

Computational Geometry

Motion Planning

Lecture #10

Thomas van Dijk

Winter Semester 2018/19





current situation, desired situation





current situation, desired situation





current situation, desired situation



sequence of steps to reach the one from the other

Path Planning



current location, desired location

Path Planning



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Path Planning



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path to reach the one from the other





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preprocessing

guerying

A First Result

Theorem: We can preprocess a set of polygonal obstacles with a total of *n* edges in $O(n \log n)$ expected time such that, given a start and a goal position, we can find a collision-free path for a point robot in O(n) time if it exists.

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What about, say, *polygonal* robots?

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2D translating, rotating robot



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3D translating robot

3D translating, rotating robot

Configuration Space



robotic arm

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robotic arm

The *configuration space* is the *d*-dimensional space of all possible (i.e., obstacle avoiding) parameter value combinations.


robotic arm

work space













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Path for a *point* through configuration space



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Path for a *point* through configuration space

path for the *robot* in the original space.



work space

configuration space

8 - 1



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8 - 2



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8 - 4



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Vector sums



Vector sums



Vector sums Algebra: $(p_x, p_y) + (q_x, q_y) = (p_x + q_x, p_y + q_y)$



Vector sums

Algebra: $(p_x, p_y) + (q_x, q_y) = (p_x + q_x, p_y + q_y)$ Geometry: place vectors head to tail



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Minkowski sums





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$$p \xrightarrow{q} p + q$$

Minkowski sums

Algebra: $S_1 \oplus S_2 = \{ p + q \mid p \in S_1, q \in S_2 \}$





Vector sums $(p_x, p_y) + (q_x, q_y) = (p_x + q_x, p_y + q_y)$ Algebra: Geometry: place vectors head to tail Minkowski sums Algebra: $S_1 \oplus S_2 = \{ p + q \mid p \in S_1, q \in S_2 \}$ Geometry: place copy of one shape at every point of the other

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Recall that $CP = \{(x, y) : \mathcal{R}(x, y) \cap P \neq \emptyset\}$ for an obstacle P.



Recall that $C\mathcal{P} = \{(x, y) : \mathcal{R}(x, y) \cap \mathcal{P} \neq \emptyset\}$ for an obstacle \mathcal{P} . In other words: $\mathcal{R}(x, y)$ intersects $\mathcal{P} \quad \Leftrightarrow \quad (x, y) \in C\mathcal{P}$.



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Pseudodisks

Definition:

A pair of planar objects o_1 and o_2 is a pair of pseudodisks if:

- $\partial o_1 \cap \operatorname{int}(o_2)$ is connected, and
- $\partial o_2 \cap \operatorname{int}(o_1)$ is connected.

Pseudodisks

Definition:

A pair of planar objects o_1 and o_2 is a pair of pseudodisks if:

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- $\partial o_1 \cap \operatorname{int}(o_2)$ is connected, and
- $\partial o_2 \cap \operatorname{int}(o_1)$ is connected.

02

Pseudodisks $o_1 \\ o_2$ o_2 o_1 o_2 o_2 o_1 o_2 o_2 o_2 o_2 o_3 o_4 o_2 o_2 o_3 o_4 o_2 o_2 o_3 o_4 o_2 o_3 o_4 o_2 o_3 o_4 o_2 o_3 o_4 o_4 o_2 o_3 o_4 o_4 o_4 o_2 o_3 o_4 o_4

 $p \in \partial o_1 \cap \partial o_2$ is a *boundary crossing* if ∂o_1 crosses at p from the interior to the exterior of o_2 .

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• $\partial o_2 \cap \operatorname{int}(o_1)$ is connected.

 $p \in \partial o_1 \cap \partial o_2$ is a *boundary crossing* if ∂o_1 crosses at p from the interior to the exterior of o_2 .

Observation: A pair of polygonal pseudodisks defines at most two boundary crossings.

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Theorem: If \mathcal{P}_1 and \mathcal{P}_2 are convex polygons with disjoint interiors, and \mathcal{R} is another convex polygon, then $\mathcal{P}_1 \oplus \mathcal{R}$ and $\mathcal{P}_2 \oplus \mathcal{R}$ is a pair of pseudodisks.

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