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# Motion Planning via Manifold Samples\* (MMS)

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\*Joint work with Michael Hemmer, Barak Raveh and Dan Halperin



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# Outline

- Background
- Hybrid Motion Planners
- Motion Planning via Manifold Samples (MMS)
- Specific Implementation

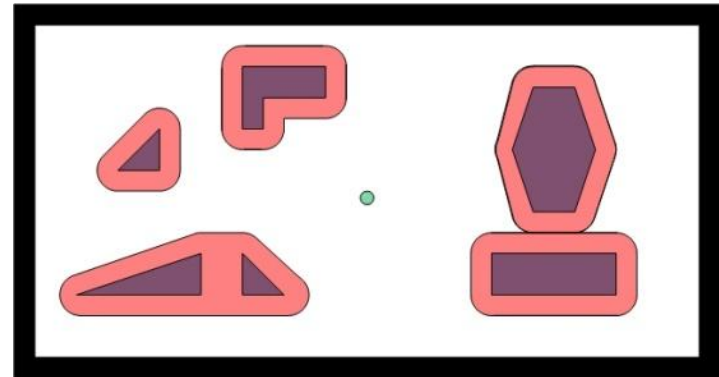
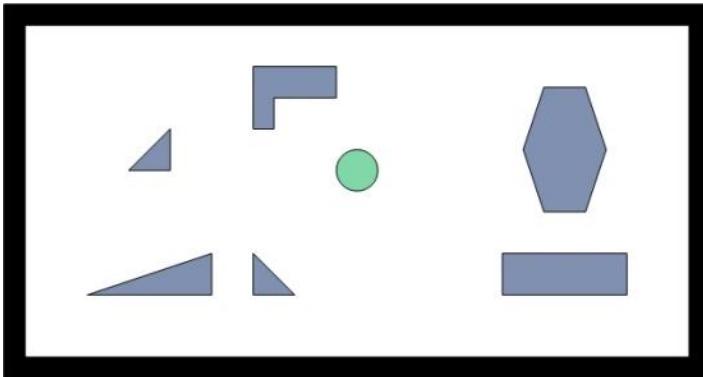
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# Motion Planning - Definitions

- **Workspace** – A description of the (2D or 3D) world consisting of a **robot** and **obstacles**
- **Configuration Space- (C)** The space of parameters that define the robot's position and orientation in the workspace

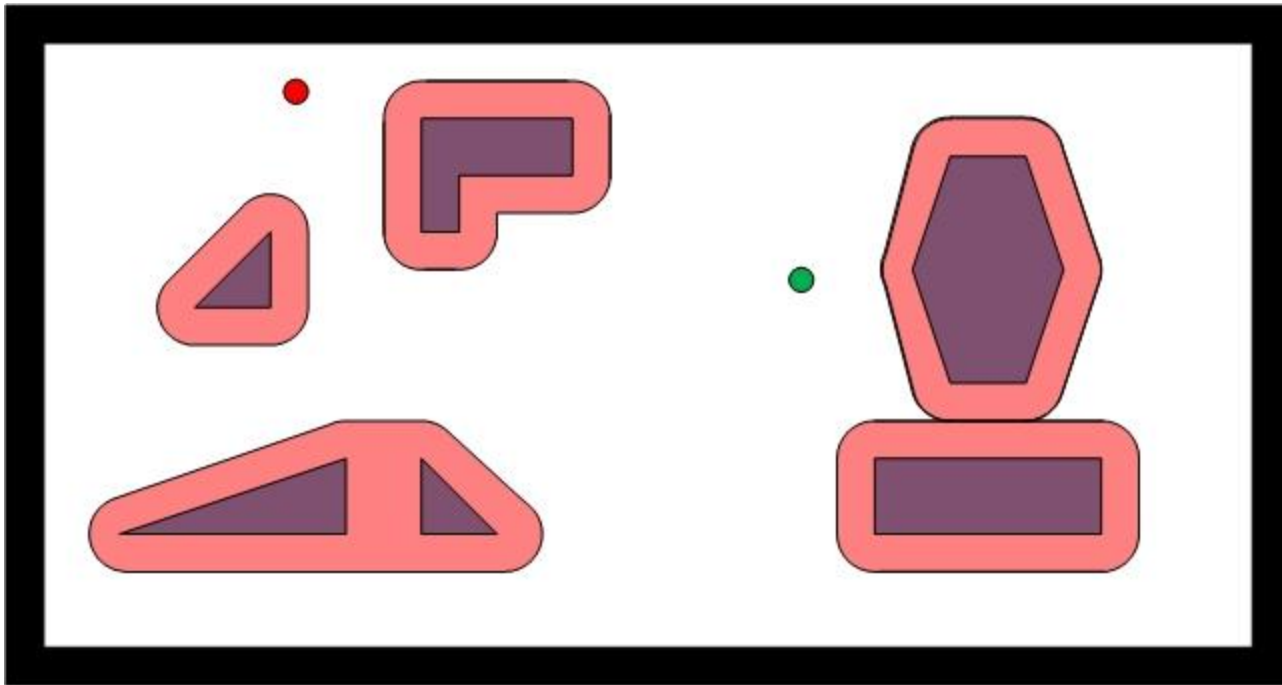


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- **Configuration Space- (C)** The space of parameters that define the robot's position and orientation in the workspace
- **Degrees of Freedom-** The minimal number of parameters required to uniquely define a position of the robot
- **Free Space ( $C_{\text{free}}$ )-** Set of collision-free configurations
- **Forbidden Space ( $C_{\text{forb}}$ )-**  $C \setminus C_{\text{free}}$

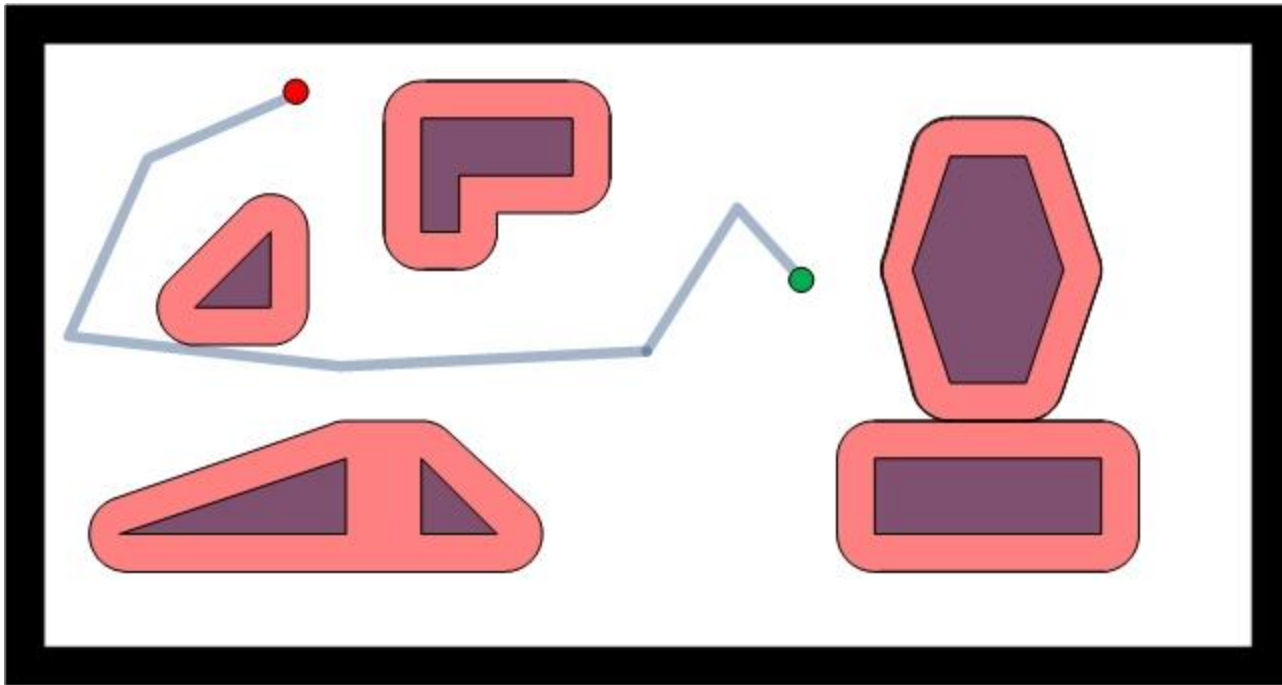
# Motion Planning - Objective

- Find a path in  $C_{\text{free}}$  from a free source configuration to a free target configuration



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# Algorithmic Approaches for Motion Planning

- Sampling-Based Planning
  - Capture connectivity of  $C_{\text{free}}$  by randomly **sampling** configurations
- Combinatorial Motion Planning
  - **Analytically** compute an explicit combinatorial representation of  $C_{\text{free}}$



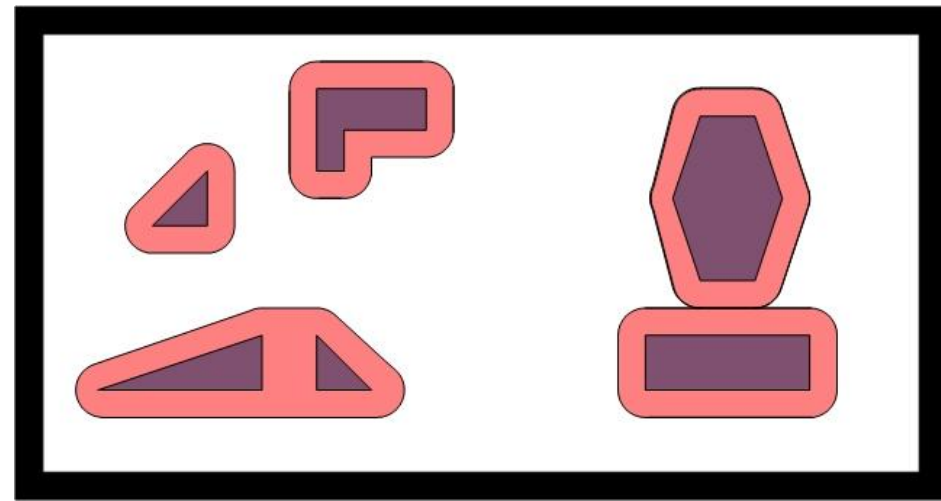
# Algorithmic Approaches for Motion Planning

## ■ Sampling-Based Planning

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- Kavraki, Svestka, Latombe, Overmars 96: Probabilistic roadmaps for path planning in high dimensional configuration spaces (PRM)
- LaValle 98: Rapidly-exploring random trees: A new tool for path planning (RRT)
- Hsu, Latombe, Motwani 99: Path planning in expansive configuration spaces (EST)

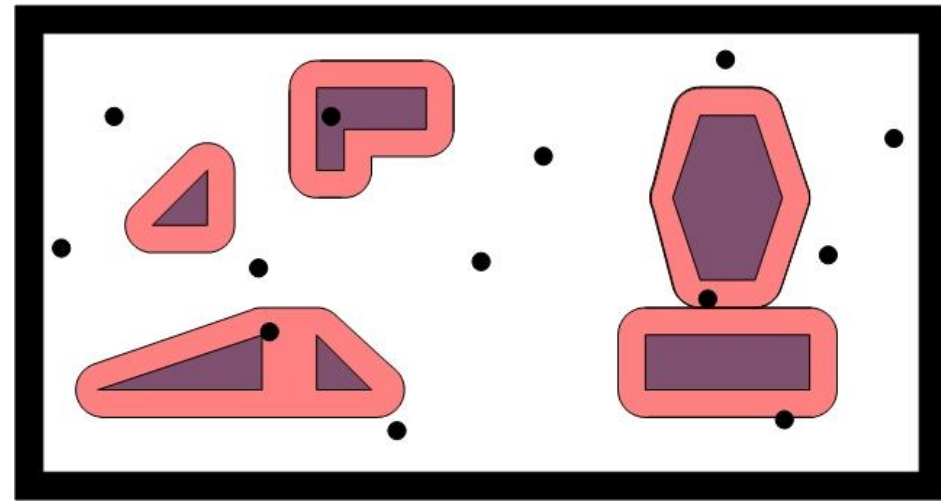
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- Multi query planner
- Preprocesses configuration space into a graph (roadmap)



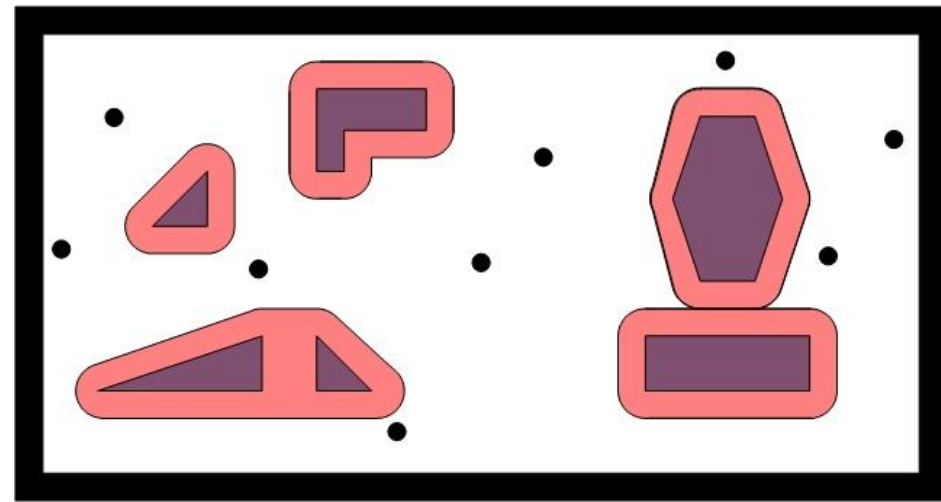
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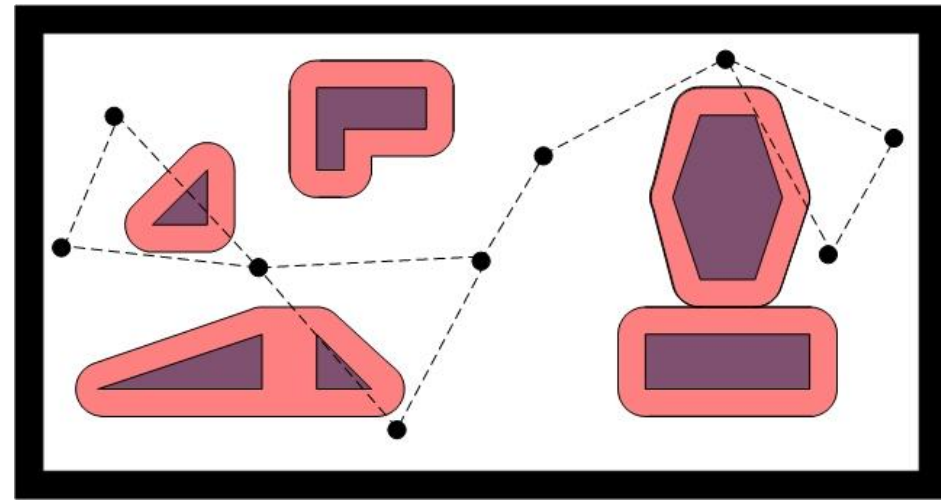
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  - Discard invalid configurations



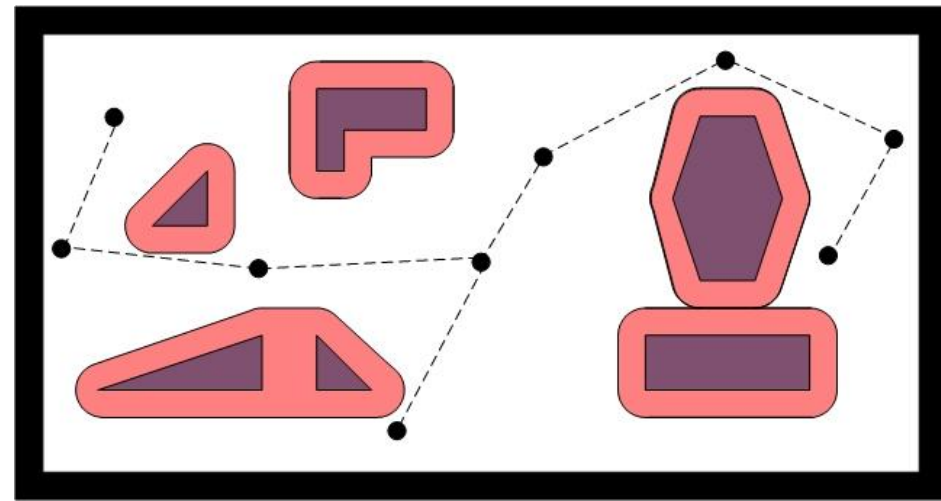
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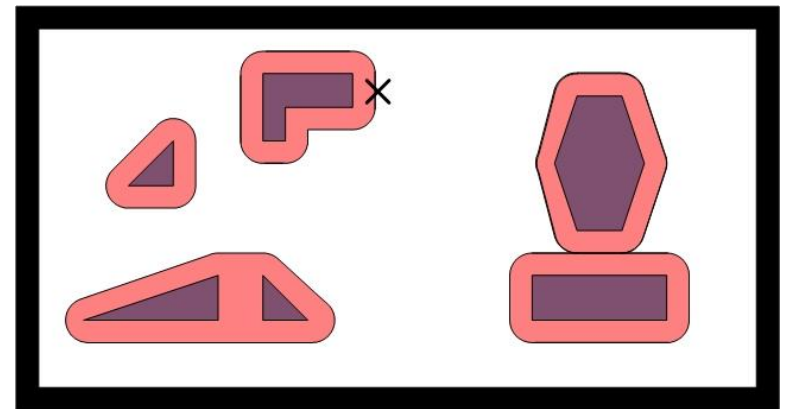
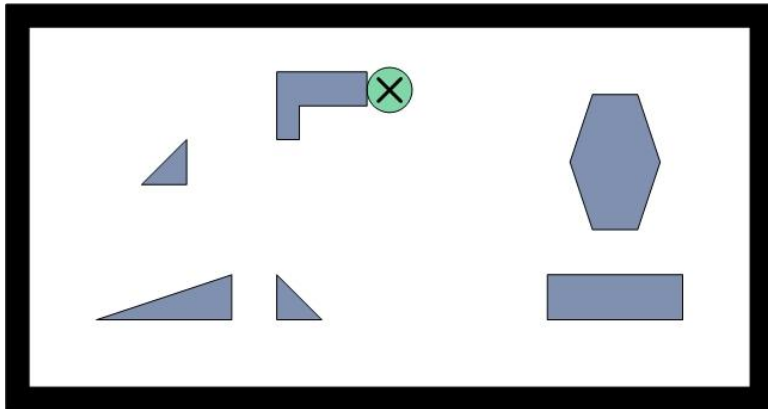
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- Multi query planner
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  - Connect close-by configurations by dense sampling (“local-planning”)
  - Discard invalid edges



# Combinatorial Motion Planning

- Analytically compute an explicit combinatorial representation of  $C_{\text{free}}$ 
  - Using critical hyper-surfaces\*

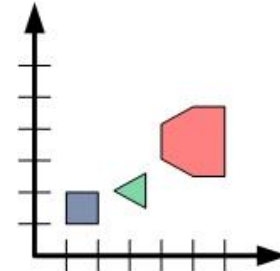


\*Schwartz, Sharir 83: On the "piano movers" problem. II. General techniques for computing topological properties of real algebraic manifolds

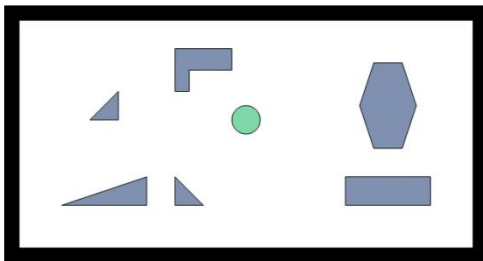
# Combinatorial Motion Planning (cont.)

- Minkowski Sums –

$$P \oplus Q = \{p + q \mid p \in P, q \in Q\}$$



- Allow representation of the configuration space of a translating robot

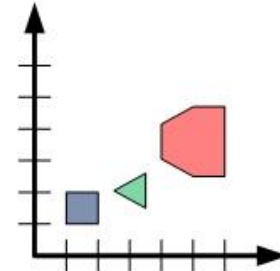




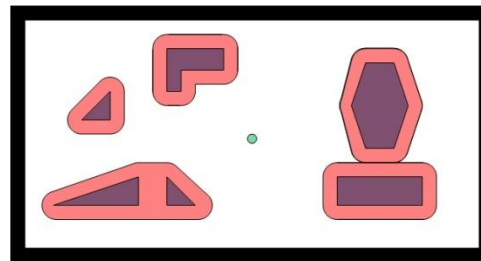
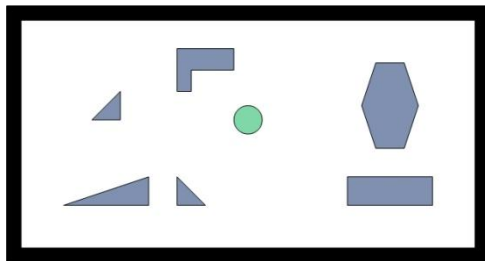
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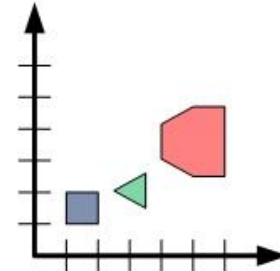
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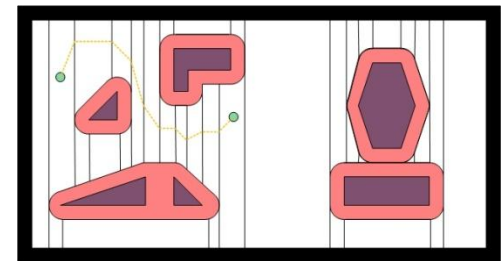
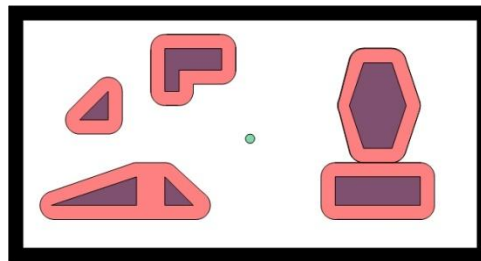
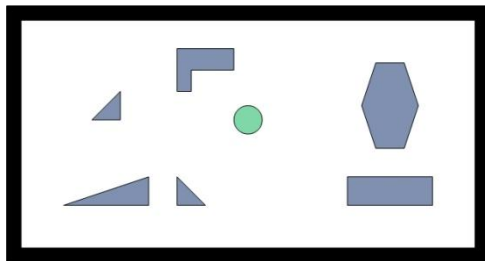
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# Advantages and Limitations of Approaches

## Probabilistic planning

- ✓ **Easy** to implement
- ✓ Applicable to **high-dimension** C-spaces
- ✗ Sensitive to **tight passages**

## Combinatorial planning

- ✗ Complex implementations
- ✗ Exponential in the number of degrees of freedom\*
- ✓ **Analytic complete** representation

\*Reif 79: Complexity of the mover's problem and generalizations

# Hybrid Planners

- S. Hirsch and D. Halperin. Hybrid motion planning: Coordinating two discs moving among polygonal obstacles in the plane. WAFR 2002 [HH02]
- Liangjun Zhang, Young J. Kim, and Dinesh Manocha. A hybrid approach for complete motion planning. IROS 2007 [ZKM07]
- Jade Yang and Elisha Sacks. RRT path planner with 3 DOF local planner. ICRA, 2006 [YS06]
- Ming Lien, J.: Hybrid motion planning using Minkowski sums. RSS 2008 [Lie08]

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# Existing Hybrid Planners - Limitations

- Applicable for low dimensions
- Applicable to specific instances

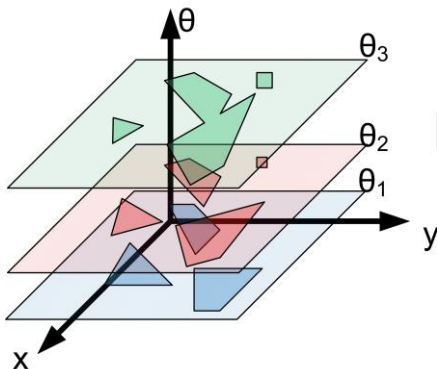
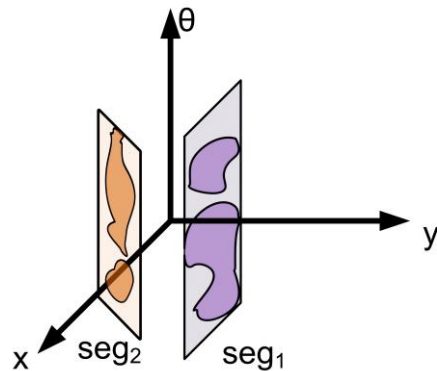
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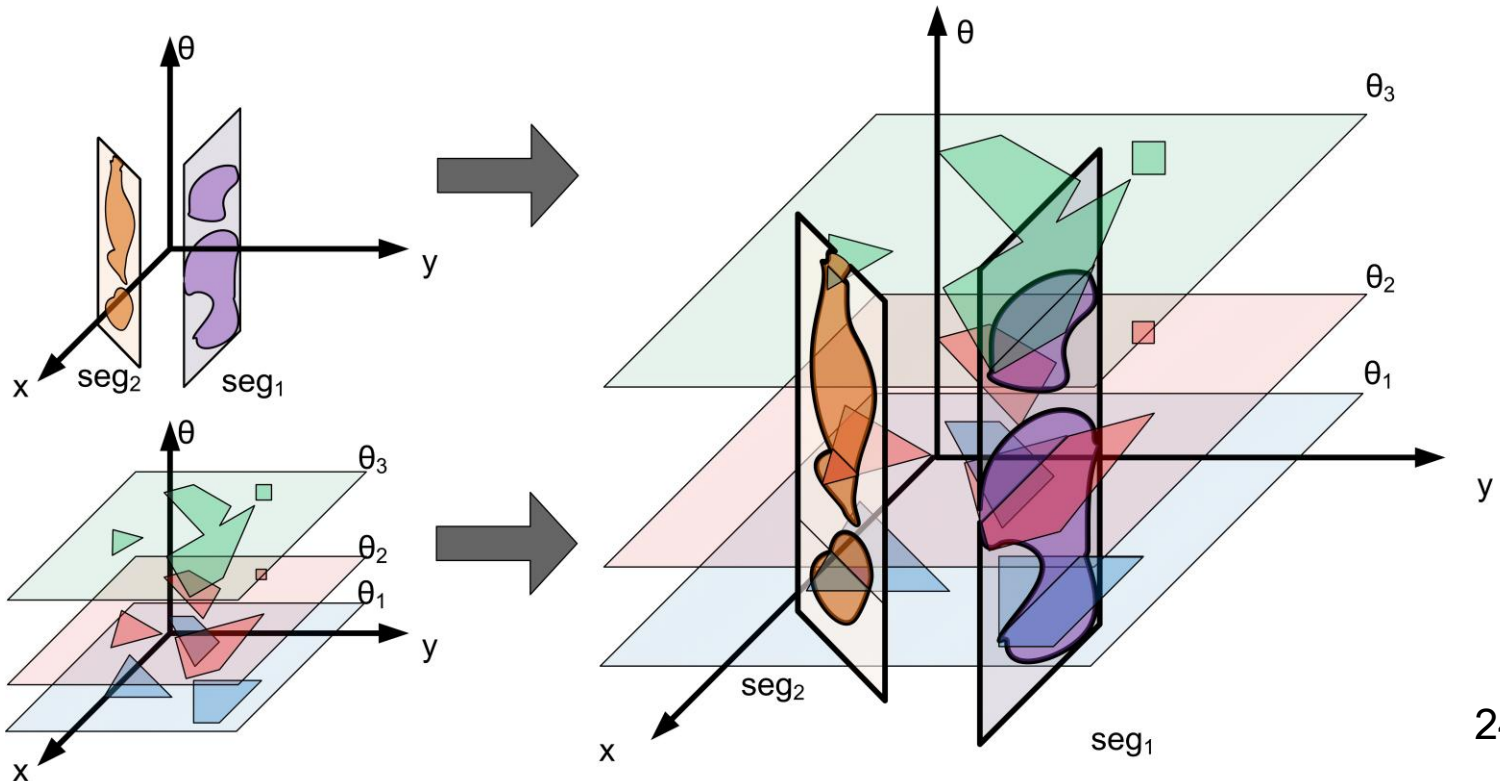
# Overview

- Sampling-based multi-query planner
- Samples are entire manifolds of low dimensions
- Manifolds are **decomposed** analytically into cells
  - A cell in  $C_{\text{free}}$  is a **Free Space Cell (FSC)**



# Overview

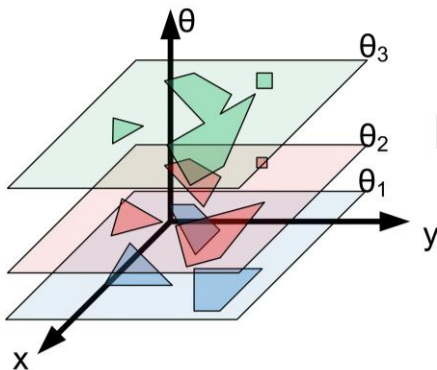
- **Preprocessing stage** - construct graph  $G = (V, E)$ 
  - $V$  – FSCs
  - $E$  – Intersecting FSCs
- **Query stage**





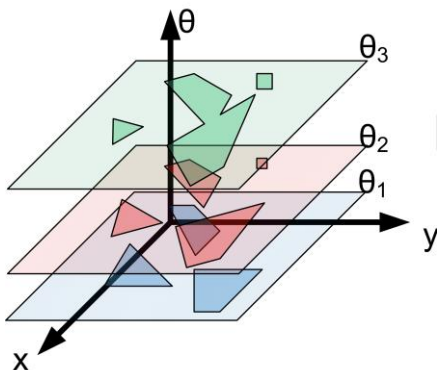
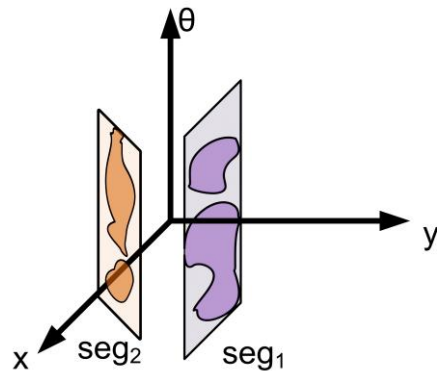
# Exploration Vs. Connection

- Manifold samples add
  - **vertices** (new connected components)



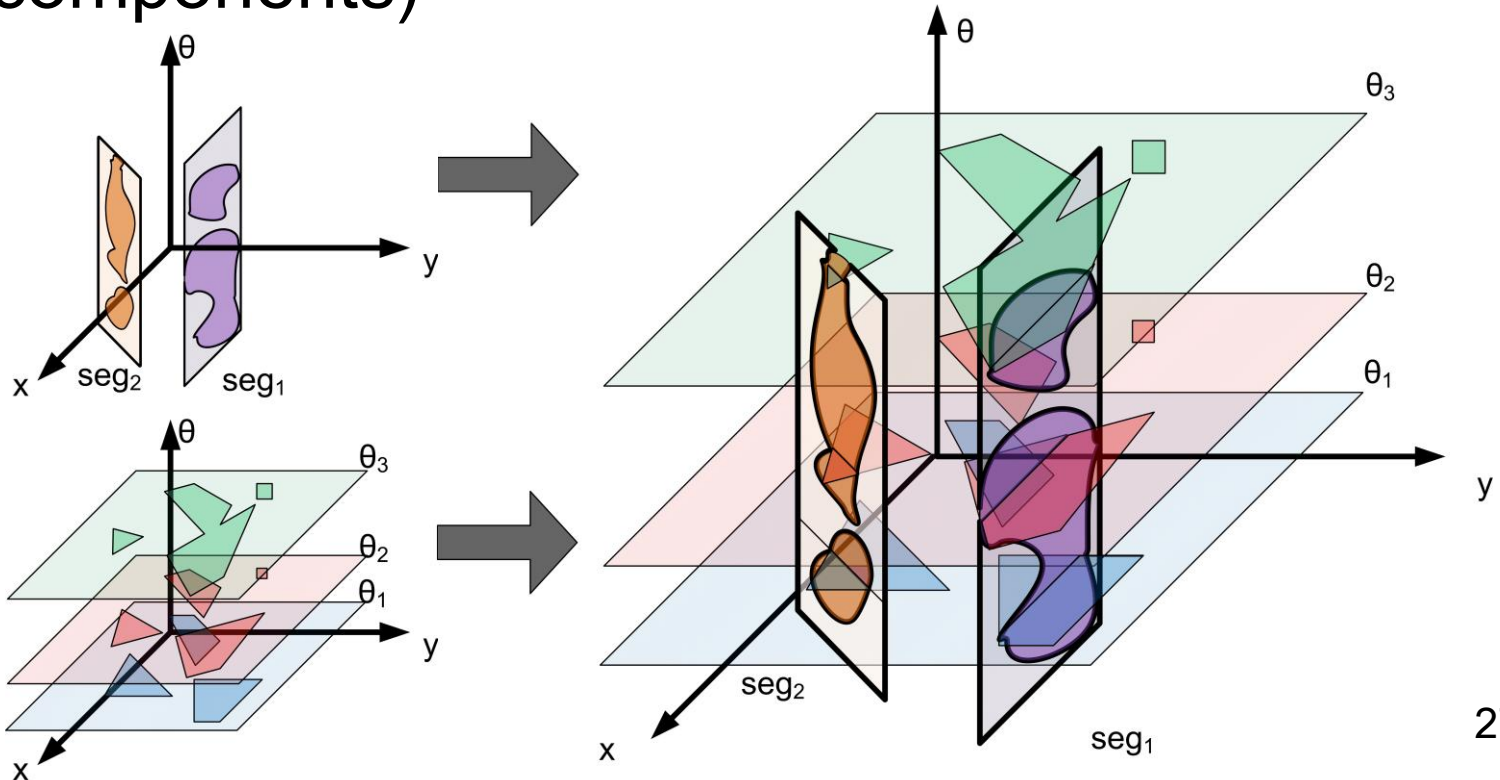
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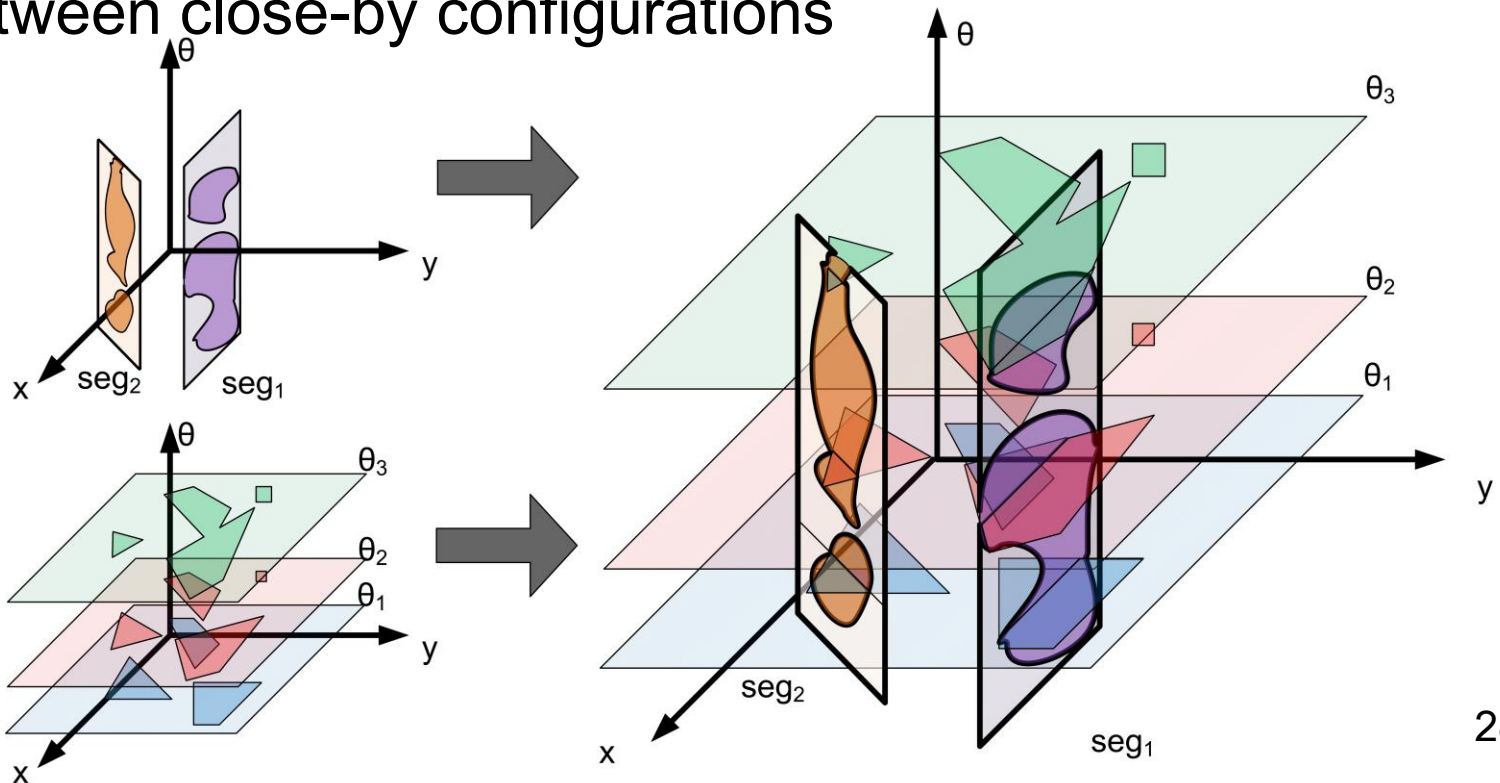
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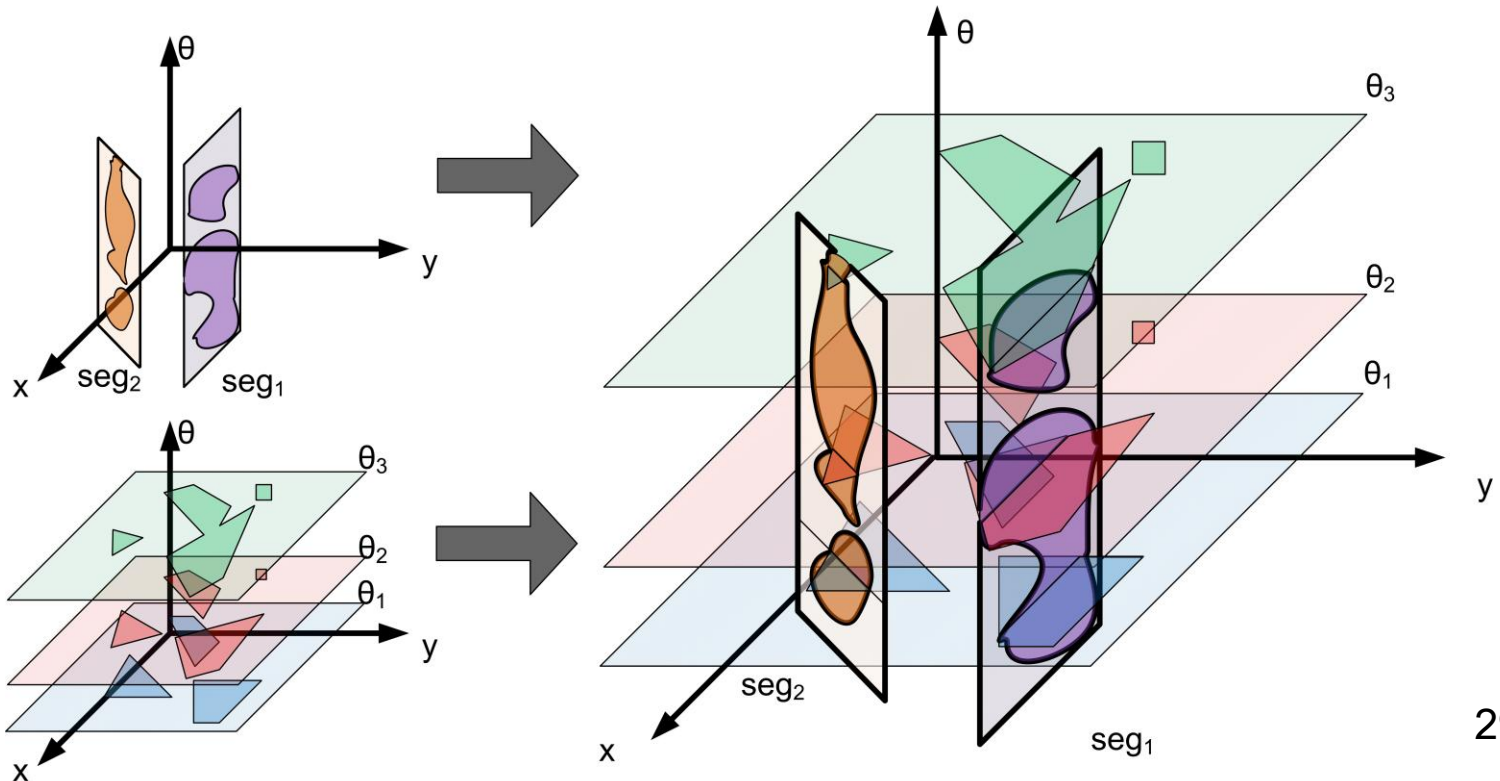
# Desired Properties of Manifolds

- **Simplicity:** Easy representation, construction and decomposition
- **Covering:** Manifolds should be dense
- **Spanning:** Manifolds should allow local connections between close-by configurations



# Comparison With PRM

	PRM	MMS
Sample type	Point	Manifold
Decomposition	Collision detector	Analytic primitive
Node type	Point	FSC
Node connection	Interpolation	Analytic (intersection)
Data structure	Roadmap graph	Geometric intersection graph of FSCs



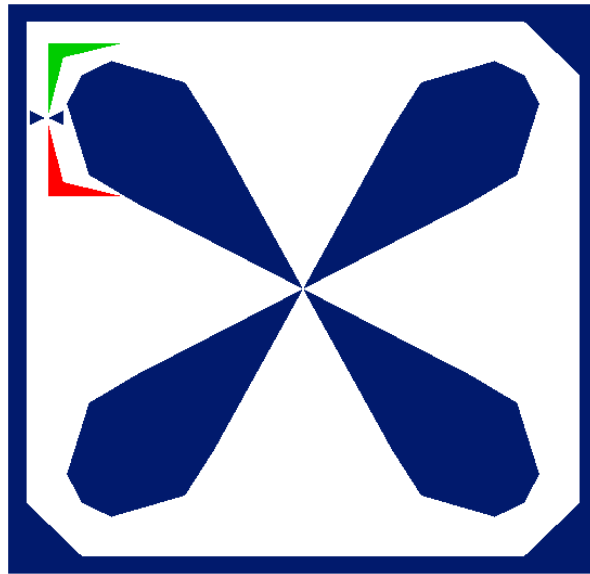
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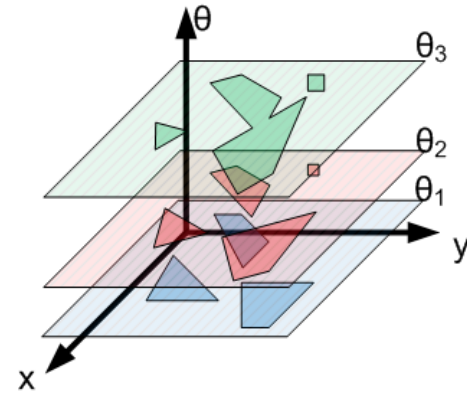
# The Setting

- Two-dimensional polygonal robot  $R$
- Three-dimensional configuration space
  - translation and rotation



# Families of Manifolds

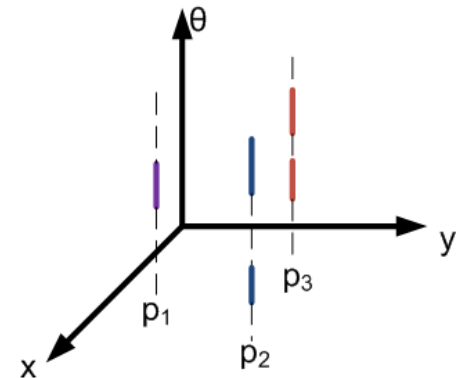
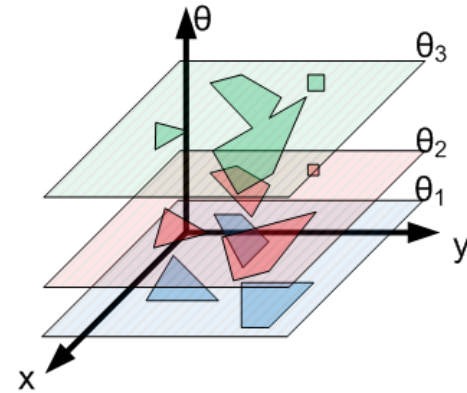
- Fixed rotation angle
  - Horizontal planes
  - Computation via Minkowski sums,





# Families of Manifolds

- **Fixed rotation angle**
  - Horizontal planes
  - Computation via Minkowski sums,
  
- **Fixed reference point**
  - Vertical lines
  - Computed analytically via critical angles



# Fixed Rotation Angle (details)

- (Thm) Let  $R(x,y)$  be a robot placed at  $x,y$  and  $P$  be an obstacle, then  $R$  intersects  $P$  iff

$$(x,y) \in P \oplus -R(0,0)$$

- If  $R(x,y)$  intersects  $P$

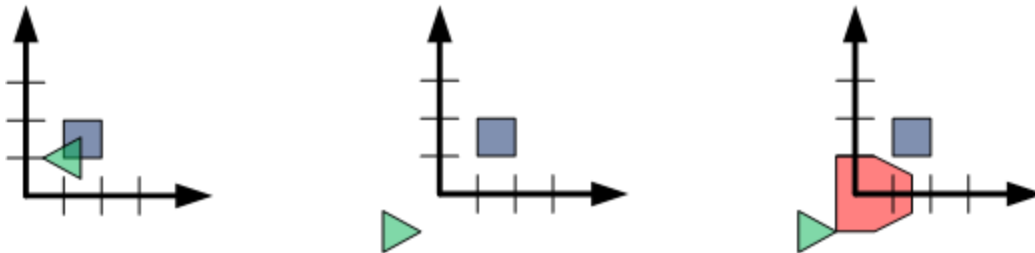
- Let  $q$  be the intersection point
- $q \in R(x,y) \Rightarrow q - (x,y) \in R(0,0) \Rightarrow -q + (x,y) \in -R(0,0)$
- $q \in P$
- Thus,  $(x,y) \in P \oplus -R(0,0)$

- If  $(x,y) \in P \oplus R(0,0)$

- There are points  $r \in R(0,0)$ ,  $p \in P$  s.t.
- $(x,y) = p - r \Rightarrow p = (x,y) + r$
- $\Rightarrow$  The robot placed at  $(x,y)$  intersects  $P$

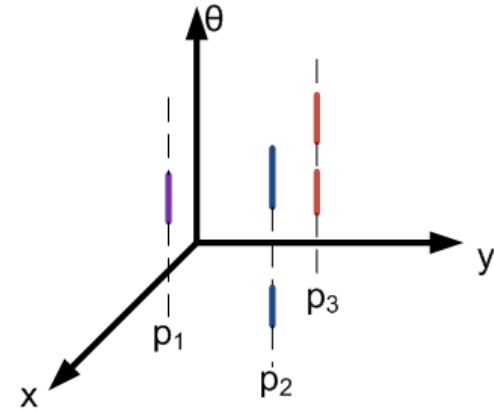
# Fixed Rotation Angle (details)

- For a robot  $R$  with its reference point at the origin and an obstacle  $O$ , the forbidden space is represented by  $-R \oplus O$



# Fixed Reference Point (details)

- Parameterization:
  - $\alpha \in [0, 1]$  reference point on segment



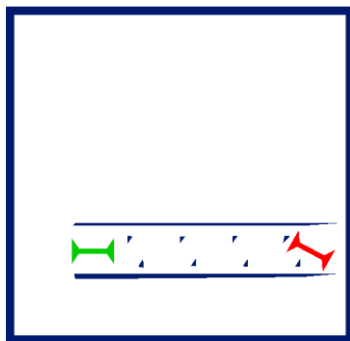
- Parameterized **critical angles** are in the form of algebraic numbers\*

\*Algebraic number - a number that is a root of a non-zero polynomial in one variable with rational coefficients

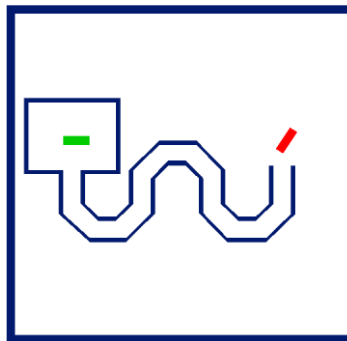
# Experimental Results

## ■ Scenarios

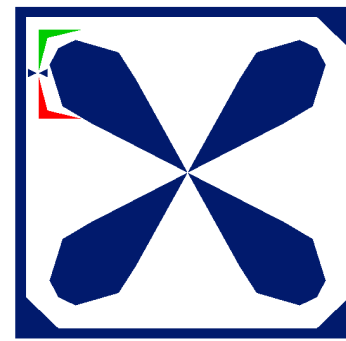
Tunnel



Snake



Flower

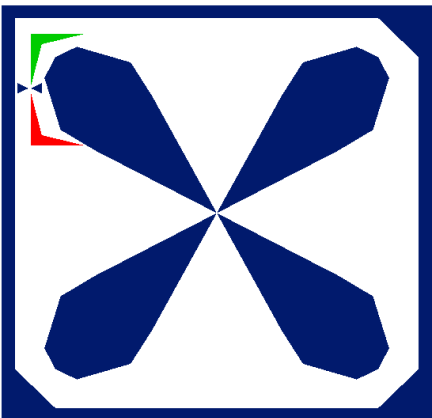
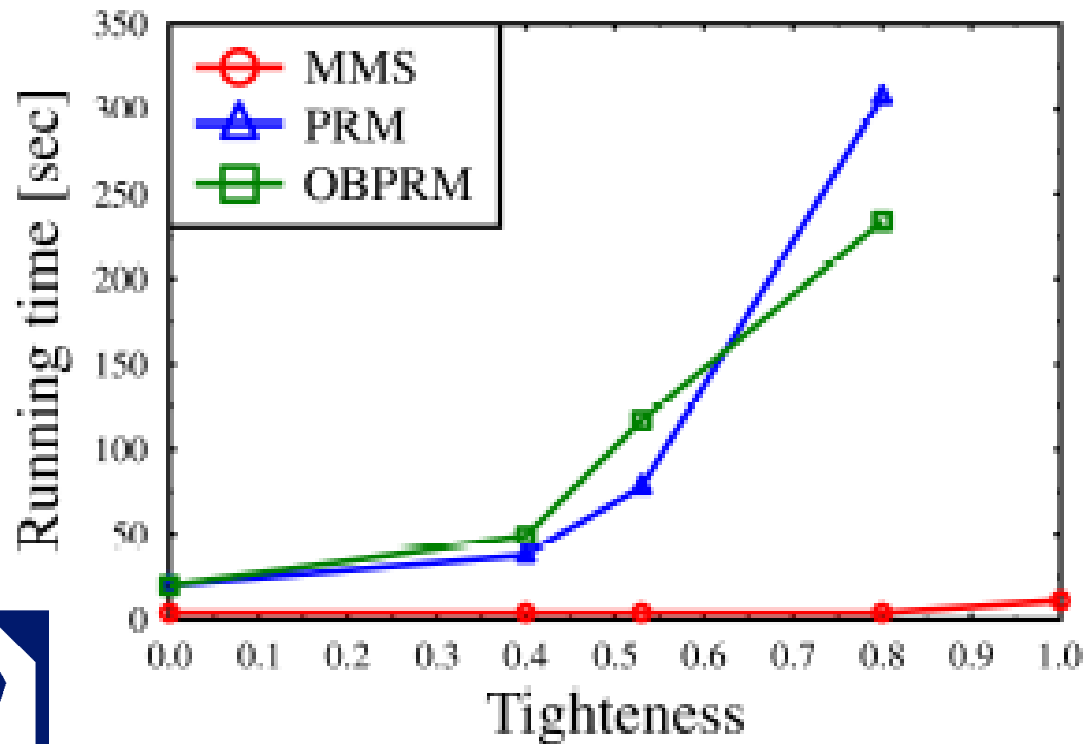


## ■ OOPSMP PRM Implementation

Scenario	MMS			PRM			OBPRM		
	$n_\theta$	$n_x$	t[sec]	k	res	t[sec]	k	res	t[sec]
Tunnel	20	128	<b>21</b>	10	0.005	<b>114</b>	10	0.01	<b>134</b>
Snake	40	256	<b>35</b>	10	0.02	<b>264</b>	10	0.01	<b>247</b>
Flower	40	256	<b>4</b>	20	0.02	<b>20</b>	14	0.01	<b>20</b>

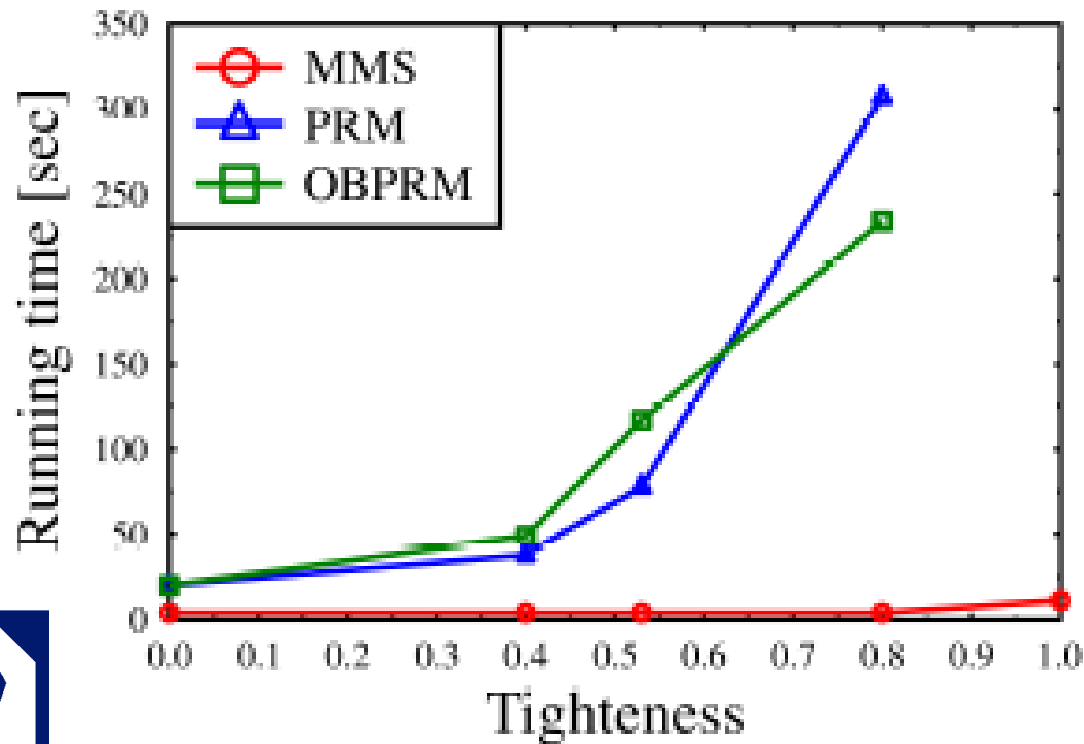
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- Tightening the configuration space



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20-fold speedup

