

ANA : Anytime Nonparametric A**

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Outline

- Motivation
- Previous Work
- The Algorithm
- Improvements of ANA* Over ARA*
- Experimental Results
- Conclusion and Future Work

Previous Work: Foundations

- Dijkstra's Algorithm
 - Shortest path from s_{start} to s_{goal} with non-negative edges
 - Maintains $g(s)$, minimum cost so far
- A^*
 - Adds heuristic to Dijkstra's
 - Admissible $h(s)$ guarantees optimality, consistent $h(s)$ guarantees cycle-free search
- Weighted A^*
 - $f(s) = g(s) + \epsilon * h(s)$
 - $\epsilon \geq 1$, bounds optimality
 - Raising ϵ trades optimality for speed

*Previous Work: Anytime A**

- Anytime Heuristic Search (AHS)
 - Continues search after solution found
 - Intermediate upper bound: G
 - Intermediate lower bound: $\min_{s \in \text{OPEN}} \{g(s)+h(s)\}$
- Anytime Repairing A* (ARA*)
 - Decreases ϵ between results, updating $f(s)$ values
 - Introduces another parameter
- Restarting Weighted A* (RWA*)
 - Restarts search when ϵ is decreased
 - Reuses best known $g(s)$ values for states

The ANA* Algorithm

- ANA*()
 $G \leftarrow \infty$; $E \leftarrow \infty$; $OPEN \leftarrow \emptyset$;
 $\forall s: g(s) \leftarrow \infty$; $g(s_{start}) \leftarrow 0$
Insert s_{start} into OPEN with
key $e(s_{start})$
while $OPEN \neq \emptyset$ do
 IMPROVESOLUTION()
 Report current E-suboptimal
 solution
 Update keys $e(s)$ in OPEN
 and prune if $g(s)+h(s) \geq G$
- IMPROVESOLUTION()
while $OPEN \neq \emptyset$ do
 $s \leftarrow \operatorname{argmax}_{s \in OPEN} \{e(s)\}$
 $OPEN \leftarrow OPEN \setminus \{s\}$
 if $e(s) < E$ then
 $E \leftarrow e(s)$
 if ISGOAL(s) then
 $G \leftarrow g(s)$
 return
 for each successor s' of s do
 if $g(s)+c(s,s') < g(s')$ then
 $g(s') \leftarrow g(s)+c(s,s')$
 $\operatorname{pred}(s') \leftarrow s$
 if $g(s')+h(s') < G$ then
 Insert or update s' in
 OPEN with key $e(s')$

The ANA Algorithm cont.*

- $e(s)$ is the maximal ε for which $f(s) \leq G$
- $e(s)$ bounds suboptimality
- G improves after each iteration

$$e(s) = \frac{G - g(s)}{h(s)}$$

*ARA** vs *ANA**

- Requires parameters ε and $\Delta\varepsilon$
- Starting ε must be finite
- Progress towards optimal solution is invariable
- If adapted to function like *ANA**, $f(s)$ keys would have to be updated for each change in ε
- Requires no parameters
- Starting G is infinite
- Progress towards optimal solution is the least possible improvement at each step
- $e(s)$ keys only need to be updated when G is reduced

Experiments: Problems

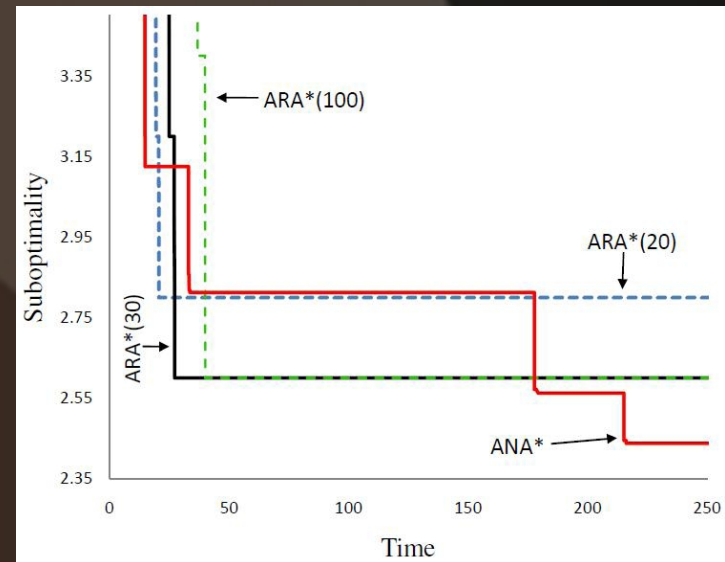
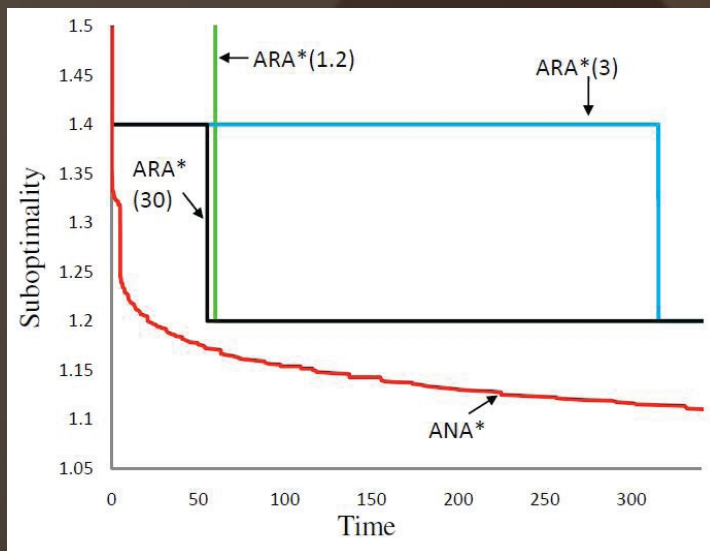
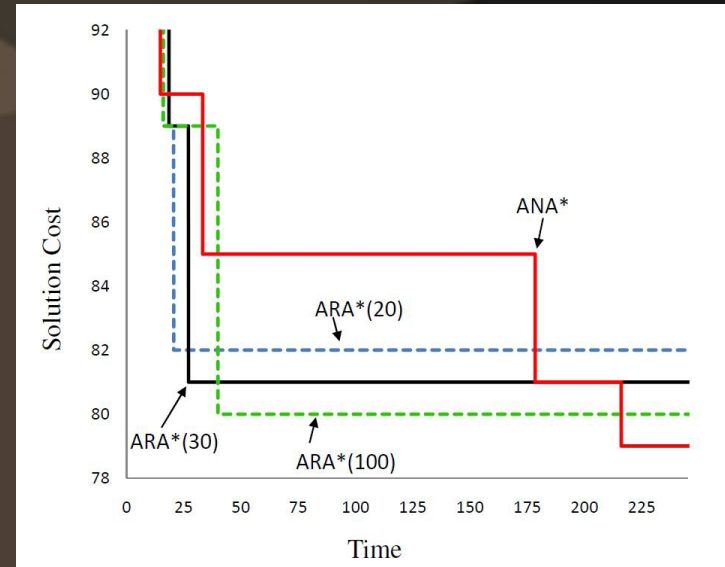
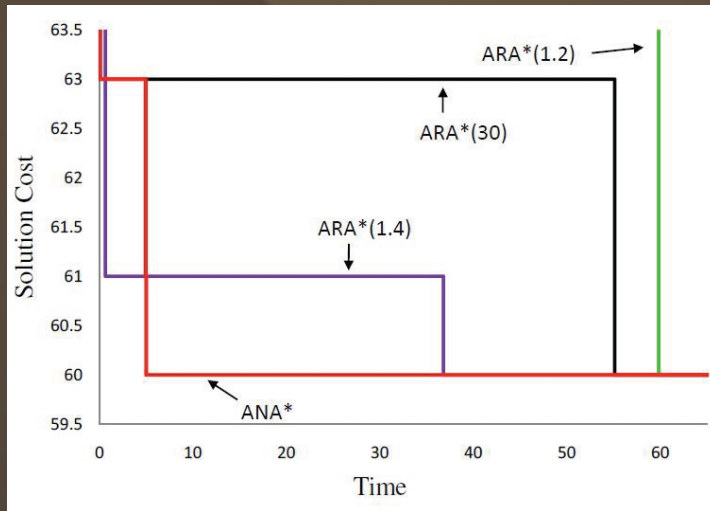
- Robot Arm: position arm to reach goal, avoiding obstacles
 - 6 or 20 degrees of freedom
 - action is a change in a joint's angle
 - $>3 \cdot 10^6$ states for 6 DOF, $>10^{26}$ states for 20
- Gridworld: navigate from start to goal in an $n \times m$ grid
 - Grid 1: 100x1200 8-connected, obstacles, uniform move cost between cells sharing a side
 - Grid 2: 5000x5000 4-connected, no obstacles, move cost randomly chosen from [1,1000]
 - Grid 3: 5000x5000 4-connected, obstacles, move cost randomly chose from [1,1000]

Experiments: Problems cont.

- Multiple Sequence Alignment: find lowest cost alignments of n proteins
 - $n=5$
 - gaps in a sequence cost 2
 - mismatched pairs cost 1

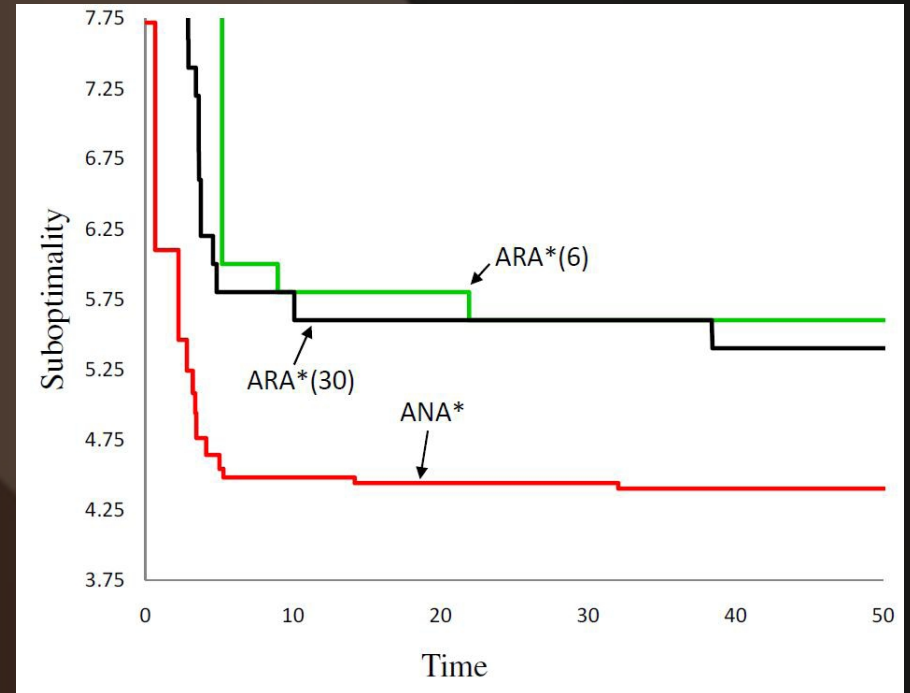
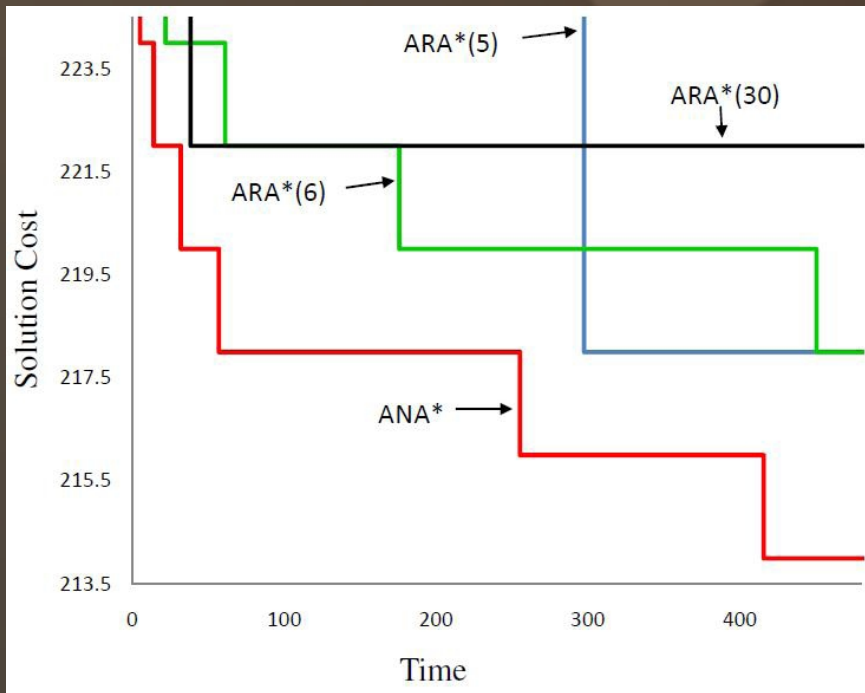
Experiments: Results-Robotic Arm

6 DOF vs 20 DOF



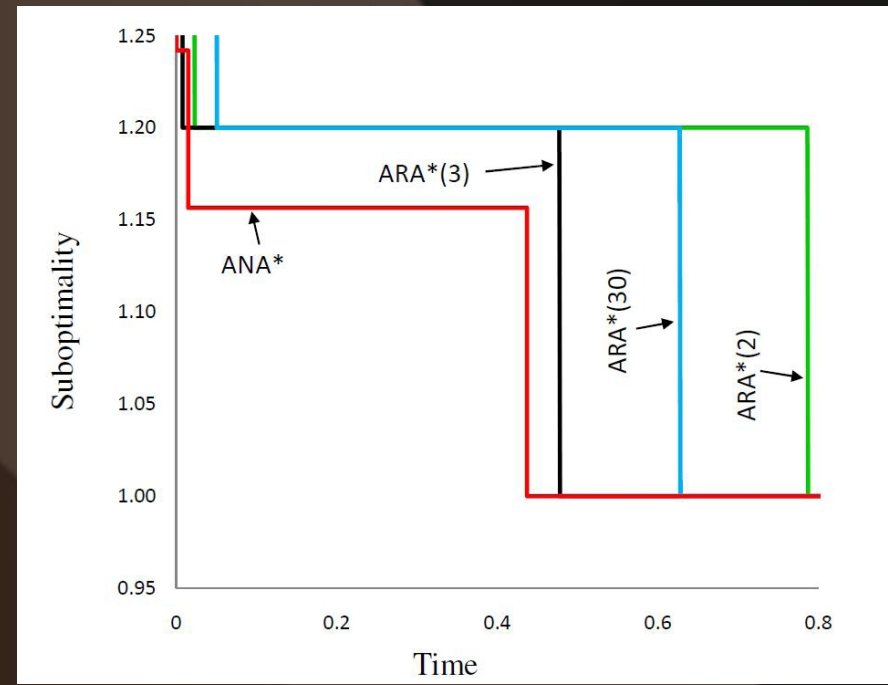
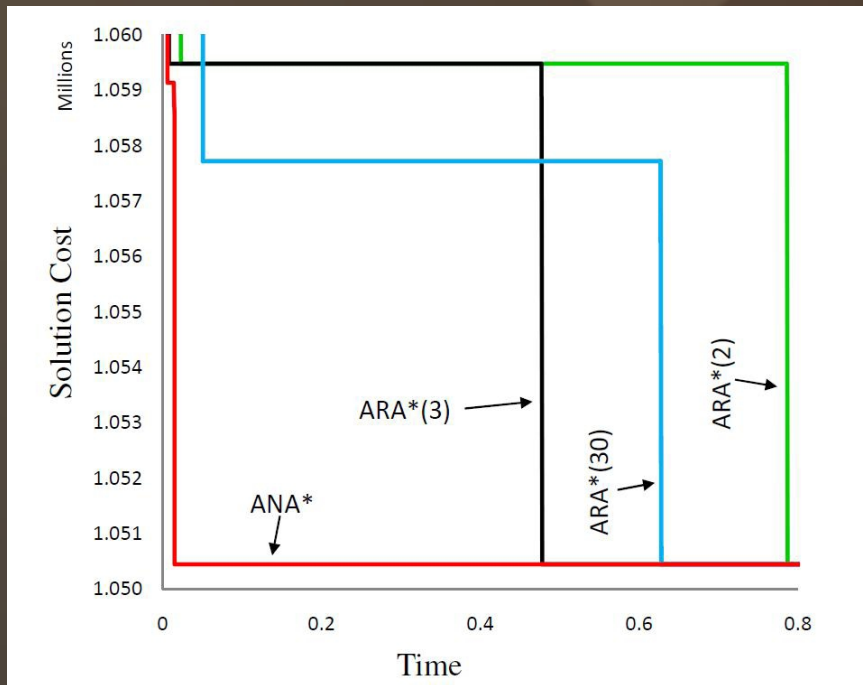
Experiments: Results-Robotic Arm

6 DOF, non-uniform cost



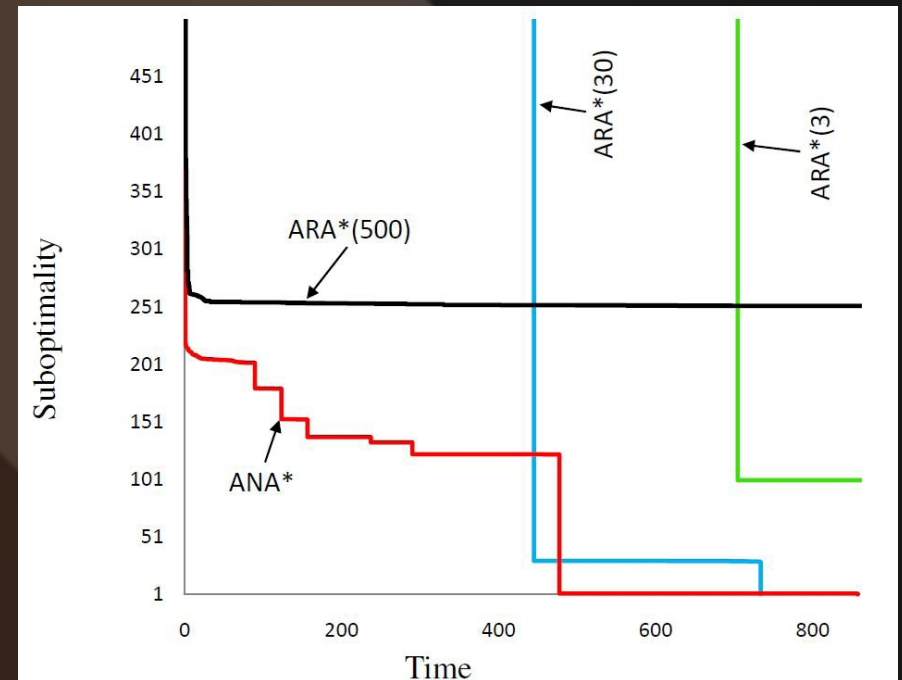
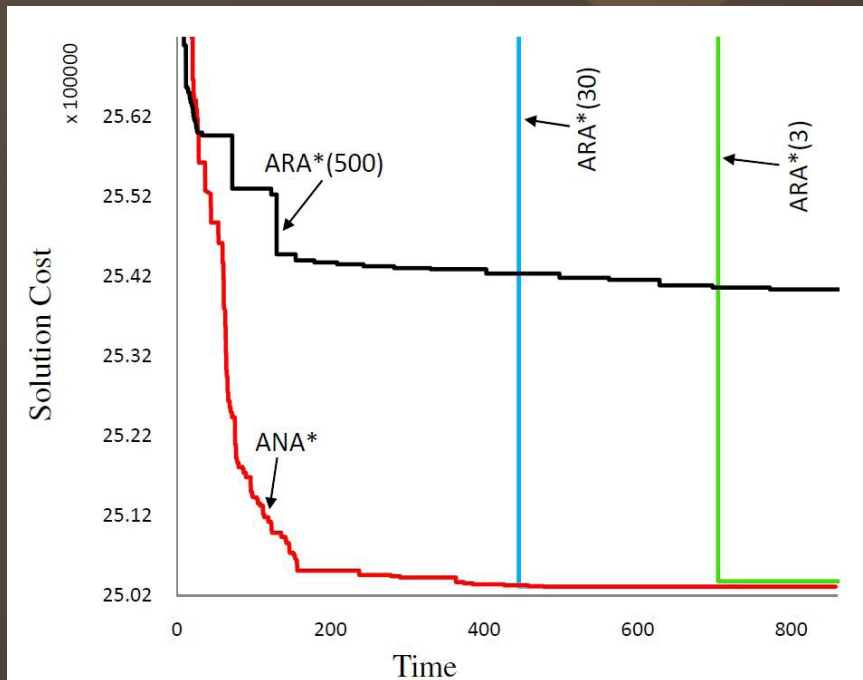
Experiments: Results-Gridworld

100x1200 with obstacles, uniform cost

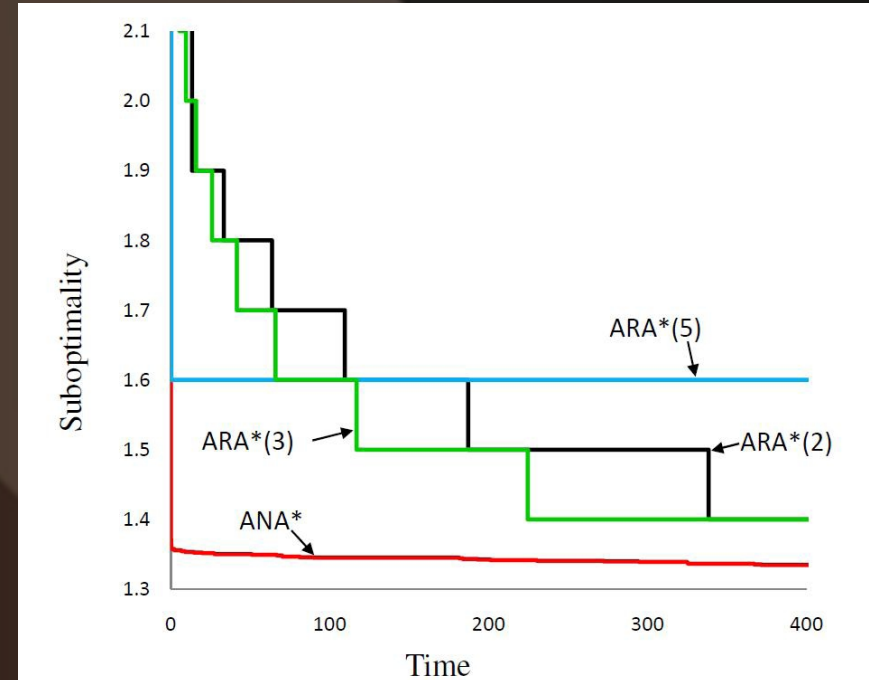
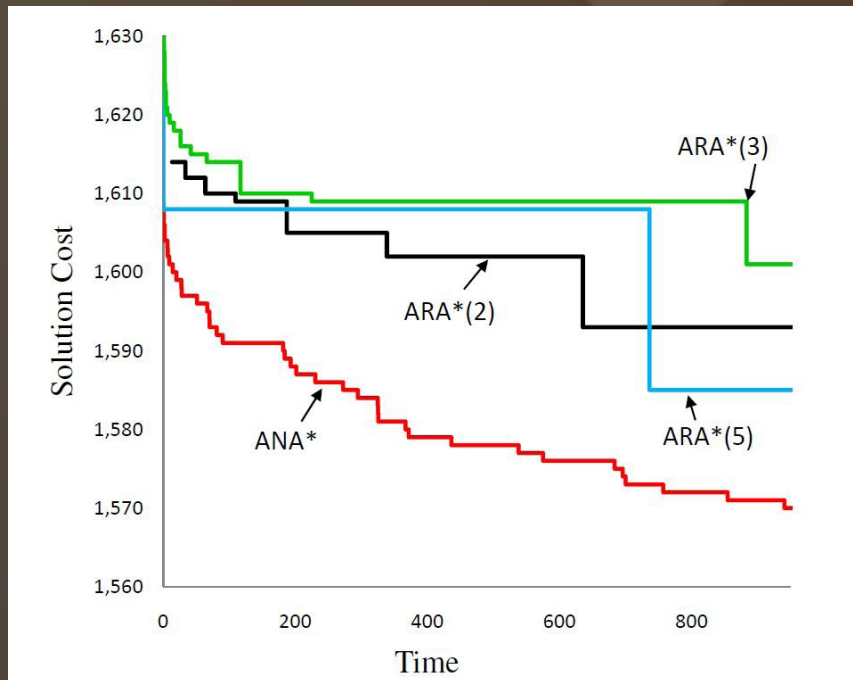


Experiments: Results-Gridworld

5000x5000 without obstacles, random cost



Experiments: Results-MSA



Conclusion and Future Work

- ANA* expands upon ARA*
- ANA* outperforms existing anytime A* algorithms both analytically and experimentally
- Future research in dynamic weight graph search

Citations

- Jur van den Berg, Rajat Shah, Arthur Huang, and Ken Goldberg, “ANA*: Anytime Nonparametric A*,” Association for the Advancement of Artificial Intelligence: Annual Conference (AAAI). San Francisco, CA. August 2011.
- Jur van den Berg, Rajat Shah, Arthur Huang, and Ken Goldberg, “ANA* *Technical Report*,” February 2011.
- Maxim Likhachev, Geoff Gordon and Sebastian Thrun, "ARA*: Anytime A* with Provable Bounds on Sub-Optimality," Advances in Neural Information Processing Systems 16 (NIPS), MIT Press, Cambridge, MA, 2004.



Questions