

# Probabilistic Motion Planning

- Configuration Space
  - Any admissible motion for the 3D mechanical system appears a collision-free path for a point in the CSpace
  - Translating the ***continuous*** problem into a ***combinatorial*** one
  - Capturing the ***topology*** of *CS<sub>free</sub>* with ***graphs***

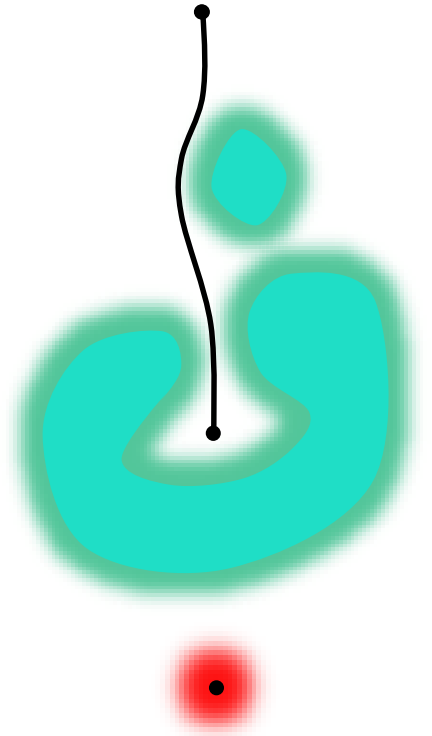
- **The Piano Mover problem**
  - **80's The complexity bottleneck**
  - **90's The amazing efficiency of empiricism**

- **The Piano Mover problem: probabilistic approaches**
  - **Gradient descents and random walks**
  - **Sampling methods: PRM**
  - **Diffusion algorithms: RRT**

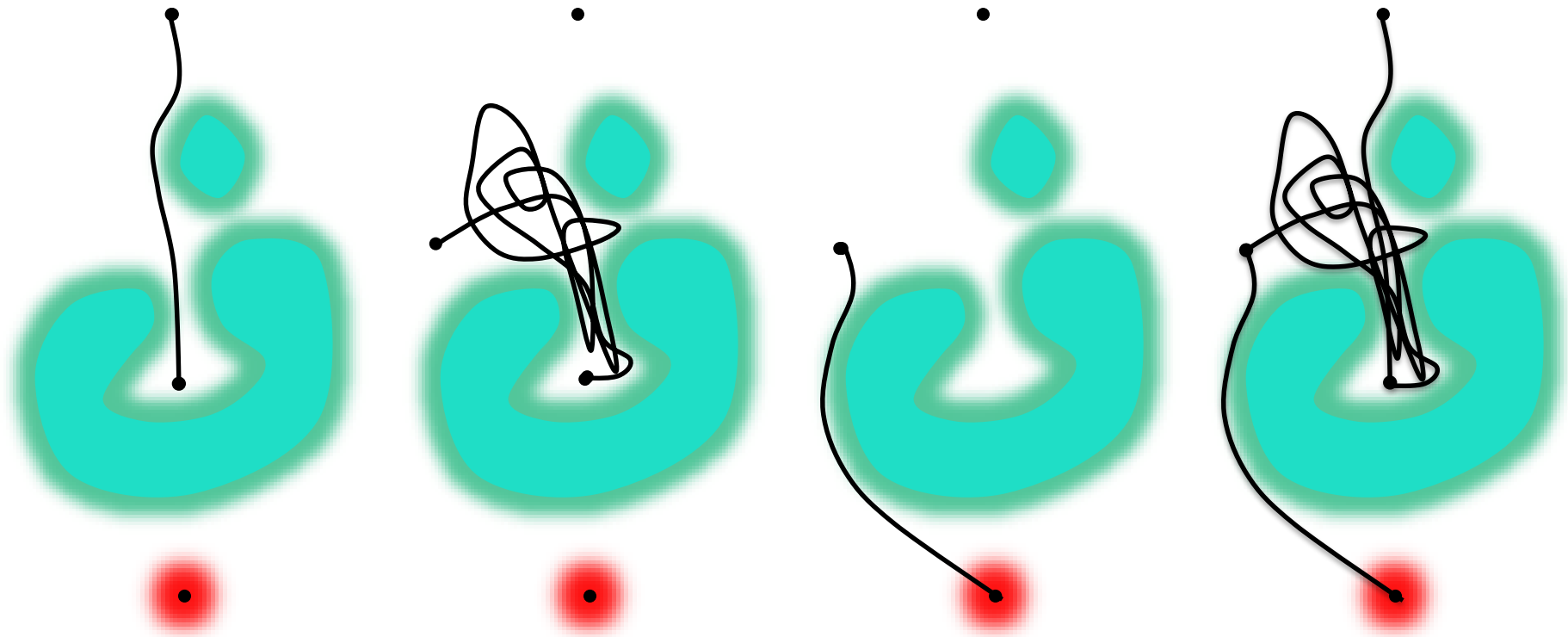
# Probabilistic Motion Planning



- **The Piano Mover problem: probabilistic approaches**
  - **Gradient descents and random walks**
  - Sampling methods: PRM
  - Diffusion algorithms: RRT



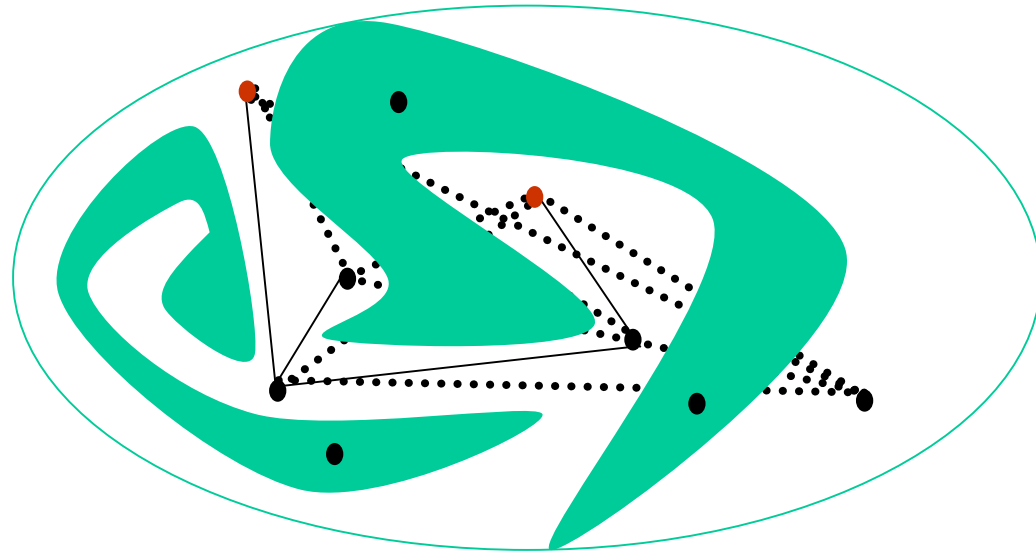
# Probabilistic Motion Planning





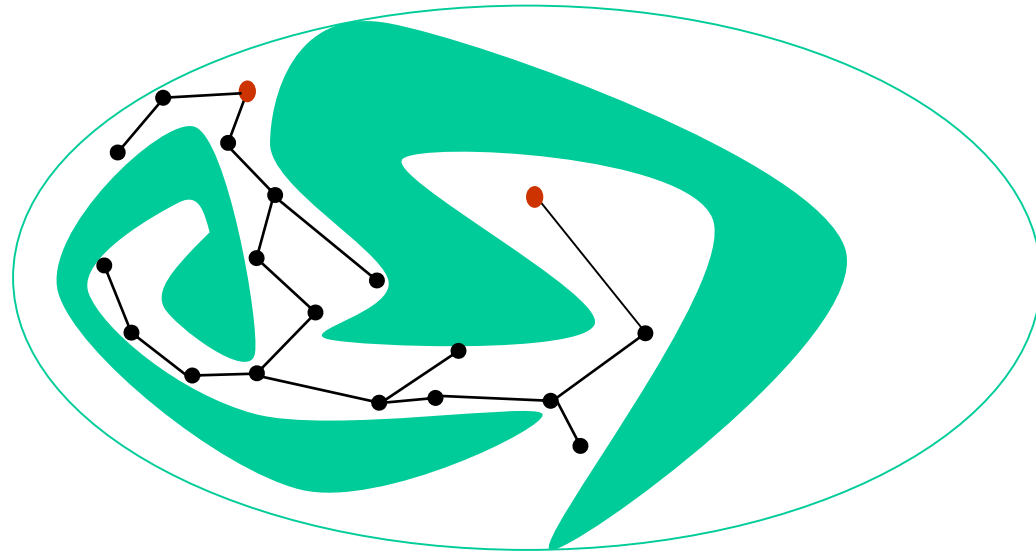
- **The Piano Mover problem: probabilistic approaches**
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  - **Sampling methods: PRM**
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# Probabilistic Motion Planning



- **The Piano Mover problem: probabilistic approaches**
  - Gradient descents and random walks
  - Sampling methods: PRM
  - **Diffusion algorithms: RRT**

# Probabilistic Motion Planning



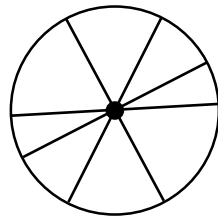
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## Combinatorial Topology and Visibility based Probabilistic Methods in Motion Planning

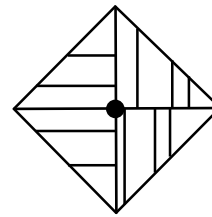
T. Siméon, J.P. Laumond, C. Nissoux,  
Visibility based probabilistic roadmaps for motion planning  
*Advanced Robotics Journal*, Vol. 14, N° 6, 2000

J.P. Laumond, T Siméon,  
Notes on visibility roadmaps and path planning  
in *Algorithmic and Computational Robotics: New Directions*  
B. Donald, K. Lynch, D. Rus Eds, A.K. Peters, 2001

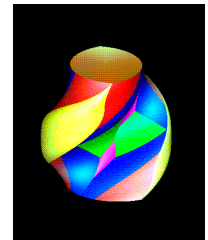
- Configuration Space Topology and Small Time Controllability
- Exemples:
  - linear interpolation,
  - Manhattan paths,
  - Reeds and Shepp paths,
  - flatness based methods



Euclidean



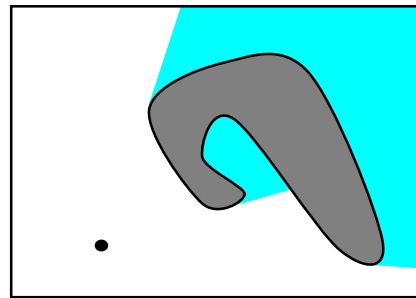
Manhattan



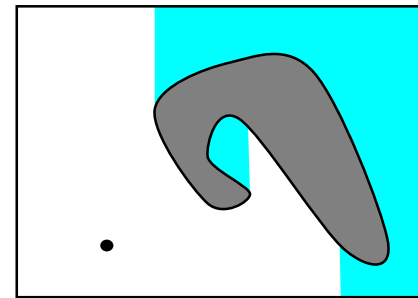
Reeds-Shepp

# Probabilistic Motion Planning

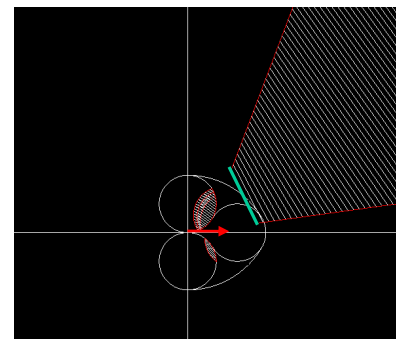
• Same topology, different visibility sets



Euclidean



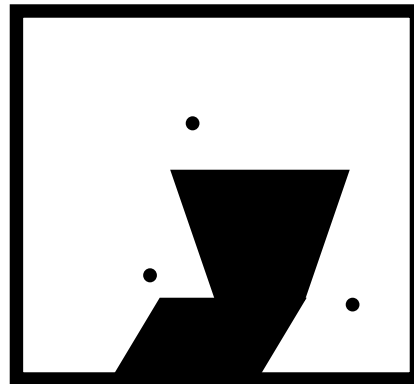
Manhattan



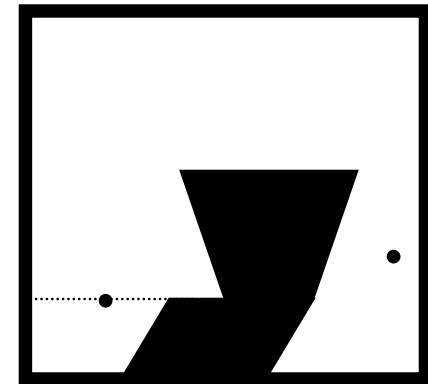
Reeds-Shepp

# Probabilistic Motion Planning

- Combinatorial topology (1)
- Existence and robustness of finite coverage



Euclidean

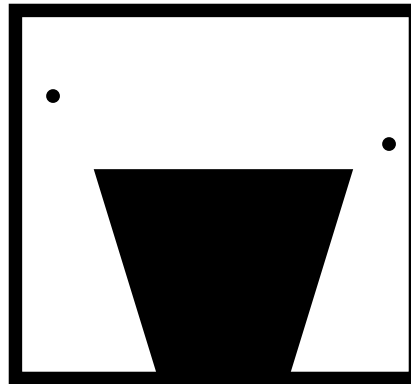


Manhattan

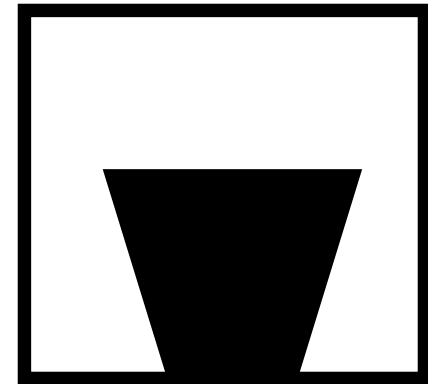


# Probabilistic Motion Planning

- Combinatorial topology (1)
- Existence and robustness of finite coverage



Euclidean



Manhattan

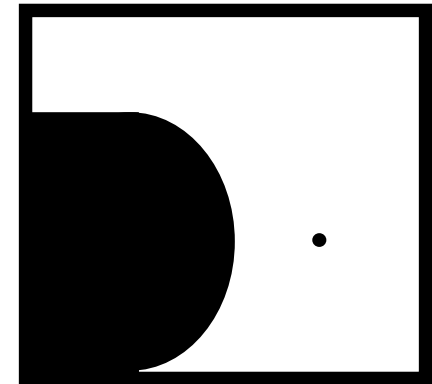
No

- Combinatorial topology (1)
- Existence and robustness of finite coverage

No

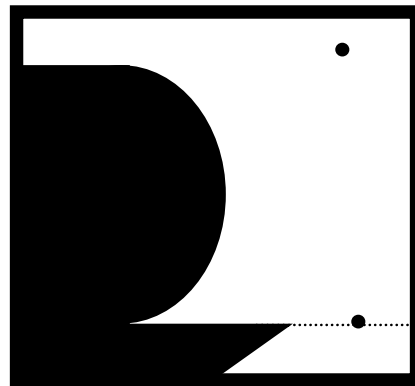


Euclidean

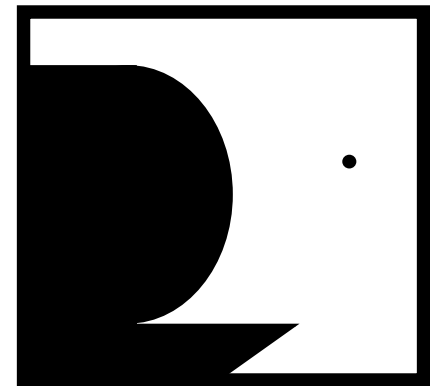


Manhattan

- Combinatorial topology (1)
- Existence and robustness of finite coverage



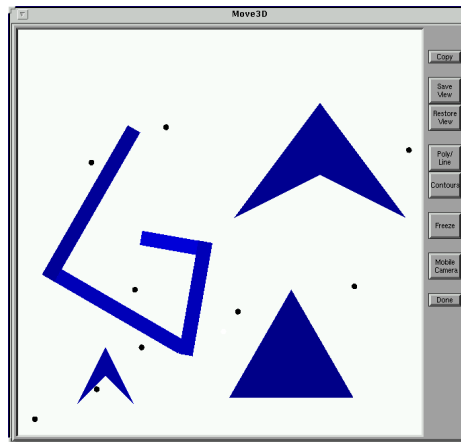
Euclidean



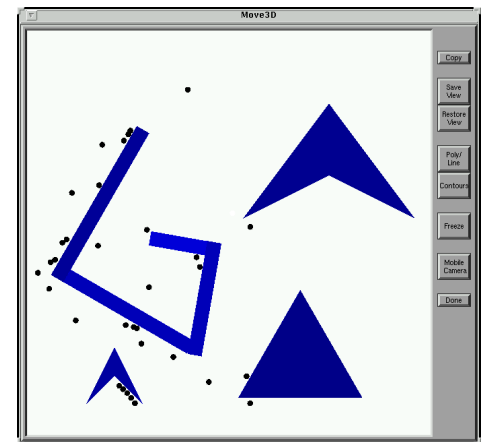
Manhattan

# Probabilistic Motion Planning

- Combinatorial topology (2)
- Optimal coverage (related to art gallery problem)



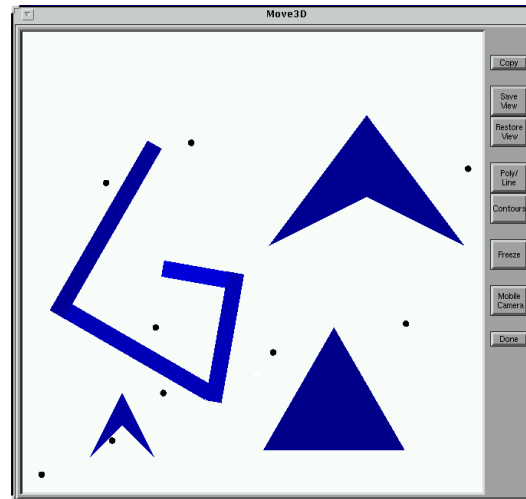
Euclidean



Manhattan

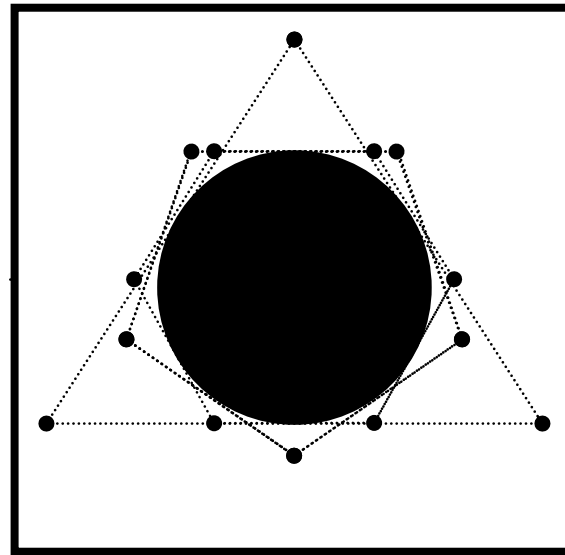
# Probabilistic Motion Planning

- Combinatorial topology (2)
- Optimal coverage: finite? bounded?



Euclidean and 2D polygonal obstacles: finite and bounded

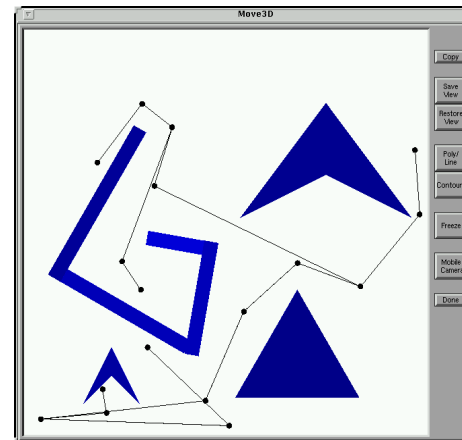
- Combinatorial topology (3)
- Optimal coverage: finite? bounded?



Euclidean: finite and unbounded

# Probabilistic Motion Planning

- Combinatorial topology (4)
- From (optimal) coverage to (visibility) roadmaps



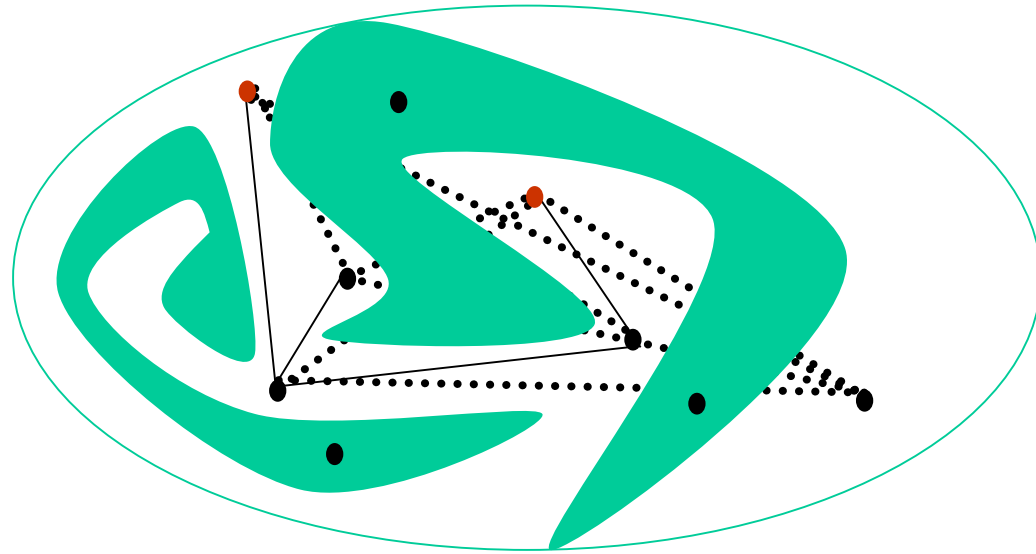
Guards + Connectors

- Computational challenge
  - No explicit knowledge of CS-obstacles
  - No explicit knowledge of visible (reachable) sets



- Computational challenge
- Probabilistic method ingredients:
  - A collision checker
  - A steering method
- Two type methods:
  - Learning CS topology by *sampling*
  - Answering single query by *diffusion*

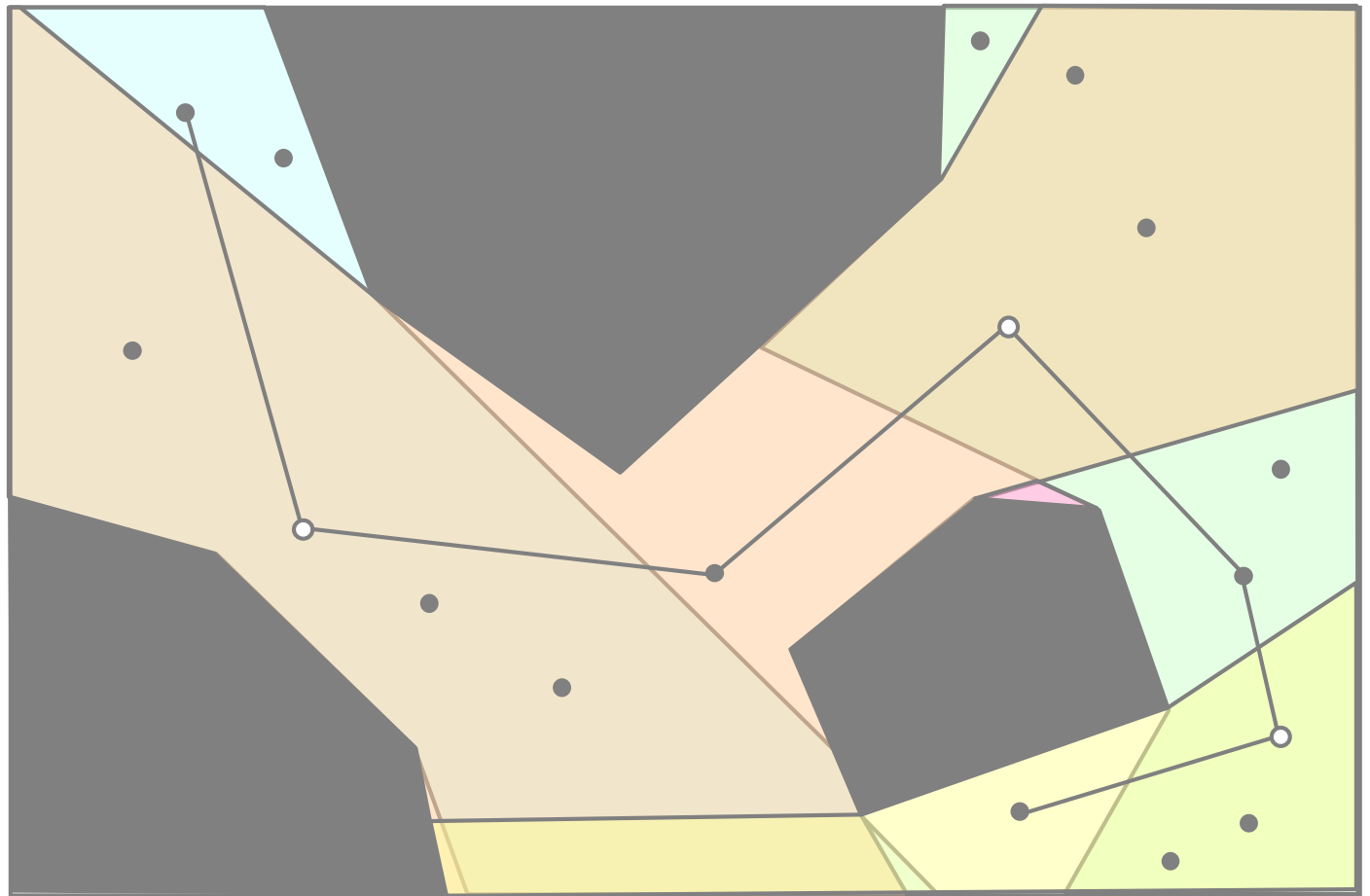
## • Computational challenge



Random sampling

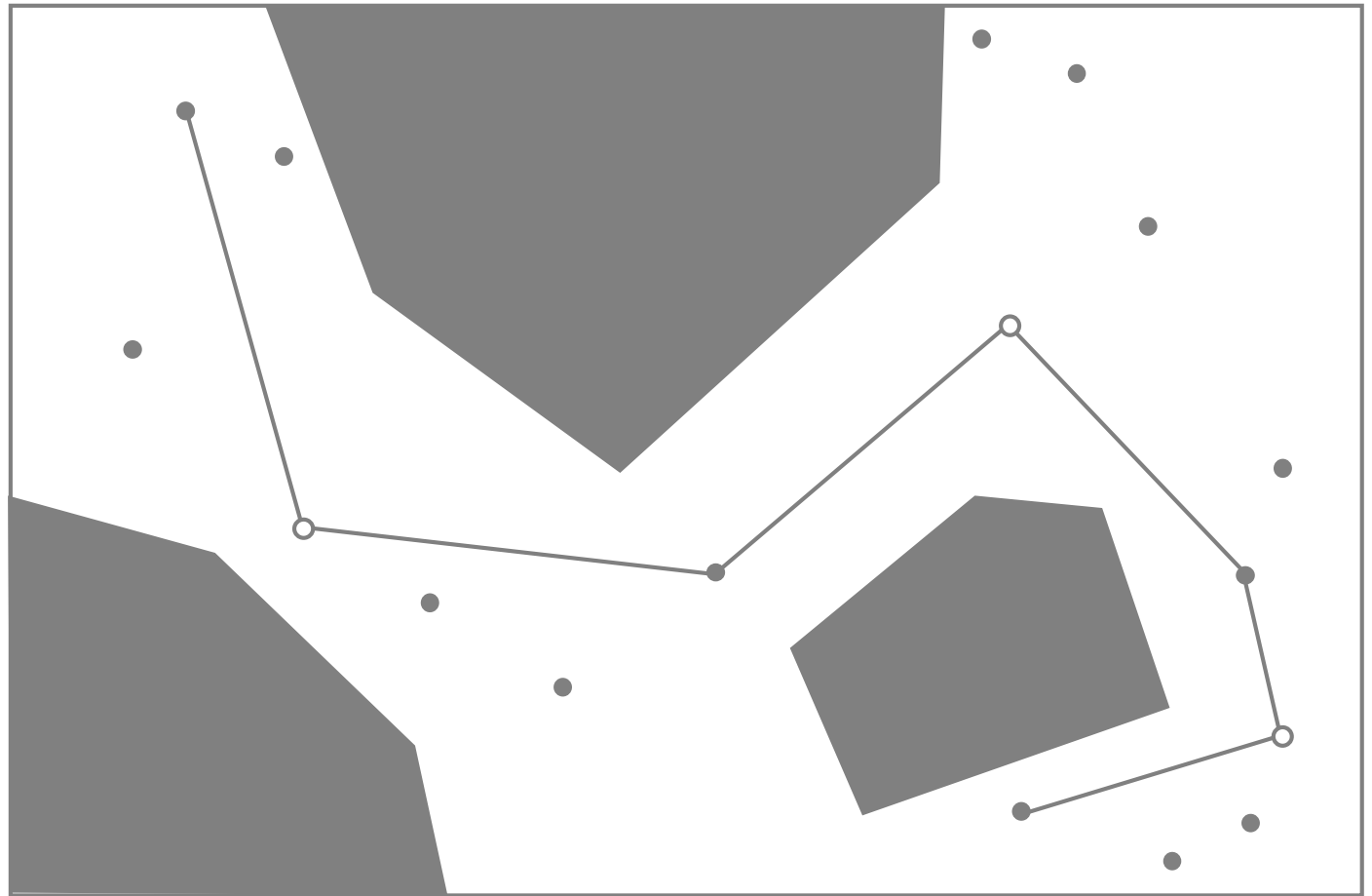
- Visibly based sampling
  - Algorithm:
    - Generate guards and connectors randomly
    - Stop after *#try* failures

## • Visibility based sampling



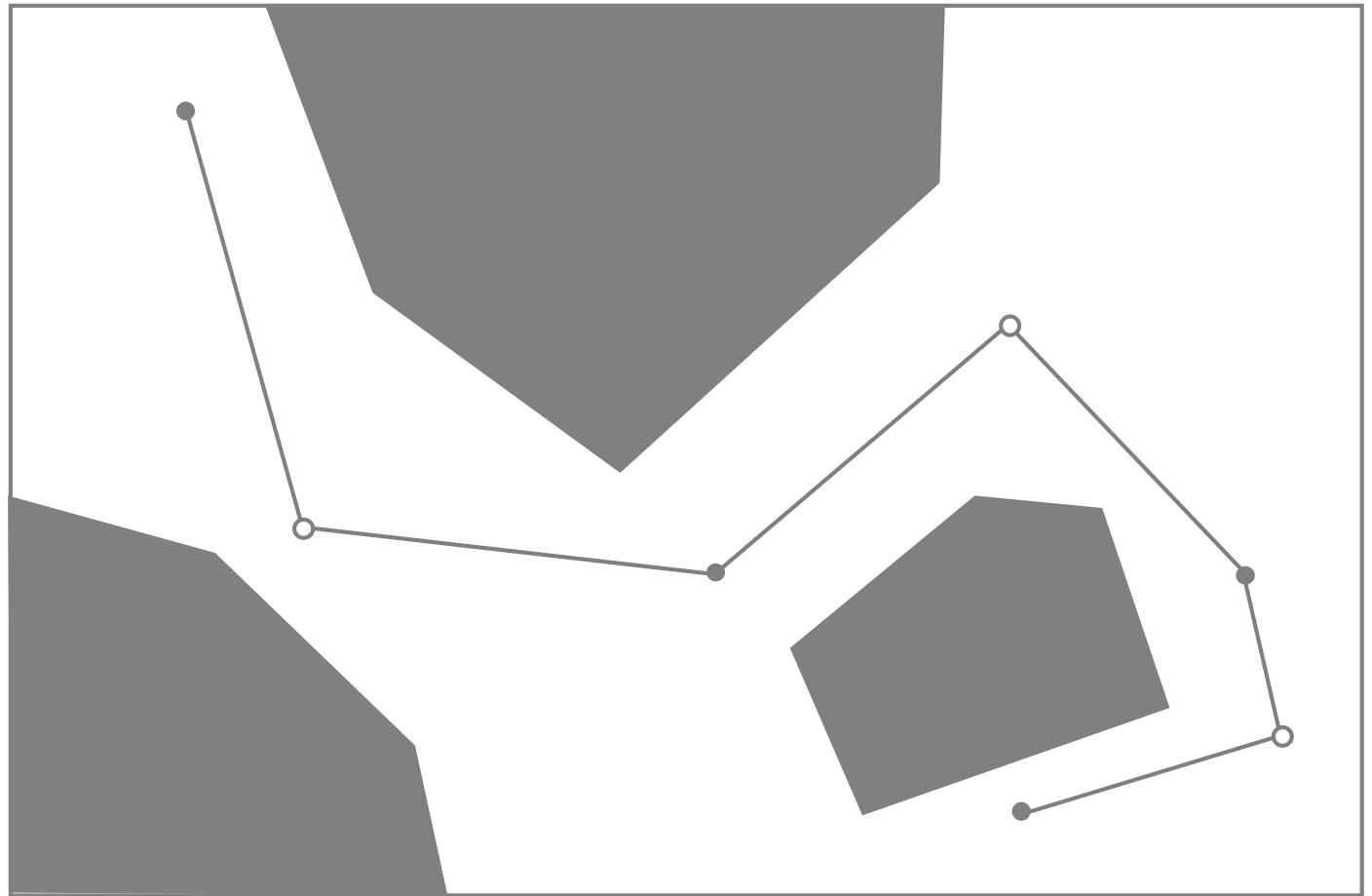
# Probabilistic Motion Planning

## • Visibility based sampling

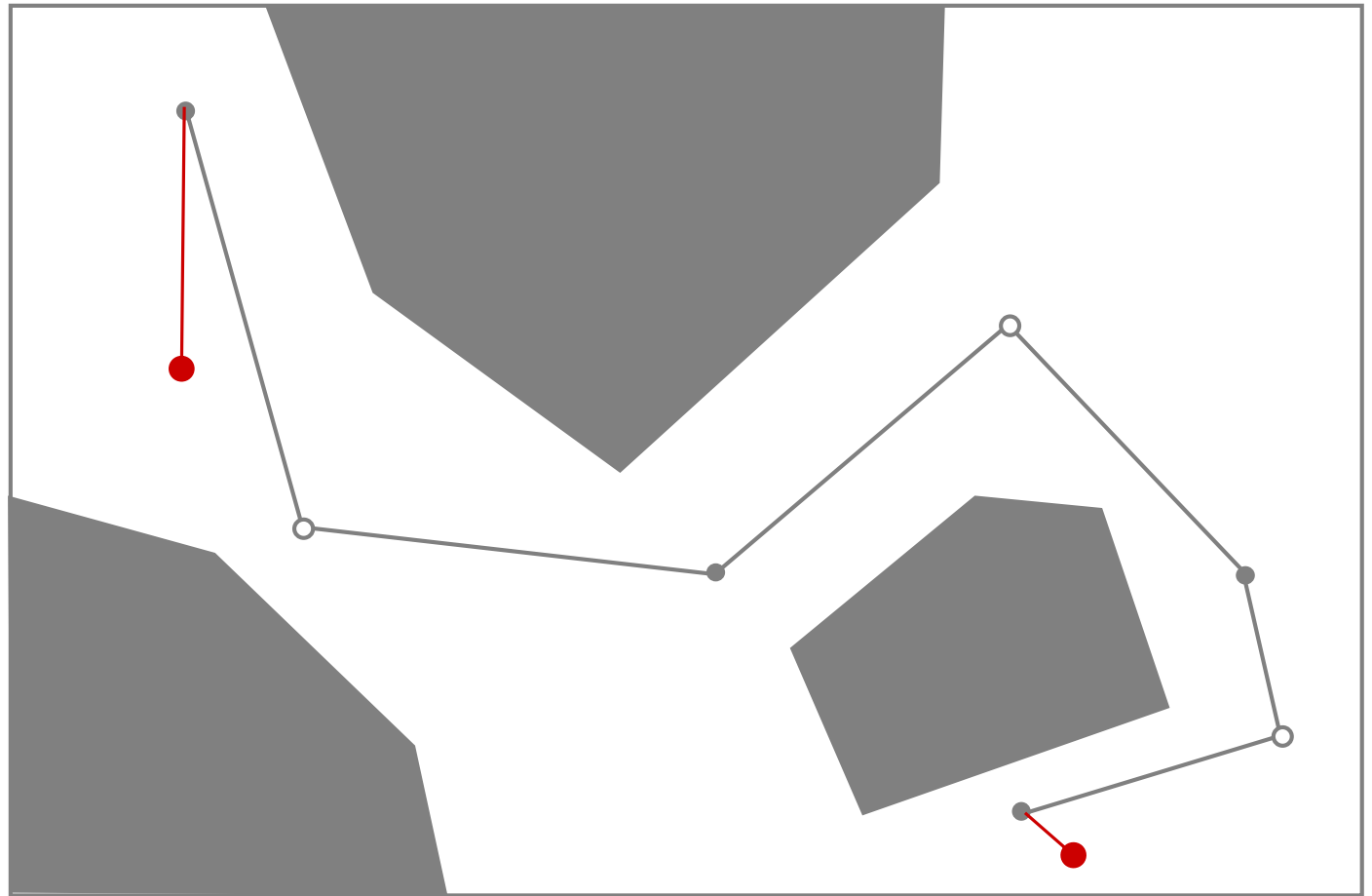


# Probabilistic Motion Planning

## • Visibility based sampling

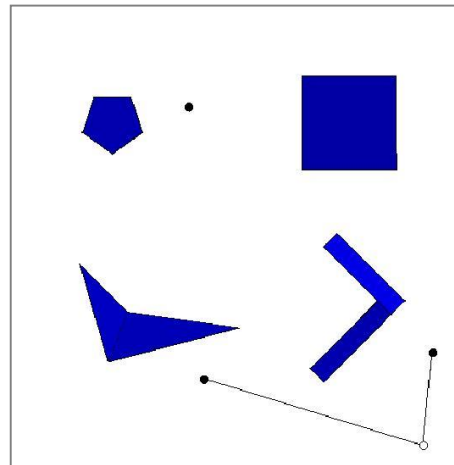


## • Visibility based sampling



# Probabilistic Motion Planning

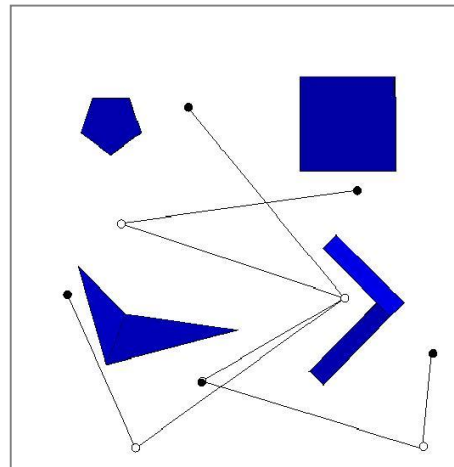
- Visibility based sampling
- Algorithm:
  - Generate guards and connectors randomly
  - Stop after *#try* failures (*#try*=10)





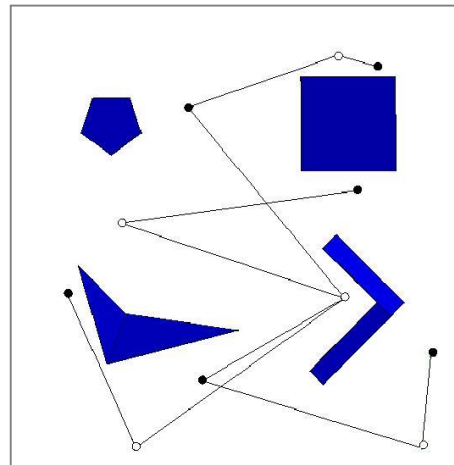
# Probabilistic Motion Planning

- Visibly based sampling
- Algorithm:
  - Generate guards and connectors randomly
  - Stop after *#try* failures (*#try*=100)



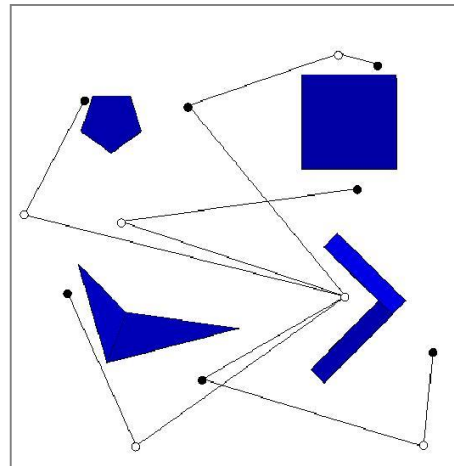
# Probabilistic Motion Planning

- Visibly based sampling
- Algorithm:
  - Generate guards and connectors randomly
  - Stop after *#try* failures (*#try*=200)



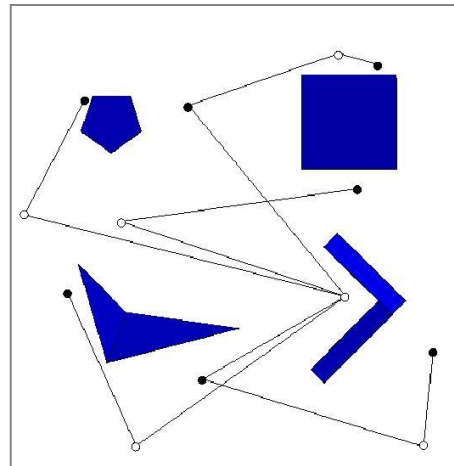
# Probabilistic Motion Planning

- Visibility based sampling
- Algorithm:
  - Generate guards and connectors randomly
  - Stop after  $\#try$  failures ( $\#try=500$ )



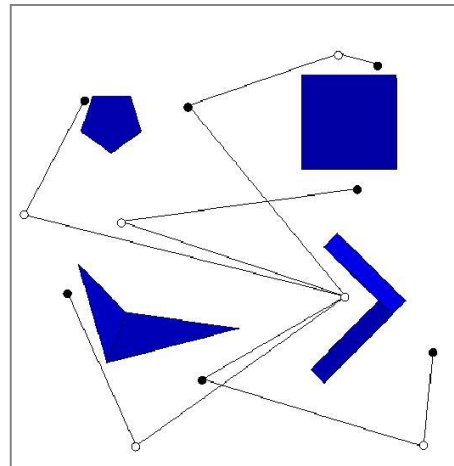
# Probabilistic Motion Planning

- Visibility based sampling
- Algorithm:
  - Generate guards and connectors randomly
  - Stop after *#try* failures (*#try*=1000)



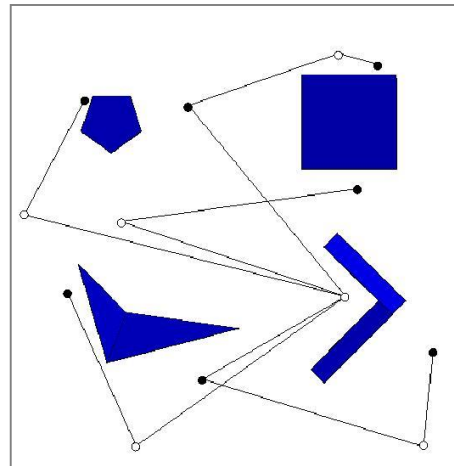
# Probabilistic Motion Planning

- Visibly based sampling
- Algorithm:
  - Generate guards and connectors randomly
  - Stop after *#try* failures (*#try*=10000)



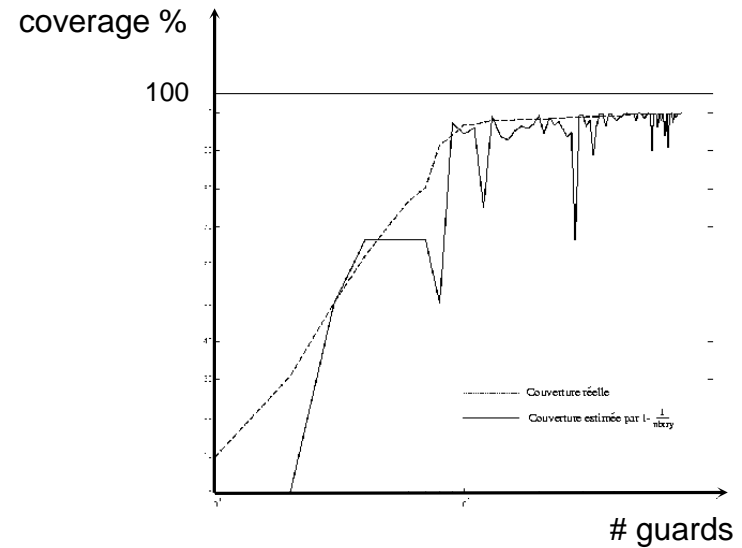
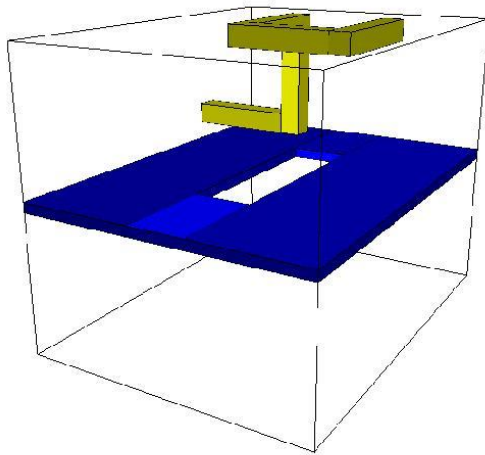
# Probabilistic Motion Planning

- Visibly based sampling
- Algorithm:
  - Generate guards and connectors randomly
  - Stop after *#try* failures (*#try*=1000000000!)



- Visibly based sampling
  - Algorithm:
    - Generate guards and connectors randomly
    - Stop after *#try* failure
  - Theorems:
    - The estimated percentage of non-covered free-space is  $\#try^{-1}$
    - Probability to find an existing path increases as an exponential function of time

- Visibly based sampling
- Possible online estimation of  $\#try$





- Visibility based sampling
- Real time demonstrations

# Probabilistic Motion Planning