



Itasdi

Innovative Teaching Approaches in development of Software
Designed Instrumentation and its application in real-time
systems

Theory of Robotics Systems

Line Fitting – Uncertainty

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Innovative Teaching Approaches in development of Software Designed Instrumentation and its application in real-time systems

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Reprezentacija nesigurnosti merenja

- Opažanje u realnom okruženju je **uvek nesigurno**.
 - Kako možemo predstaviti i kvantifikovati nesigurnost?
 - Kako se propagira nesigurnost?
 - Kako se ovo sve odnosi na mobilnu robotiku?

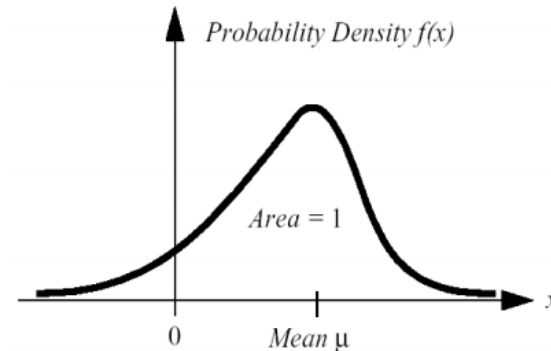
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Reprezentacija nesigurnosti merenja

- Koristi se Funkcionalna Gustina Verovatnoće (FGV) za opisivanje karakteristika promenljive x :



$$\int_{-\infty}^{\infty} f(x) dx = 1$$

- Očekivana vrednost promenljive: $\mu = E[X] = \int_{-\infty}^{\infty} xf(x) dx$
- Varijansa promenljive: $\sigma^2 = \int_{-\infty}^{\infty} (x - \mu)^2 f(x) dx$

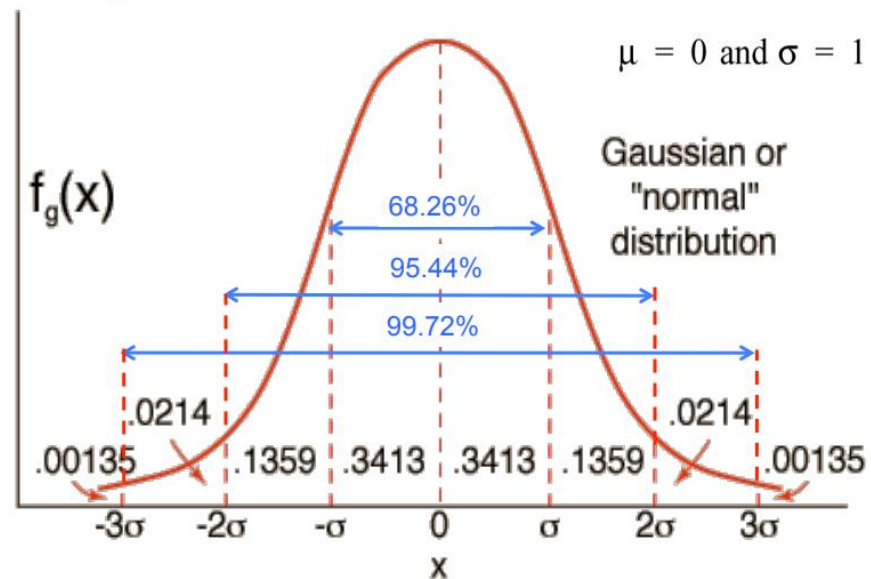




Gausovska FGV

- Najčešće korišćena FGV za karakterizaciju nesigurnosti je Gausovska.

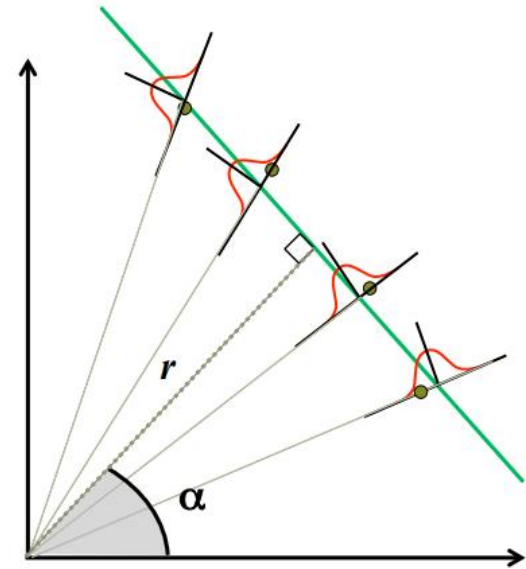
$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$$





Zakon propagiranja greške

- Posmatramo problem izdvajanja linije baziranom na merenjima koja imaju određenu nesigurnost.
- Model parametara je predstavljen u polarnim koordinatama.
- Koja je nesigurnost za izdvojenu liniju ukoliko znamo nesigurnost tačkaka koje čine tu liniju?



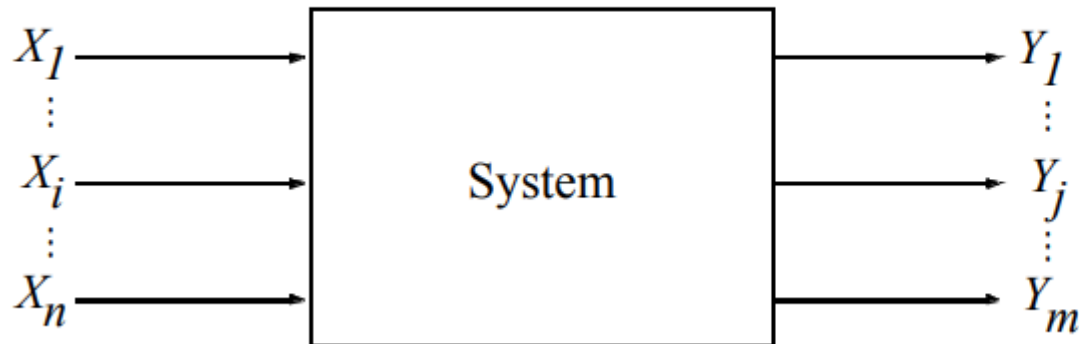


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Zakon propagiranja greške

- Propagacija greške je sistem sa više ulaza i više izlaza.



$$Y_j = f_j(X_1 \dots X_n)$$

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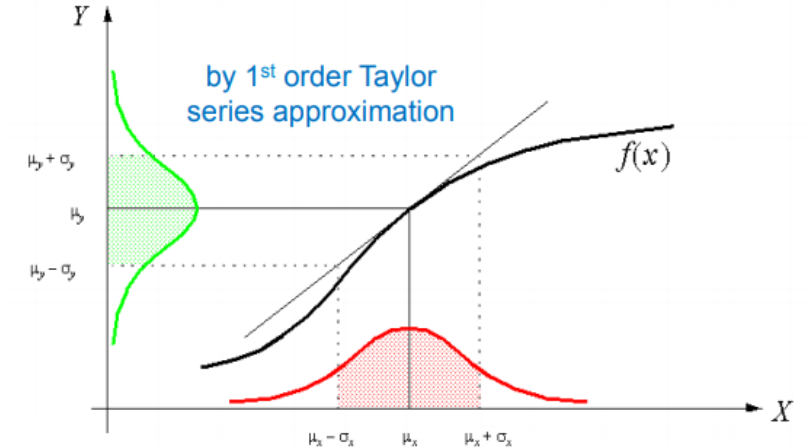




Zakon propagiranja greške

- 1D slučaj nelinearne propagacije greške.
- Može se pokazati da je izlazna kovarijaciona matrica: $C_{YY} = F_{YX} C_{XX} F_{XY}$, $F_{XY} = F_{YX}^T$
- Gde je:
 - C_{XX} kovarijaciona matrica ulaznih nesigurnosti
 - C_{YY} kovarijaciona matrica propagiranih nesigurnosti
 - F_{YX} Jacobian matrica

$$F_{YX} = \begin{pmatrix} \frac{\partial f_1}{\partial X_1} & \dots & \frac{\partial f_m}{\partial X_1} \\ \vdots & \dots & \vdots \\ \frac{\partial f_m}{\partial X_1} & \dots & \frac{\partial f_m}{\partial X_n} \end{pmatrix}$$





Zakon propagiranja greške – izdvajanje linije

- Rastojanje tačke od linije
- Ukoliko pretpostavimo da su sva merenja podjednako nesigurna

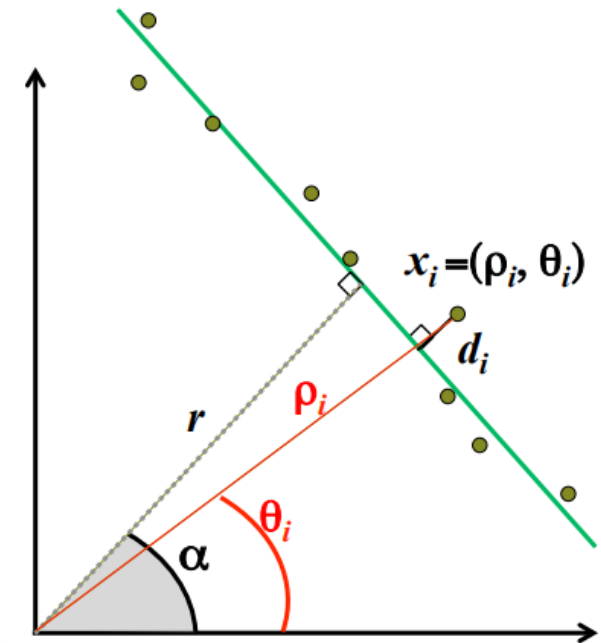
$$\rho_i \cos(\theta_i - \alpha) - r = d_i$$

$$S = \sum_i d_i^2 = \sum_i (\rho_i \cos(\theta_i - \alpha) - r)^2$$

- Cilj: minimizovati **S** da dobijemo što tačnije (r, α) :

$$\frac{\partial S}{\partial \alpha} = 0 \quad \frac{\partial S}{\partial r} = 0$$

- Unweighted Least Squares





Zakon propagiranja greške – izdvajanje linije

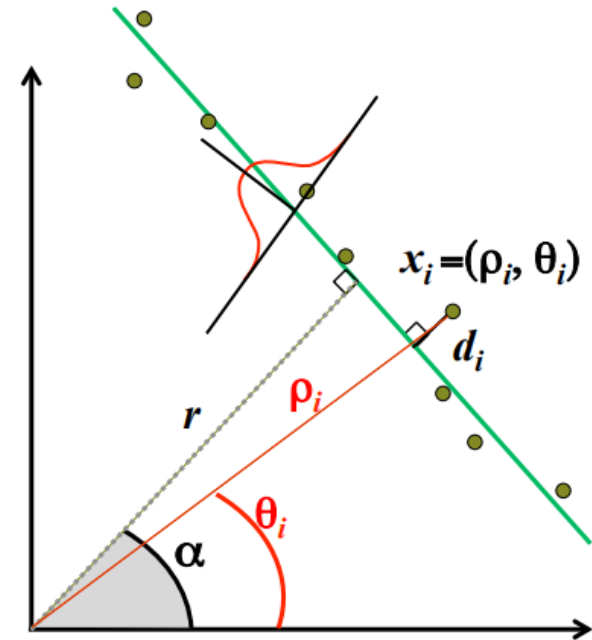
- Rastojanje tačke od linije
- Merenja mogu imati jedinstvene nesigurnosti

$$\rho_i \cos(\theta_i - \alpha) - r = d_i$$
$$S = \sum w_i d_i^2 = \sum w_i (\rho_i \cos(\theta_i - \alpha) - r)^2$$
$$w_i = 1/\sigma_i^2$$

- Cilj: minimizovati **S** da dobijemo što tačnije (r, α) :

$$\frac{\partial S}{\partial \alpha} = 0 \quad \frac{\partial S}{\partial r} = 0$$

- **Weighted Least Squares**



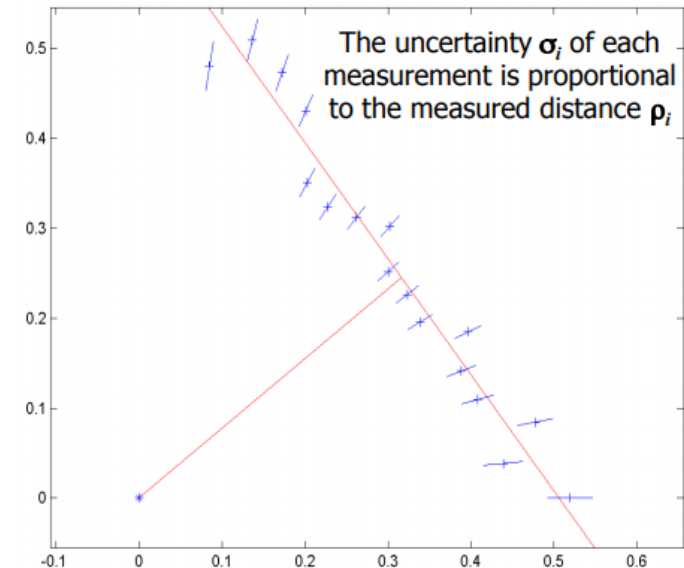


Zakon propagiranja greške – izdvajanje linije

- Rešavamo sistem: $\frac{\partial S}{\partial \alpha} = 0$ $\frac{\partial S}{\partial r} = 0$
- Dobijamo parametre linija:

$$\alpha = \frac{1}{2} \operatorname{atan} \left(\frac{\sum w_i \rho_i^2 \sin 2\theta_i - \frac{2}{\sum w_i} \sum \sum w_i w_j \rho_i \rho_j \cos \theta_i \sin \theta_j}{\sum w_i \rho_i^2 \cos 2\theta_i - \frac{1}{\sum w_i} \sum \sum w_i w_j \rho_i \rho_j \cos(\theta_i + \theta_j)} \right) \quad r = \frac{\sum w_i \rho_i \cos(\theta_i - \alpha)}{\sum w_i}$$

- Ako svaki od ulaznih parametra ima svoju nesigurnost kako predstaviti nesigurnost linije?





Zakon propagiranja greške – izdvajanje linije

- Nesigurnost svakog merenja je predstavljena kovarijacionom matricom
- Nesigurnost linije je predstavljena svojom kovarijacionom matricom

$$C_{xx_i} = \begin{bmatrix} \sigma_{\rho_i}^2 & 0 \\ 0 & \sigma_{\theta_i}^2 \end{bmatrix}$$

$$C_{ll} = \begin{bmatrix} \sigma_a^2 & \sigma_{ar} \\ \sigma_{ra} & \sigma_r^2 \end{bmatrix} = ?$$





Zakon propagiranja greške – izdvajanje linije

- Kovarijacionu matricu svih merenja definišemo kao:

$$C_{xx} = \begin{bmatrix} \text{diag}(\sigma_{\rho}^2) & 0 \\ 0 & \text{diag}(\sigma_{\theta}^2) \end{bmatrix} = \begin{bmatrix} \dots & 0 & 0 & \dots & 0 & 0 & \dots \\ \dots & \sigma_{\rho_i}^2 & 0 & \dots & 0 & 0 & \dots \\ \dots & 0 & \sigma_{\rho_{i+1}}^2 & \dots & 0 & 0 & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & 0 & 0 & \dots & \sigma_{\theta_i}^2 & 0 & \dots \\ \dots & 0 & 0 & \dots & 0 & \sigma_{\theta_{i+1}}^2 & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \end{bmatrix}_{2n \times 2n}$$

- Zakon propagacije greške:

$$F_{lx} = \begin{bmatrix} \dots & \frac{\partial \alpha}{\partial \rho_i} & \frac{\partial \alpha}{\partial \rho_{i+1}} & \dots & \frac{\partial \alpha}{\partial \theta_i} & \frac{\partial \alpha}{\partial \theta_{i+1}} & \dots \\ \dots & \frac{\partial r}{\partial \rho_i} & \frac{\partial r}{\partial \rho_{i+1}} & \dots & \frac{\partial r}{\partial \theta_i} & \frac{\partial r}{\partial \theta_{i+1}} & \dots \end{bmatrix}$$

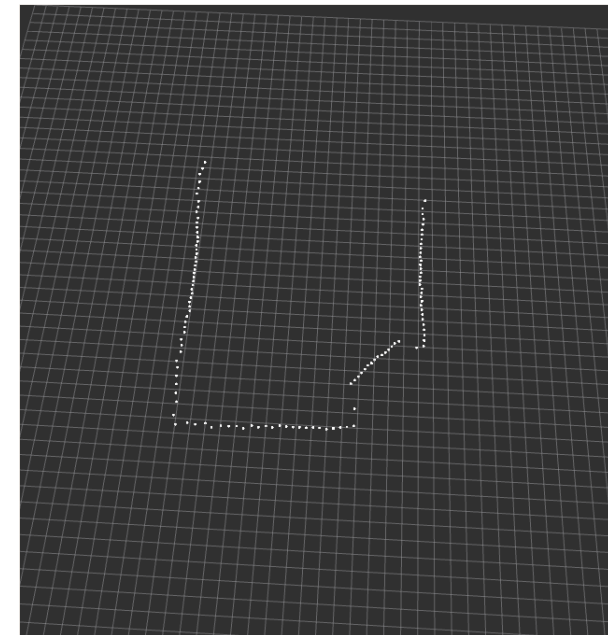
$$C_{ll} = F_{lx} C_{xx} F_{xl}$$





Izdvajanje linije iz oblaka tačkaka

- Tri glavna problema:
 - Koliko linija se postoji?
 - **Segmentacija:** Koje tačke pripadaju kojoj liniji?
 - **Fitovaje linije:** Kako estimirati parametre linije?
- Najpoznatiji algoritmi:
 - Split-and-merge
 - Linearna regresija
 - RANSAC





Split-and-Merge

- Popularan algoritam, originalno korišćen u kompjuterskoj viziji.
- Predstavlja rekurzivnu proceduru za izdvajanje i spajanje linja.

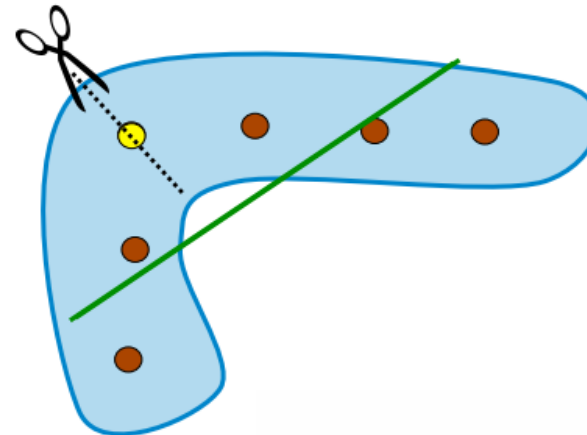
Let **S** be the set of all data points

Split

- Fit a line to points in current set **S**
- Find the most distant point to the line
- If distance $>$ threshold \Rightarrow split set & repeat with left and right point sets

Merge

- If two consecutive segments are collinear enough, obtain the common line and find the most distant point
- If distance \leq threshold, merge both segments





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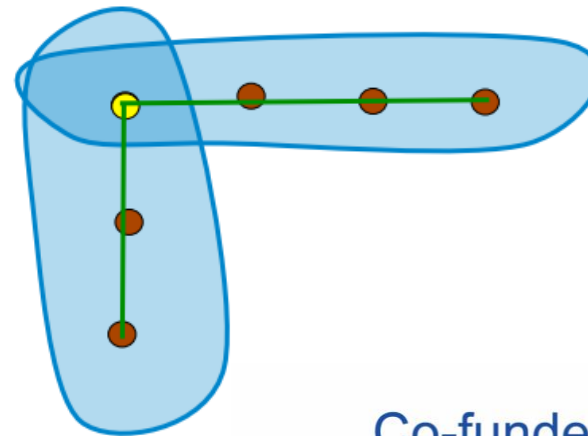
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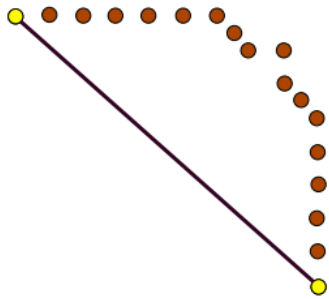
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Split-and-Merge (iterativni)

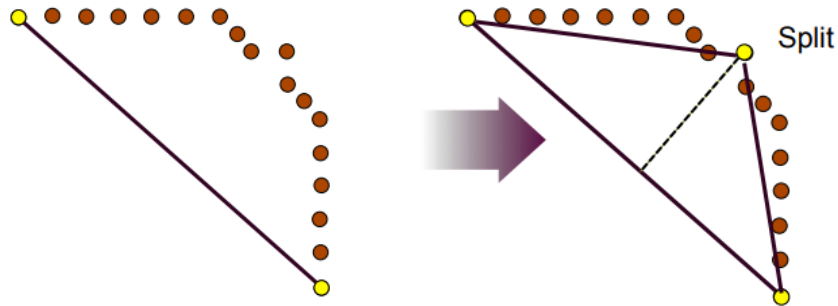
- Povežu se prva i poslednja tačka





Split-and-Merge (iterativni)

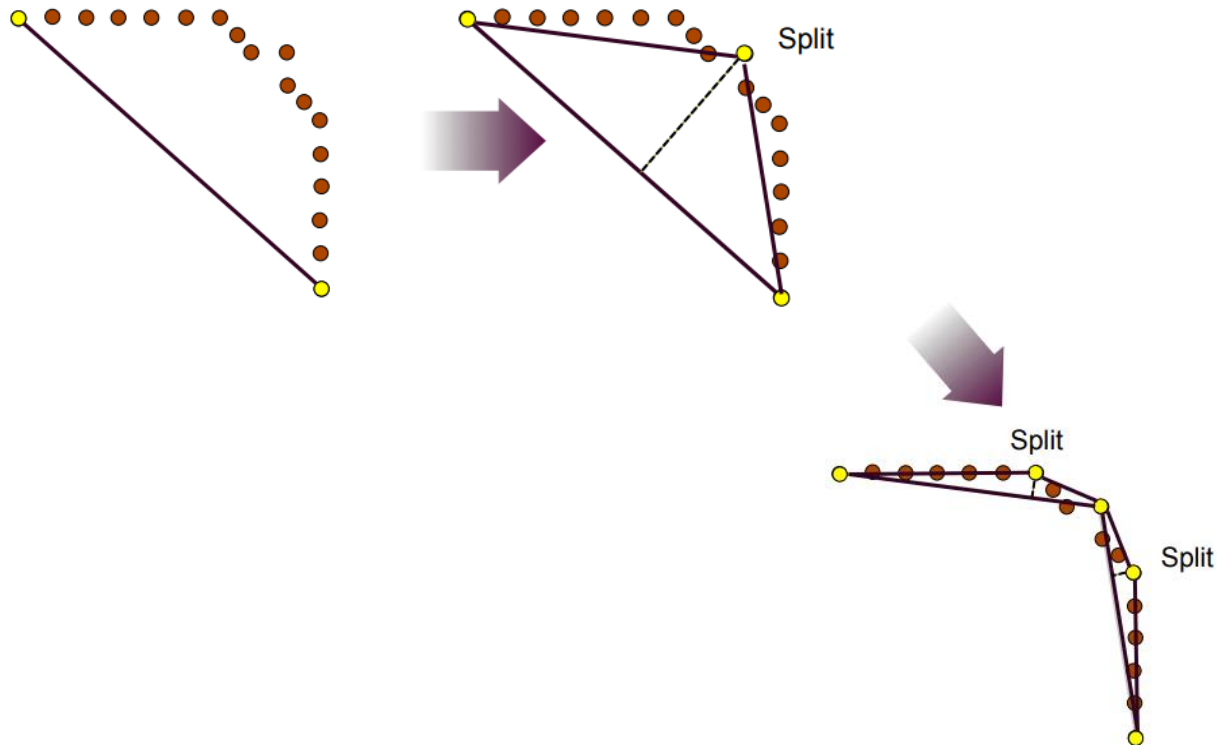
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Split-and-Merge (iterativni)

- Povežu se prva i poslednja tačka



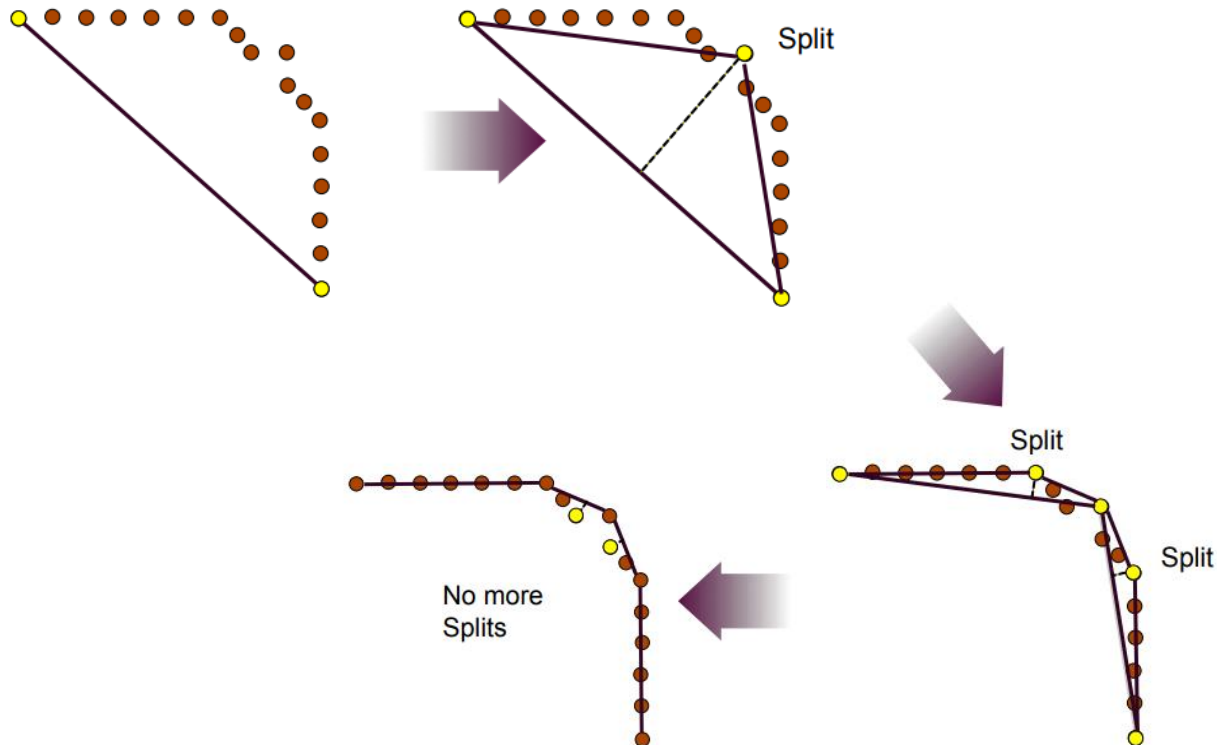


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Split-and-Merge (iterativni)

- Povežu se prva i poslednja tačka



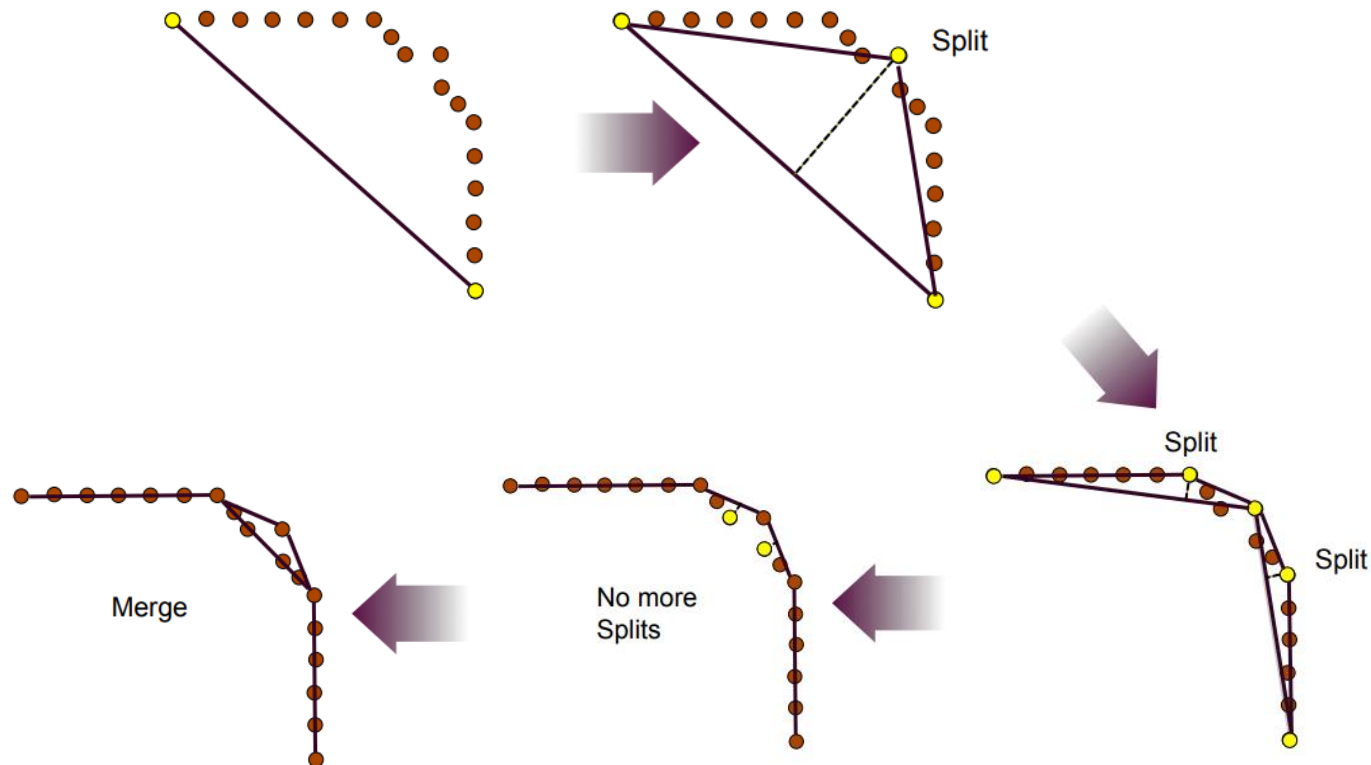
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Split-and-Merge (iterativni)

- Povežu se prva i poslednja tačka



Izdvajanje linija - poređenje

	Complexity	Speed (Hz)	False positives	Precision
Split-and-Merge	$N \log N$	1500	10%	+++
Incremental	$S N$	600	6%	+++
Line-Regression	$N N_f$	400	10%	+++
RANSAC	$S N k$	30	30%	++++
Hough-Transform	$S N N_C + S N_R N_C$	10	30%	++++
Expectation Maximization	$S N_1 N_2 N$	1	50%	++++



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