii) $B_{1M_\sigma}(W) \subseteq B_{2M_\sigma}(W)$
(iii) $POS_{2M_\sigma}(W) \subseteq POS_{1M_\sigma}(W)$
(iv) $NEG_{2M_\sigma}(W) \subseteq NEG_{1M_\sigma}(W)$

5 Medical applications

At present, COVID-19 pandemic occupies a great interest in all areas of the world because of its rapid negative influences which affect the health system, society, economy, and even politics. According to the experts’ speech, the physical contact or nearness between individuals is the main way of transmission of COVID-19. Regrettably, there is no successful remedy yet. Only you can stay safe following some simple precautions according to WHO’s recommendations such as wearing a mask, physical distancing (preserve at least a one-meter distance between yourself and others), avoiding crowds, and cleaning your hands. World Health Organization showed that the suspicious individual is quarantined for fourteen days. Then, if no symptoms develop through the quarantine period, the procedures of infection prevention and control measures are continued. On the other hand, isolation continues if the symptoms develop through the quarantine period.

Herein, we applied $\mathcal{M}_\sigma$-neighborhoods to classify the community sample under study (such as patients, medical staff, school staff, etc.) in terms of suspicion of infected them by COVID-19. Note that we can apply this technique to any contagious diseases such as influenza.

To explain our model to quarantine the COVID-19 patients, let us consider $B = \{ a_1, a_2, \ldots, a_n \}$ as a group of individuals who works at a specific facility such as school, hospital, bank, etc. The facility administration does a periodic check for its employees from time to time (this period is depending on available capabilities) to check their medical status with respect to COVID-19.
Let $X \subseteq B$ be a group of individuals having a positive COVID-19 test. Directly, this group is quarantined. Now, it is natural to ask what about $W = B \setminus X$? The answer to this question according to our model is as follows; we determine the individuals of $W$ who had contact with individuals of $X$ and vice versa. As we know the contact between individuals leads to spread COVID-19 among them, so that, we consider $\theta$ as a relation on $B$ as following: $\theta_i \theta_j$ iff $\theta_i$ had a contact with $\theta_j$, where $i, j \in \{1, 2, \ldots, m\}$. $\theta$ is a symmetric relation (but is not transitive) because $\theta_i$ is in contact with $\theta_j$ iff $\theta_j$ is in contact with $\theta_i$. According to Proposition 8, there are only two kinds of $M$-neighborhoods. Now, we compute $\mathcal{N}_\sigma$-neighborhood and $\mathcal{M}_\sigma$-neighborhood for all members in the infected set $X$. Then, we divide the suspected set $W$ into three groups $W_1$, $W_2$ and $W_3$.

Group 1 (under high suspicion): The individuals $W_1$ of $W$ who belong to $\mathcal{M}_\sigma$-neighborhood of a member $\alpha \in X$, where $\sigma \in \{r, l, i, u\}$. This implies that for each $w \in W_1$ belongs to $\mathcal{M}_\sigma(\alpha)$ all individuals of $X$ who contact infected person $\alpha$ are in contact with $w$ as well. This matter increases the probability of infection of $w$ with COVID-19. Group 2 (under suspicion): The individuals $W_2$ of $W$ who belong to $\mathcal{M}_\sigma$-neighborhood\ $\mathcal{M}_\sigma(\alpha)$ of a member $\alpha \in X$ (\mathcal{M}_\sigma(\alpha) \setminus \mathcal{M}_\sigma(\alpha))$, where $\sigma \in \{r, l, i, u\}$. This implies that for each $w \in W_2$ belongs to $\mathcal{M}_\sigma(\alpha) \setminus \mathcal{M}_\sigma(\alpha)$ there exists an individual of $X$ who is in contact with the infected person $\alpha$ contacts $w$ as well. Group 3 (no suspicion): The individuals $W_3$ of $W$ who do not belong to $\mathcal{M}_\sigma$-neighborhood of any member $\alpha \in X$. This implies that the individuals of $W_3$ do not contact any suspected person of $X$ or his/her contacts. Therefore, we can say that the individuals of $W_3$ do not have the COVID-19.

After this classification, the quarantine is applied according to the capabilities available; for instance, it suffices to quarantine the individuals who belong to the high suspicion group if the available capabilities is not enough. Otherwise, we quarantine the two suspicion groups (high suspicion and suspicion). In this technique, our role stops at the classification of suspicion individuals into three groups.

Finally, we present the algorithm (Algorithm 1) arising from the discussion above, in order to classify the members with respect to the degree/probability of their infection with COVID-19.

### Algorithm 1

The algorithm for determining the risk of infection.

**Require:** The individuals of a specific facility $B$, the group of individuals $X$ who have a positive COVID-19 test in a given time, the group of individuals $W = B \setminus X$ who have a negative COVID-19 test in the same given time.

**Ensure:** The classification of $W$ in two risk groups and a safe group.

1. Classify the group $B$ depending on COVID-19 test into two groups: positive (denoted by $X$) and negative (denoted by $W$).
2. if $X = \emptyset$ then
3.   return all members in $B$ are classified as non-suspicious.
4. else
5.   Set the group $W \subseteq B$ who is in contact with the group of positive COVID-19 individuals $X$.
6.   Define the binary relation on the set $B$, that takes into account the contacts among the given facility members.
7.   for $\alpha \in X \subseteq B$ do
8.     compute $\mathcal{N}_\sigma$-neighborhoods.
9.     compute $\mathcal{M}_\sigma(\alpha)$-neighborhood.
10.    compute $\mathcal{M}_\sigma(\alpha)$-neighborhood.
11. end if
12. Classify the elements of $W$ in terms of $\mathcal{M}_\sigma$ and $\mathcal{M}_\sigma(\alpha)$ neighborhoods into three groups:
13. High suspicion: The individuals who belong to $\mathcal{M}_\sigma(\alpha)$-neighborhood of at least one $\alpha \in X$
14. Susicion: The individuals who belong to $\mathcal{M}_\sigma(\alpha)$-neighborhood of at least one $\alpha \in X$
15. No suspicion: The individuals who do not belong to $\mathcal{M}_\sigma(\alpha)$-neighborhood of any $\alpha \in X$
16. return the obtained classification.

### 6 Discussions: strengths and limitations

- **Strengths**
1. Determination of the smallest and largest $\mathcal{M}_\sigma$-neighborhoods from all $\sigma$ under any arbitrary relation (see, item (iii) of Proposition 6 and Corollary 7). In fact, this is a unique characterization of $\mathcal{M}_\sigma$-neighborhoods. The previous types of neighborhoods, given in (Abd El-Monsef et al. 2014; Abotabl 2011; Allam et al. 2005, 2006; Al-shami 2021a; Al-shami et al. 2021b; Yao 1996, 1998), do not have this characterization; their determination is limited on the disjoint two sets $\{i, r, l, u\}$ and $\{(i), (r), (l), (u)\}$. This matter leads to compare $\mathcal{M}_\sigma$-approximations and $\mathcal{M}_\sigma$-accuracy measures as investigated in Corollary 23 and to compare $A_\sigma$-accuracy and $\mathcal{M}_\sigma$-accuracy measures as investigated in Corollary 24. Whereas, we cannot compare between these two types of neighborhoods (approximations, accuracy measures) generated from $(\mathcal{N}_\sigma, \mathcal{N}_\sigma^\top)$, $(\mathcal{E}_\sigma, \mathcal{E}_\sigma^\top)$, and $(\mathcal{C}_\sigma, \mathcal{C}_\sigma^\top)$ under an arbitrary relation. To validate this matter, let $(\mathcal{B}, \theta, \zeta_\sigma)$ be a $\sigma$-NS as given in Example 1. Then $\mathcal{N}_\sigma^\top(\delta) = \mathcal{E}_\sigma^\top(\delta) = \{\gamma\}$ and $\mathcal{N}_\sigma^\top(\delta) = \mathcal{E}_\sigma^\top(\delta) = \{\beta, \delta\}$ are incomparable; also, $\mathcal{C}_\sigma^\top(\delta) = \{\delta, \gamma\}$ and $\mathcal{C}_\sigma^\top(\delta) = \{\alpha, \beta, \delta\}$ are incomparable.

2. Our method keeps the monotonic property for the accuracy and roughness values under any arbitrary relation as illustrated in Proposition 23 and Corollary 26. Whereas, this property is losing or keeping under strict conditions in some foregoing methods. This is due to that our method is depending only on the union of $\mathcal{N}_\sigma$-neighborhoods which is proportional to the size of the given relations. In contrast, the relations used in the definitions of some methods (i.e.; equality, subset, superset) randomly works with respect to the size of the given relations. To illustrate this matter in case of $\mathcal{C}_\sigma$-neighborhoods, let $\theta_1 = \{(\alpha, \alpha), (\alpha, \beta), (\beta, \delta)\}$ and $\theta_2 = \theta_1 \cup \{(\alpha, \delta), (\delta, \alpha)\}$ be two binary relations on $\mathcal{B} = \{\alpha, \beta, \delta\}$. Then we have Table 5 below. Taking $V = \{\alpha, \beta\}$ and $W = \{\beta, \delta\}$. By calculation, we obtain $\mathcal{H}_{1\mathcal{C}}(V) = \mathcal{H}_{1\mathcal{S}}(V) = \mathcal{H}_{1\mathcal{C}}^\mathcal{V}(V) = \mathcal{H}_{1\mathcal{S}}^\mathcal{V}(V) = V$, $\mathcal{H}_{2\mathcal{C}}(V) = V$, $\mathcal{H}_{2\mathcal{S}}(V) = \{\alpha, \delta\}$, $\mathcal{H}_{3\mathcal{C}}(V) = \mathcal{H}_{3\mathcal{S}}(V) = \mathcal{B}$, $\mathcal{H}_{1\mathcal{C}}(W) = \mathcal{H}_{1\mathcal{S}}(W) = \{\delta\}$, $\mathcal{H}_{1\mathcal{C}}^\mathcal{V}(W) = \mathcal{H}_{1\mathcal{S}}^\mathcal{V}(W) = W$, $\mathcal{H}_{2\mathcal{C}}(W) = W$, $\mathcal{H}_{2\mathcal{S}}(W) = \{\delta\}$ , and $\mathcal{H}_{3\mathcal{S}}(W) = W$. Now, $A_{2\mathcal{C}}(V) = A_{2\mathcal{S}}(V) = \frac{2}{3} < A_{1\mathcal{C}}(V) = A_{1\mathcal{S}}(V) = 1$, whereas $A_{2\mathcal{C}}(W) = \frac{2}{3} > A_{1\mathcal{C}}(W) = \frac{1}{3}$ and $A_{2\mathcal{S}}(W) = \frac{1}{2} > A_{1\mathcal{S}}(W) = \frac{1}{3}$.

3. The $\mathcal{M}_\sigma$-lower and $\mathcal{M}_\sigma$-upper approximations keep most Pawlak’s properties as we investigated in Proposition 6, Corollary 16, Proposition 17 and Corollary 18. Also, the properties of Pawlak which are losing by $\mathcal{M}_\sigma$-approximations are preserved under a reflexive condition.

4. Our method produces approximations and accuracy values better than the methods given in (Dai et al. 2018) in cases of $\sigma \in \{i, (i), (r)\}$. Also, the approximations and accuracy values generated by our methods are equal to those generated by $\mathcal{E}_\sigma$-neighborhoods under a symmetric relation for $\sigma \in \{i, r, l, u\}$ (see, item (ii) of Corollary 10).

- Limitations

1. The approximations and accuracy measures generated from the methods of $\mathcal{N}_\sigma$-neighborhoods and $\mathcal{C}_\sigma$-neighborhoods (Al-shami 2021a) are better than our method under a reflexive relation.

2. Dai et al.’s method (Dai et al. 2018) produces approximations and accuracy values better than our method in case of $\sigma = u$.

Table 5 $\mathcal{N}_\sigma, \mathcal{C}_\sigma$- and $\mathcal{S}_i$-neighborhoods

|   | $\alpha$ | $\beta$ | $\delta$ |
|---|---------|---------|---------|
| $\mathcal{N}_{1i}$ | $\{\alpha\}$ | $\{\alpha\}$ | $\{\beta\}$ |
| $\mathcal{C}_{1i}$ | $\{\alpha, \beta\}$ | $\{\alpha, \beta\}$ | $\{\delta\}$ |
| $\mathcal{S}_{1i}$ | $\{\alpha, \beta\}$ | $\{\alpha, \beta\}$ | $\{\delta\}$ |
| $\mathcal{N}_{2i}$ | $\{\alpha, \delta\}$ | $\{\alpha\}$ | $\{\alpha, \beta\}$ |
| $\mathcal{C}_{2i}$ | $\{\alpha, \beta\}$ | $\{\beta\}$ | $\{\beta, \delta\}$ |
| $\mathcal{S}_{2i}$ | $\{\alpha\}$ | $\mathcal{B}$ | $\{\beta\}$ |
7 Conclusion and future work

One of the considerable approaches to address the issues of vagueness and uncertain knowledge is rough set theory. In this article, we have established novel kinds of neighborhood systems namely $\mathcal{M}_\sigma$-neighborhoods. We have deliberated some properties concomitant with $\mathcal{M}_\sigma$-neighborhoods and elucidated their relationships with some neighborhoods types introduced in the literature. Then, we have exploited them to introduce some types of lower and upper approximations. We have compared between them and showed that an $\mathcal{M}_{(0)}$-neighborhood produces the best approximations and highest accuracy measure. Also, we have demonstrated that $\mathcal{M}_\sigma$-accuracy and $\mathcal{M}_\sigma$-roughness values are monotonic under any arbitrary relation, where $\sigma \in \{r, l, i, u\}$.

Moreover, we have compared the followed technique with its counterparts induced from the method given in (Dai et al. 2018) in terms of approximations and accuracy measures of subsets. Also, we have proved the identity between our technique and those given in (Amer et al. 2017) under a symmetric relation. Finally, we applied our approach to protect people from infection with the corona-virus COVID-19.

Possible forthcoming works are

- Benefit from $\mathcal{M}_\sigma$-neighborhoods to generate different topologies which are applied to establish new approximations similar to those given in (Abo-Tabl 2013; Dai et al. 2018; Sun et al. 2019);
- Explore $\mathcal{M}_\sigma$-neighborhoods system in the areas of soft rough set and fuzzy rough sets.

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