

Introduction

Asymptotically optimal motion planning algorithms improve the quality of the solution over time.

Overview:

- Standard sampling-based motion planning algorithms tend to produce paths that have lots of extra motion.
- Asymptotically optimal (AO) motion planning algorithms overcome this problem by **rewiring** the graph as more vertices are added.
- The most widely used AO algorithm is called **RRT***.

Aside: Path smoothing

Path smoothing can help, but its limited by the **homotopy class** of the initial solution.

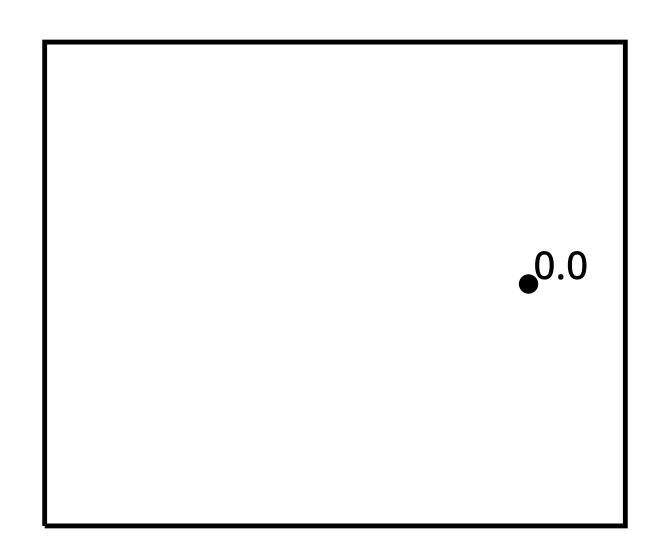
RRT*: Main idea

RRT* differs from RRT in two ways:

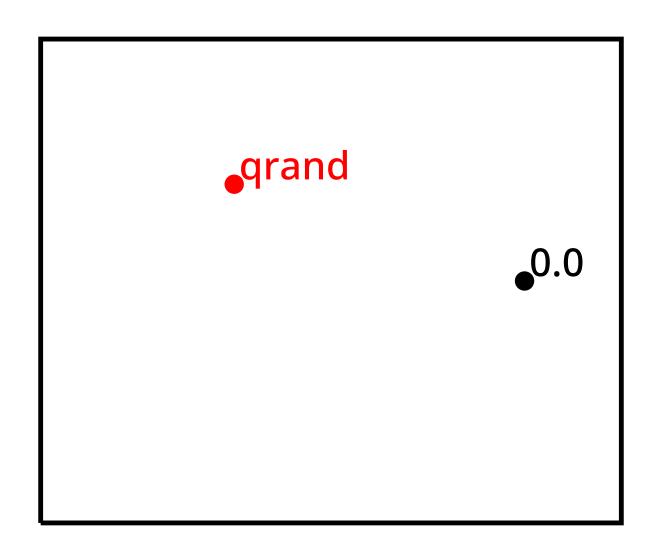
Change 1: Each new node connects to the **best** nearby parent, determined by the new node's distance to the root. Not necessarily the node that was nearest neighbor.

Change 2: For each new node, check nearby nodes to see if any would be have smaller distances with the new node as their parent. If so, **rewire** the tree.

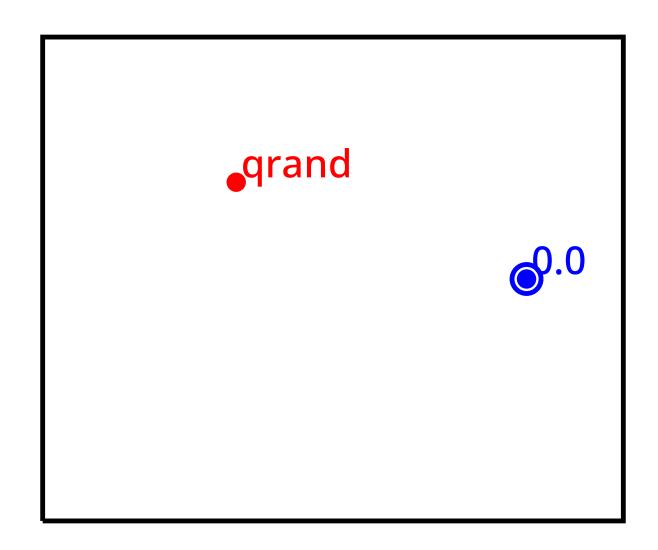
Start from a tree with just the start configuration.



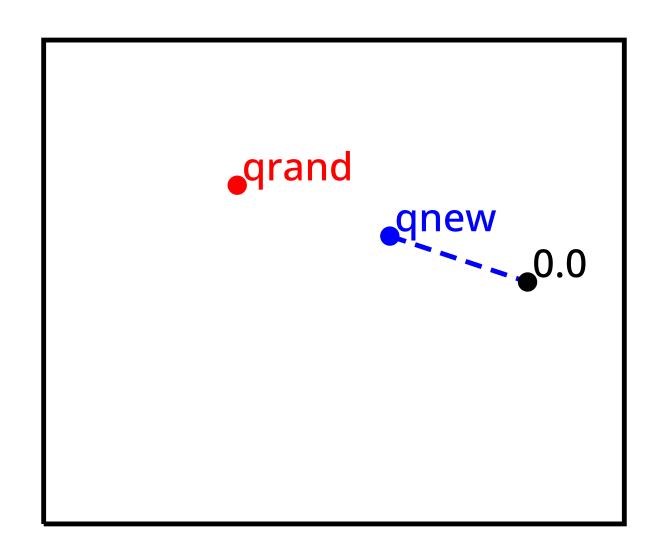
Choose a random sample.



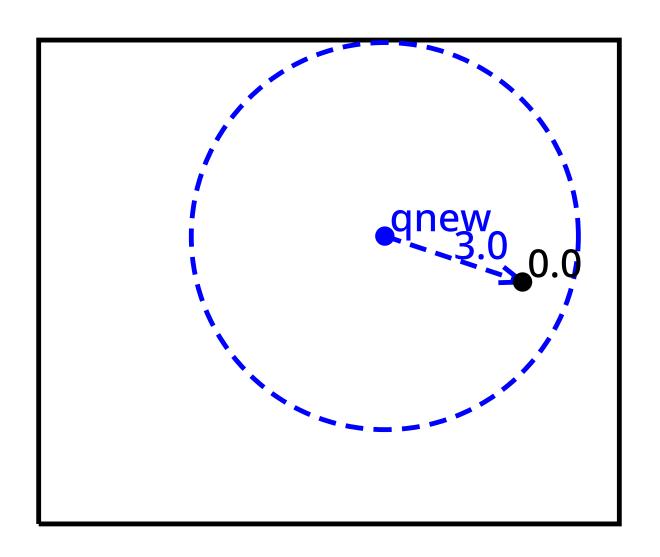
Find the nearest neighbor.



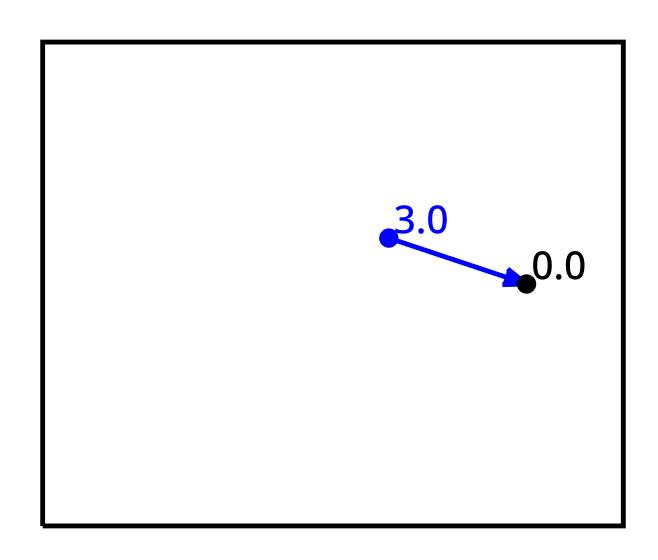
Extend from the nearest neighbor toward the sample.



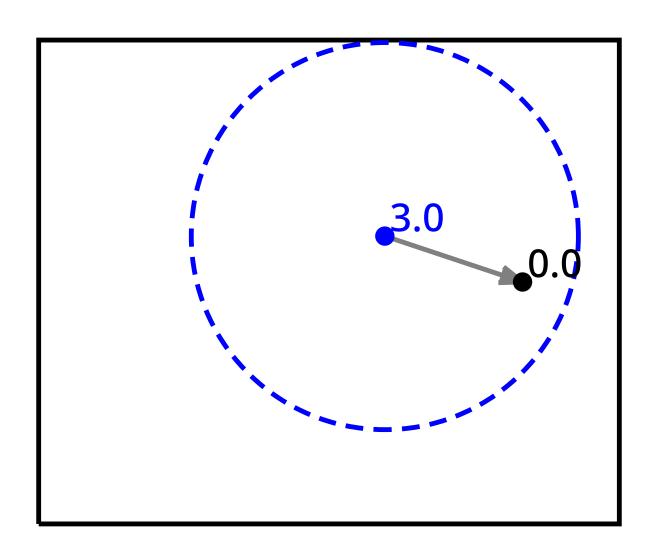
Find candidate parents for the new node; 1 candidate this time.



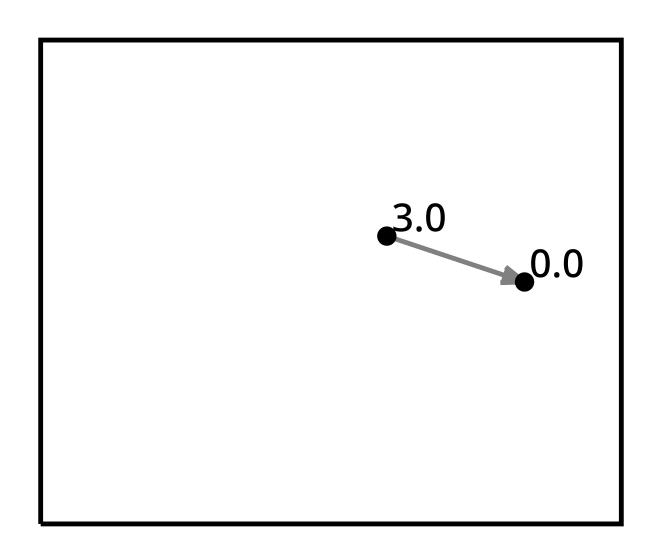
Choose the best parent for the new node.



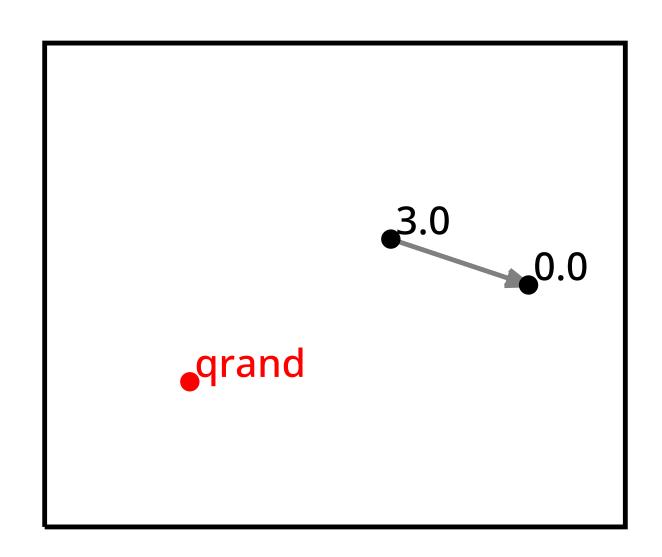
Find rewiring candidates near the new node; 0 candidates this time.



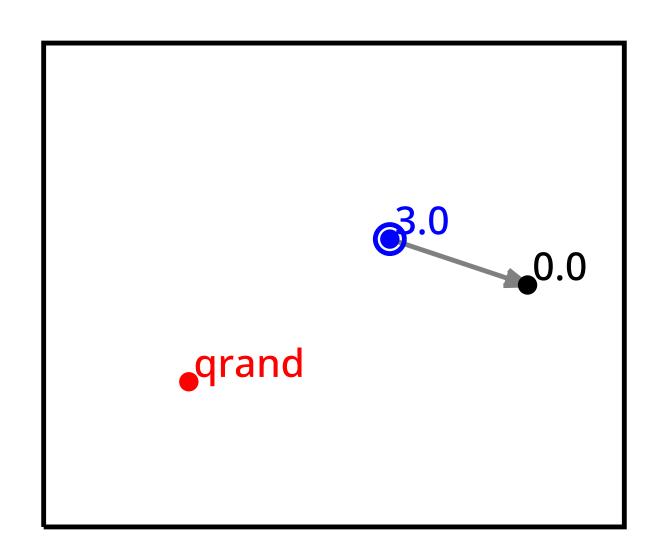
Rewire if needed; 0 changes this time.



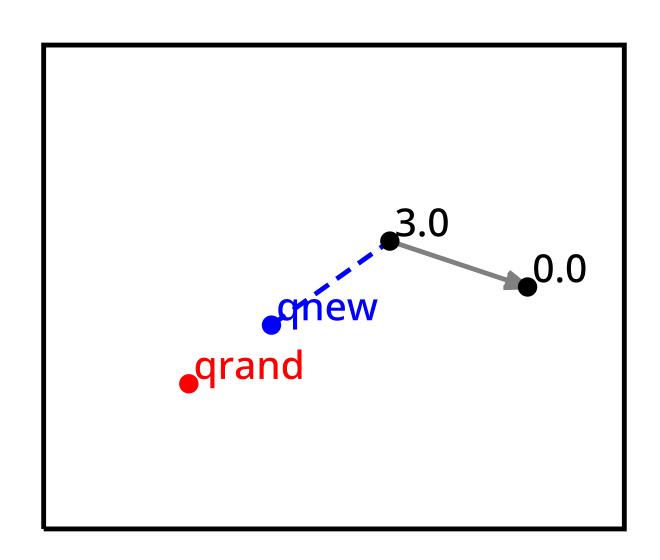
Choose a random sample.



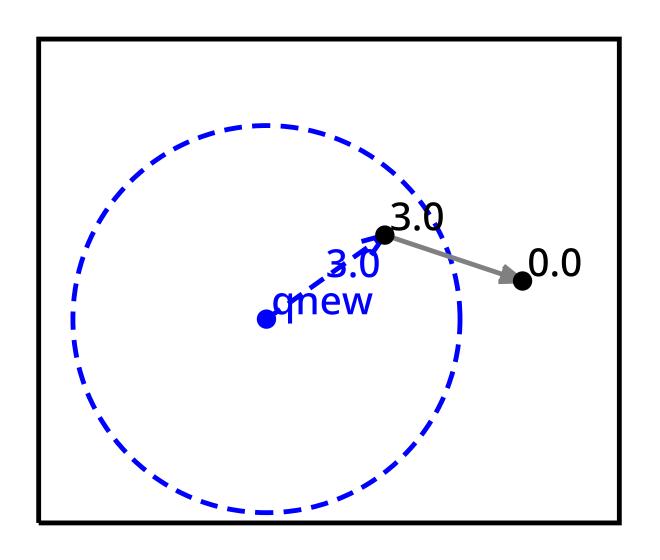
Find the nearest neighbor.



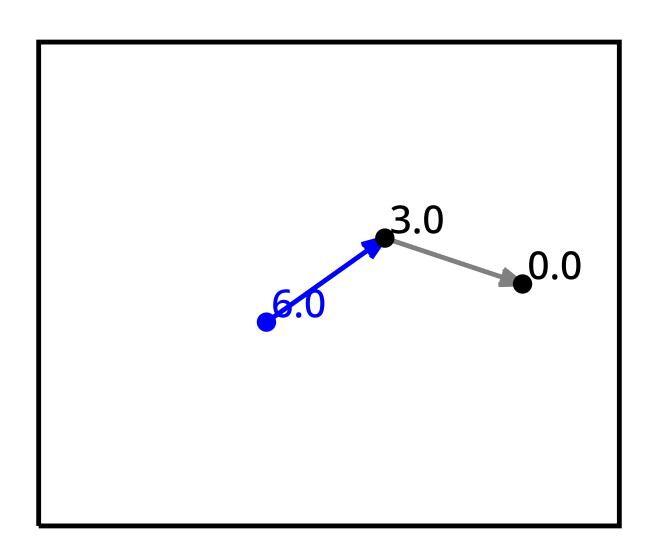
Extend from the nearest neighbor toward the sample.



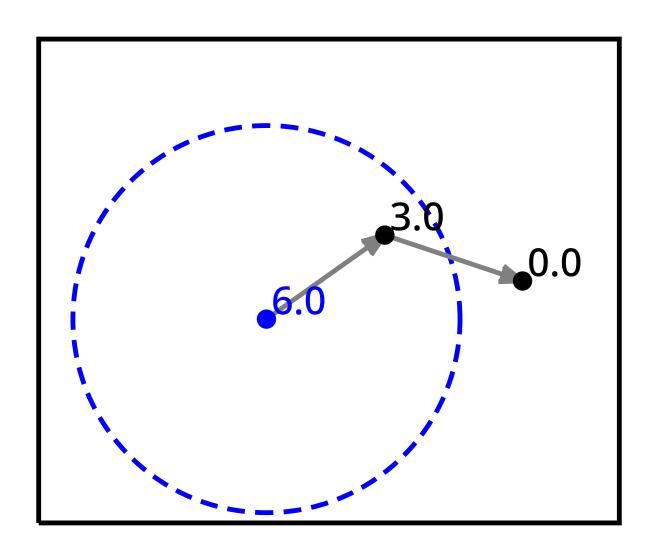
Find candidate parents for the new node; 1 candidate this time.



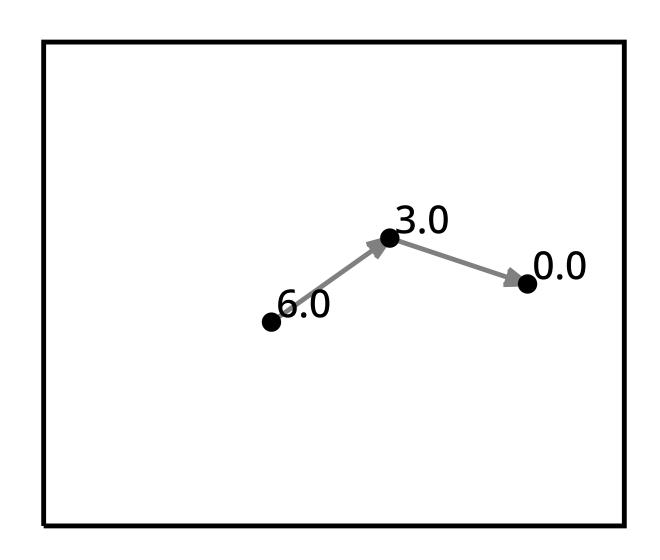
Choose the best parent for the new node.



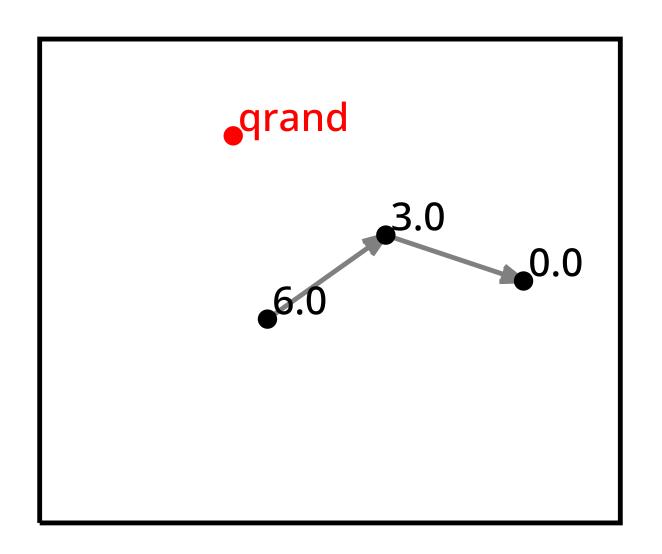
Find rewiring candidates near the new node; 0 candidates this time.



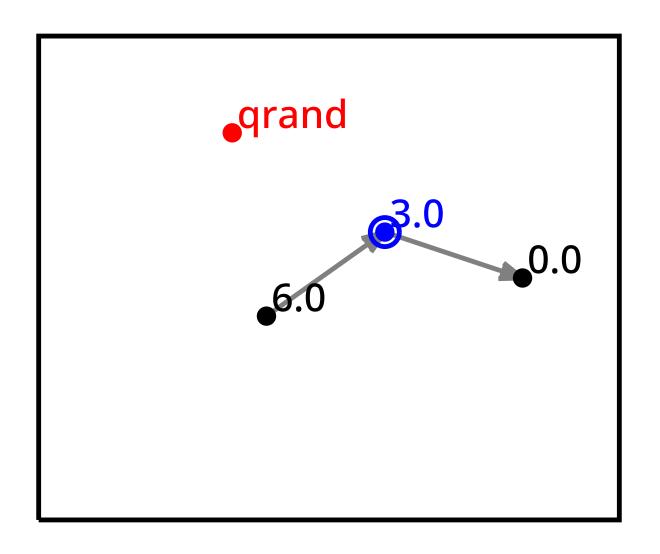
Rewire if needed; 0 changes this time.



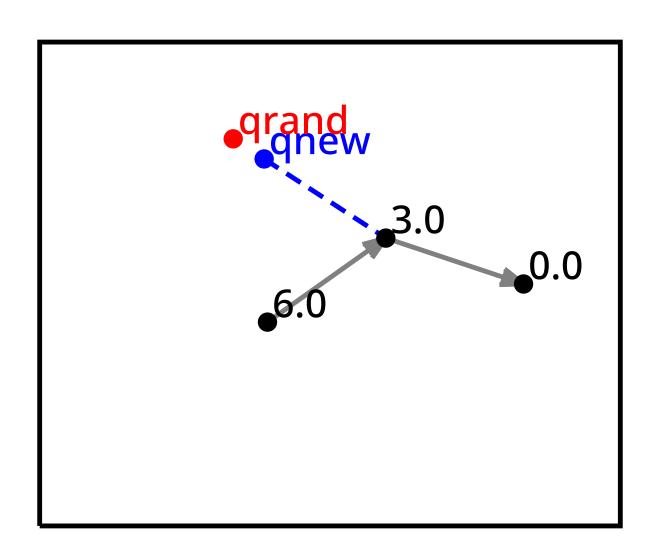
Choose a random sample.



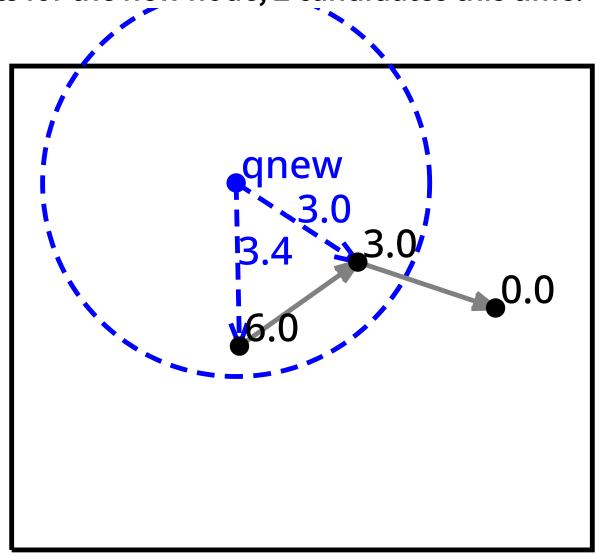
Find the nearest neighbor.



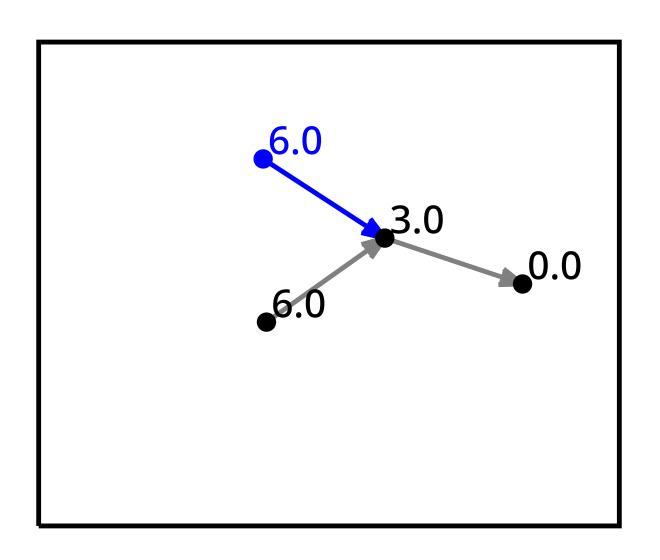
Extend from the nearest neighbor toward the sample.



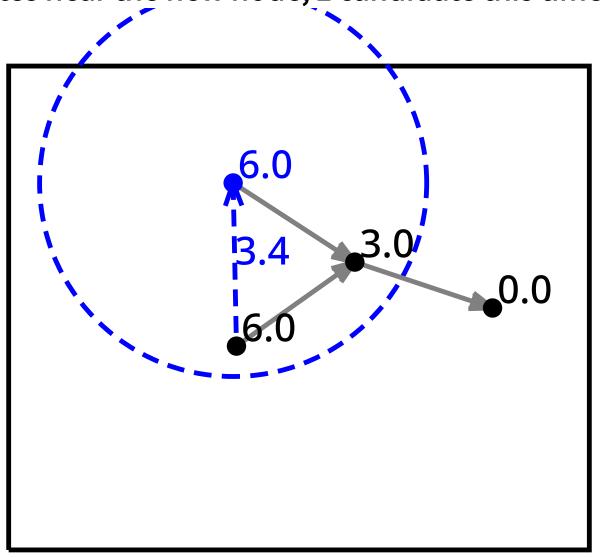
Find candidate parents for the new node; 2 candidates this time.



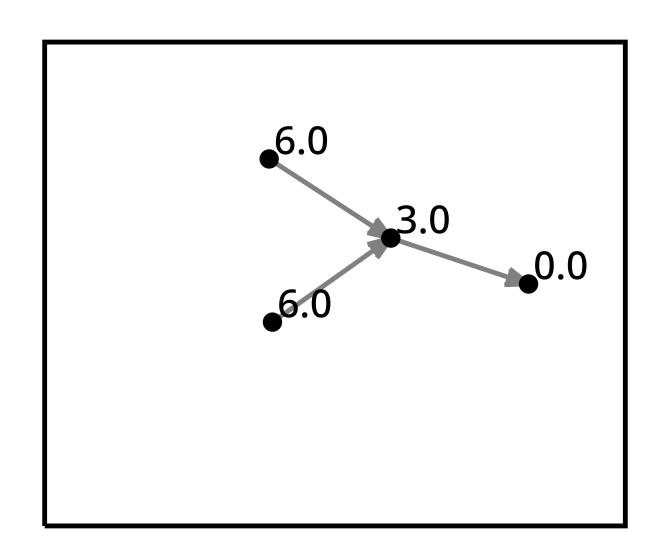
Choose the best parent for the new node.



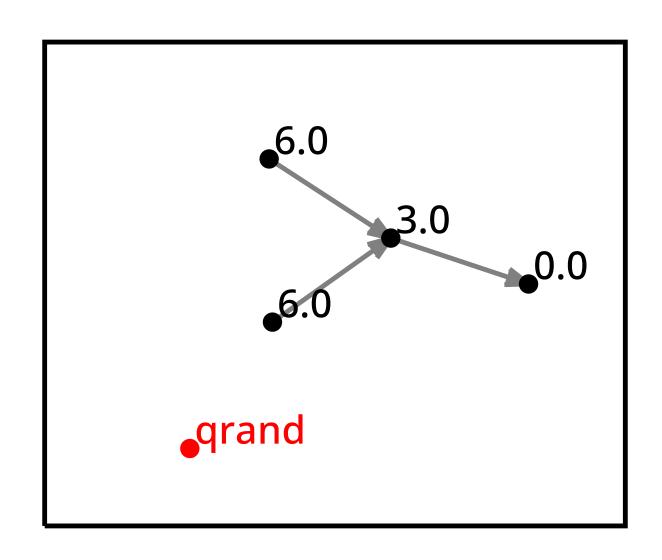
Find rewiring candidates near the new node; 1 candidate this time.



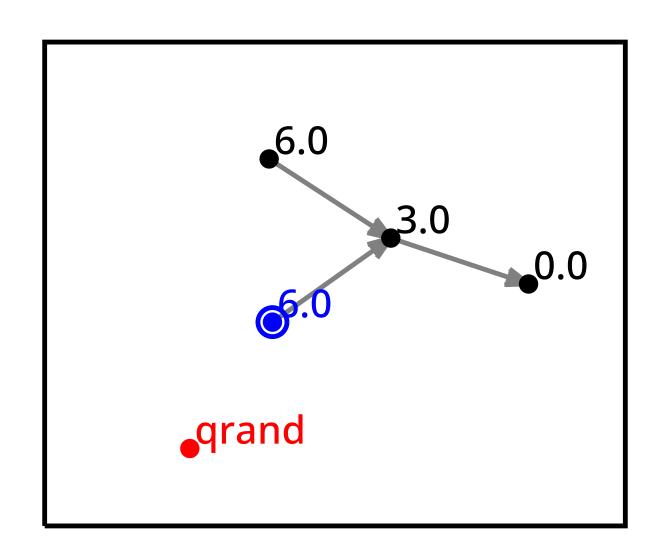
Rewire if needed; 0 changes this time.



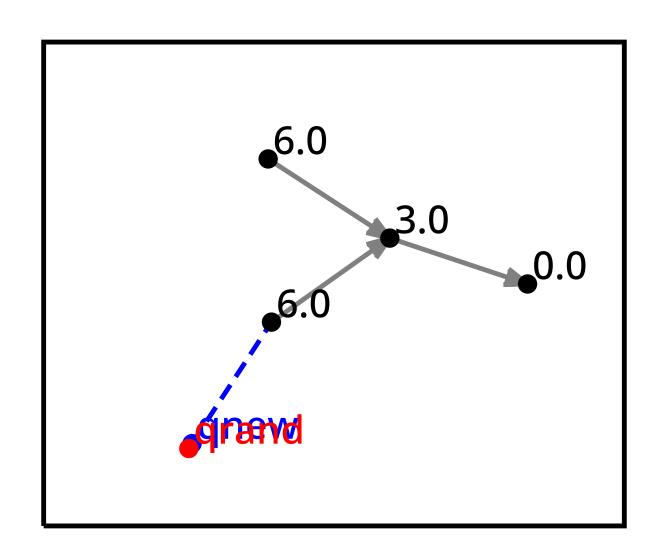
Choose a random sample.



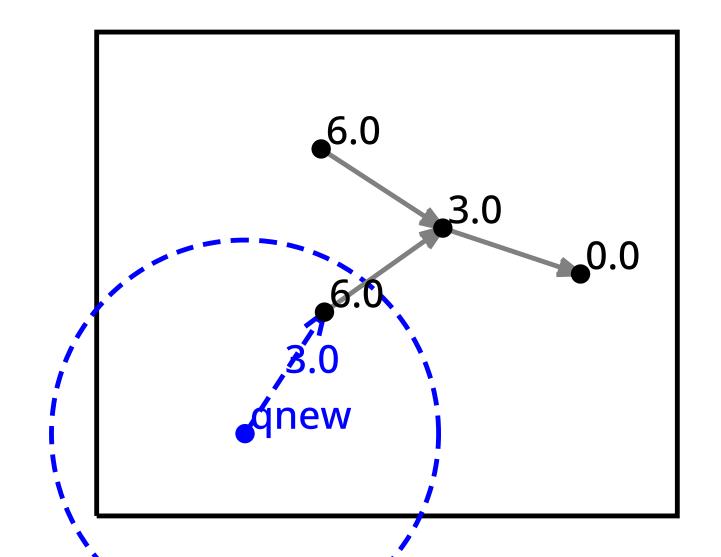
Find the nearest neighbor.



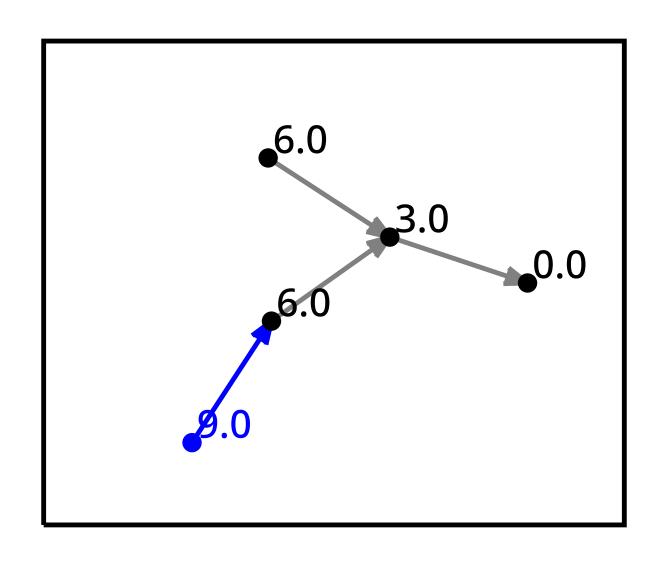
Extend from the nearest neighbor toward the sample.



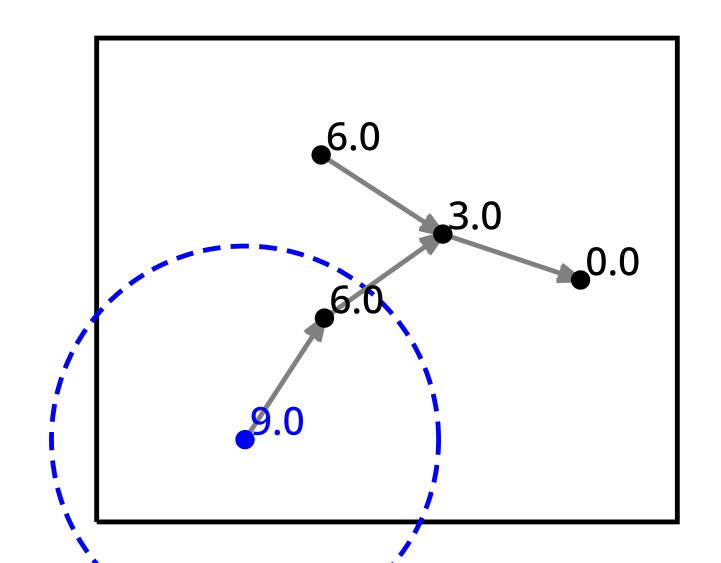
Find candidate parents for the new node; 1 candidate this time.



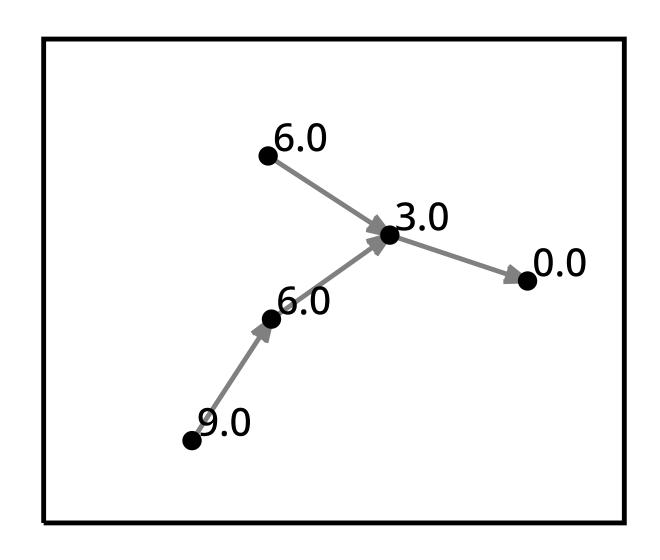
Choose the best parent for the new node.



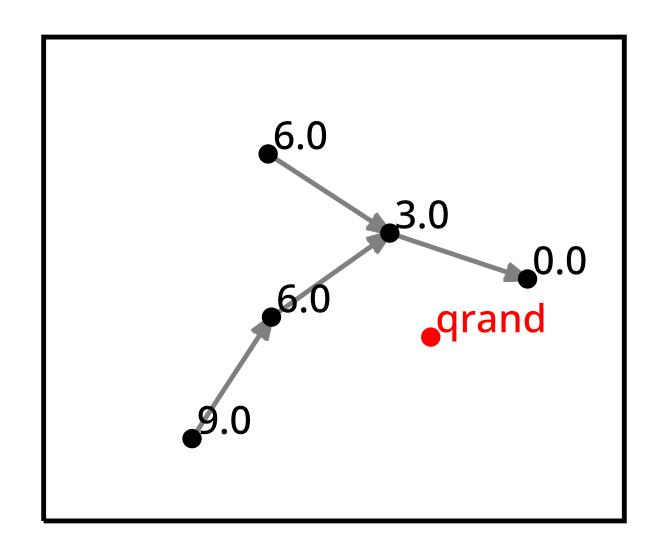
Find rewiring candidates near the new node; 0 candidates this time.



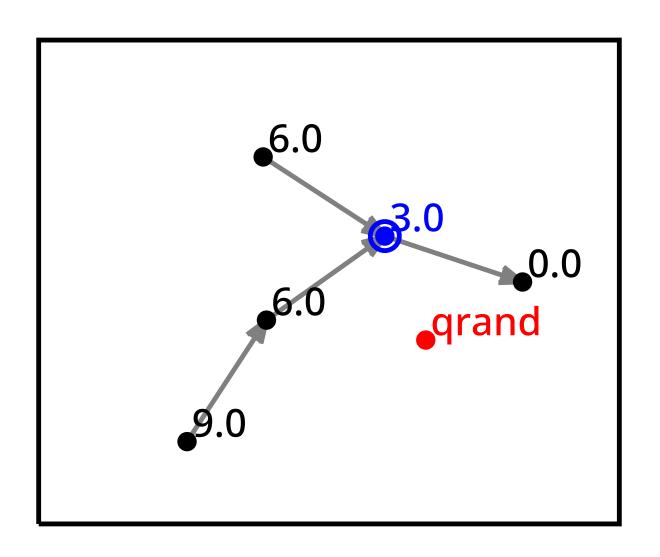
Rewire if needed; 0 changes this time.



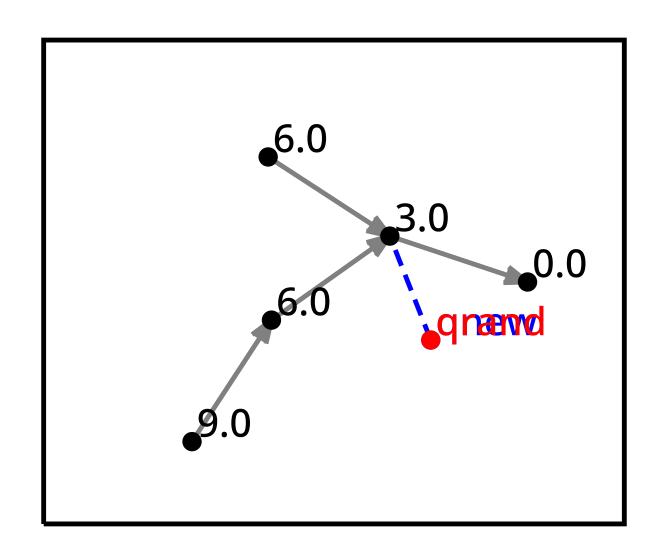
Choose a random sample.



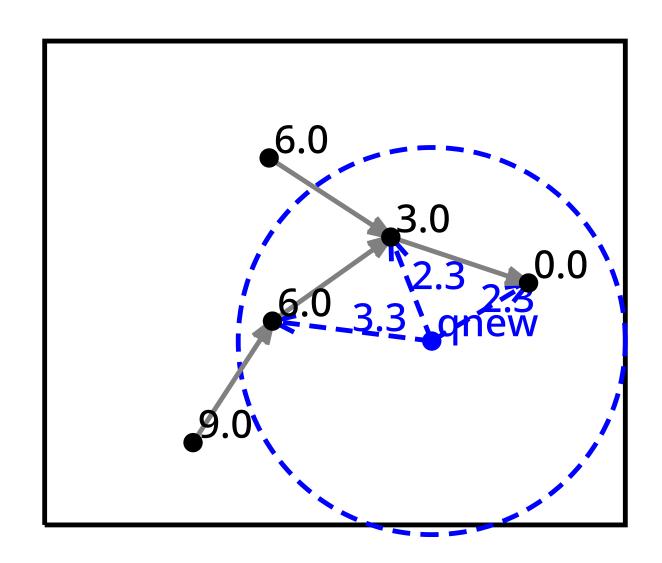
Find the nearest neighbor.



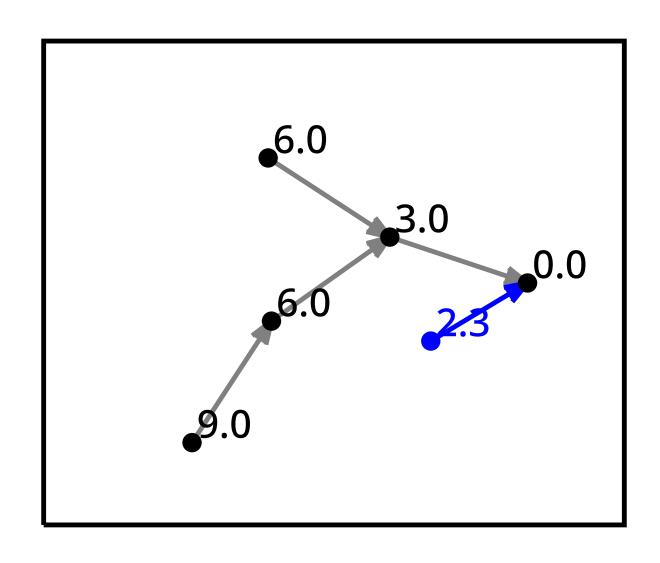
Extend from the nearest neighbor toward the sample.



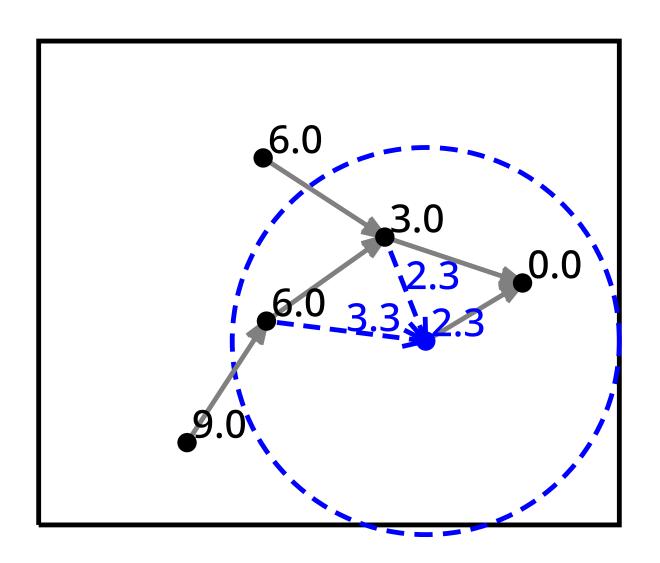
Find candidate parents for the new node; 3 candidates this time.



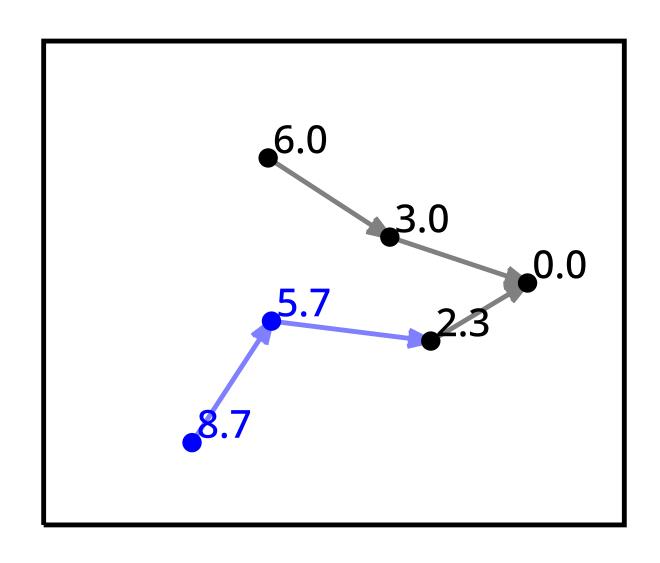
Choose the best parent for the new node.



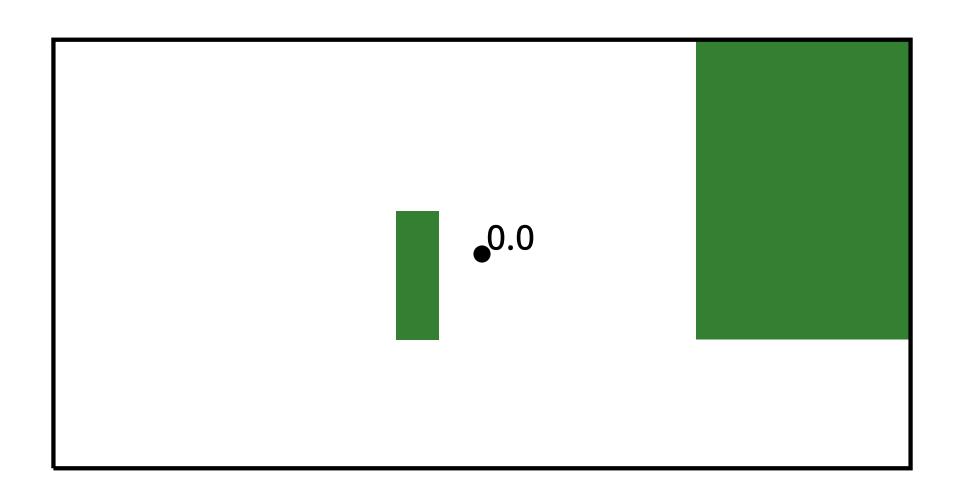
Find rewiring candidates near the new node; 2 candidates this time.



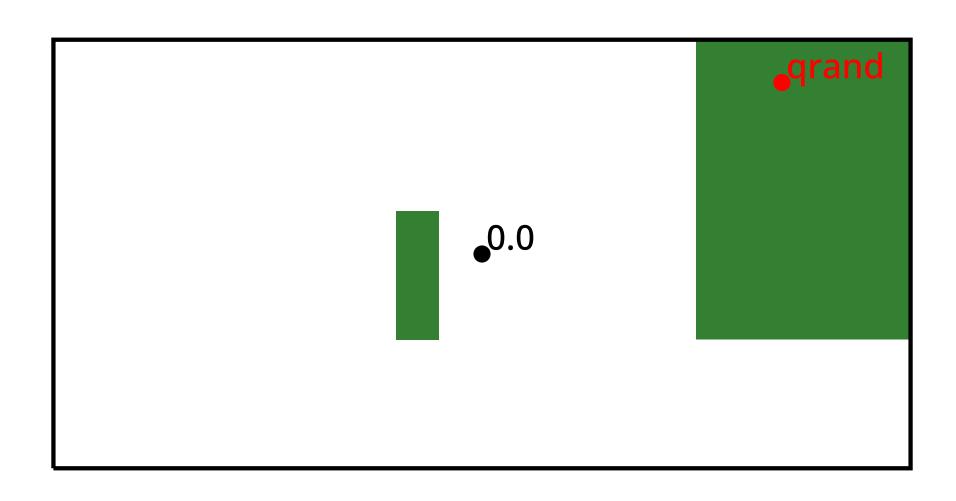
Rewire if needed; 2 changes this time.



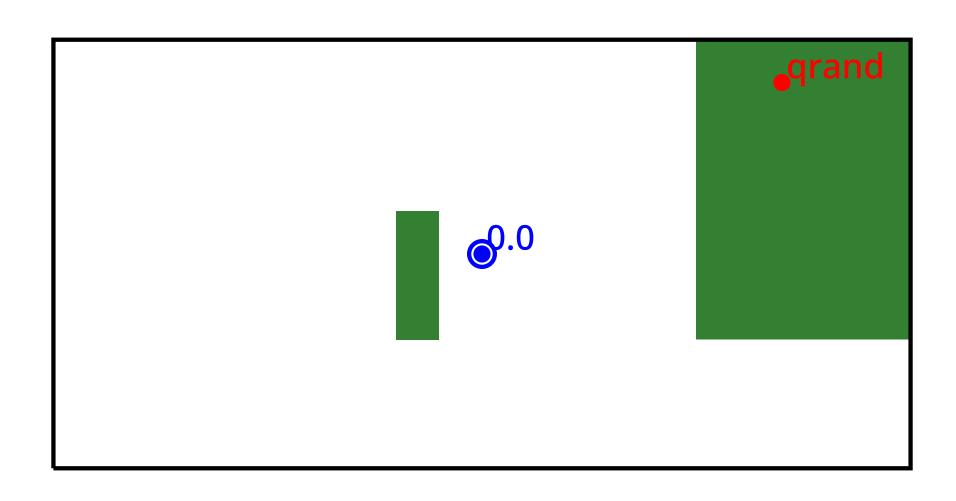
Start from a tree with just the start configuration.



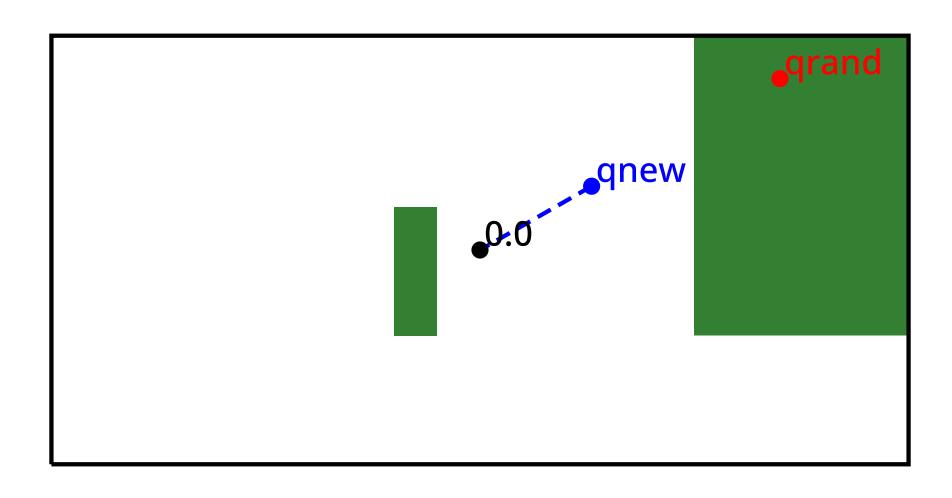
Choose a random sample.



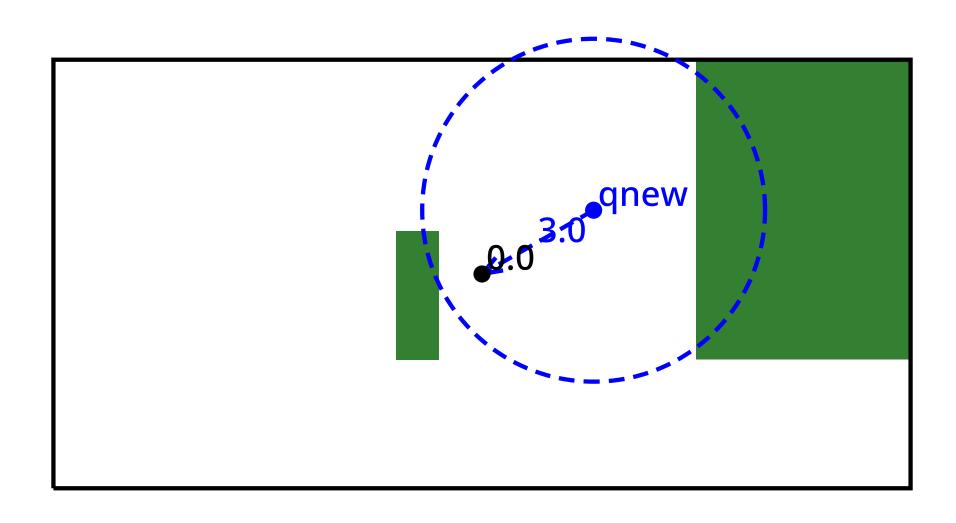
Find the nearest neighbor.



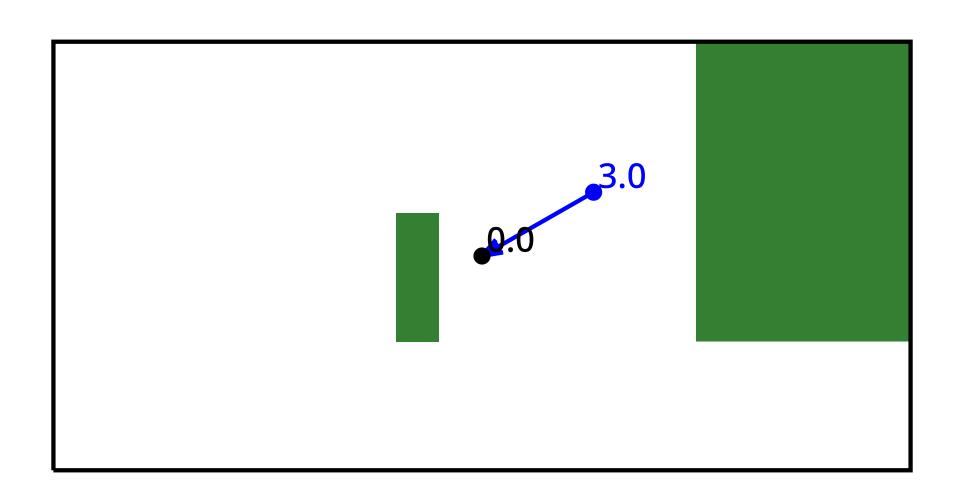
Extend from the nearest neighbor toward the sample.



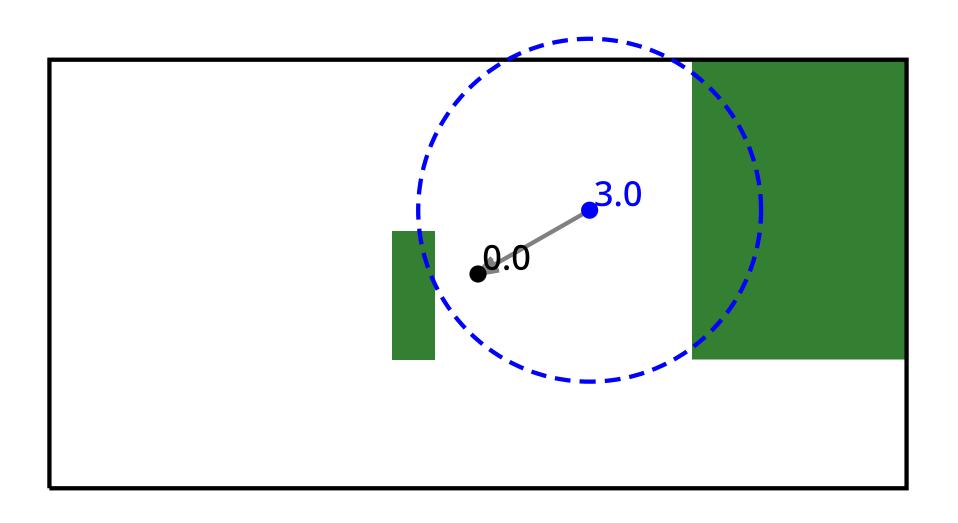
Find candidate parents for the new node; 1 candidate this time.



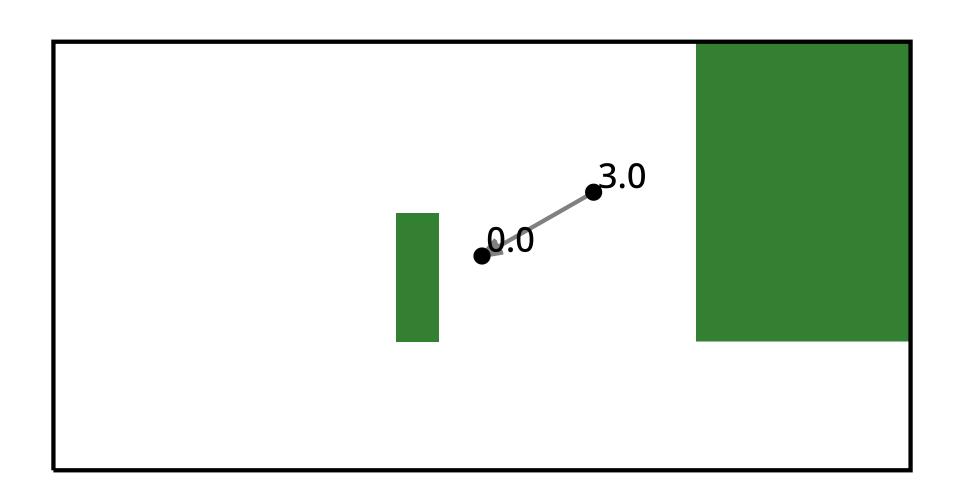
Choose the best parent for the new node.



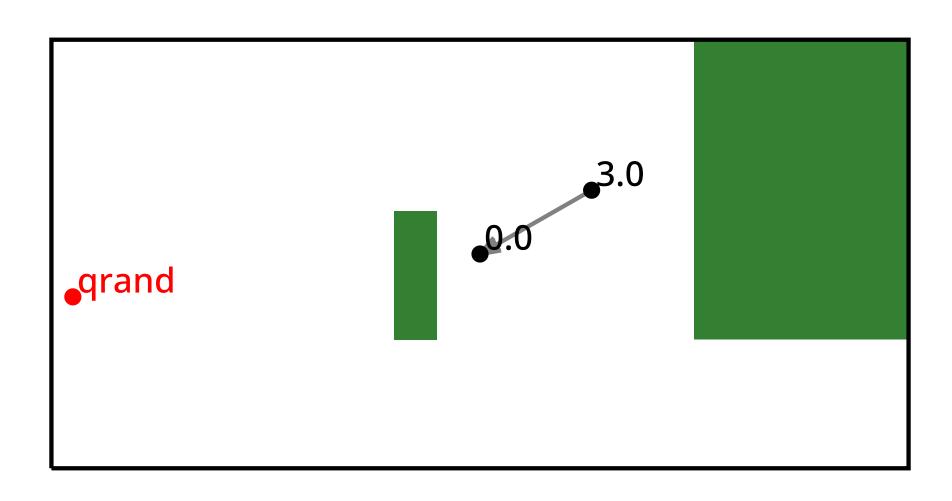
Find rewiring candidates near the new node; 0 candidates this time.



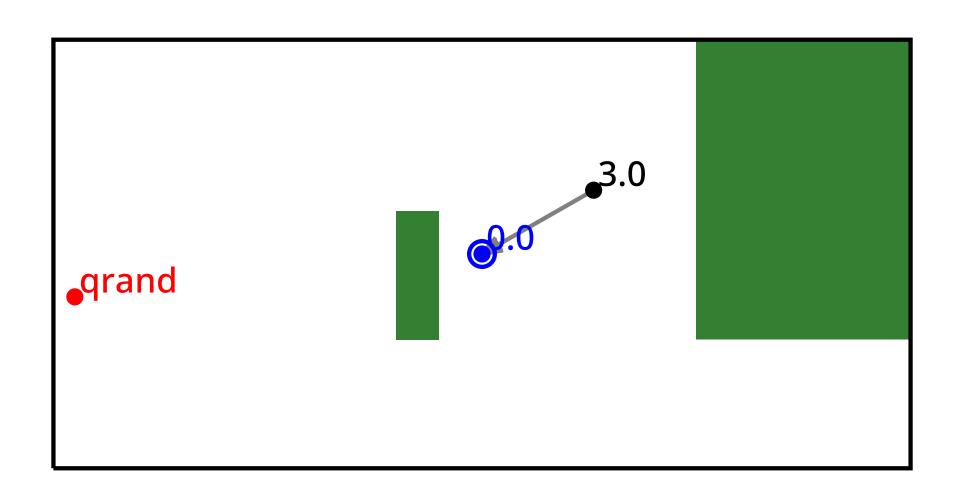
Rewire if needed; 0 changes this time.



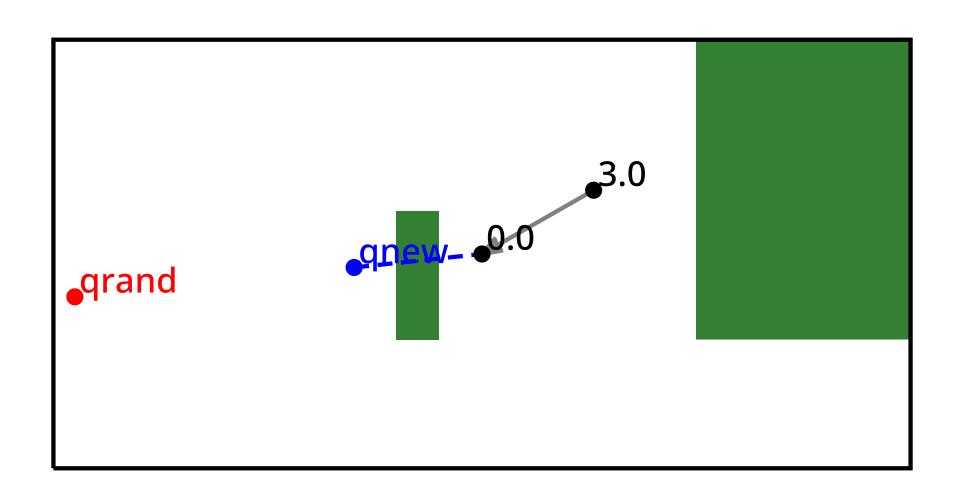
Choose a random sample.



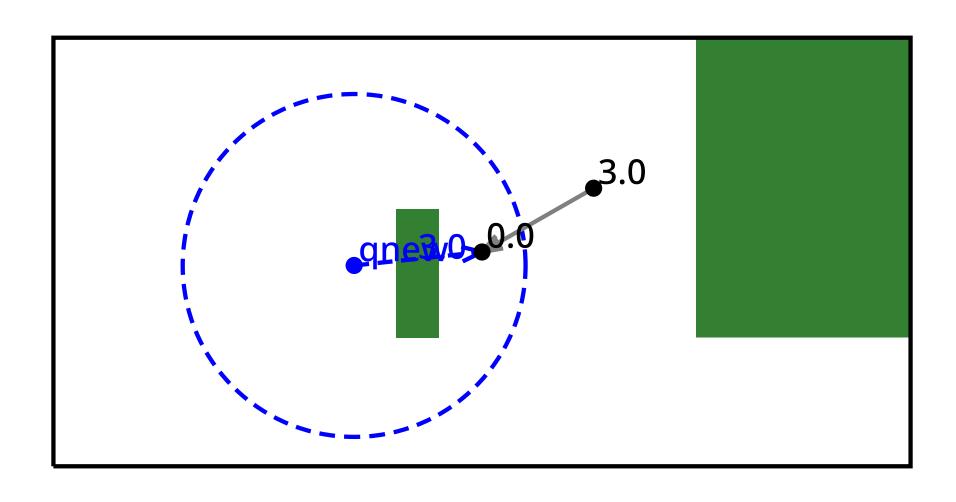
Find the nearest neighbor.



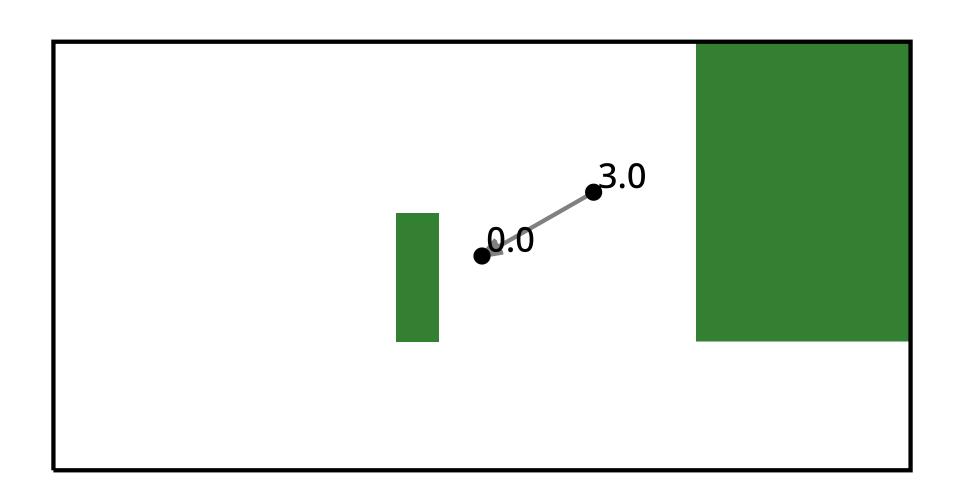
Extend from the nearest neighbor toward the sample.



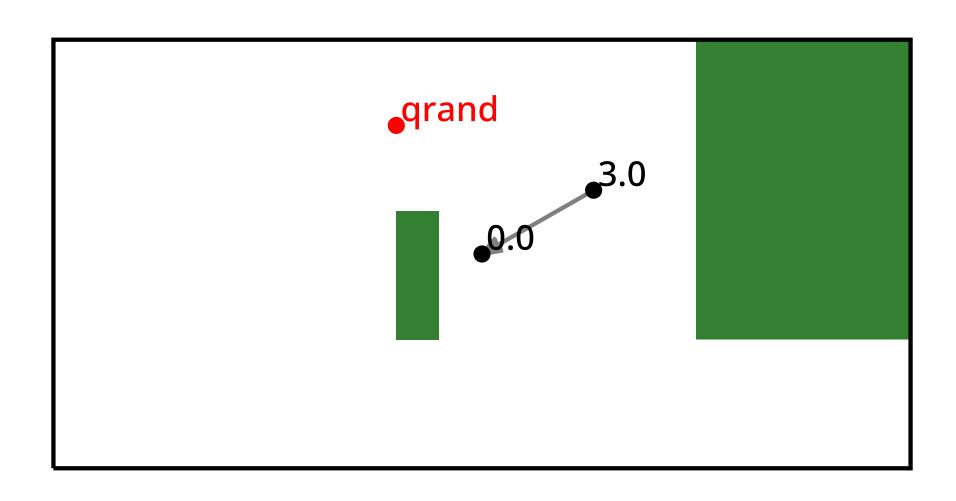
Find candidate parents for the new node; 1 candidate this time.



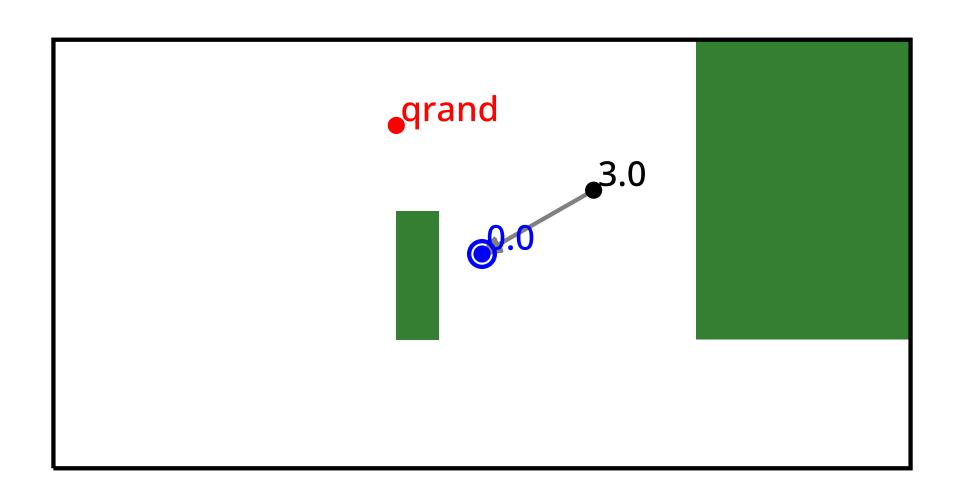
No valid parent available for the new node.



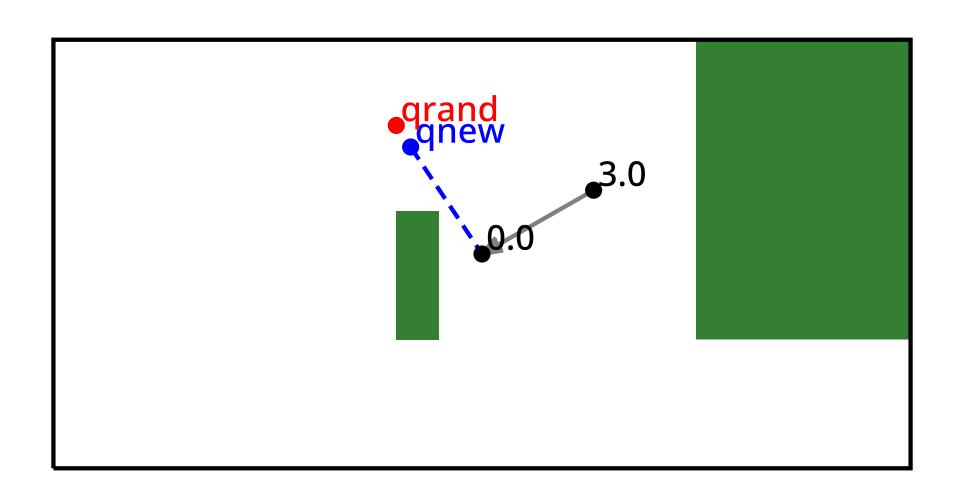
Choose a random sample.



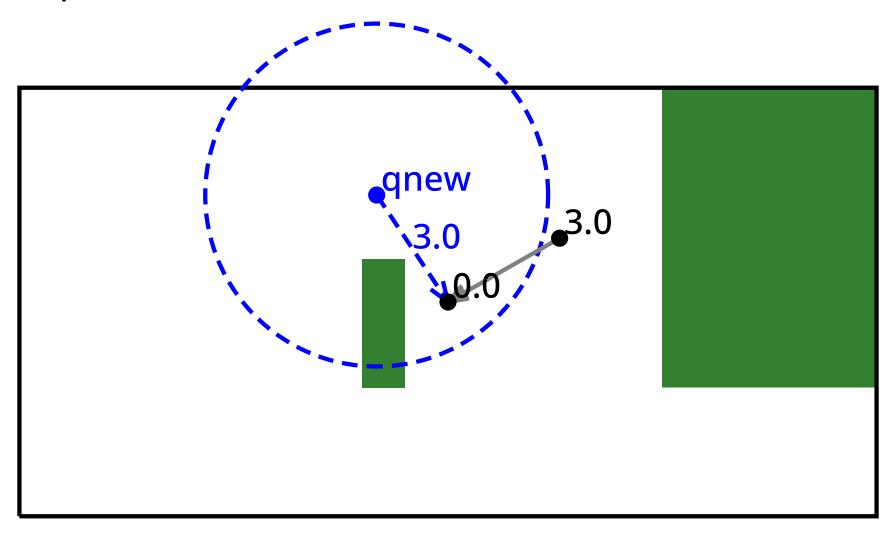
Find the nearest neighbor.



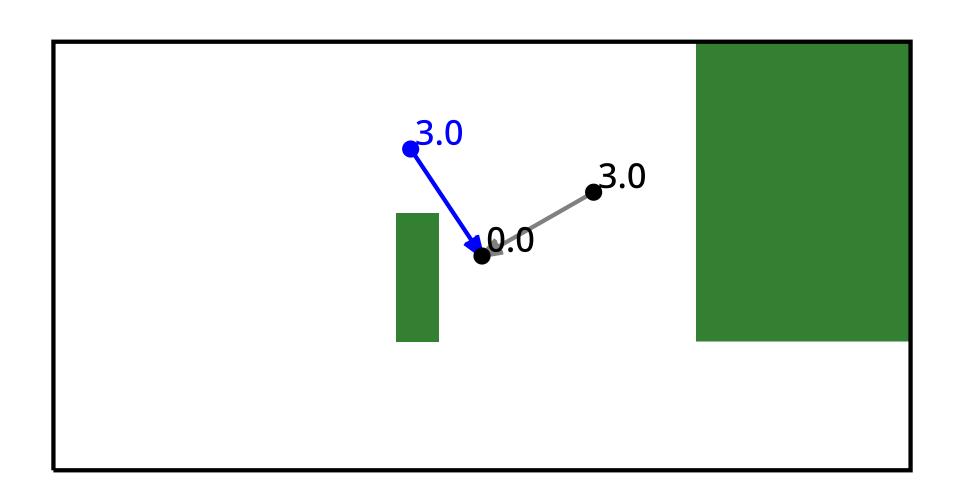
Extend from the nearest neighbor toward the sample.



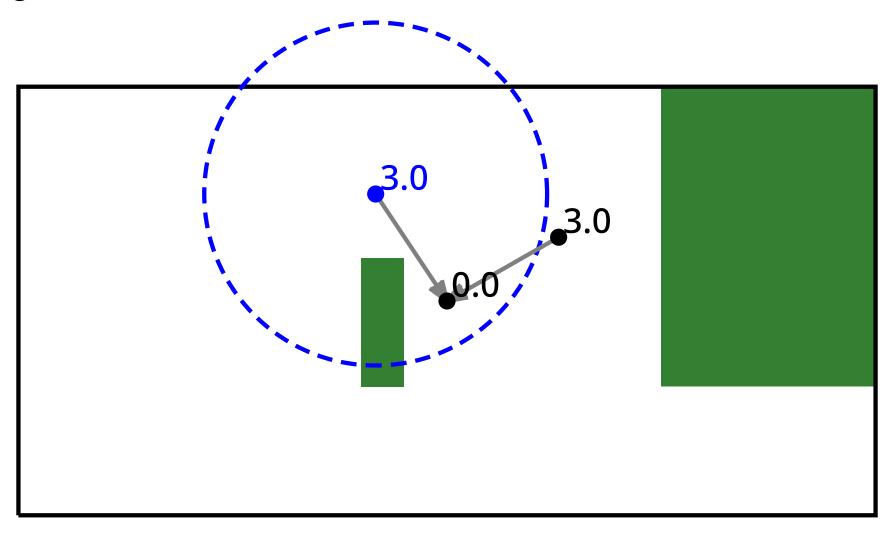
Find candidate parents for the new node; 1 candidate this time.



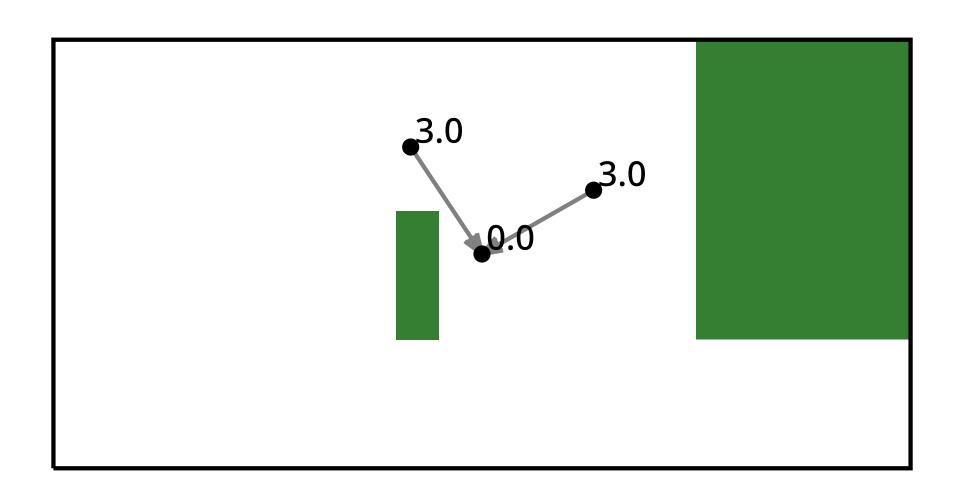
Choose the best parent for the new node.



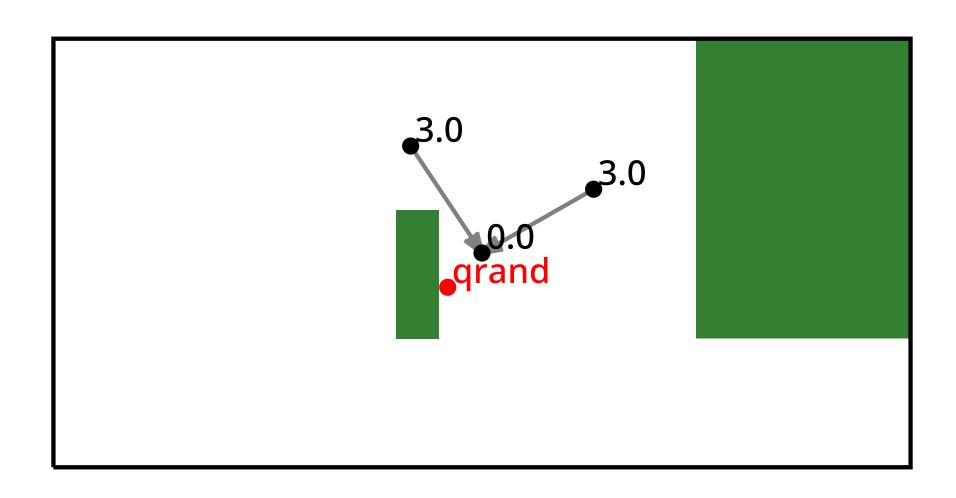
Find rewiring candidates near the new node; 0 candidates this time.



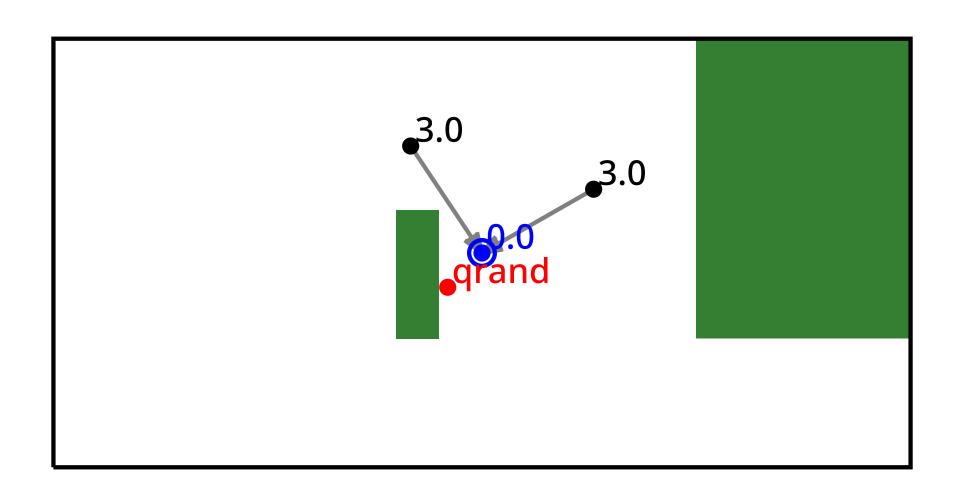
Rewire if needed; 0 changes this time.



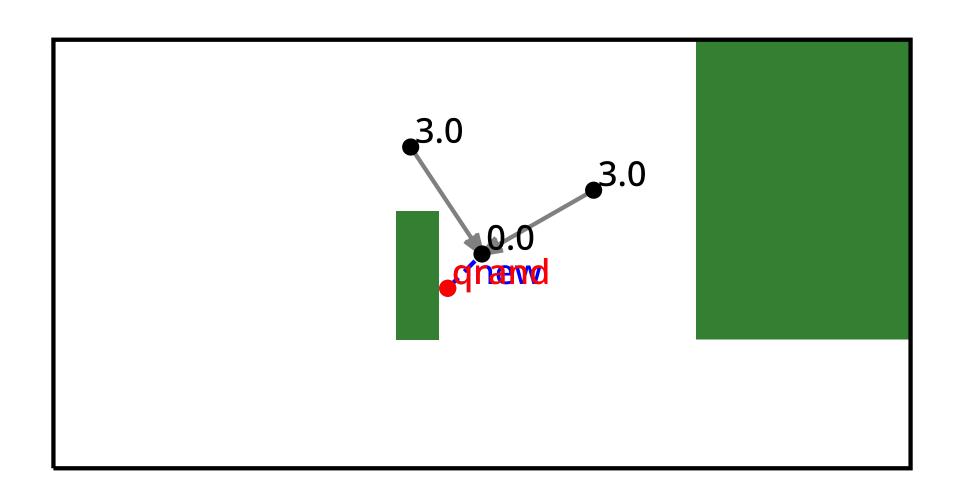
Choose a random sample.



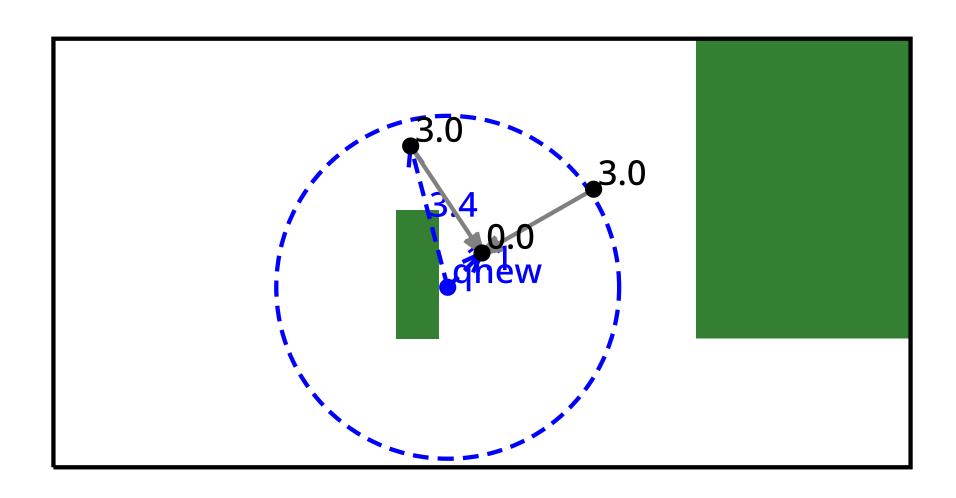
Find the nearest neighbor.



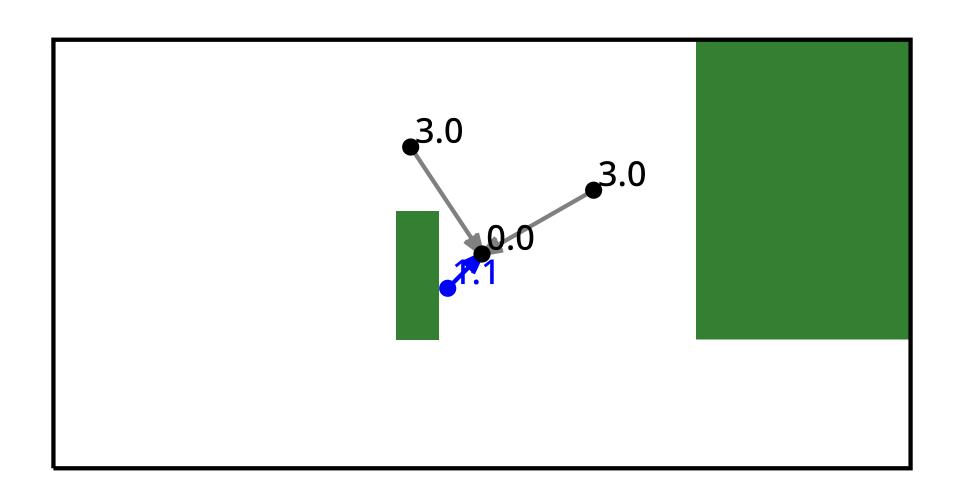
Extend from the nearest neighbor toward the sample.



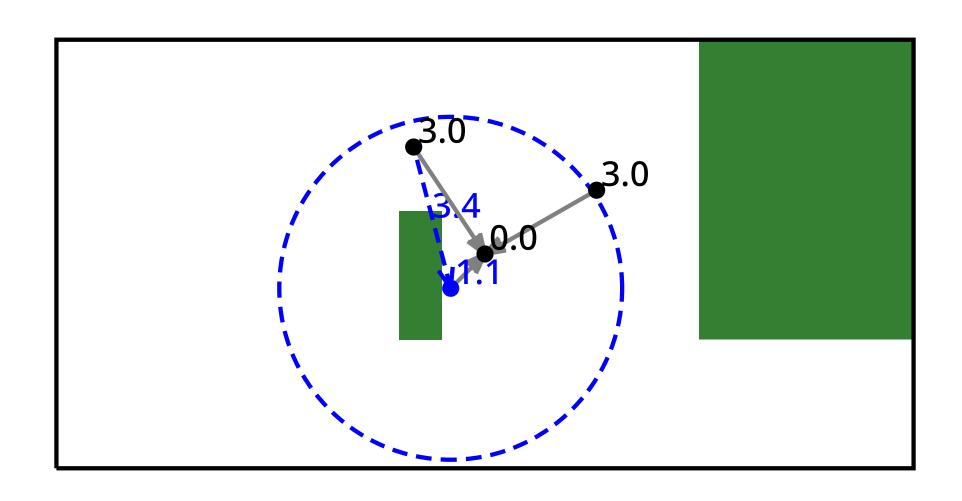
Find candidate parents for the new node; 2 candidates this time.



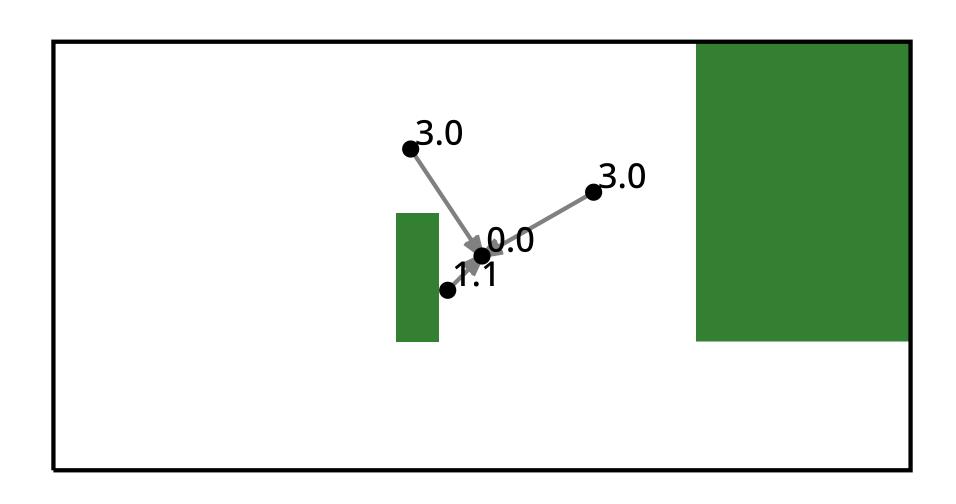
Choose the best parent for the new node.



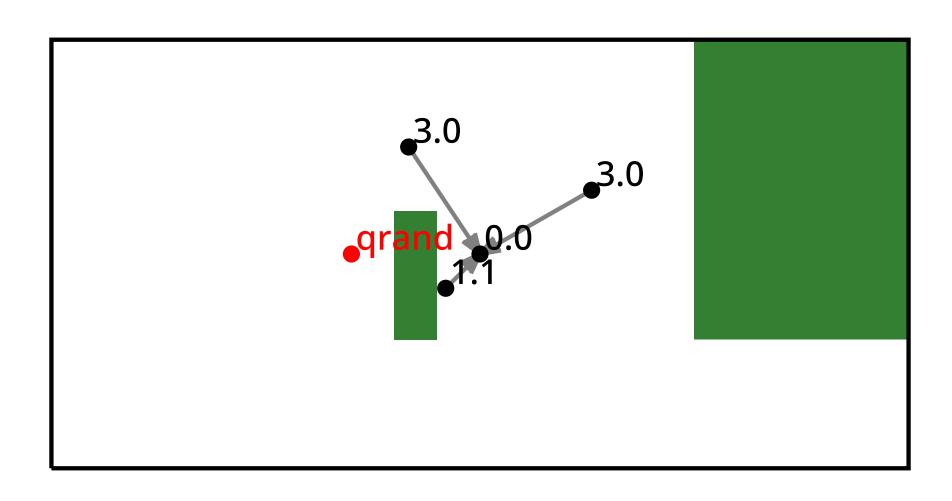
Find rewiring candidates near the new node; 1 candidate this time.



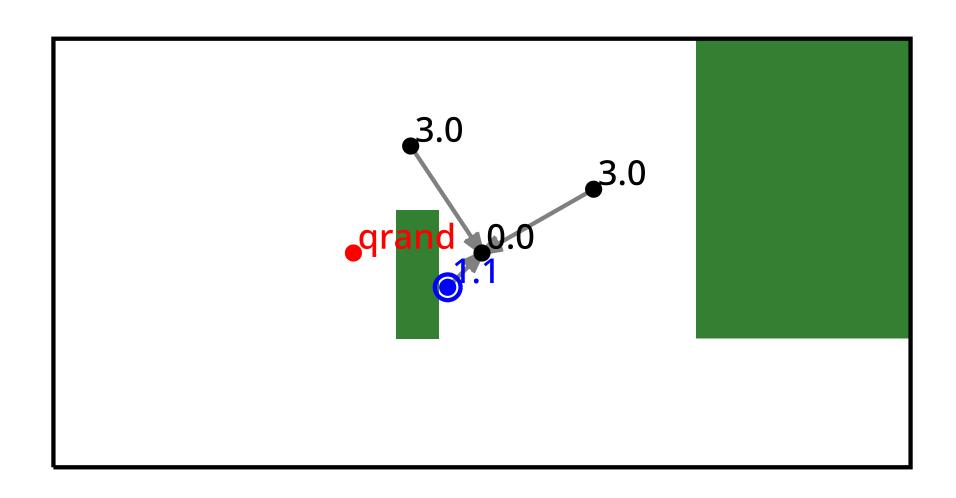
Rewire if needed; 0 changes this time.



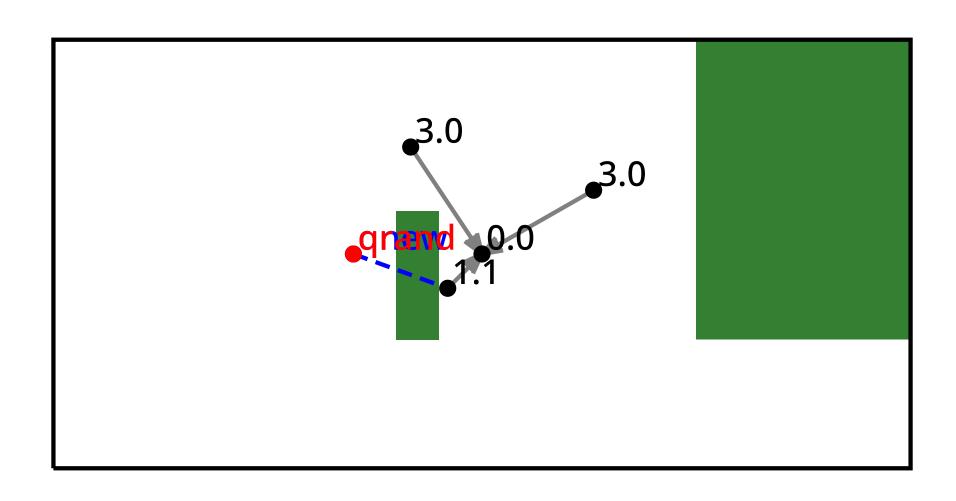
Choose a random sample.



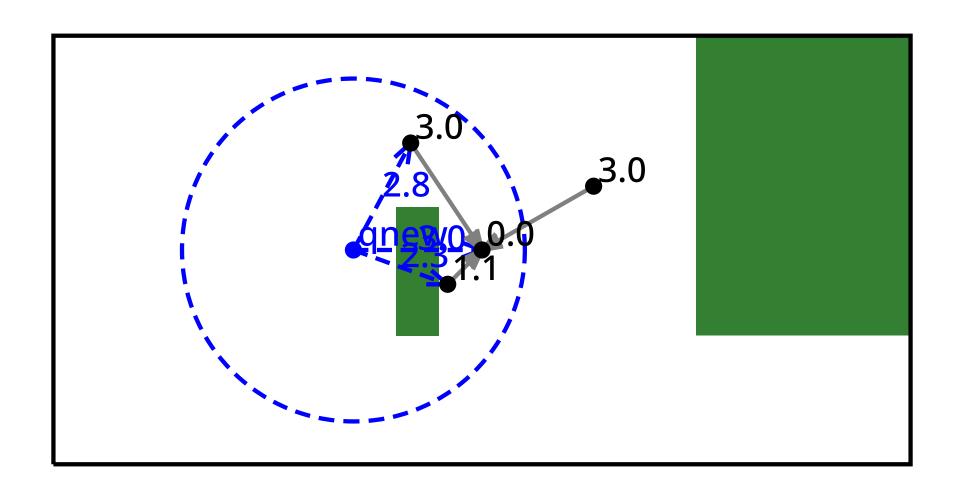
Find the nearest neighbor.



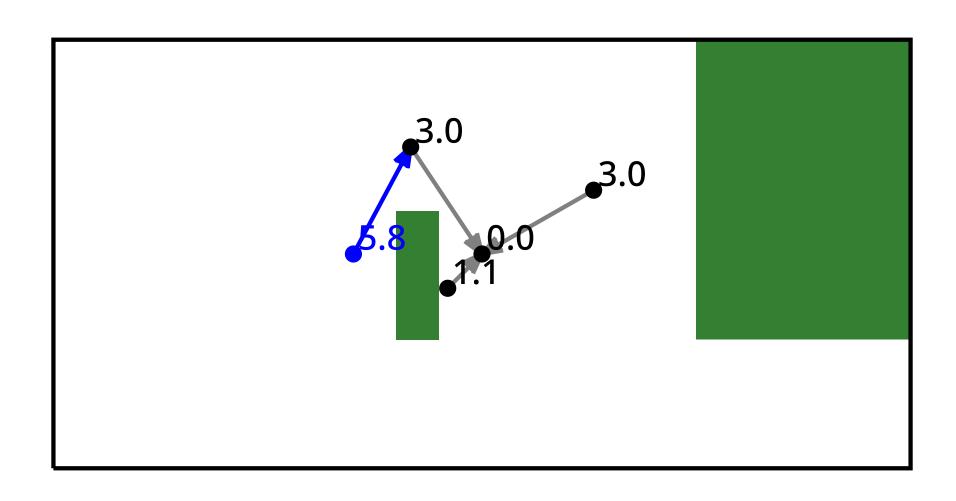
Extend from the nearest neighbor toward the sample.



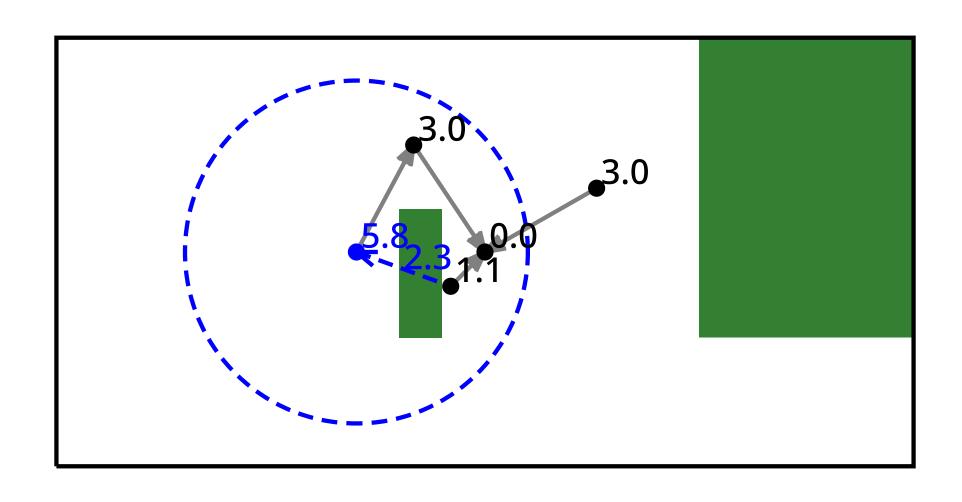
Find candidate parents for the new node; 3 candidates this time.



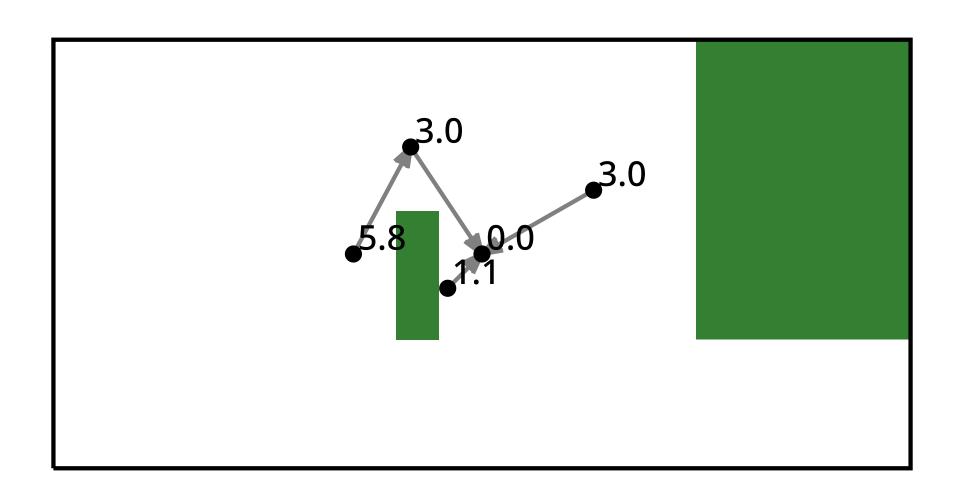
Choose the best parent for the new node.



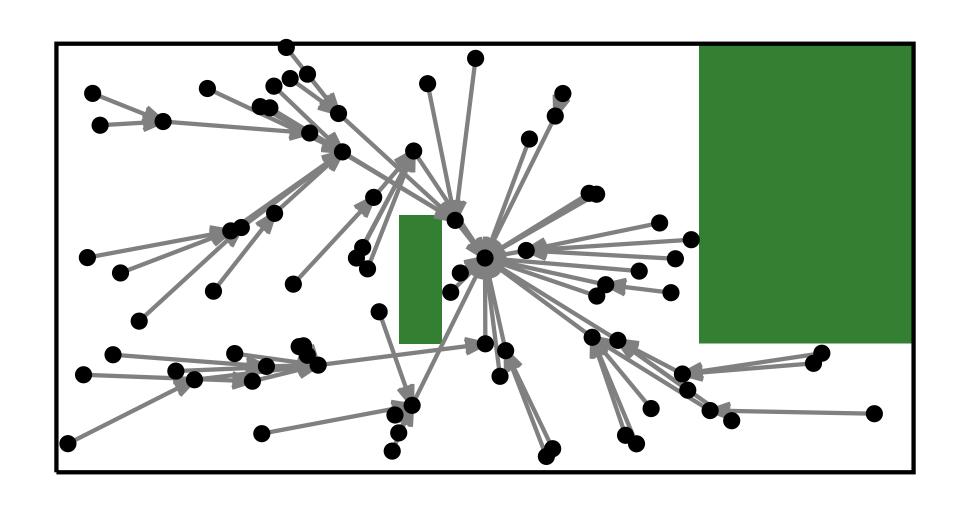
Find rewiring candidates near the new node; 1 candidate this time.



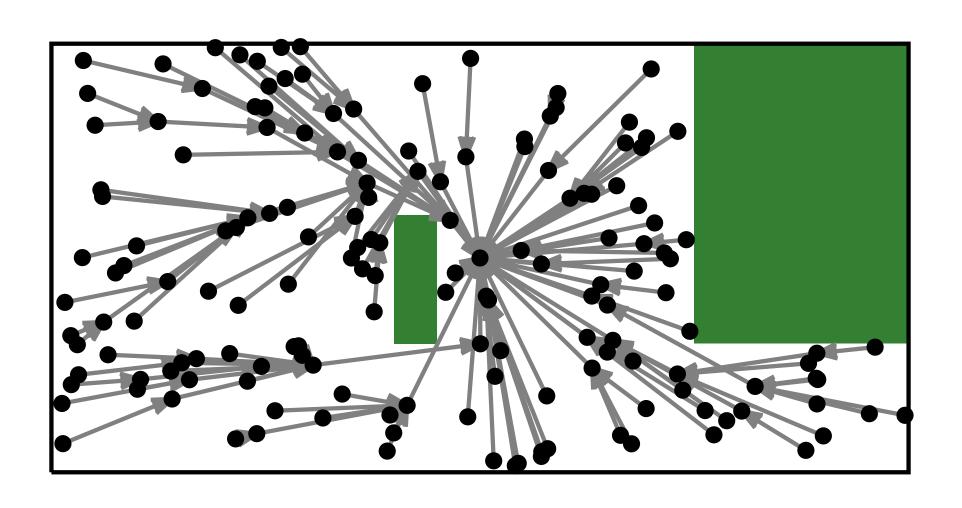
Rewire if needed; 0 changes this time.



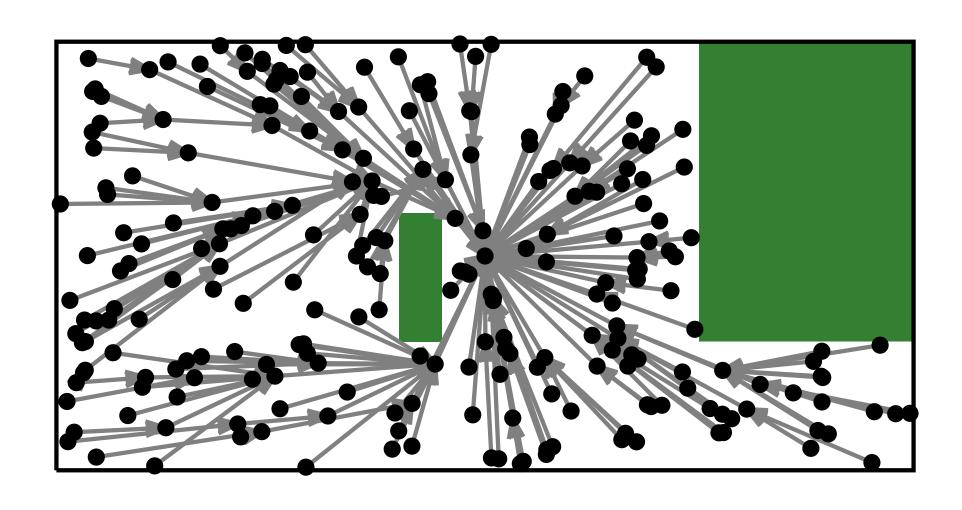
After 100 iterations



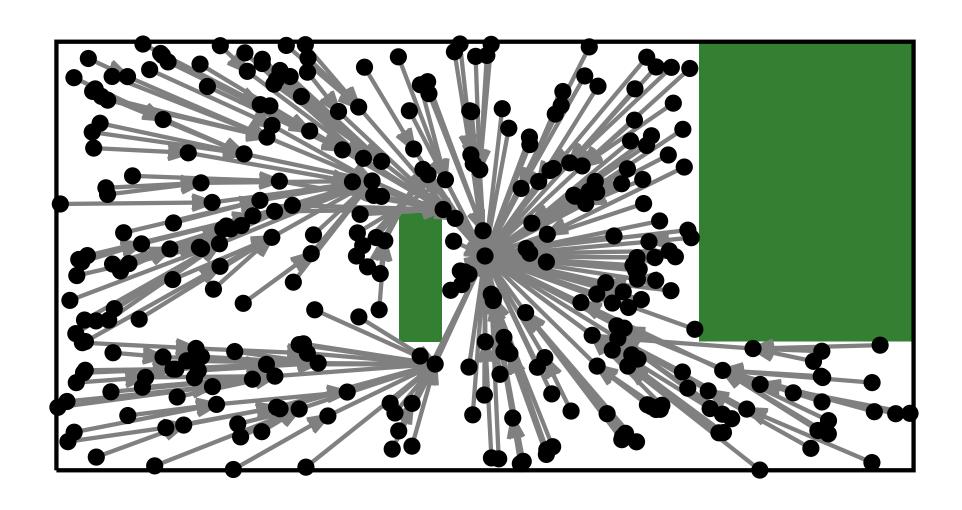
After 200 iterations



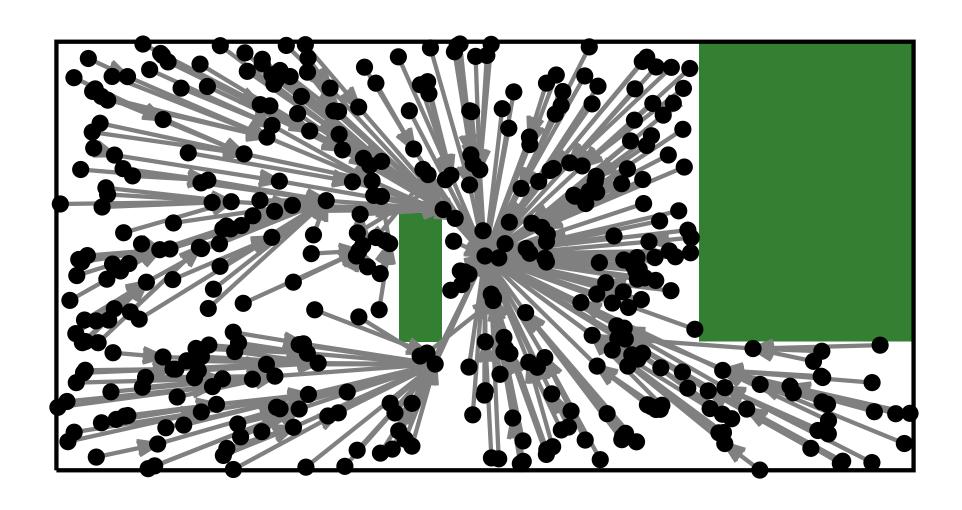
After 300 iterations



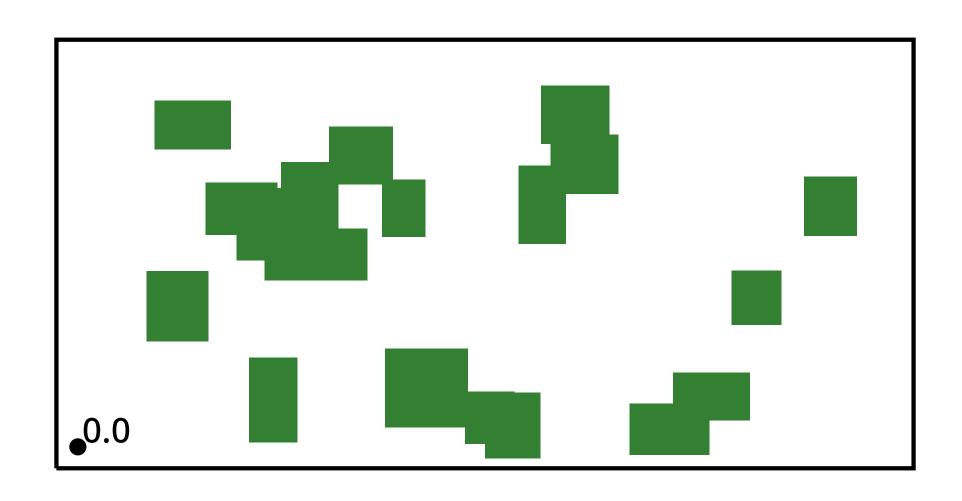
After 400 iterations



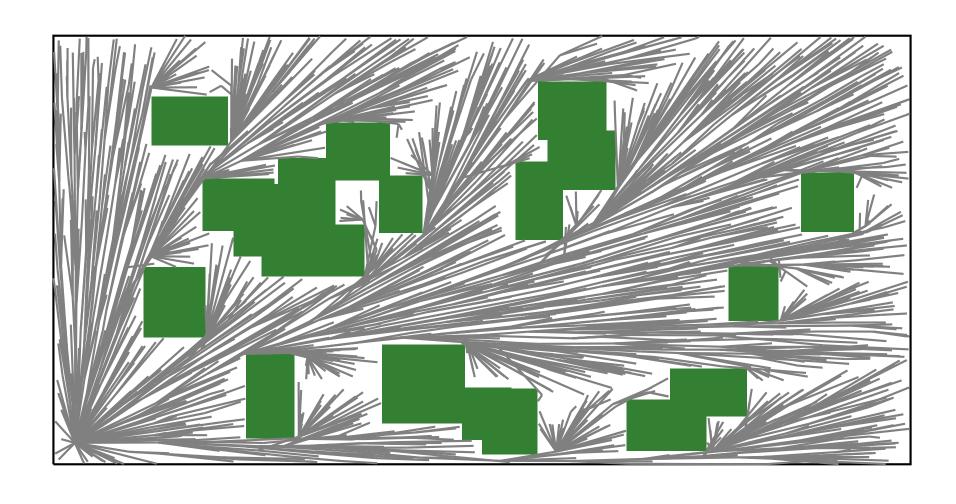
After 500 iterations



Start from a tree with just the start configuration.



After 3000 iterations



Asymptotic optimality

Definition A probabilistically complete motion planner is **asymptotically optimal** if the expected length of the solution it produces converges to the optimal length as the number of samples increases.

$$\lim_{n o\infty} E[c(P)] = c^*$$

Theorem RRT* is asymptotically optimal.