

Motion planning ST5 Autonomous robotics

Francis Colas

2022-09-23

Introduction

Motion

- generated by motors
 - electric (AC/DC), pneumatic or hydraulic
 - stepper motor: fixed positions
 - servomotors: motor+integrated sensor for position control
- controlled to perform a given trajectory



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Motion planning

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- to reach a defined goal
- while following constraints (collision avoidance, dynamics...)



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Aim of the session

configuration space

2 - Francis Colas - Autonomous robotics - Motion planning - 2022-09-23





Trajectory

Path

- sequence of poses
- typically from a start to a goal location



Trajectory

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Trajectory

- mapping from time to configuration
- allows to compute velocities and commands



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Trajectory

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- allows to compute velocities and commands

Constraints

- minimum length/distance
- minimum cost (application-dependent)
- security/distance to obstacles
- kinematics or dynamics



Configuration space

- Workspace \mathcal{W}
- space in which robot evolves
 - \blacktriangleright in general: $\mathcal{W} = \mathbb{R}^3$
 - sometimes: $\mathcal{W} = \mathbb{R}^2$



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Configuration space ${\mathcal C}$

- \blacktriangleright set of feasible configurations $oldsymbol{q} \in \mathcal{C}$
- taking into account physical constraints
- \blacktriangleright in general $\mathcal{C}
 eq \mathcal{W}$



-



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- Free space ${\mathcal E}$
- configurations not in collision
- $\blacktriangleright \ \mathcal{E} = \mathcal{C} \setminus \mathcal{O}$
- adapted for planning









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Configuration space ${\mathcal C}$

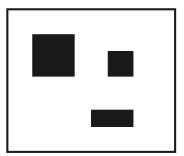
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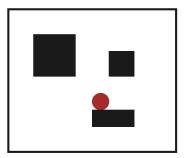


Workspace



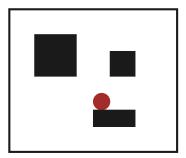


Workspace and robot

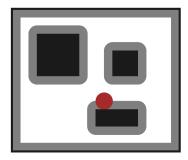




Workspace and robot

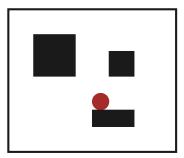


Obstacle inflation (Minkowski sum)

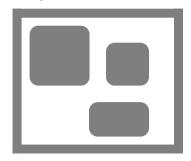




Workspace and robot



Free space





Robotic arm

Robotic arm

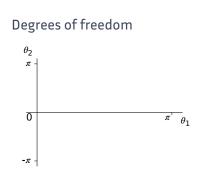




Robotic arm

2 degrees-of-freedom arm





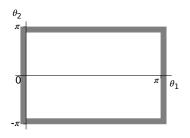


Robotic arm

2 degrees-of-freedom arm



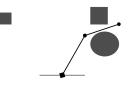
Configuration space



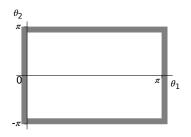


Robotic arm

Workspace



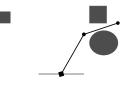
Configuration space



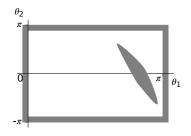


Robotic arm

Workspace



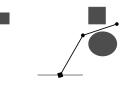
First obstacle



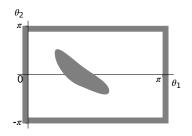


Robotic arm

Workspace



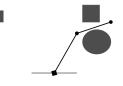
Second obstacle



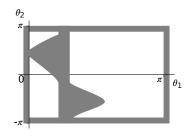


Robotic arm

Workspace



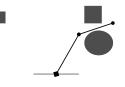
Third obstacle



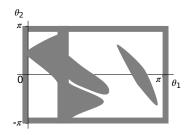


Robotic arm

Workspace



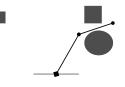
Free space



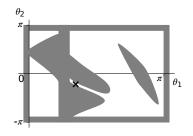


Robotic arm

Workspace



Free space





Conclusion on configuration space

Configuration space

- unification of mobile robots and robotic arms
- free space (no collision)
- adapted for planning

Building

- Minkowski sum for a mobile robot
- collision test for an arm

Limits

- only geometric constraints
- no kinematics nor dynamics
- potentially high-dimension space



02

Planning algorithms

Algorithms

Approaches

- space decomposition
 - 🕨 grid
 - cell decomposition
- sampling
- potential fields
- geometric resolution
- path refinement



Path planning with a grid

Path planning with a grid

- adapted to occupancy grids
- neighborhood graph
- graph search (Dijkstra, A*)

Result

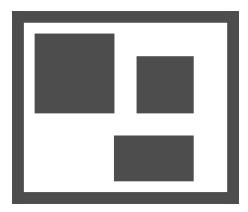
- 🕨 path
- discretized orientation
- not necessarily optimal in distance
- high complexity in medium to high dimension



Planning algorithms

Example

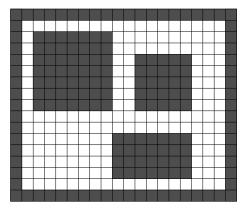
Free space





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Grid decomposition

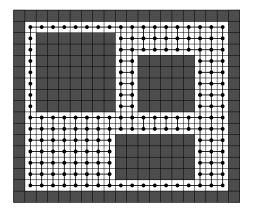




Planning algorithms

Example

Cell graph

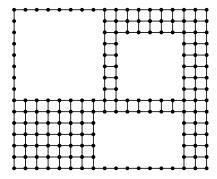




Planning algorithms

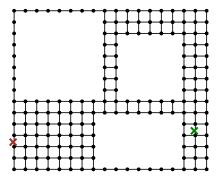
Example

Cell graph



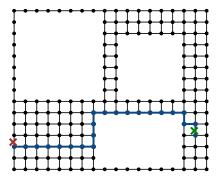


Start and goal locations



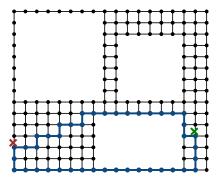


Graph search (A* or other)





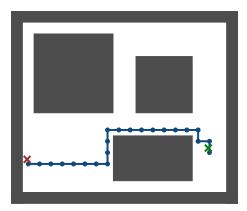
Graph search (A* or other)



Shortest path not unique!



Example





Planning with cell decomposition

Cell decomposition

various tessellation methods

Voronoi diagram

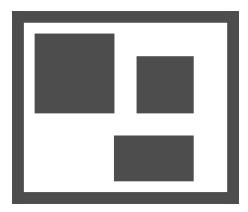
- tiling based on distance to obstacles
- dual of Delaunay triangulation
- path following of the edges of the cells

- 🕨 path
- as far from the obstacles as possible
- difficult to build in high dimensions
- not distance optimal



Example

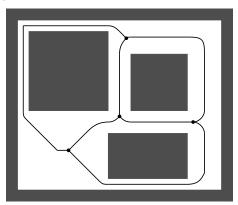
Free space





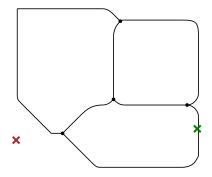
14 - Francis Colas - Autonomous robotics - Motion planning - 2022-09-23

Voronoi diagram





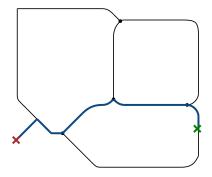
Voronoi diagram





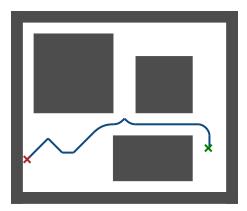
Example

Graph search





Example





Sampling-based planning

Rapidly-expanding Random Trees

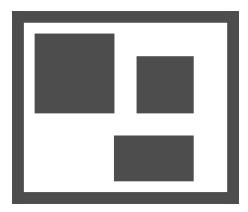
- ▶ RRT, RRT*...
- stochastic algorithm by sampling the space
- connection tree building
- until goal is found
- refinement of the path

- 🕨 path
- distance optimal for infinite samples
- quick and anytime as soon as the path is found



Example

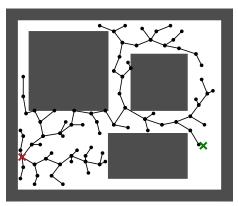
Free space





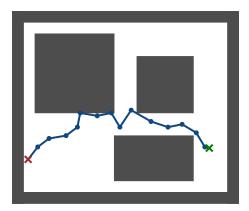
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Graph expansion





Example





Algorithm

RRT

$$\begin{split} V &\leftarrow \{x_{\text{init}}\}; E \leftarrow \emptyset \\ \text{for } i = 1, \ldots, n \text{ do} \\ x_{\text{rand}} \leftarrow \text{SampleFree}() \\ x_{\text{nearest}} \leftarrow \text{Nearest} \left(G = \left(V, E\right), x_{\text{rand}}\right) \\ x_{\text{new}} \leftarrow \text{Steer} \left(x_{\text{nearest}}, x_{\text{rand}}\right) \\ \text{if CollFree} \left(x_{\text{nearest}}, x_{\text{new}}\right) \\ V \leftarrow V \cup \{x_{\text{new}}\} \\ E \leftarrow E \cup \left\{\left(x_{\text{nearest}}, x_{\text{new}}\right)\right\} \\ \text{end if} \\ \text{end for} \\ \text{return } G = \left(V, E\right) \end{split}$$

Functions

- SampleFree(): sample point in free space
- Nearest(G, x): point in G nearest from x
- Steer(x₁, x₂): point toward x₂ at a given distance from x₁
- CollFree(x₁, x₂): no obstacle between x₁ and x₂



Algorithm

Improvement of RRT

$$\begin{split} V \leftarrow \{x_{\text{init}}\}; E \leftarrow \emptyset \\ \text{for } i = 1, \dots, n \text{ do} \\ x_{\text{rand}} \leftarrow \text{SampleFree}() \\ x_{\text{nearest}} \leftarrow \text{Nearest} \left(G = \left(V, E\right), x_{\text{rand}}\right) \\ x_{\text{new}} \leftarrow \text{Steer} \left(x_{\text{nearest}}, x_{\text{rand}}\right) \\ \text{if CollFree} \left(x_{\text{nearest}}, x_{\text{new}}\right) \text{ then} \\ x_{\text{near}} \leftarrow \text{Near} \left(G = \left(V, E\right), x_{\text{new}}, \delta\right) \\ x_{\text{min}} \leftarrow \arg\min \left(C \right) + c \left(x, x_{\text{new}}\right) \\ v \leftarrow V \cup \{x_{\text{new}}\} \\ E \leftarrow E \cup \left\{\left(x_{\text{min}}, x_{\text{new}}\right)\right\} \\ \text{end if} \\ \text{end for} \\ \text{return } G = \left(V, E\right) \end{split}$$

Functions

- SampleFree(): sample point in free space
- Nearest(G, x): point in G nearest from x
- Steer(x₁, x₂): point toward x₂ at a given distance from x₁
- CollFree(x₁, x₂): no obstacle between x₁ and x₂
- Near(G, x, d): points in G at a distance from x less than d
- C(x): cost between x_{init} and x, walking up the graph
- ► $c(x_1, x_2)$: cost between x_1 and x_2



Algorithm RRT*

$$V \leftarrow \{x_{init}\}; E \leftarrow \emptyset$$

for $i = 1, \dots, n$ do
 $x_{rand} \leftarrow SampleFree()$
 $x_{nearest} \leftarrow Nearest (G = (V, E), x_{rand})$
 $x_{new} \leftarrow Steer (x_{nearest}, x_{rand})$
if CollFree $(x_{nearest}, x_{new})$ then
 $X_{near} \leftarrow Near (G = (V, E), x_{new}, \delta)$
 $x_{min} \leftarrow \arg \min C (x) + c (x, x_{new})$
 $x \in x_{near}$
 $V \leftarrow V \cup \{x_{new}\}$
 $E \leftarrow E \cup \{(x_{min}, x_{new})\}$
for all $x \in X_{near}$ do
if $C(x_{new}) + c (x_{new}, x) < C(x)$ then
 $E \leftarrow E \cup \{(P(x), x)\}$
 $end if$
end for
end if
end for
return $G = (V, E)$
 $V - Francic Cold - Autonomous (V, E)$

Functions

- SampleFree(): sample point in free space
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- Steer(x₁, x₂): point toward x₂ at a given distance from x₁
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- C(x): cost between x_{init} and x, walking up the graph
- $c(x_1, x_2)$: cost between x_1 and x_2
- P(x): parent of x.



Planning with potential fields

Potential fields

- repulsive field around obstacles
- attractive field around goal
- combination of both
- gradient descent

Result

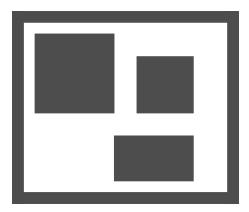
path

- away from obstacles
- quick to compute
- 🕨 local minima



Example

Free space

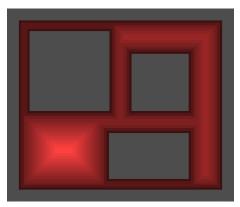




19 - Francis Colas - Autonomous robotics - Motion planning - 2022-09-23

Example

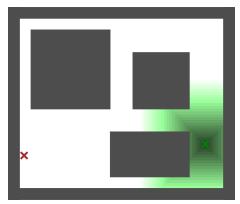
Repulsive field





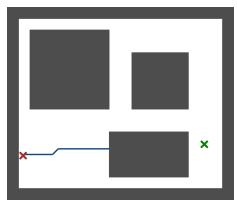
Example

Attractive field





Gradient descent



Local minimum!



Planning with visibility graph

Visibility graph

- nodes: vertices of obstacles
- edge: iff visibility between nodes
- start and goal as nodes
- graph search

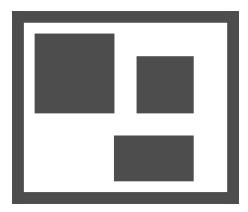


- 🕨 distance optimal
- can follow obstacle edges
- need polygonal/polyhedral obstacles



Example

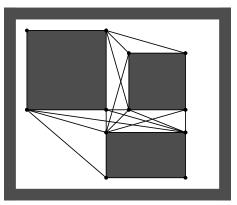
Free space





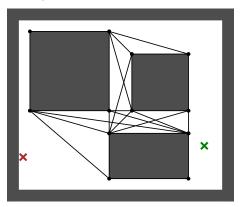
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Visibility graph of obstacles



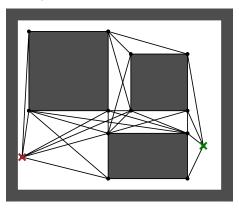


Include start and goal





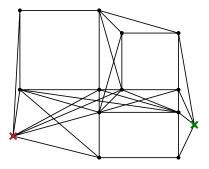
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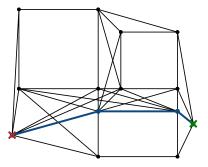
Example

Complete graph



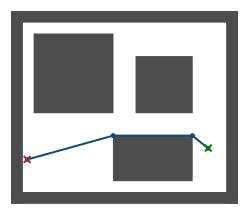


Graph search (A* or other)





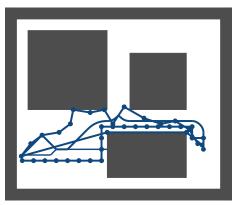
Example





Comparison

Comparison of the paths





03

Conclusion

Conclusion

Configuration space

- space free from collision
- adapted to planning
- similar for mobile and articulated robots

Planning algorithms

- different families with different internal representations
- various optimization criteria

Limits

- 🕨 known map
- static obstacles



Bibliography

RRT*, PRM*, etc.

 Karaman and Frazzoli, Sampling-based algorithms for optimal motion planning, IJRR, 2011.

Books

- Latombe, Robot Motion Planning, Kluwer Academic Publishers, 1991.
- Lavalle, Planning Algorithms, Cambridge University Press, 2006.
- Siciliano et al., Springer Handbook of Robotics, 2nd ed., Springer, 2016.





Thanks for your attention Questions?