

Project 1: Vehicle Route-Finding

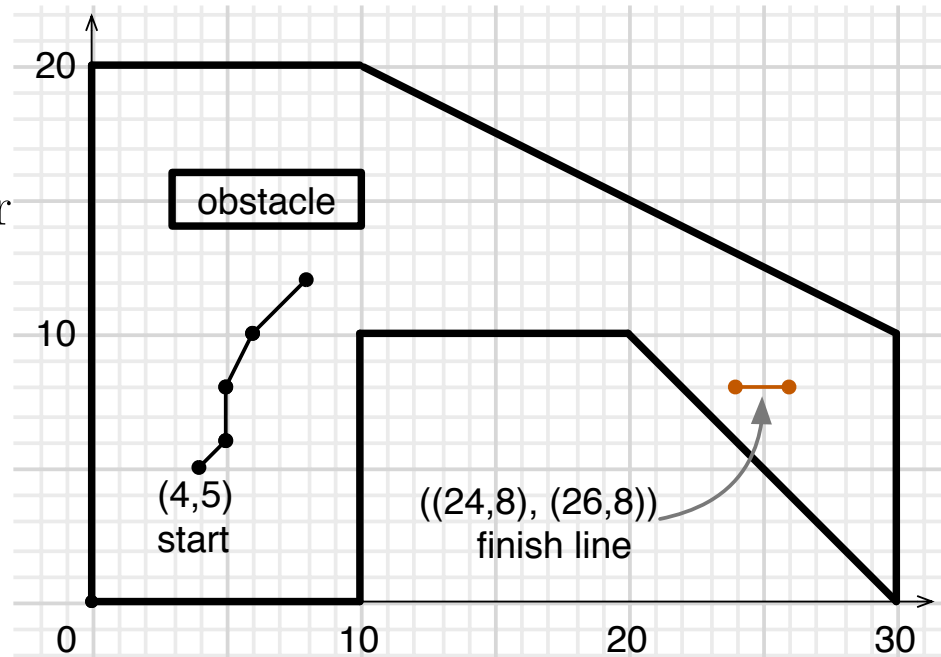
CMSC 421, Fall 2019

Last update September 11, 2019

- ▶ Due date: Sept 28
- ▶ Late date (20% off): Oct 1

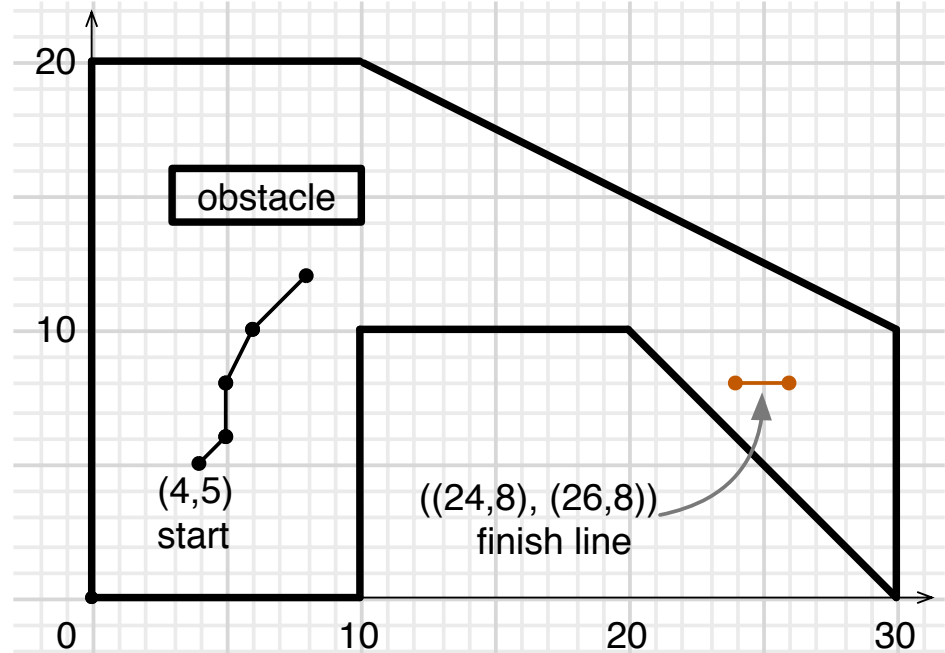
Problem domain

- Modified version of [Racetrack](#)
 - Invented in early 1970s
 - played by hand on graph paper
- 2-D polygonal region
 - Inside are a starting point, finish line, maybe obstacles
- All walls are straight lines
- All coordinates are nonnegative integers
- Robot vehicle begins at starting point, can make certain kinds of moves
- Want to move it to the finish line as quickly as possible
 - Without crashing into any walls
 - Need to come to a complete stop on the finish line



Moving the vehicle

- Before the i 'th move, current state is
 $s_{i-1} = (p_{i-1}, z_{i-1})$
 - location $p_{i-1} = (x_{i-1}, y_{i-1})$, nonnegative integers
 - velocity $z_{i-1} = (u_{i-1}, v_{i-1})$, integers



- To move the vehicle
 - First choose a new velocity $z_i = (u_i, v_i)$, where

$$u_i \in \{u_{i-1} - 1, u_{i-1}, u_{i-1} + 1\}, \quad (1)$$

$$v_i \in \{v_{i-1} - 1, v_{i-1}, v_{i-1} + 1\}. \quad (2)$$

- New location: $p_i = (x_{i-1} + u_i, y_{i-1} + v_i)$
 - New state: $s_i = (p_i, z_i)$

Example

- Initial state:

$$p_0 = (4, 5)$$

$$z_0 = (0, 0)$$

$$s_0 = (p_0, z_0) = ((4, 5), (0, 0))$$

- First move:

$$z_1 = (0, 0) + (1, 1) = (1, 1)$$

$$p_1 = (4, 5) + (1, 1) = (5, 6)$$

$$s_1 = ((5, 6), (1, 1))$$

- Second move:

$$z_2 = (1, 1) + (-1, 1) = (0, 2)$$

$$p_2 = (5, 6) + (0, 2) = (5, 8)$$

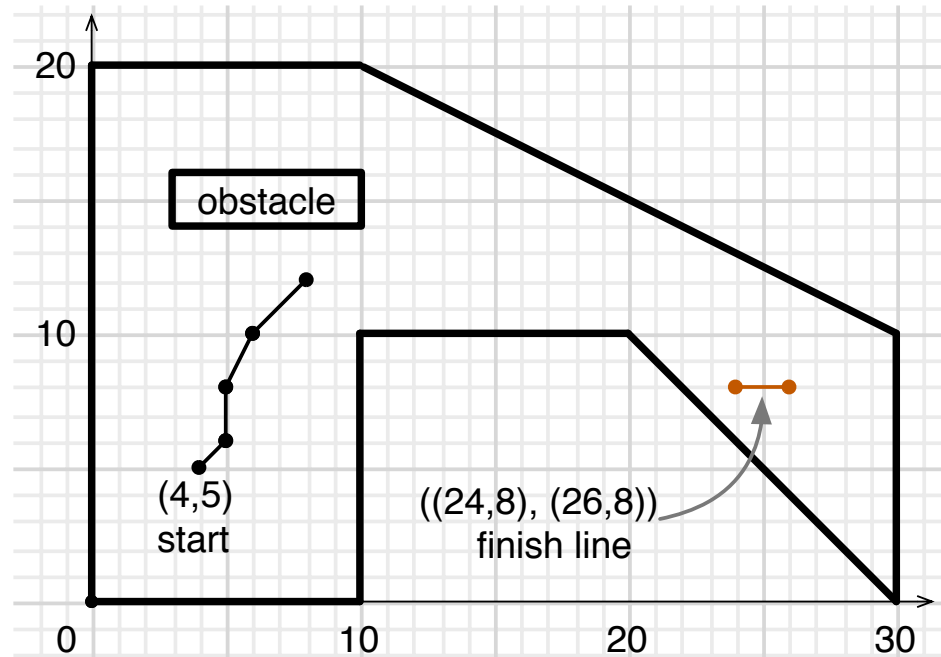
$$s_2 = ((5, 8), (0, 2))$$

- Third move:

$$z_3 = (0, 2) + (1, 0) = (1, 2)$$

$$p_3 = (5, 8) + (1, 2) = (6, 10)$$

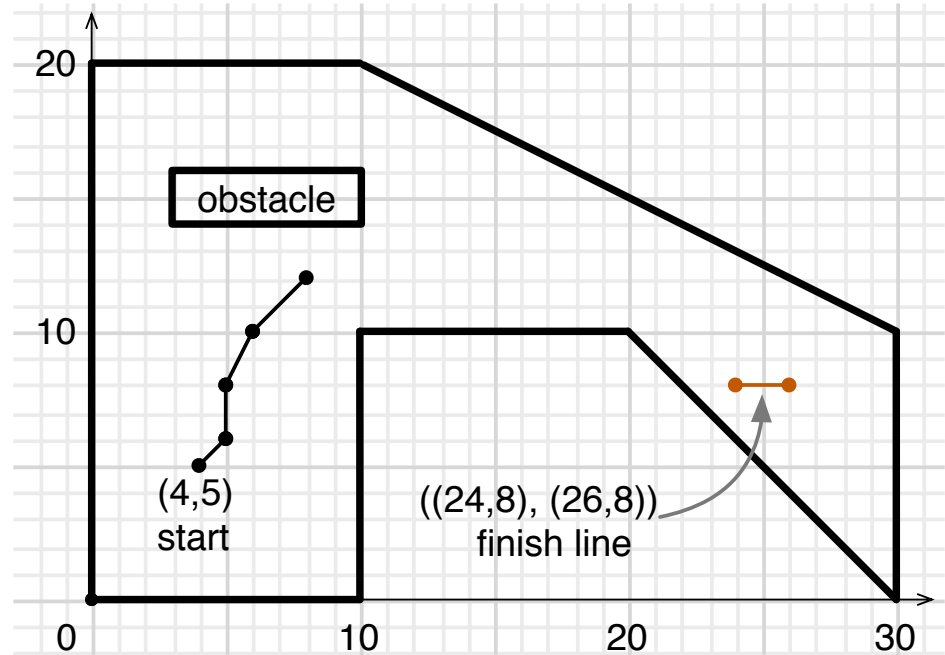
$$s_3 = ((6, 10), (1, 2))$$



Walls

- *edge*: a pair of points (p, q)
 - $p = (x, y), q = (x', y')$
 - coordinates are nonnegative integers
- *wall*: an edge that the vehicle can't cross
- List of walls in the example:

$[((0, 0), (10, 0)), ((10, 0), (10, 10)), ((10, 10), (20, 10)),$
 $((20, 10), (30, 0)), ((30, 0), (30, 10)), ((30, 10), (10, 20)),$
 $((10, 20), (0, 20)), ((0, 20), (0, 0)), ((3, 14), (10, 14)),$
 $((10, 14), (10, 16)), ((10, 16), (3, 16)), ((3, 16), (3, 14))]$



Moves and paths

- *move*: an edge $m = (p_{i-1}, p_i)$
 - ▶ $p_{i-1} = (x_{i-1}, y_{i-1})$
 - ▶ $p_i = (x_i, y_i)$
 - ▶ represents change in location from time $i - 1$ to time i

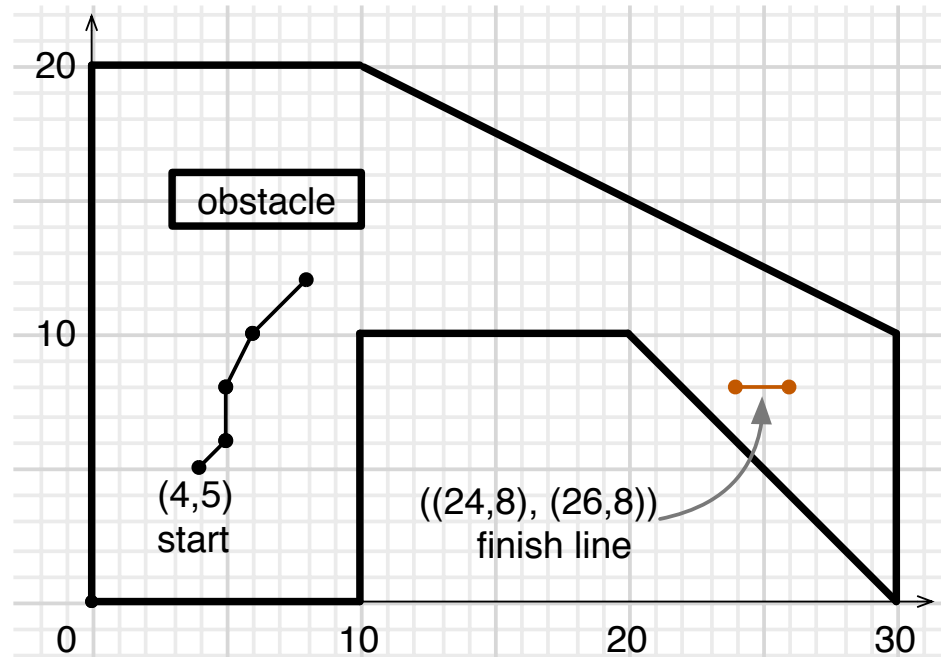
- Example:

$$m_1 = ((4, 5), (5, 6))$$

$$m_2 = ((5, 6), (5, 8))$$

$$m_3 = ((5, 8), (6, 10))$$

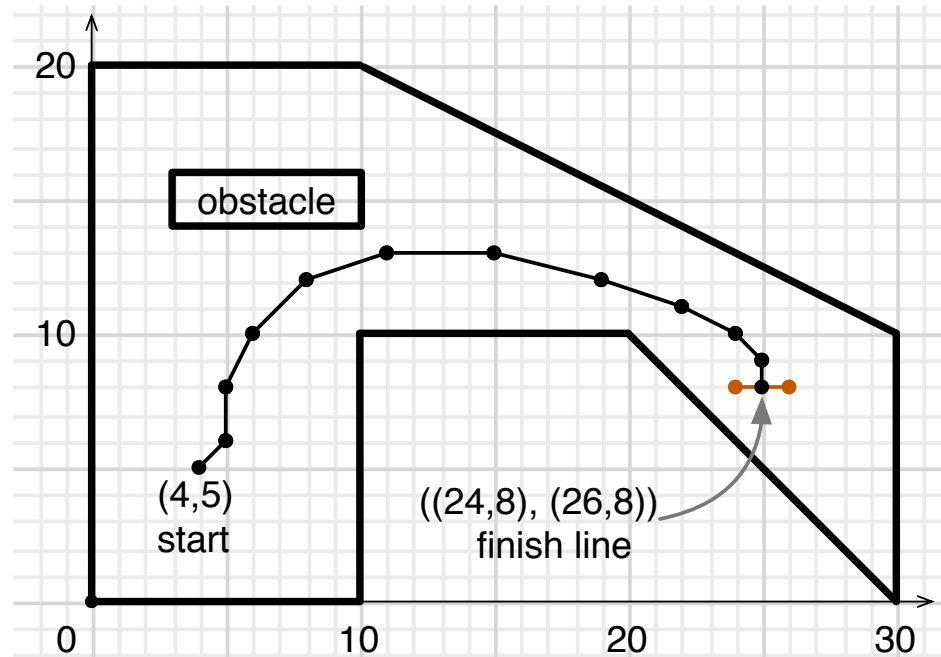
$$m_4 = ((6, 10), (8, 12))$$



- *path*: list of locations $[p_0, p_1, p_2, \dots, p_n]$
 - ▶ represents sequence of moves $(p_0, p_1), (p_1, p_2), (p_2, p_3), \dots, (p_{n-1}, p_n)$
 - ▶ Example: $[(4, 5), (5, 6), (5, 8), (6, 10), (8, 12)]$
- If a move or path intersects a wall, it *crashes*, otherwise it is *safe*

Objective

- *Finish line*:
 - ▶ an edge $f = ((q, r), (q', r'))$
 - ▶ always horizontal or vertical
- Want to reach the finish line with as few moves as possible
- Find a path $[p_0, p_1, \dots, p_n]$
 - ▶ p_n must be on the finish line
 - $\exists t, 0 \leq t \leq 1$, such that $p_n = t(q, r) + (1 - t)(q', r')$
 - ▶ Final velocity must be $(0, 0)$
 - Thus $p_{n-1} = p_n$



Example: $[(4, 5), (5, 6), (5, 8), (6, 10), (8, 12), (11, 13), (15, 13),$
 $(19, 12), (22, 11), (24, 10), (25, 9), (25, 8), (25, 8)]$

Things I'll provide

I'll post a zip archive that includes the following:

- ▶ `fsearch.py` – domain-independent forward search algorithm
 - can do depth first, best first, uniform cost, A^* , and GBFS
 - has hooks for calling a drawing package to draw search spaces
- ▶ `tdraw.py` – code to draw search spaces for racetrack problems
- ▶ `racetrack.py` – code to run `fsearch.py` on racetrack problems
- ▶ `sample_probs.py` – Some racetrack problems I created by hand
- ▶ `make_random_probs.py` – Code to generate random racetrack problems
 - not very good; we probably won't use it
- ▶ `sample_heuristics.py` – Some heuristic functions for Racetrack problems
- ▶ `nmoves.py` – An admissible heuristic function for Racetrack problems
- ▶ `run_tests.bash` – a customizable shell script to run experiments
 - you *must* customize it in order to use it

Here are some details ...

fsearch.py

Domain-independent forward-search algorithm

- ▶ Implementation of **Graph-Search-Redo**
- `main(s0, next_states, goal_test, strategy, h=None, verbose=2, draw_edges=None)`
 - ▶ `s0` – initial state
 - ▶ `next_states(s)` – function that returns the possible next states after s
 - ▶ `goal_test(s)` – function that returns `True` if s is a goal state, else `False`
 - ▶ `strategy` – one of 'bf', 'df', 'uc', 'gbf', 'a*'
 - ▶ `h(s)` – heuristic function, should return an estimate of $h^*(s)$
 - ▶ `verbose` – one of 0, 1, 2, 3, 4
 - how much information to print out (see documentation in the file)
 - ▶ `draw_edges` – function to draw edges in the search space

racetrack.py

Code to run `fsearch.main` on racetrack problems

- `main(problem, strategy, h, verbose=0, draw=0, title='')`
 - ▶ `problem` – `[s0, finish_line, walls]`
 - ▶ `strategy` – one of `'bf'`, `'df'`, `'uc'`, `'gbf'`, `'a*'`
 - ▶ `h(s, f, w)` – heuristic function for racetrack problems
 - s = state, f = finish line, w = list of walls
 - `racetrack.py` converts this to the $h(s)$ function that `fsearch.main` needs
 - ▶ `verbose` – one of 0, 1, 2, 3, 4 (same as for `fsearch.py`)
 - ▶ `draw` – either 0 (draw nothing)
or 1 (draw problems, node expansions, solutions)
 - ▶ `title` – a title to use at the top of the graphics window
 - default is the names of the strategy and heuristic
- Some subroutines that may be useful ...

racetrack.py (continued)

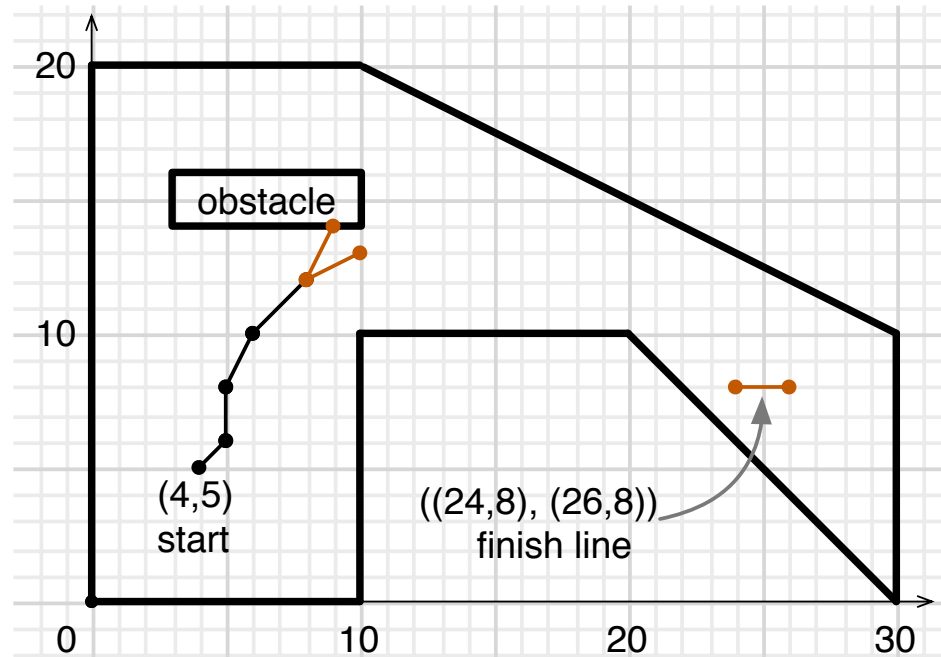
- `intersect(e1,e2)` returns `True` if edges `e1` and `e2` intersect, `False` otherwise
 - `intersect([(0,0),(1,1)], [(0,1),(1,0)])` returns `True`
 - `intersect([(0,0),(0,1)], [(1,0),(1,1)])` returns `False`
 - `intersect([(0,0),(2,0)], [(0,0),(0,5)])` returns `True`
 - `intersect([(1,1),(6,6)], [(5,5),(8,8)])` returns `True`
 - `intersect([(1,1),(5,5)], [(6,6),(8,8)])` returns `False`

Basic idea (except for some special cases)

- ▶ Suppose $e1 = (p_1, p'_1)$, $e2 = (p_2, p'_2)$
- ▶ Calculate the lines that contain the edges
 - $y = m_1x + b_1$; $y = m_2x + b_2$
- ▶ If $m_1 = m_2$ and $b_1 \neq b_2$ then parallel, don't intersect
- ▶ If $m_1 = m_2$ and $b_1 = b_2$ then collinear \Rightarrow check for overlap
 - Does either edge have an endpoint that's inside the other edge?
- ▶ If $m_1 \neq m_2$ then calculate the intersection point p
 - The edges intersect if they both contain p

racetrack.py (continued)

- `crash(e,walls)`
 - `e` is an edge
 - `walls` is a list of walls
 - True if `e` intersects at least one wall in `walls`, else False



- Example:

```
crash([(8,12),(10,13)],walls) returns False
```

```
crash([(8,12),(9,14)],walls) returns True
```

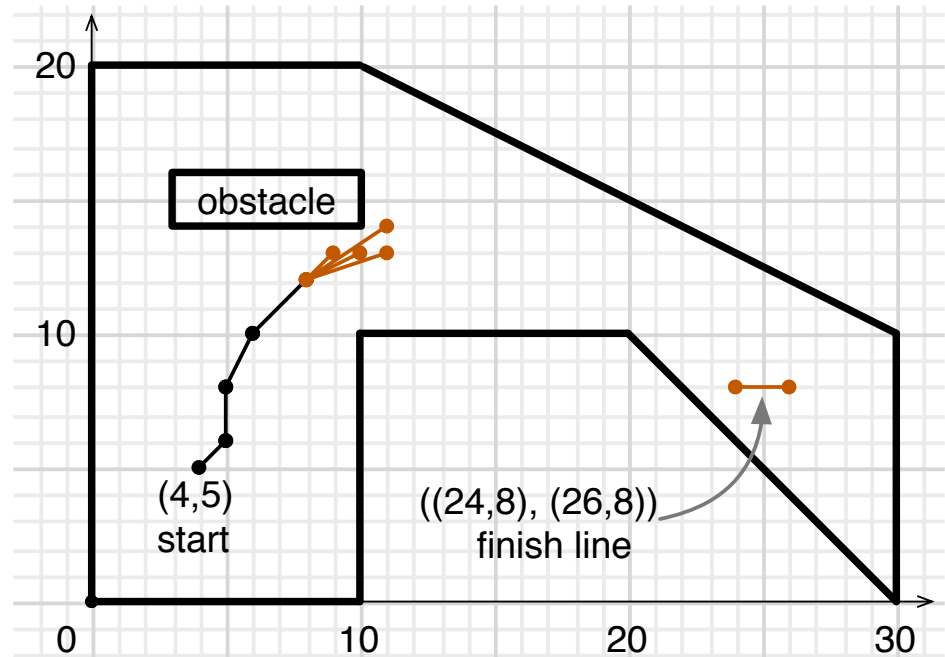
racetrack.py (continued)

- `children(state, walls)`

- ▶ state, list of walls
- ▶ Returns a list $[s_1, s_2, \dots, s_n]$
 - each s_i is a state that we can move to from `state` without crashing

- Example:

- ▶ current state is $((8, 12), (2, 2))$
- ▶ 9 possible states, 5 of them crash into the obstacle
- ▶ `children(((8,12),(2,2)), walls)` returns
 $[((9,13), (1,1)), ((10,13), (2,1)), ((11,13), (3,1)), ((11,14), (3,2))]$



tdraw.py

- `racetrack.py` draws search graphs using Python's turtle graphics
 - ▶ You won't interact with turtle graphics directly
- In most cases it should run with no problem
- If it doesn't:
 - ▶ Do your computer and your Python installation have **Tk** support?
 - ▶ If not, probably you can install **Tk** on your computer and re-install Python
 - ▶ If that doesn't work, you can run `racetrack.main` with `draw = 0`

sample_heuristics.py

Three heuristic functions for the Racetrack domain:

- `h_edist(s, f, walls)` returns Euclidean distance from *s* to the goal, ignoring walls
 - ▶ Not admissible, but finds near-optimal solutions
 - ▶ Problems
 - can go in the wrong direction because it ignores walls
 - can overshoot because it ignores the number of moves needed to stop
- `h_esdist(s, f, walls)` is a modified version of `h_edist`
 - ▶ includes an estimate of how many moves it will take to stop
 - avoids overshooting
 - ▶ still can go in the wrong direction because it ignores walls

sample_heuristics.py (continued)

- `h_walldist(s, f, walls):`
 - ▶ The first time it's called:
 - For each gridpoint that's not inside a wall, cache a rough estimate of the distance to the finish line
 - Breadth-first search backwards from the finish line, one gridpoint at a time
 - ▶ On all subsequent calls
 - Retrieve the cached value
 - Add an estimate of how many moves needed to stop

nmoves.py

Another heuristic function:

- `h_nmoves(s, f, walls)`:
 - ▶ Calculates how many moves would be needed to reach the finish line if there were no walls
 - ▶ It's admissible, but I don't recommend using it
 - A* finds optimal solutions (vs. near-optimal with `h_esdist`), but does about 10 times as many node expansions
 - With `h_esdist`, GBFS expands about the same number of nodes, and `h_esdist` has lower overhead
 - simple calculations vs. a recursive computation
 - Like `h_esdist`, `h_nmoves` ignores walls, so can go in the wrong direction

What to do

- Write a better heuristic function than `h_walldist`
 - ▶ *Don't* just make minor modifications to `h_walldist`, you need to write something of your own
- Look for something that works just as well, but runs faster
 - ▶ Avoid caching distances for *all* the gridpoints
 - ▶ Some of them don't matter; others don't matter *much*
 - ▶ How to tell which ones?
 - Experiment to see what works best
- Another possibility might be to try improving the heuristic estimates so that A^* expands fewer nodes or GBFS finds shorter solutions
 - ▶ *Warning*: I don't advise doing this one
 - ▶ I doubt you'll be able to get significant improvements

What to Submit

- One file, `proj1.py`
 - ▶ In the file, your heuristic function should be named `h_proj1`
 - ▶ In the docstring at the start of the file, include a copy of the honor pledge:

```
I pledge on my honor that I have not given or received  
any unauthorized assistance on this project.
```

```
<your name>
```

- Submit it at the submit server, `submit.cs.umd.edu`

Grading

Evaluation criteria:

35% correctness: – whether your heuristic works correctly,
whether your submission follows the instructions

15% programming style – see the following

- Style guide: <https://www.python.org/dev/peps/pep-0008/>
- Python essays: <https://www.python.org/doc/essays/>

15% documentation

- Docstrings at the start of the file and the start of each function
- Comments elsewhere

35% performance

- A* and GBFS using your heuristic function:
- Running time, length of solution path, number of nodes generated
- Top n performers ($n \approx 5$) will get extra credit