## CSCE 420 - Fall 2022

Programming Assignment PA1
due: Thurs, Sep 28, 2023, 11:59pm, pushed to your class Github account

## Objective

The overall goal of this assignment is to implement A* Search and use it to solve the classic "Blocksworld" (BW) Al search problem. This problem involves stacking blocks from a random initial configuration into a target configuration. You are to write a program to solve random instances of this problem using your own implementation of $\mathbf{A}^{*}$ in $\mathrm{C}++$ or Python, based on the iterative algorithm (called BestFirstSearch) in the textbook using a priority queue. A critical part of this assignment is to develop a heuristic, $\mathbf{h}(\mathbf{n})$, for the Blocksworld that can be used by $A^{*}$ to improve the efficiency of the search, enabling it to solve harder problems with longer solution paths.

A BW problem instance is given as two states: an initial state and a goal state. An example is shown below. The operator for generating moves in this problem can only pick up the top block on any stack and move it to the top of any other stack. Each action has equal (unit) cost. The objective is to find a sequence of actions that will transform the initial state into the goal state (preferably with the fewest moves, hence the shortest path to the goal.) Your program should be able to handle problems with different numbers of stacks and blocks, and the goal state will not always have blocks stacked in order on the first stack - it could be an arbitrary configuration. (This instance requires 10 moves to solve, starting with moving D onto E (making stack 1 empty), and then A into stack 1, and so on...)


Recall that a heuristic $h(n)$ is an estimate of the distance from a node in the search space to the goal. A* uses the heuristic score, in combination with the pathcost to $n, g(n)$, to keep the nodes sorted in the frontier by $f(n)=g(n)+h(n)$ using a priority queue. In the Blocksworld, the pathcost is just the depth of a node in the search tree. The heuristic can be thought of as a function of 2 states: the current state, and the goal state, which is constant for a given run, but could be different for each problem instance.

## Program Interface

The compiled program called 'blocksworld’ should be run from the command line with an interface like this:

```
> blocksworld test1.bwp
or
> python blocksworld.py test1.bwp
```

The filename argument is a problem description in the file format described below. You might also want to add other flags on the command line. For example, you might want to add a flag specifying which heuristic function to use. By default, the program should use your best heuristic; however, you can change it with the '-H' flag; this is useful for simulating BFS using HO.
usage: blocksworld <filename> [-H HO|H1|...] [-MAX_ITERS <int>]
One critical aspect of the program is that it should have a built-in limit on the maximum number of iterations (in the main loop in BestFirstSearch). If that limit is reached for a given problem, the program should give up and report failure. In initial recommendation for a default for max iterations is either 100,000 or $1,000,000$, depending on the speed of your computer. (Practically speaking, set it to search for a couple minutes before giving up.) Optionally, you can also include a command-line flag to adjust the limit (e.g. MAX_ITERS). Be sure to document this in your README file.

It is OK if your program cannot solve some harder problems, e.g. with solution paths of length 10-20. These are hard problems and the search trees become enormous. You will only be able to solve them by developing a sophisticated heuristic. That is the real objective of this project - to make the program more capable by implementing a heuristic that estimates as accurately as possible that true distance from any state to the goal. Thus, the power comes not just from the $A^{*}$ algorithm, but from the intelligence you build-in to the heuristic function, which leads $A^{*}$ more efficiently toward the goal.

When your program finishes, it should print out the solution path (sequence of states between the initial state and goal), and it should print out a line of summary statistics like this (also see Transcript below):
statistics: probB05.bwp method Astar planlen 5 iters 9 maxq 139
'iters' is a count of the number of times through the main loop; 'maxq' is the maximum size of the queue. If it is not able to find a solution before reach MAX_ITERS, set planlen to 'FAILED'.

## File Format for Blocksworld Problems

We will use an input format for each problem, defined as follows. Line 1 indicates the number of stacks $\mathbf{S}$ and number of blocks $\mathbf{B}$ and number of moves $\mathbf{M}$ (if generated, else -1). This is followed by a separator line with 10 greater-than ('>') characters. Then the next $\mathbf{S}$ lines give the configuration of the initial state by listing the stacks horizontally, on their side. Blocks (assumed to be single characters, A-Z) are listed left-to-right in order from bottom to top, with no spaces. Finally, the last S lines give the goal configuration. Then, another separator line, S
more lines for the goal configuration, and a final separator line. Here is an example with 3 stacks and 5 blocks (generated from 10 random moves):

```
### blocks1.bwp ###
3 10
>>>>>>>>>>>
D
CA
BE
>>>>>>>>>>
ABCDE
```

$\ggg \ggg \ggg \gg$

This describes the same problem as shown in the figure on page 1. Notice the two empty stacks in the goal. Do not assume the goal will not always be to have the blocks stacked in order in a single stack - any goal configuration is possible.

On the course website, we will post 2 problem sets: probA03-probA11 (with 3 stacks and 5 blocks), and probB03-probB20 (with 5 stacks and 10 blocks). The 'probA' set are easier, and can be solved even with BFS (or A* using $h_{0}$ ). The 'probB' set are much harder, because the branching factor is higher. You will probably only be able to solve up to probB07 with BFS before the program begins reaching the MAX_ITERS limit and fails. The challenge is to design an intelligent heuristic that allows $\mathrm{A}^{*}$ to efficiently discover solution for as many of these problems as possible.

As part of your program, you will have to read in a problem file given as a command-line argument and use it to create objects representing the initial and goal states. You can assume that input files will always comply with this specification; you don't have to put a lot of errorchecking in your function that reads these input files. Don't worry about checking for extra spaces or characters, or things like incorrect number of lines or duplicate blocks, etc.

## Hints

- The first thing you will have to do is write a simple parser to read the input file (.bwp) with the problem specification, and construct the initial and goal state objects.
- You will probably want to implement a State class, which stores basic information about the state (stacks of blocks), and operations on them.
- You will probably also want to implement a Node class. A Node is like a wrapper around State; it also contains other things like a pointer to the parent from which it the state was generated (which is useful for printing out the solution path at the end when you find the goal node), the depth in the tree, etc.
- An important function to write is successors() which returns a list of all the successors of a given state by any legal move. This generates the branches in the search tree. Don't forget to keep a pointer to the parent Node.
- you will have to implement the GraphSearch version of BestFirstSearch for efficiency, to keep track of visited (or reached) states. One way to do this is to create a unique
string representing any given state as a hash key, and then store them in an unordered_map or dict.
- You don't have to reimplement priority queues from scratch; you can use priority_queue in the STL (for C++ users). (Hint: put Node* pointers in the queue, not Node objects; and don't worry about memory deallocation for this project). You will have to figure out how to get it to sort Nodes on $\mathrm{f}(\mathrm{n})$ by using a comparator function. For python users, you can use any priority-queue package or implementation you want, such as queue. PriorityQueue. Pay attention to direction the queue is sorted; make sure you pop Nodes with the lowest $f(n)$ score first.
- While you are developing your code, it will be helpful to simulate BreadthFirstSearch (BFS) at first, by setting $h(n)$ to 0 , which we can call $h_{0}(\mathrm{n})$ (or ' HO '). Thus $\mathrm{f}(\mathrm{n})=\mathrm{g}(\mathrm{n})=$ depth will be used to sort the priority queue. Later, after you get BFS working on simpler problems (e.g that require only a couple of moves), you can develop your heuristic function to see if you can make it more efficient so it can solve harder problems (e.g. with solution paths of 10-20 steps). Remember that the simplest non-trivial heuristic function is $h_{1}(\mathrm{n})$, which is "number of blocks out of place". So start with that, then try to do better.
- The numbers/statistics in your output do not have to exactly match those shown in my example transcripts. Number of iterations and queue size will be slightly different for everybody, because it depends on a number of implementation details. So don't worry if you can't exactly reproduce my output. But the trends should be similar (e.g. more iterations for harder problems with longer solution paths). Also pay attention to the path length of the solutions; they should be at least close to optimal (as defined by BFS).


## What to Turn In

Create a subdirectory in your (private) Github project for the class called 'PA1'. All programming assignments and homeworks for this course will be checked into the same github project, using different folders.

In PA1/, you should have the following files checked-in:

- README - should explain how your program works (e.g. command-line arguments), and any important parameters or constraints
- If your program takes optional flags on the command line, such as for selecting different heuristics, simulating BFS, or setting an iteration limit, document them.
- If your program has any known limitations, such as that it only works with a certain number of stacks, or number of maximum blocks, or can only solve problems of a certain type or difficulty, document all that.
- blocksworld.cpp - your main program should be called Blocksworld.cpp, but you might also want to have other .cpp or .hpp files, e.g. if you want to split your code into multiple source files based on classes for modularity
- makefile - we must be able to compile your C++ code on compute.cs.tamu.edu by simply calling 'make'
- if you are using python, then the program will be called blocksworld.py
- RESULTS.txt (or RESULTS.docx) - this is a text file
- Describe your heuristic. Explain how it works, or what it computes. Show two examples of states with their $\mathrm{h}(\mathrm{n})$ score to illustrate your heuristic.
- Give a table summarizing the performance of your best heuristic on the test cases provided (probA03-A11, probB03-B20. Report the solution length with your best heuristic, number of iterations (or FAILED), and maximum queue size. See the example 'statistics' shown below.
- Transcripts - Show the output for at least 3 test cases, including the solution path. If the tracing information is too large, you can just truncate it. (name these files like this, for example: probB03.transcript.txt)

The date you commit your files and push them to the Github server will be used to determine whether it is turned in on time, or whether any late penalties will be applied.

When pushing the final version of your PAs, also run the command:

## git tag "paX" \&\& git push origin "paX"

where ' X ' is PA1 or PA2. This will be used to record time of submission for late penalty when applicable. If you want to resubmit, simply delete the old git tag, and then recreate it on your latest commit. <git push --delete origin "paX" \&\& git tag --delete "paX">.

## Grading

The materials you turn in will be graded according to the following criteria:

- $20 \%$ - Does it compile on compute.cs.tamu.edu?
- $20 \%$ - Does it run (on simple test cases)?
- $20 \%$ - Does the implementation of $A^{*}$ (based on priority queue), successor function, heuristic, checking for visited states, etc look correct?
- $20 \%$ - Does the heuristic described in RESULTS.txt look like it is designed well? Does it look like it captures important aspects of how close a state is to being solved?
- $20 \%$ - Performance on test problems (probA03-probA11, probB03-probB20), as summarized in the RESULTS.txt file you submit. Are solution path lengths, number of iterations, max queue size, and other metrics reasonable? How do they scale up with harder problems?

Final reminders:

- Your program MUST have an iterative function that looks like BestFirstSearch, and a heuristic function that, given any node, estimates the remaining distance to the goal.
- Your goal is to develop a heuristic that will allow solving as many of the challenge problems as possible, in the fewest iterations.
- Don't forget to try simulating BFS BreadthFirstSearch first, by setting the heuristic to H 0 .
- You must include a written document (RESULTS.txt or .docx or .pdf) than explains the idea behind your heuristic and summarizes its performance.
- Your program does not have to exactly reproduce the output shown in my examples.


## Example Transcript

```
> cat probB05.bwp
5 10 7
>>>>>>>>>>>
AE
CF
D
GHIJ
B
>>>>>>>>>>
AEG
CFI
DJB
H
>>>>>>>>>>>
> blocksworld probB05.bwp
    (note: you can print out whatever tracing info you want here, or turn it off,
or truncate it, or only print it every 1000 iterations...)
iter=1, depth=0, pathcost=0, heuristic=15, score=0, children=20, Qsize=0
iter=2, depth=1, pathcost=1, heuristic=12, score=13, children=20, Qsize=19
iter=3, depth=2, pathcost=2, heuristic=9, score=11, children=20, Qsize=37
iter=4, depth=3, pathcost=3, heuristic=8, score=11, children=20, Qsize=55
iter=5, depth=3, pathcost=3, heuristic=8, score=11, children=20, Qsize=73
iter=6, depth=3, pathcost=3, heuristic=8, score=11, children=20, Qsize=90
iter=7, depth=3, pathcost=3, heuristic=8, score=11, children=16, Qsize=106
iter=8, depth=4, pathcost=4, heuristic=6, score=10, children=20, Qsize=120
success! iter=9, cost=5, depth=5, max queue size=139
move 0, pathcost=0, heuristic=15, f(n)=g(n)+h(n)=0
AE
CF
```

D
GHIJ
B
>>>>>>>>>>
move 1, pathcost=1, heuristic=12, $f(n)=g(n)+h(n)=13$
AE
CF
DJ
GHI
B
>>>>>>>>>>
move 2, pathcost=2, heuristic=9, $f(n)=g(n)+h(n)=11$
AE
CFI
DJ
GH
B
>>>>>>>>>>
move 3, pathcost=3, heuristic=8, $f(n)=g(n)+h(n)=11$
AE
CFI
DJB
GH
>>>>>>>>>>
move 4, pathcost=4, heuristic=6, $f(n)=g(n)+h(n)=10$
AE
CFI
DJB
G
H
>>>>>>>>>>
move 5, pathcost=5, heuristic=5, $f(n)=g(n)+h(n)=10$
AEG
CFI
DJB

H
>>>>>>>>>>
statistics: probs/probB05.bwp method Astar planlen 5 iter 9 maxq 139

## RESULTS

## using BFS...

```
statistics: probA03.bwp method BFS planlen 3 iter 74 maxq 91
statistics: probA04.bwp method BFS planlen 4 iter }87\mathrm{ maxq 113
statistics: probA05.bwp method BFS planlen 5 iter 256 maxq 260
statistics: probA06.bwp method BFS planlen 6 iter 415 maxq 350
statistics: probA07.bwp method BFS planlen 7 iter 981 maxq 540
statistics: probA08.bwp method BFS planlen 8 iter 1304 maxq 595
statistics: probA09.bwp method BFS planlen 9 iter 1801 maxq 646
statistics: probA10.bwp method BFS planlen 10 iter 2462 maxq 646
statistics: probA11.bwp method BFS planlen 11 iter 2348 maxq 602
statistics: probB03.bwp method BFS planlen 3 iter 1181 maxq 6428
```

```
statistics: probB04.bwp method BFS planlen 4 iter 9700 maxq 50442
statistics: probB05.bwp method BFS planlen 5 iter }71358\mathrm{ maxq }30167
statistics: probB06.bwp method BFS planlen 6 iter }186812\mathrm{ maxq }74662
statistics: probB07.bwp method BFS planlen 7 iter }936119\mathrm{ maxq 3101719
statistics: probB08.bwp method BFS planlen FAILED iter 1000001 maxq 3465665
statistics: probB09.bwp method BFS planlen FAILED iter 1000001 maxq 3559226
...BFS also failed for B10-B20
```


## using $A^{*}$ with my best heuristic...

```
statistics: probA03.bwp method Astar planlen 3 iter 5 maxq 16
statistics: probA04.bwp method Astar planlen 4 iter 7 maxq 19
statistics: probA05.bwp method Astar planlen 5 iter 9 maxq 29
statistics: probA06.bwp method Astar planlen 6 iter 10 maxq 29
statistics: probA07.bwp method Astar planlen 7 iter 11 maxq 34
statistics: probA08.bwp method Astar planlen 8 iter 28 maxq 75
statistics: probA09.bwp method Astar planlen 9 iter 27 maxq 82
statistics: probAl0.bwp method Astar planlen 10 iter 148 maxq 358
statistics: probA11.bwp method Astar planlen 12 iter 22 maxq 67 *
* (note: my solution for probA11 has sub-optimal length, min should be 11)
statistics: probB03.bwp method Astar planlen 3 iter 6 maxq 75
statistics: probB04.bwp method Astar planlen 4 iter 6 maxq 88
statistics: probB05.bwp method Astar planlen 5 iter 9 maxq 139
statistics: probB06.bwp method Astar planlen 6 iter 7 maxq }8
statistics: probB07.bwp method Astar planlen 7 iter 13 maxq 171
statistics: probB08.bwp method Astar planlen 8 iter 18 maxq 249
statistics: probB09.bwp method Astar planlen 9 iter 131 maxq 1929
statistics: probB10.bwp method Astar planlen 10 iter 49 maxq 713
statistics: probB11.bwp method Astar planlen 11 iter 60 maxq 954
statistics: probB12.bwp method Astar planlen 12 iter 61 maxq 996
statistics: probB13.bwp method Astar planlen 13 iter 72 maxq 1025
statistics: probB14.bwp method Astar planlen 14 iter 112 maxq 1725
statistics: probB15.bwp method Astar planlen 15 iter 144 maxq 2303
statistics: probB16.bwp method Astar planlen 16 iter 136 maxq 2106
statistics: probB17.bwp method Astar planlen 17 iter 104 maxq 1649
statistics: probB18.bwp method Astar planlen 18 iter 3641 maxq 50778
statistics: probB19.bwp method Astar planlen 19 iter }7777\mathrm{ maxq 107512
statistics: probB20.bwp method Astar planlen 20 iter 326 maxq 5208
```

