

Analysis and Detection in V-Formations with Outliers¹

Análisis y detección de formaciones en V con individuos atípicamente separados²

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Abstract

A diversity of movement patterns may be identified when studying a set of moving entities. One of these patterns is called V-formation since its shape resembles such letter. Informally, a set of entities shows a V-formation if they are located in one of its two characteristic lines. In this paper, we propose a model for identifying V-formations with outliers in a list of moving entities. An outlier is an entity of the formation that is apart from its characteristic lines. We present the formal rules of our model and an algorithm for detecting outliers. Our model was validated in NetLogo, a programming and modeling environment for simulating natural and social phenomena.

Keywords

Moving objects, movement patterns, V-formation, outliers.

Keywords plus

Outlier, movement patterns, V training, atypically separated values.

Resumen

Una gran variedad de patrones de movimiento puede ser identificada cuando se estudia un conjunto de entidades móviles. Uno de estos patrones se denomina *formación en V*, ya que su forma se asemeja a dicha letra. Informalmente, un conjunto de entidades exhibe una formación en V si está ubicado en una de sus dos líneas características. En este artículo se propone un modelo para identificar formaciones en V con *outliers* en una lista de entidades móviles. Un *outlier* es una entidad de la formación que está separada de sus líneas características. Se presentan las reglas formales del modelo y un algoritmo para la detección de este tipo de entidades. El modelo se validó mediante NetLogo, un ambiente de programación y de modelamiento para simular fenómenos naturales y sociales.

Palabras clave

Objetos móviles, patrones de movimiento, formación en V, valores atípicamente separados.

Palabras clave descriptores

Valor atípico, patrones de movimiento, formación en v, valores atípicamente separados.

Introduction

A diversity of movement patterns may be identified when studying a set of moving entities, for example, a flock of birds (Dodge *et al.*, 2008). One of these patterns is called V-formation since its shape resembles such letter. Informally, a set of entities shows a V-formation if they are located in one of its two characteristic lines. The lines converge in a position where a unique entity is located, being this one considered the leader of the formation (Cattivelli and Sayed, 2009).

Several authors have analyzed this type of formations. Seiler *et al.* (2002) and Sewatkar *et al.* (2010) try to explain, from a physical point of view, the reasons why some species of birds such as Canada Goose (*Branta canadensis*), Red Knot (*Calidris canutus*) and the Dunlin (*Calidris alpina*), tend to fly this way.

Other authors try to simulate this type of formations to a computational level. For example, Nathan and Barbosa (2008) propose a model based on rules which allow the generation of V-formations according to determined parameters. The authors validated the mentioned model in NetLogo (Wilensky and Rand, 2007), a programming and modeling environment for simulating natural and social phenomena.

On the other hand, Andersson *et al.* (2007) provides a model to identify the leader entity in a group of moving entities. However, his model is not appropriate to V-formations as it is shown in (Moreno *et al.*, 2012 and 2013). There, Andersson's model is scaled up in order to identify V-formations. Nevertheless, the extended model does not consider entities which are part from the formation, but are apart from its characteristic lines, that is, outliers entities.

In this paper, the model of Moreno *et al.* (2013) is extended in order to identify V-formations considering the possible presence of outliers. For achieving this goal, a group of formal rules and an algorithm are presented. Besides, the degree of linearity of the formations is analyzed in more detail, an aspect that was not deepened in the original model.

The analysis of this type of formations can be applied in areas such as economics where formations of this type can be presented in stock prices (Rueda,

2005), in zoology to analyze the birds movement (Cativelli and Sayed, 2009), and in the military and video games fields where ships and aircrafts squadrons often adopt this type of formations (Cativelli and Sayed, 2009).

This paper is organized as follows: In section 1, the model of Moreno *et al.* (2013) is presented to detect aligned V-formations. In section 2, this model is extended to introduce the mechanisms to detect non-aligned V-formations; and then, in section 3, a method to detect outliers in this type of formations is presented. In section 4, experiments are presented, and finally the paper is concluded and future works are proposed.

1. Aligned V-Formations

In this section, the essential elements of the model of Moreno *et al.* (2013) are presented, which at the same time, are an extension of the models of Andersson *et al.* (2007) and Moreno *et al.* (2012).

Consider a group E of n entities $\{e_1, e_2, \dots, e_n\}$ that moves in a space during a time interval (t_1, t_p) . This space is represented by the Euclidian plane. Furthermore, T_s represents the set of time points $\{t_1, t_2, \dots, t_p\}$. Every $t_i \in T_s$, represents a time-step and corresponds to a time point in which the position of the entities was registered.

The entity coordinates at a time point are given by a couple of functions $xpos$ and $ypos$, both with signature: $E \times T_s \rightarrow \text{Real}$. It is assumed that between two consecutive time-steps, the entity moves through a straight line and with a constant speed (Behr and Güting, 2005; Cativelli and Sayed, 2009).

Let $F = \langle e_1, e_2, \dots, e_n \rangle$ be a list of moving entities at a time point t . Let $e_k \in F$ be the leader entity where $k = \lceil n/2 \rceil$ and let be $e_m \in F$, $1 \leq m \leq n$, $m \neq k$. The angle between e_k and e_m at t , denoted $\angle_t(e_k, e_m)$, is defined by the line that goes from $(xpos(e_k, t), ypos(e_k, t))$ to $(xpos(e_m, t), ypos(e_m, t))$. $\angle_t(e_k, e_m) \in [0, 2\pi)$. F is an aligned V-formation at a time point $t \in T_s$ if:

- i. $\forall e_i, e_j \in F, 1 \leq i < k, 1 \leq j < k, \angle_t(e_k, e_i) = \angle_t(e_k, e_j)$.
- ii. $\forall e_p, e_q \in F, k < p \leq n, k < q \leq n, \angle_t(e_k, e_p) = \angle_t(e_k, e_q)$.
- iii. $\forall e_i, e_p \in F, 1 \leq i < k, k \leq p \leq n, \angle_t(e_k, e_i) \neq \angle_t(e_k, e_p)$.
- iv. $\forall e_i, e_j \in F, 1 \leq i \leq n, 1 \leq j \leq n, i \neq j, (xpos(e_i, t), ypos(e_i, t)) \neq (xpos(e_j, t), ypos(e_j, t))$.

The condition i) states that the angles formed by the leader entity e_k and each of the preceding entities must be equal. The condition ii) is analogous and it is applied to every entity succeeding the leader entity e_k in the list. The condition iii) establishes that the angles of the conditions i) and ii) must

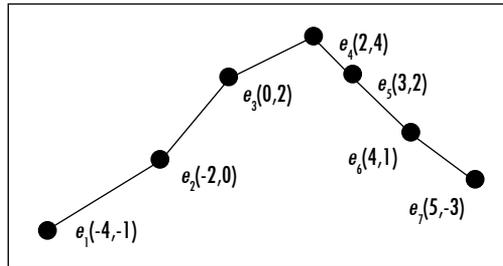
be different. The condition iv) states that two formation entities in the same position cannot exist.

In the original model (Moreno *et al.* 2013), additional rules to make the model more flexible are introduced: i) allow the number of entities in every formation line not to be necessarily equal, ii) limit the distance between entities, iii) limit the opening angle of the formation, and iv) limit the entities orientation regarding the direction of the leader entity. Besides, some incipient elements which lead to model V-formations not necessarily aligned are presented. The following section is focused on this aspect.

2. Not Necessarily Aligned V-Formations

Consider Figure 1, a formation without aligned entities is observed; however, this formation tends to show a configuration in V. Consequently, the previous section conditions are extended in order to identify and consider this type of formations as V-formations.

Figure 1. Formation which tends to be an Aligned V-Formation



Source: Authors' own presentation.

Let $F = \langle e_1, e_2, \dots, e_n \rangle$ be a list of moving entities at a time point t . Let $e_k \in F$ be the leader entity, $1 < k < n$. F is a V-formation (not necessarily aligned) if:

- i. The entities e_i , $1 \leq i \leq k$, tend to form straight line l_1 .
- ii. The entities e_j , $k \leq j \leq n$, tend to form a straight line l_2 .
- iii. The straight lines l_1 and l_2 converge in position ($x_{pos}(e_k, t)$, $y_{pos}(e_k, t)$).
- iv. $\angle \alpha_t > 0$ (the lowest angle defined by the straight lines l_1 and l_2).

Regarding the conditions i) and ii), to state if a set of entities tend to form a straight line, the Pearson's correlation coefficient r (Calderon and Sherry, 2011) is used. Thus, given a set of points $\{(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)\}$, r indicates how

much these points fit a straight line (degree of linearity). $r \in (-1, 1)$, if $|r| \approx 1$ data then tend to form a straight line. r is calculated as shown in the equation (1).

$$r = \frac{n * (\sum_{i=1}^n x_i * y_i) - \sum_{i=1}^n x_i * \sum_{i=1}^n y_i}{\sqrt{[n * \sum_{i=1}^n x_i^2 - (\sum_{i=1}^n x_i)^2] * [n * \sum_{i=1}^n y_i^2 - (\sum_{i=1}^n y_i)^2]}} \tag{1}$$

It can be specified a threshold b to indicate the degree of linearity required to the formation lines, that is, $|r| \geq b$.

Consider a set of points $\{(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)\}$ that represents the positions of the entities of a characteristic line at a time point. To obtain the equation of the straight line $y = mx + b$ that best fits this set of points we can apply equations (2) and (3).

$$m = \frac{n * (\sum_{i=1}^n x_i * y_i) - \sum_{i=1}^n x_i * \sum_{i=1}^n y_i}{n * \sum_{i=1}^n x_i^2 - (\sum_{i=1}^n x_i)^2} \tag{2} \quad b = \frac{\sum_{i=1}^n y_i - m * \sum_{i=1}^n x_i}{n} \tag{3}$$

In Table 1, the results for the formation of Figure 1 are presented.

Table 1. Characteristic Lines: Results for the Formation of Figure 1

| Characteristic Line | Entities in the Characteristic Line | Equation of the Characteristic Line | Pearson's Coefficient | Entity | Entity Coordinates | Entity Coordinates Calculated with the Equation of the Characteristic Line |
|---------------------|-------------------------------------|-------------------------------------|-----------------------|--------|--------------------|--|
| l_1 | e_1, e_2, e_3, e_4 | $y = 2.1 + 0.85x$ | 0.99 | e_1 | (-4,-1) | (-4,-1.3) |
| | | | | e_2 | (-2,0) | (-2,0.4) |
| | | | | e_3 | (0,2) | (0,2.1) |
| | | | | e_4 | (2,4) | (2,3.8) |

Continúa

| Characteristic Line | Entities in the Characteristic Line | Equation of the Characteristic Line | Pearson's Coefficient | Entity | Entity Coordinates | Entity Coordinates Calculated with the Equation of the Characteristic Line |
|---------------------|-------------------------------------|-------------------------------------|-----------------------|--------|--------------------|--|
| l_2 | e_4, e_5, e_6, e_7 | $y = 8.7 - 2.2x$ | 0.97 | e_4 | (2,4) | (2,4.3) |
| | | | | e_5 | (3,2) | (3,2.1) |
| | | | | e_6 | (4,1) | (4,0.1) |
| | | | | e_7 | (5,-3) | (5,2.3) |

Source: Authors' own presentation.

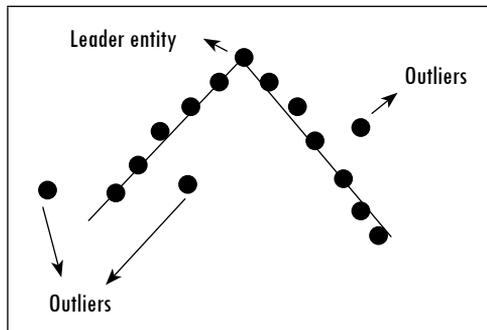
Regarding the condition iv), $\angle ap_i$ it is calculated as follows: the straight lines l_1 and l_2 of the formation are obtained, and the lower angle among them is obtained like this: let a be the angle of the leader entity towards l_1 , b the angle of the leader entity towards l_2 and $w = |a - b|$, then $\angle ap_i = w$ if $w \leq \pi$ and $\angle ap_i = 2\pi - w$, otherwise. To illustrate this $a = 40.36$, $b = 294.44$ and $w = 254.08$; therefore, $\angle ap_i = 105.91$.

3. Outliers

There exist groups of entities that, although tending to show a V-formation, can have distant entities from its characteristic straight lines at a time point t and; therefore, they affect Pearson's coefficient. These entities are called outliers (Hawkins, 1984; Ben-Gal, 2005).

There are numerous methods to detect outliers in different fields (Papadimitriou, 2003). Consequently, an algorithm, that receives a set of entities m (*lineMembers* array) forming a characteristic straight line of a formation at a time point t , is presented. The algorithm determines if when removing a maximum number of entities from the group, the Pearson's coefficient exceed a given threshold μ_p . For instance, if a maximum of two entities are taken away from a straight line of the formation, the entities from Figure 2 are considered to show a V-formation which has two outliers in one straight line and one outlier in the other one. The algorithm then receives the minimum value from the Pearson's coefficient μ_p and the maximum percentage of entities (*percentageOutliers*) that can be taken away from the group of entities. This percentage is calculated considering the total number of entities m .

Figure 2. Example of a V-Formation with Two Outliers



Source: Authors' own presentation.

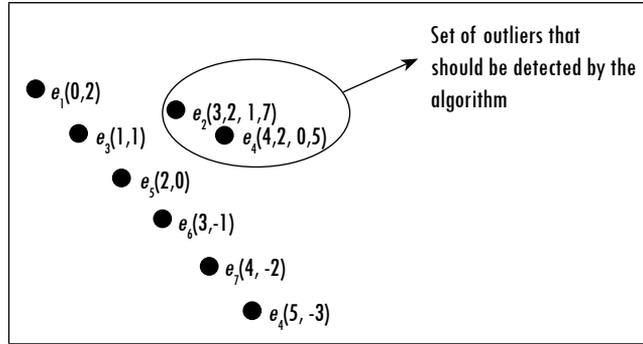
List 1. Algorithm to Detect Outliers

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ALGORITHM: Detection of outliers in a characteristic line
INPUT: lineMembers = [ $e_1, e_2, \dots, e_m$ ] //Array of  $m$  entities
 $\mu_p$  //Threshold for Pearson's coefficient
percentageOutliers //Maximum percentage of outliers allowed in lineMembers array
OUTPUT: outliers //Array of outliers
BEGIN
//Maximum number of outliers allowed in lineMembers array
maxOutliers = ceil( $m * \text{percentageOutliers}/100$ );
FOR  $k = 1$  TO maxOutliers LOOP
//Find out all the possible combinations of  $k$  elements from the lineMembers array
combinationsMatrix = combinations( $k$ , lineMembers);
FOR  $i = 1$  TO size(combinationsMatrix) LOOP
outliers = combinationsMatrix( $i$ ); //Get the current combination
/*Copy into auxMembers the lineMembers array but without the elements of
the current combination*/
auxMembers = remove(lineMembers, outliers); newPearson = PearsonCoefficient(auxMembers);
IF ( $\text{newPearson} \geq \mu_p$ ) THEN //If threshold  $\mu_p$  is satisfied
RETURN outliers;
END IF
END FOR
END FOR
PRINT "It was not possible to meet Pearson's threshold: " $\mu_p$ ";
END
    
```

Example: Consider the set of entities from Figure 3 which form a straight line if the entities e_2 and e_4 are not considered, that is, e_2 and e_3 are outliers.

Figure 3. Characteristic Straight Line of a V-Formation with two Outliers



Source: Authors' own presentation.

Input parameters of the algorithm:

$lineMembers = [e_1, e_2, \dots, e_8]$, $\mu_p = 0.99$ and $percentageOutliers = 30\%$. So $maxOutliers = \lceil 8 * 0.3 \rceil = \lceil 2.4 \rceil = 3$, that is, a maximum of three outliers in the array of the entities is allowed.

The algorithm calculates all the combinations of 1 element in 8. In Table 2, each of the generated combinations are shown, the contents of the *auxMembers* array and the value of Pearson's coefficient (*newPearson*) corresponding the positions of the entities of such array.

Table 2. Combinations of an Element and Pearson's Coefficient of Entities in *auxMembers*

| Combination | <i>auxMembers</i> Array | <i>NewPearson</i> |
|-------------|---------------------------------------|-------------------|
| $\{e_1\}$ | $[e_2, e_3, e_4, e_5, e_6, e_7, e_8]$ | 0.90 |
| $\{e_2\}$ | $[e_1, e_3, e_4, e_5, e_6, e_7, e_8]$ | 0.90 |
| $\{e_3\}$ | $[e_1, e_2, e_4, e_5, e_6, e_7, e_8]$ | 0.90 |
| $\{e_4\}$ | $[e_1, e_2, e_3, e_5, e_6, e_7, e_8]$ | 0.94 |
| $\{e_5\}$ | $[e_1, e_2, e_3, e_4, e_6, e_7, e_8]$ | 0.90 |
| $\{e_6\}$ | $[e_1, e_2, e_3, e_4, e_5, e_7, e_8]$ | 0.97 |
| $\{e_7\}$ | $[e_1, e_2, e_3, e_4, e_5, e_6, e_8]$ | 0.90 |
| $\{e_8\}$ | $[e_1, e_2, e_3, e_4, e_5, e_6, e_7]$ | 0.90 |

Source: Authors' own presentation.

In the absence of a value of *newPearson* which meets the threshold $\mu_p = 0.99$, see Table 2, the algorithm calculates all the combinations of 2 elements in 8. The results are shown in Table 3. When the algorithm evaluates the combination $\{e_2, e_4\}$, the *auxMembers* array = $\{e_1, e_3, e_5, e_6, e_7, e_8\}$ and *newPearson* = 1. As this value is higher than the required threshold ($\mu_p = 0.99$), the algorithm determines that the entities e_2 and e_4 are outliers. It is then concluded that the entities e_1, e_3, e_5, e_6, e_7 , and e_8 tend to form a straight line with a Pearson's coefficient over 0.99.

Table 3. Combinations of two elements and Pearson's Coefficient of the entities in *auxMembers* (in gray the combinations that satisfies $\mu_p = 0.99$)

| Combination | <i>auxMembers</i> Array | <i>newPearson</i> |
|----------------|------------------------------------|-------------------|
| $\{e_1, e_2\}$ | $\{e_3, e_4, e_5, e_6, e_7, e_8\}$ | 0.93 |
| $\{e_1, e_3\}$ | $\{e_2, e_4, e_5, e_6, e_7, e_8\}$ | 0.91 |
| $\{e_1, e_4\}$ | $\{e_2, e_3, e_5, e_6, e_7, e_8\}$ | 0.93 |
| $\{e_1, e_5\}$ | $\{e_2, e_3, e_4, e_6, e_7, e_8\}$ | 0.91 |
| $\{e_1, e_6\}$ | $\{e_2, e_3, e_4, e_5, e_7, e_8\}$ | 0.91 |
| $\{e_1, e_7\}$ | $\{e_2, e_3, e_4, e_5, e_6, e_8\}$ | 0.91 |
| $\{e_1, e_8\}$ | $\{e_2, e_3, e_4, e_5, e_6, e_7\}$ | 0.91 |
| $\{e_2, e_3\}$ | $\{e_1, e_4, e_5, e_6, e_7, e_8\}$ | 0.93 |
| $\{e_2, e_4\}$ | $\{e_1, e_3, e_5, e_6, e_7, e_8\}$ | 1 |
| ... | | |

Source: Authors' own presentation

4. Experiments

For the experiments, NetLogo was used to allow the exploration of the relationship between the behavior at a micro level of the individuals and the patterns at a macro level of the groups, an approach followed in previous works (Andersson *et al.*, 2007; Miller *et al.*, 2010). A model by (Nathan and Barbosa, 2008) was used, which was conceived specifically to generate V-formations in NetLogo.

A population of 15 moving objects was used. The dimensions of the space were $(-35, 35)$ for both axis x and axis y . To create V-formations, parameters of the Table 4 were used. These parameters allow defining aspects related to speed, visual range, and the proximity of the entities. Furthermore, these parameters manage to stabilize the creation of V-formations within a reasonable amount

of time (around 4000 *ticks*) since such model requires a stabilizing time. A *tick* is the time unit in NetLogo, and at a normal speed it is equivalent to 0.5 seconds; however, in NetLogo it is possible to change and program the speed, this is why the value in seconds of a tick is relative.

Table 4. Used Parameters to Generate V-Formations in NetLogo

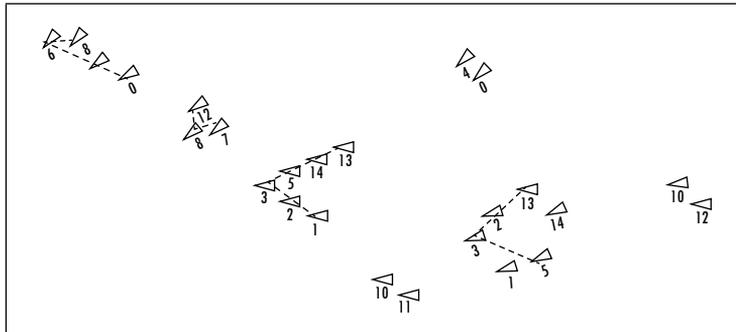
| | |
|--------------------------------------|------|
| Number of birds | 15 |
| Vision Parameters | |
| Vision Distance | 9 |
| Vision Cone | 103° |
| Obstruction Cone | 43° |
| Movement Parameters | |
| Speed | 0,2 |
| Speed Change Factor | 0,15 |
| Vertical distance of movement | 9 |
| Minimum allowed distance among birds | 3,1 |
| Maximum allowed turn | 8° |

Source: Authors' own presentation

The experiments started from tick 40199 and finished in tick 40400 (for this moment, the model that generates V-formations in NetLogo had already been stabilized). The time-steps were taken at 20 ticks intervals, because if we considered a shorter step, it will become practically imperceptible to the change of position and direction of the entities. Thus, the total number of steps was 10. In Figure 4, the position of objects in ticks 40200 and 40400 is shown, and in Table 5 the results are summarized.

In accordance with the results of Table 5, the algorithm classified several groups of entities as V-formations. These results are in accord with what was expected since it was supposed that the model of Nathan & Barbosa (2008) had been already stabilized; and that V- formations were effectively been generated at that moment. Besides, only in one tick (in the tick 40380, for formation 1) it was required to apply the algorithm of detection of outliers ($\mu_p > 0.92$ and $percentageOutliers < 4$). On the other hand, it was found that the formations present a high degree of linearity since the Pearson's coefficients, in all cases, were higher than 0.92.

Figure 4. Position of Entities in Ticks: a) 40200 and b) 40400



Source: Authors' own presentation

Conclusions and Future Works

In this paper, a formal model to identify V-formations with outliers was proposed. The model considers the location of the entities in order to determine if they constitute this type of formations. The rules of the model are flexible since they allow V-formations that are not necessarily aligned as it is in the real world. In addition, the possible presence of outliers entities was considered, that is, members of the formation that can be apart from its characteristic lines. For this reason, an algorithm, where the analyst can specify the maximum number of allowed outliers and the required degree of linearity, was presented. The model was validated using the programming environment NetLogo. The results demonstrate the validity of the proposed model since it was possible to identify the V formations generated by NetLogo.

Concerning future works, we plan to apply our model to the stock market where formations of this type are presented (Rueda, 2005). We also plan to extend our model to identify other kinds of patterns, for instance, identifying *isolated entities*, which are entities that although being considered members of a group, they follow their own way and do not have any influence over the other entities (Laube and Imfeld, 2002); *convergence*, that is, to identify a set of entities which converge or get closer to a site; *divergence*, that is, to identify a set of entities that are spread-out from a site (Gudmundsson *et al.*, 2004), and *self-organization*, that is, a set of entities moving in groups without a leader entity (Cañizo *et al.*, 2010) or that the members do not recognize the leader (Zongyao and Dongbing, 2007; Su *et al.*, 2013).

Table 5. Results of Experiments in NetLogo

| Tick | Groups of Entities (Leader Underlined) | Leader | Leader Coordinates | Pearson's Coefficient Straight Line 1 | Pearson's Coefficient Straight Line 2 | Outliers | Opening Angle | Is it V? |
|-------|--|--------|--------------------|---------------------------------------|---------------------------------------|----------|---------------|----------|
| 40200 | 13,14,5,3,2,1 | 3 | (2.8,2.0) | {13,14,5,3} à 0,99 | {3,2,1} à 0,99 | | 86.49059° | Yes |
| | 12,9,7 | 9 | (-10.1,11.8) | {12,9}à1 | {9,7}à1 | | 75.91894° | Yes |
| | 8,6,4,0 | 6 | (-9.1,33.9) | {8,6}à1 | {6,4,0}à0,98 | | 29.96832° | Yes |
| | 11,10 | NA | NA | {11,10}à1 | NA | | NA | No |
| 40220 | 13,14,5,3,2,1 | 3 | (-1.3,2.1) | {13,14,5,3}à 0,99 | {3,2,1} à 0,99 | | 86.49059° | Yes |
| | 12,9,7 | 9 | (-13.9,9.8) | {12,9}à1 | {9,7}à1 | | 75.91894° | Yes |
| | 8,6,4,0 | 6 | (-15.6,31.4) | {8,6}à1 | {6,4,0}à0,98 | | 29.96832° | Yes |
| | 11,10 | NA | NA | {11,10}à1 | NA | | NA | No |
| ... | | | | | | | | |
| 40360 | 13,14,5,3,2,1 | 3 | (-28.3,-4.3) | {13,14,5,3}à0,95 | {3,2,1}à0,96 | | 40.98386° | Yes |
| | 12,9,7 | 9 | (32.2,-3.1) | {12,9}à1 | {9,7}à1 | | 75.91894° | Yes |
| | 8,6,4,0 | 6 | (-35.0,14.8) | {8,6}à1 | {6,4,0}à0,98 | | 29.96832° | Yes |
| | 11,10 | NA | NA | {11,10}à1 | NA | | NA | No |
| 40380 | 3,2,1,5,14,13 | 1 | (-28.7,8.5) | {2,1}à1 | {1,5,14,13}à0,95 | 3 | 45.129837° | Yes |
| | 12,9,7 | 9 | (28.7,-4.9) | {12,9}à1 | {9,7}à1 | | 75.91894° | Yes |
| | 8,6,4,0 | 4 | (-31.2,9.8) | {8,6}à1 | {4,0}à1 | | 153.73763° | Yes |
| | 11,10 | NA | NA | {11,10}à1 | NA | | NA | No |
| | 6 | NA | NA | NA | NA | | NA | NA |
| 40400 | 13,2,3,1,5,14 | 3 | (-35.4,-8.1) | {13,2,3}à0,98 | {3,1,5,14}à0,93 | | 49.5018° | Yes |
| | 12,9,7 | 9 | (25.2,-6.8) | {12,9}à1 | {9,7}à1 | | 75.91894° | Yes |
| | 0,4 | NA | NA | {4,0}à1 | NA | | NA | No |
| | 11,10 | NA | NA | {11,10}à1 | NA | | NA | No |
| | 6,8 | NA | NA | {6,8}à1 | NA | | NA | No |

Source: authors' own presentation.

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