

A fast and efficient method for solving the multiple generalized circle detection problem

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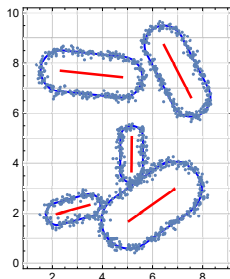
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Problem statement

$$\mathcal{A} = \{a^i = (x_i, y_i) \in \Delta \subset \mathbb{R}^2: \Delta = [\alpha_1, \beta_1] \times [\alpha_2, \beta_2], i = 1, \dots, m\}$$

(set of data points coming from a number of generalized circles in the plane not known in advance that should be reconstructed or detected)

Assumption: Number of data points coming from a generalized circle \mathcal{C} depends on $|\mathcal{C}|$



$$\mathcal{C}(\mu, \nu, r) = \{x \in \mathbb{R}^2: D(\text{seg}(\mu, \nu), x) = r\},$$

$$\text{seg}(\mu, \nu) = \{x \in \mathbb{R}^2: x = \lambda\mu + (1 - \lambda)\nu, \lambda \in [0, 1]\},$$

Applications – may arise in research in various areas like biology, medicine, etc.

Problem statement - center-based clustering

$$\Pi(\mathcal{A}) = \{\pi_1, \dots, \pi_k\}, \quad \mathcal{A} \subset \mathbb{R}^n \quad (k\text{-partition})$$

$$\bigcup_{i=1}^k \pi_i = \mathcal{A}, \quad \pi_r \cap \pi_s = \emptyset, \quad r \neq s, \quad |\pi_j| \geq 1, \quad j = 1, \dots, k$$

$$d: \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}_+, \quad d(u, v) = \|u - v\|_2^2 \quad (\text{LS-distance-like function})$$

$$c_j := \operatorname{argmin}_{x \in \operatorname{conv}(\pi_j)} \sum_{a^i \in \pi_j} d(x, a^i) \quad (\text{centroid of the cluster } \pi_j)$$

Global optimization problem (GOP)

$$\operatorname{argmin}_{\Pi \in \mathcal{P}(\mathcal{A}; k)} \mathcal{F}(\Pi), \quad \mathcal{F}(\Pi) = \sum_{j=1}^k \sum_{a^i \in \pi_j} d(c_j, a^i)$$

$$\mathcal{P}(\mathcal{A}; k) \quad (\text{set of all partitions})$$

Problem statement - center-based clustering

Conversely, let $z_1, \dots, z_k \in \mathbb{R}^n$ be the set of points

Minimal distance principle gives the partition $\Pi = \{\pi(z_1), \dots, \pi(z_k)\}$

$$\pi(z_j) = \{a \in \mathcal{A} : d(z_j, a) \leq d(z_s, a), \forall s = 1, \dots, k\}, \quad j = 1, \dots, k.$$

Global optimization problem:

$$\operatorname{argmin}_{z \in \Delta^k} F(z), \quad F(z) = \sum_{i=1}^m \min_{1 \leq j \leq k} d(z_j, a^i) \quad z = (z_1, \dots, z_k).$$

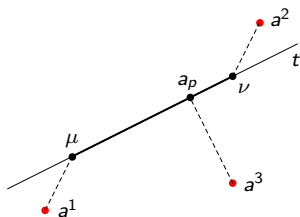
In the case when cluster-centers are generalized circles (\mathcal{C} -cluster centers)

$$\operatorname{argmin}_{\mu, \nu \in \Delta^k, r \in (0, R]^k} F(\mu, \nu, r), \quad F(\mu, \nu, r) = \sum_{i=1}^m \min_{1 \leq j \leq k} \mathfrak{D}(\mathcal{C}_j(\mu_j, \nu_j, r_j), a^i),$$

$$R = \frac{1}{2} \min\{\alpha_2 - \alpha_1, \beta_2 - \beta_1\},$$

$\mathfrak{D}(\mathcal{C}_j, a^i)$ (distance from the point a^i to the generalized circle \mathcal{C}_j)

Problem statement - distance \mathfrak{D} from the point to the generalized circle



$$D(\text{seg}(\mu, \nu), a) = \min_{t \in \text{seg}[\mu, \nu]} \|t - a\|_2 = \begin{cases} \|a - \mu\|_2, & (\nu - \mu)^T (a - \mu) \leq 0, \\ \|a - \nu\|_2, & (\mu - \nu)^T (a - \nu) < 0, \\ \left\| \mu \frac{(\nu - \mu)^T (\nu - a)}{\|\mu - \nu\|_2^2} + \left(1 - \frac{(\nu - \mu)^T (\nu - a)}{\|\mu - \nu\|_2^2}\right) \nu \right\|_2, & \text{else.} \end{cases}$$

$$\mathfrak{D}_1(\mathcal{C}(\mu, \nu, r), a) = |D(\text{seg}(\mu, \nu), a) - r| \quad (\text{LAD-distance})$$

$$\mathfrak{D}_2(\mathcal{C}(\mu, \nu, r), a) = (D(\text{seg}(\mu, \nu), a) - r)^2 \quad (\text{TLS-distance})$$

$$\mathfrak{D}(\mathcal{C}(\mu, \nu, r), a) = (D(\text{seg}(\mu, \nu), a)^2 - r^2)^2 \quad (\text{algebraic distance})$$

Multiple generalized circle detection problem:

$$\operatorname{argmin}_{\mu, \nu \in \Delta^k, r \in (0, R]^k} F(\mu, \nu, r), \quad F(\mu, \nu, r) = \sum_{i=1}^m \min_{1 \leq j \leq k} \mathfrak{D}(\mathcal{C}_j(\mu_j, \nu_j, r_j), a^i)$$

Possibilities:

- (i) Application of global optimization algorithm DIRECT [1-2];
- (ii) Modification of k -means algorithm for the case of \mathcal{C} -cluster-centers [3-4];
 - Our proposal: (ii) with a very carefully chosen initial approximation

[1] D.R.Jones, C.D.Perttunen, B.E.Stuckman, *Journal of Optimization Theory and Applications* **79**(1993)

[2] D.E.Finkel, DIRECT Optimization Algorithm User Guide (2003),
http://www4.ncsu.edu/~ctk/Finkel_Direct/DirectUserGuide_pdf.pdf

[3] T.Marošević, R.Scitovski, *Croatian Operational Research Review* **6**(2015)

[4] R.Grbić, D.Grahovac, R.Scitovski, *Pattern Recognition* **60**(2016)

Algorithm 1

Step A: (Assignment step) For each set of mutually different generalized circles $\mathcal{C}_1(\mu_1, \nu_1, r_1), \dots, \mathcal{C}_k(\mu_k, \nu_k, r_k)$, the data set \mathcal{A} should be partitioned into k disjoint nonempty clusters π_1, \dots, π_k by using the minimal distance principle for $j = 1, \dots, k$

$$\pi_j := \{a \in \mathcal{A} : \mathfrak{D}(\mathcal{C}_j(\mu_j, \nu_j, r_j), a) \leq \mathfrak{D}(\mathcal{C}_s(\mu_s, \nu_s, r_s), a), s \neq j\};$$

Step B: (Update step) For the given partition $\Pi = \{\pi_1, \dots, \pi_k\}$ of the set \mathcal{A} , we should determine the corresponding \mathcal{C} -cluster-centers $\hat{\mathcal{C}}_j(\hat{\mu}_j, \hat{\nu}_j, \hat{r}_j)$, $j = 1, \dots, k$ by solving the following GOPs:

$$\operatorname{argmin}_{\mu, \nu \in \Delta, r \in [0, R]} \sum_{a \in \pi_j} \mathfrak{D}(\mathcal{C}(\mu, \nu, r), a), \quad j = 1, \dots, k.$$

Stopping criterion according to A.M.Bagirov, :

$$\frac{F_{k-1} - F_k}{F_1} < \epsilon_B, \quad F_k = F(\mu^{(k)}, \nu^{(k)}, r^{(k)}), \quad \epsilon_B > 0$$

[1] A.M.Bagirov, Pattern Recognition, 2008

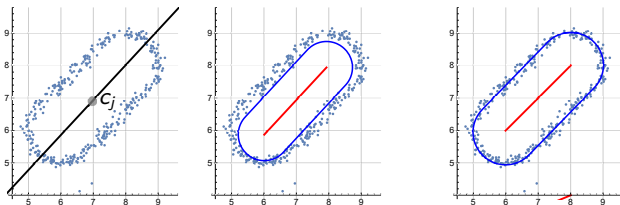
Searching for the best \mathcal{C}_j -cluster-center

$$\operatorname{argmin}_{\mu, \nu \in \Delta, r \in [0, R]} \sum_{a \in \pi_j} \mathfrak{D}(\mathcal{C}(\mu, \nu, r), a)$$

Estimation

- of the length of the line-segment (see [1]): $\delta_j = \frac{2}{|\pi_j|} \sum_{s \in \pi_j} |(a^i - c_j) \cdot v_1^{(j)}|$;
- of the line-segment $\operatorname{seg}(\mu_j, \nu_j) \approx [c_j - \frac{1}{2}\delta_j v_1^{(j)}, c_j + \frac{1}{2}\delta_j v_1^{(j)}]$;
- of the radius of the generalized circle \mathcal{C}_j : $r_j \approx \frac{1}{|\pi_j|} \sum_{s \in \pi_j} |(a^i - c_j) \cdot v_2^{(j)}|$;

DIRECT-domain: $K_{\mu_j} \times K_{\nu_j} \times K_{r_j}$, where $K_u = \{x \in \mathbb{R}^2: \|u - x\|_\infty \leq \frac{1}{4}\delta_j\}$, $u \in \{\mu_j, \nu_j\}$ and $K_{r_j} = \{x \in \mathbb{R}: |x - r_j| \leq \frac{1}{4}\delta_j\}$



[1] J.C.R.Thomas, Progress in Pattern Recognition, Image Analysis, Computer Vision, and Applications, 2011

The proposed method

Application of KMCC algorithm

- with a good enough initial partition (DBSCAN);
- with a good enough approximation of \mathcal{C} -cluster-centers (Incremental algorithm for line detection)

Application of KMCC algorithm by using a good initial partition

Estimation of the parameter ϵ

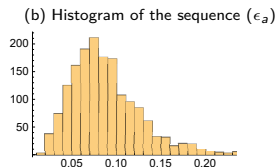
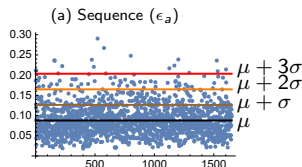
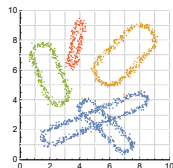
Application of DBSCAN (Density-Based Spatial Clustering of Applications with Noise) algorithm with parameters $MinPts > 2$, and $\epsilon > 0$

Estimation of the parameter ϵ

ϵ_a , $a \in \mathcal{A}$ - radius of the circle containing $MinPts > 2$ elements from \mathcal{A}

$$\mathcal{R} = \{\epsilon_a : a \in \mathcal{A}\}$$

$$\epsilon = \mu + 3\sigma, \quad \mu = \text{Mean}(\mathcal{R}), \quad \sigma^2 = \frac{1}{m-1} \sum_{a \in \mathcal{A}} (\epsilon_a - \mu)^2$$



- [1] M.Ester and H.P.Kriegel, J.Sander, *2nd International Conference on Knowledge Discovery and Data Mining (KDD-96)*, 1996
- [2] P.Viswanath, V.S.Babu, *Pattern Recognition Letters* **30**(2009)
- [3] Y.Zhu, K.M.Ting, M.J.Carman, *Pattern Recognition* **60**(2016)

Algorithm 3 (KMCC by using a good initial partition)

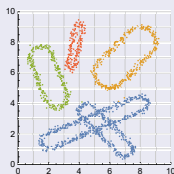
Input: $\mathcal{A} \subset \mathbb{R}^2$; $MinPts$;

- 1: Determine parameter ϵ in the previously described way;
- 2: By using the DBSCAN algorithm determine the initial partition $\Pi^0 = \{\pi_1, \dots, \pi_\kappa\}$;
- 3: By using the KMCC algorithm and the initial partition Π^0 determine generalized circles $\mathcal{C}_1^*, \dots, \mathcal{C}_\kappa^*$;

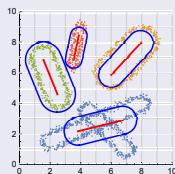
Output: $\{\mathcal{C}_1^*, \dots, \mathcal{C}_\kappa^*\}$.

Example 1

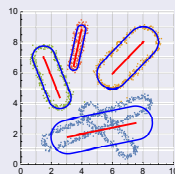
(a) DBSCAN ($\epsilon = .203$)



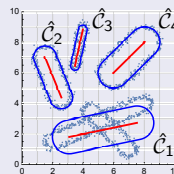
(b) Approximation



(c) DIRECT



(d) KMCC



$$QAD = (0.831, 0.139, 0.174, 0.155)$$

The question is, when does a calculated circle detect some generalized circle, and when not?

A new density-based index for recognizing the detected generalized circle

- $\mathcal{A} \subset \mathbb{R}^2$ - a set of data points coming from a number of generalized circles
 $\mathcal{C} = \{\mathcal{C}_1, \dots, \mathcal{C}_k\}$ not known in advance
- $\Pi = \{\hat{\pi}_1, \dots, \hat{\pi}_\kappa\}$ - an optimal κ -partition of the set \mathcal{A}
- $\hat{\mathcal{C}}_1, \dots, \hat{\mathcal{C}}_\kappa$ - \mathcal{C} -cluster-centers
- $\epsilon > 0$ - parameter from DBSCAN-algorithm

$V_j = \{\mathcal{D}_1(\hat{\mathcal{C}}_j, a) : a \in \hat{\pi}_j\}$ - set of LAD-distances

$\text{QAD}[j] := \mu_j + 2\sigma_j$, $\mu_j = \text{Mean}(V_j)$, $\sigma_j^2 = \frac{1}{|V_j|-1} \sum_{v_j \in V_j} (v_j - \mu_j)^2$ - quantile absolute deviation

Due to the homogeneity property of the set \mathcal{A} , if $\text{QAD}[j] < \epsilon$, then at least 95% of points of the cluster $\hat{\pi}_j$ have a smaller or equal LAD-distances from the circle $\hat{\mathcal{C}}_j$

If for some $j_0 \in \{1, \dots, \kappa\}$, $\text{QAD}[j_0] > \epsilon$, the generalized circle $\hat{\mathcal{C}}_{j_0}$ has not been recognized as one of generalized circles from \mathcal{C}

Density-Based Clustering index (DBC):

$$\text{DBC}(\kappa) := \max_{j=1, \dots, \kappa} \text{QAD}[j]$$

Example

Application of KMCC algorithm by using good initial \mathcal{C} -cluster-centers

Let us assume that among the original generalized circles $\mathcal{C}_1, \dots, \mathcal{C}_k$ there are some that touch or intersect.

Algorithm 4 (KMCC by using a good initial \mathcal{C} -cluster-centers)

Input: $\mathcal{A} \subset \mathbb{R}^2$;

1: According to (Nievergelt, 1994), determine the best TLS-line of the set \mathcal{A} ;

2: Solve $\text{GOP} \quad \underset{\alpha \in [-\pi/2, \pi/2], \delta \in [0, \beta_2]}{\text{argmin}} \quad \Phi(\alpha, \delta), \quad \Phi(\alpha, \delta) = \sum_{i=1}^m \min\{d_1^{(i)}, \mathfrak{D}_H(\ell(\alpha, \delta), a^i)\}$, where

$$d_1^{(i)} = \frac{(u_1 x_i + v_1 y_i + z_1)^2}{u_1^2 + v_1^2}, \text{ and } \mathfrak{D}_H(\ell(\alpha, \delta), a^i) = (x_i \cos \alpha + y_i \sin \alpha - \delta)^2 \text{ and}$$

$\ell(\alpha, \delta) \equiv x \cos \alpha + y \sin \alpha - \delta = 0$ is line in the Hesse normal form. Denote the solution by \mathcal{L}_2 ;

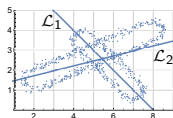
3: By using KMCL-algorithm determine lines $\hat{\mathcal{L}}_1, \hat{\mathcal{L}}_2$;

4: By using Algorithm 2 determine the corresponding \mathcal{C} -cluster-centers $\hat{\mathcal{C}}_1, \hat{\mathcal{C}}_2$;

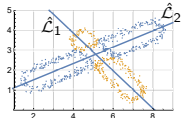
5: By using the KMCC algorithm and initial centers $\hat{\mathcal{C}}_1, \hat{\mathcal{C}}_2$ determine generalized circles $\mathcal{C}_1^*, \mathcal{C}_2^*$;

Output: $\{\mathcal{C}_1^*, \mathcal{C}_2^*\}$.

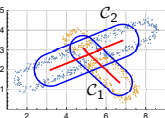
(a) Incremental alg.



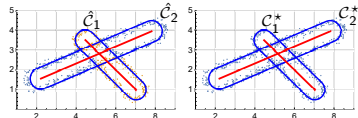
(b) KMCL algorithm



(c) Application of Algorithm 2



(e) KMCC algorithm

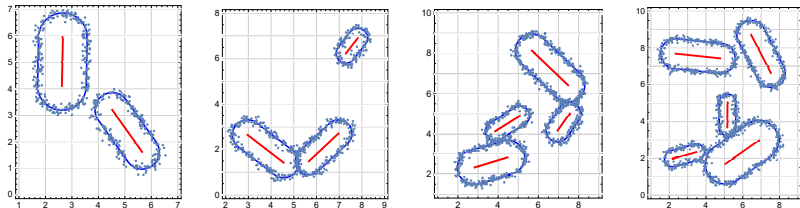


Numerical experiments

Let us first choose integer $k \in \{2, 3, 4, 5\}$ and real numbers $\gamma = 2.25$ and $\delta = 1.5$. Choose line-segments $\text{seg}(\mu_j, \nu_j)$, $j = 1, \dots, k$, such that $\mu_j, \nu_j \in [1.2, 8.8]^2 \subset \Delta$, $\Delta \subset [0, 10] \times [0, 10]$ and that the following holds:

- (i) $\|\mu_j - \nu_j\| < \gamma$,
- (ii) mutual distances between midpoints of line-segments are greater than δ ,
- (iii) distances between vertices of each segment and other segments are greater than δ .
- (iv) radii of generalized circles $r_j \in \mathcal{U}(0.25, 1.0)$.

On the generalized circle \mathcal{C}_j , depending on its length, we choose $m_j = \lceil 30|\mathcal{C}_j| \rceil$ uniformly distributed points and then to each point in the direction of the normal vector we add a random error from normal distribution with expectation 0 and the variance $\sigma^2 = 0.1$.



Result analysis for generalized circles

No. of circles	Generalized circles detected							CPU-time		
	0	1	2	3	4	5	6	(DIRECT)	(KMCC)	Total
2	0	0	29	0	0	0	0	.123	.102	0.225
3	0	0	0	30	2	0	0	.175	.156	0.331
4	0	0	0	1	26	0	0	.253	.295	0.548
5	0	0	1	0	0	11	1	.322	.311	0.633

Construction of data in the neighborhood of a generalized circle

$m = 30|\mathcal{C}|$ - number of uniformly distributed random points in the neighbourhood of the generalized circle $\mathcal{C}(\text{seg}(\mu, \nu), r)$

- Random points in the neighbourhood of the circle $C(S_0, r)$, $S_0 = \frac{1}{2}(\mu + \nu)$ are translated into the left and the right edge of the generalized circle \mathcal{C} ;
- Random points in the neighbourhood of line-segments \overline{AB} , \overline{CD}

$$T_i = \lambda_i A + (1 - \lambda_i)B + \epsilon_i, \quad A = \mu - n_0 r, \quad B = \nu - n_0 r$$
$$\lambda_i \in \mathcal{U}(0, 1), \quad \epsilon_i \sim \mathcal{N}(0, \sigma^2 I)$$

