# Soft Subdivision Motion Planning for Complex Planar Robots<sup>^</sup>

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# **Motion Planning**

- A central problem in robotics
- There is a fixed rigid robot:  $R_0 \subseteq R^k$  (k = 2,3)
- Configuration: pos. & orientation of a point p in  $R_0$

#### **INPUT** : (*α*, *β*, Ω)

- Start and Goal configurations  $\alpha$ ,  $\beta$
- Polyhedral obstacle set  $\Omega \subseteq R^k (k = 2,3)$ OUTPUT:
- A path from  $\alpha$  to  $\beta$  avoiding all obstacles in  $\Omega$ , if it exists.
- Else report "NO PATH".





#### State of the Art

#### (A) Exact Methods

- + Strong theoretical guarantees
- High complexity
- e.g., roadmap is single exponential time [Canny 93] basic path planning is semi-algebraic (book of [Basu-Pollack-Roy])
- Complex to implement & expensive to compute (rarely implemented and not practical)
- (B) Subdivision Methods (e.g., [Zhu-Latombe91], [Zhang et al 08]) Fairly popular but ``does not scale'' Often degenerate into "grid method"



## State of the Art (cont.)

#### (C) Sampling Methods

- \* Probabilistic Road Map (PRM) [Kravraki 96]; many variants: EST, RRT, SRT, etc.
- \* Dominate the field in the last 2 decades.

Major Issue: Halting Problem ("Narrow Passage" problem) ---Don't know how to halt when there is no path (except for artificial cut-off)

• Some subdivision work (e.g., [Zhang et al 08]) can detect non-existence of paths, but cannot guarantee to always detect that (sol. is partial).



## State of the Art (cont.)

#### **Resolution-Exact Algorithms**

- We initiated in [Wang-Chiang-Yap SoCG13], [Yap 13]
- Use subdivision and soft predicates --- Soft Subdivion Search (SSS)
- Avoid exact computation, easy to implement correctly, run fast, always halt, with theoretical guarantees (see paper for details).
- Further extended for 2-link planar robot with 4 degrees of freedom (4 DOFs) [Luo-Chiang-Yap WAFR14], [Chee-Luo-Hsu WAFR16], 5-DOF 3D robots [Hsu-Chiang-Yap 18].
- In this paper, we work on 2D complex robot under this framework.



#### New Results: SSS for Complex Robots

- 2D rigid complex robots with arbitrary complexity (*m*-sided polygon, m>=5).
- Use triangulation.
- Our previous [SoCG 13] method
   for triangle robot does not work (c) C-shaped (d) S-shaped since the triangles in a complex robot must share a common origin (rotation center).





## Review of SSS: Resolution Exactness

- An resolution-exact planner takes an extra input parameter  $\epsilon > 0$ . It always halts and outputs either a path or NO-PATH. The output satisfies:
- There is an accuracy constant K > 1, s.t.
- If exists a path of clearance  $K\epsilon$ , it must output a path;
- If there is no path of clearance  $\varepsilon/K$ , it must output NO-PATH.
- Indeterminacy allowed (small price for avoiding exact computation)



# Review of SSS: Search Framework

- Maintain a subdivision tree T rooted at box  $B_0$  (input domain)
- Each internal node is a box *B*, which is split into  $2^i$  ( $1 \le i \le d$ ) congruent subboxes (T/R-split: see later)
- Each box B is classified as free (each t ∈ B is a free configuration), stuck (each t ∈ B is in the exterior of the free space), or mixed (otherwise).
- We maintain connected components of free boxes via a Union-Find data structure ( $\alpha$ ,  $\beta \in$  same component  $\rightarrow$  path found)
- Priority Queue Q of mixed boxes to be expanded later



A star-shaped region *R*: there exists a point  $A \in R$  s.t. A can "see" every point in *R*. We call *A* a **center** of *R*.



When a robot  $R_0$  is star-shaped, we decompose  $R_0$  into a set of triangles that share a **common vertex** at a center *A*.

We need a predicate that can easily classify boxes B as free/stuck/mixed.



- Triangular Set:  $T = H_1 \cap H_2 \cap H_3$  (intersection of three half-spaces) T can be bounded (triangle) or unbounded (Figure (a) below)
- Apex: distinguished vertex (red)
- Truncated Triangular Set (TTS):  $TTS = T \cap D = H_1 \cap H_2 \cap H_3 \cap D$ *T* intersects with a disc D centered at *A* (Figure (b) below)





- Angular Range:  $\Theta = [\alpha, \beta]$
- Swept area  $T_0[\Theta] = T_0[\alpha, \beta]$  for triangle  $T_0(A, B, C)$  where A is the apex)
- Nice swept area (Figure (c) below) and not nice (Figure (d) below)
- Nice triangle:  $b \ge \pi / 2 = 90^{\circ}$

(where a,b,c are the angles of vertices A,B,C resp.)





(d) sweeping [ABC] to [AB'C']



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• Lemma 1: *T* is nice  $\Leftrightarrow$  Footprints *T*[0,  $\alpha$ ] and *T*[- $\alpha$ , 0] are TTS for all  $\alpha$ where 0 <  $\alpha$  < $\pi$  - a



 Lemma 2: A star-shaped robot R<sub>0</sub> (an n-gon) can be decomposed into an essentially disjoint union of at most 2n nice triangles sharing apex A





- Complex Predicates:  $R_0 = \bigcup_{j=1}^m T_j$  $\tilde{C}(B) = \begin{cases} FREE & \text{if each } \tilde{C}_j(B) \text{ is FREE} \\ STUCK & \text{if some } \tilde{C}_j(B) \text{ is STUCK} \\ MIXED & \text{otherwise} \end{cases}$
- T/R Splitting
- Do *translational* (T) splitting only and keep the rotational component full until the box width is ≤ ε (ε-small) --- top: quad tree
- Do *rotational* (R) splitting if the box is *ε*-small --- bottom: binary tree



### Feature Set of a Box

- Each box B: we use its feature set  $\tilde{\phi}(B)$  to classify B as free/stuck/mixed.
- Obstacles  $\Omega$ : polygonal set  $\Omega \subseteq \mathbb{R}^2$
- Look at boundary of  $\Omega$ . Feature *f* : corner or edge of boundary of  $\Omega$
- Feature set  $\tilde{\phi}(B)$ : contains all *f* that are potentially in conflict with robot  $R_0$ when its configuration is in B.
- As we split a box into subboxes, the feature sets become smaller.
- Classification:

 $\tilde{\phi}(B)$  is non-empty: B is **mixed**.

 $\tilde{\phi}(B)$  is empty: B is in no conflict with obstacle boundary  $\rightarrow$ B is free or stuck (use parent feature set to decide)



## Feature Sets for Star-Shaped Robots (I)

Soft Predicates for classification

- When B is a T-split box (we only split its translational box; quad-tree part) Its features set φ̃(B) comprises those features f such that Sep(m<sub>B</sub>, f) ≤ r<sub>B</sub> + r<sub>0</sub> where m<sub>B</sub> and r<sub>B</sub> are the midpoint and radius of translational box of B,
  - $r_0$  is the radiuis of robot R<sub>0</sub>



# Feature Sets for Star-Shaped Robots (II)

When B is an R-split box (we only split its rotational box; binary-tree part)

- $\tilde{\phi}(B)$ : a collection of  $\tilde{\phi}_i(B)$  for each nice triangle  $T_i$
- $TTS_i$ : apex is at the box center  $m_B$
- Let Q be a shape and s be a real number, s-expansion of Q is defined as the Minkowski sum of Q with the Disc(s) of radius s centered at the origin
- φ̃<sub>j</sub>(B) comprises those features *f* satisfying

   (1) Sep(m<sub>B</sub>, f) ≤ r<sub>B</sub> + r<sub>j</sub>
   (2) f also intersects the r<sub>B</sub>-expansion of TTS<sub>j</sub> (yellow: super set)
   Condition (2) can be easily checked





- $R_0$  is a general polygon, we can still decompose  $R_0$  into a set of triangles  $T_j$  (j = 1, ..., m)
- The rotation of these triangles are relative to a fixed point O
- We will define  $T_i$  to be "nice relative to a point O"



- Let T = [A, B, C], O be the origin (outside of T)
- Let  $0 \le ||A|| \le ||B|| \le ||C||$  where ||A|| is the Euclidean norm of a vector A
- We say T is nice if  $<A, B-A> \ge 0,$ and  $<A, C-A> \ge 0,$ and  $<B, C-B> \ge 0.$





• If T is a nice triangle,  $T[\alpha, \beta]$  is called a nicely swept set (NSS).

We want an easy way to detect the intersection between an s-expansion of NSS and any feature (point or edge)

- We define a subset of R<sup>2</sup> as a:
- O-basic shape: half-space, a disc or complement of a disc
- 1-basic shape: finite intersection of 0-basic shapes
- 2-basic shape: finite union of 1-basic shapes





- E.g. 1-basic shapes:
- Triangles (ABC)
- Sectors (A'C'C")
- Truncated strips (ACC"A' --- shown in yellow)
- The s-expansion of a sector / truncated strip / triangle is 2-basic.

**Theorem**: Let  $T[\alpha, \beta]$  be a nicely swept set where  $[\alpha, \beta]$  has width  $\leq \pi/2$ . It can be decomposed into a triangle, a sector and a truncated strip. The s-expansion of  $T[\alpha, \beta]$  has a basic decomposition into 2-basic shapes.

Testing intersection of 2-basic shapes with any feature is O(1).







- Partitioning an *n*-gon into Nice Triangles
- First triangulate into *n*-2 triangles
- For the one contains the origin *O*, split into 6 nice triangles using the star-shaped technique
- Lemma: If T is an arbitrary triangle and O is exterior to T, then we can partition T into at most 4 nice triangles.

**Theorem:** Given any triangulation of *P* into n - 2 triangles, we can refine it into  $\leq 4n - 6$  nice triangles.





- Soft Predicates: similar to the technique for star-shaped robots
- *φ̃<sub>j</sub>*(*B*) comprises those features *f* satisfying

   (1) Sep(*m<sub>B</sub>*, *f*) ≤ *r<sub>B</sub>* + *r<sub>j</sub>* (2) *f* also intersects the *r<sub>B</sub>*-expansion of *T*TS<sub>*j*</sub> NSS<sub>*j*</sub>



# **Experimental Results**

• Created challenging environments with several complex robots.







# Summary of Experimental Results

Comparing with several sampling methods (PRM, RRT, EST, KPIECE) in open-source library OMPL.

- OMPL planners often have unsuccessful runs and have to time out even when there is a path.
- Our algorithms perform in real time, often much faster than OMPL planners, in addition to being able to report NO-PATH.



# Video Demo

- Video is available at (link given in the paper) <u>https://cs.nyu.edu/exact/gallery/complex/complex-robot-demo.mp4</u>
- Code is available (link given in the paper): Core Library <u>https://cs.nyu.edu/exact/core\_pages/downloads.html</u>



# Conclusions

- We extended our SSS resolution-exact approach to challenging planning problems where no exact algorithms exist.
- Experiments show that our methods typically outperform OMPL sampling methods.
- Open Problems:
- (1) Optimal decomposition of m-gons into nice triangles?
- (2) Complex rigid robots in 3D?