COMS 4995 PARALLEL FUNCTIONAL PROGRAMMING Stephen Edwards

Rubik's Cube Solver using IDA*

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Cube Representation in Haskell


```
initCube :: Int -> Cube
initCube n = Cube {
   up = replicate n (replicate n W'),
    down = replicate n (replicate n 'Y'),
     left = replicate n (replicate n 'O'),
   right = replicate n (replicate n 'R'),
    front = replicate n (replicate n 'G'),
   back = replicate n (replicate n 'B')
```
In our Haskell implementation, a Rubik's Cube is represented by a custom Cube data type with six labeled faces: up, down, left, right, front, and back. Each face is a 2D list of colors represented by characters ('R', 'G', 'B', 'Y', 'O', 'W'). The initCube function initializes each face with a uniform color. This structure allows easy manipulation, display, and transformation of the cube's state.

Move Representation in Haskell

-- Define possible moves

data Move = F | Fi | R | Ri | U | Ui | B | Bi | L | Li | D | Di deriving (Eq. Show)

-- Function to map a Move to its corresponding Cube -> Cube function applyMove :: Move -> Cube -> Cube

```
-- Function to apply a list of Moves sequentially to a Cube
applyMoves :: [Move] -> Cube -> Cube
applyMoves moves cube = foldl (\c{m} -> applyMove m c) cube moves
```

```
-- Perform a move that rotates the front face clockwise
moveF :: Cube -> Cube
moveF cube = cube {
     front = rotateFaceClockwise (front cube),
    up = replaceRow (up cube) (n-1) (reverse (getCol (left cube) (n-1))),
     left = replaceCol (left cube) (n-1) downFirstRow,
    down = replaceRow (down cube) 0 (reverse (getCol (right cube) \emptyset),
    right = replaceCol (right cube) \theta upLastRow
```
where

}

```
 n = length (front cube)
upLastRow = up cube \lfloor (n-1)
downFirstRow = down cube !! @
```
The move logic for manipulating a virtual Rubik's Cube in Haskell involves defining a Move data type for possible moves (e.g., F, Fi, R, Ri, etc.) and implementing functions to rotate faces 90 degrees clockwise or counterclockwise. Each move function, such as moveF, updates the cube's state by manipulating two-dimensional arrays that represent each face of the cube.

Helper functions like replaceRow, replaceCol, and getCol facilitate these updates, ensuring that when a face is rotated, the adjacent faces are also correctly adjusted, maintaining the integrity of the cube's state.

IDA* Algorithm Using Pattern Database

What is IDA*?

- **Iterative Deepening A**^{*} combines the space efficiency of depth-first search with the optimality of A*.
- Searches to a specific threshold $f = depth+h$, then increases *f* iteratively if no solution is found.
- Uses a heuristic function $h(n)$ to guide the search.

Applying IDA* with PDBs to the Rubik's Cube

- 1. **Precompute PDB** for subsets of the cube.
- 2. During search, estimate cost using the PDB heuristic.
- 3. Perform an IDA* search using this heuristic.
- 4. Iteratively deepen the search until the optimal solution is found.
	- a. This is guaranteed as the heuristic is admissible

Pattern Databases (PDBs)

- **Precomputed databases** of optimal distances (minimum number of moves) to the goal for subsets of the cube's pieces.
- Built by performing BFS from the solved state.

Why Use a PDB?

Provides admissible and consistent heuristics: The heuristic never overestimates the cost to the goal. For unseen states, a fallback value is used (e.g. 8 for states not in a PDB with a 7-moves distance limit)

Admissibility guarantees that solutions found by A^*/IDA^* are optimal

Details of PDB Caching

- Color Conversion* Functions convert Rubik's Cube face colors (chars) to Word8 values and vice versa for efficient storage.
- Cube State Representation: cubeToKey converts a Rubik's Cube state into a Word8Vector, a compact vector of Word8 values, for efficient manipulation and comparison.
- PDB Generation: generatePDB creates a pattern database mapping cube states to their distances from the solved state using BFS.
- PDB Storage: The pattern database (PDB) is stored as a Map from Word8Vector to Int, allowing for efficient lookups and insertions of cube states and their distances.
- File Operations: savePDB and loadPDB handle saving and loading the PDB to/from files using binary serialization for efficient storage and retrieval.


```
colorToWord8 :: Color -> Word8
word8ToColor :: Word8 -> Color
cubeToKey :: Cube -> Word8Vector
```
-- Bin.Binary instance: Defines how to serialize and deserialize Word8Vector.

instance Bin.Binary Word8Vector where

 put (Word8Vector vec) = Bin.put (V.toList vec) get = Word8Vector . V.fromList <\$> Bin.get

-- Int representing the distance to the solved state. type PDB = Map.Map Word8Vector Int

```
savePDB :: FilePath -> PDB -> IO ()
savePDB file pdb = Bin.encodeFile file pdb
```
loadPDB :: FilePath -> IO PDB loadPDB file = Bin.decodeFile file

IDA* Algorithm Using Pattern Database Results

./IDAStarBatch pdb_2x2_7.dat 2 scrambles_3000.txt +RTS -N1 -ls -RTS

pdb_2x2_7.dat: The pattern database for the 2x2 cube contains all the states that are within 7 moves of the solved state

scrambles_3000.txt An input file containing 3000 scrambled cubes, each is obtained from performing 30 random moves from the solved state.

Parallelization of IDA*

We attempted to parallelize the solving process of the 2x2 Rubik's Cube using IDA*, but this did not improve efficiency. The main challenges included: -- Shared Visited States

- The speed at which the 2 x 2 is solved with a linear algorithm did not leave much room for improvement
	- \circ A huge percentage of the solution time was loading in the PDB which has to be done sequentially so as per Amdahl's law we didn't see much benefit
- Solving a 3x3 cube is challenging due to the vast state space; our PDB doesn't cover enough states. We will elaborate on this more later in the presentation.
- Maintaining a shared visited set is difficult and leads to contention from frequent read/write operations. This was necessary to avoid threads were repeatedly searching the same nodes/states.

import qualified Data.HashTable.IO as H type VisitedStates = H.BasicHashTable Cube Bool initVisitedStates :: IO (MVar VisitedStates) initVisitedStates = do visitedStates <- H.new :: IO VisitedStates newMVar visitedStates

Parallelizing the solve step on a single cube offers little benefit because the 2-second PDB load time dominates the process. Even with perfect parallelization, the load time remains a sequential bottleneck.

Parallelization by Cube

- When running the IDA* algorithm to solve the 2x2 Rubik's Cube multiple times, we did observe significant speedup by assigning each cube to a different core using Control.Concurrent.Async (forConcurrently)
	- This function is very similar to ParMap but has support for IO operations which made debugging much simpler
- Utilized multiple cores effectively, with each solving a separate instance.
- Diminishing returns caused by thread overhead, memory contention, and limited parallelism.
- Speedup peaked at 8 threads; performance leveled off or declined beyond that.

The 3x3 case: Possible States of the cube

The (solvable) states of the Rubik's cube are determined by:

- 1) Corner arrangement
	- a) *Corner permutations* = Spatial arrangement of the 8 corners, within their 8 available slots
	- b) *Corner orientations* = Whether a corner is twisted correctly, CW, or CCW

2) Edge arrangement

- a) *Edge permutations* = Spatial arrangement of the 12 edges, within their 12 available slots
- b) *Edge orientations* = Whether an edge is flipped or not

Haskell representation:

```
data CubeState = CubeState {<br>edgesPermutation :: [Int],
                                              -- Edge indices (0 to 11) representing their current positions
      edgesOrientation :: [Bool], -- Edge flips (False = correct, True = flipped)
     cornersPermutation :: [Int], - - Corner indices (0 to 7) representing their current positions cornersOrientation :: [Int] - Corner twists (0 = correct, 1 = 120° CW, 2 = 120° CCW)
                                           - Corner twists (0 = correct, 1 = 120° CW, 2 = 120° CCW)
 deriving (Eq, Show)
```
The number of (solvable) states explodes with dimensions

In a 2x2 cube we can only permute and orient the corners, yielding

 $8! \times 3^7$ / 24 $\sim 3,674,160$ (solvable) states

However, in a 3x3 cube we can also permute and orient all the edges, yielding

$8! \times 3^7 \times 12! \times 2^{11}$ / 2 ~ 43,252,003,274,489,856,000

A 4x4 cube – "Rubik's Revenge" would have **~ 10⁴⁵** solvable states…

Size of Rubik's Cube

State Space:

- 2x2 Cube: **3.6 million states**
- 3x3 Cube: **43 quintillion states**

Even with linear IDA* and state pruning techniques, traversing such an enormous number of states is impractical.

Hence, whilst the 2x2 Cube can be "brute forced," it is clear that the 3x3 Cube requires a much more sophisticated algorithm.

Heuristic Issues

- 1. **Pattern Database Limitation**: We couldn't store a full pattern database because the state space is enormous, making it impractical to generate and store all possible configurations.
- 2. **Ineffective Heuristics**: Simpler heuristics (e.g., number of misplaced pieces or Manhattan distance) are ineffective because they **underestimate** the number of moves required and fail to capture the complexity of the cube's state transitions.

Avg solve time of 2 x 2 or linear solution threadscope here

(Kociemba) Two-Phase-Algorithm

1) Solve the orientations

In this first run of IDA*, reach **G1** state – any state that can be generated from the solved state by the orientation-preserving moves **<U,D,R2,L2,F2,B2>**. There are ~ 8! x 12! such G1 states.

Essentially, we want to reach a cube state where all the edge/corner orientations are zero (solved).

2) Solve the permutations

In this second run of IDA*, apply purely **G1** moves – keeping the orientations locked to the solution – to reach the overall solved state.

Complex: requires two separate heuristic functions, stored in several MB's of tables, for each IDA* application, and efficient manipulation of the ever-changing cube state

IDA* : A parallelization attempt

Initialize global search coordinator and search state Initialize current bound with h(root) Initialize worker threads Enqueue initial task (starting node) into the task queue

For each worker thread: – *parallel threads* Fetch a task from the task queue Terminate worker thread if no tasks are left Process the current task:

> If solution found, notify search coordinator If bound exceeded, update candidate bound Enqueue successor tasks (nodes) to the task queue

IDA* : The "main loop"

Read results from worker threads: If a solution is found in search coordinator Terminate workers and return path If a new minimum bound candidate is found: Update new bound if exceeds current bound If all tasks are processed: *– workers cannot proceed within bound* Update current bound Kill worker threads *– we need to start a DFS search anew* Reinitialize root task for new bound Relaunch new worker threads

See the big performance inefficiency here?

IDA* : Sequential vs Parallel

 $==$ Using parallel IDA* implementation $==$

0.17335s F2 L B2 R U' B' L2 D' F L2 U' R2 U L2 B2 L2 B2 D L2 U F2 U' F2 Parallel IDA* is solving the cube...

 $==$ Using parallel IDA* implementation $==$

 $==$ Using parallel IDA* implementation $==$ 0.05048s U2 L2 B2 D B' U' R F' B R' D2 F2 U L2 B2 U' B2 R2 U' F2 L2 D' Parallel IDA* is solving the cube...

 $==$ Using parallel IDA* implementation $==$

 $==$ Using parallel IDA* implementation $==$ 0.05477s LUF'LK2DB'U2DLUB2U'B2R2U'F2D2L2UF2R2D'L2 Parallel IDA* is solving the cube...

 $==$ Using parallel IDA* implementation $==$

=== Using parallel IDA* implementation === $1.54758s$ L U' B R2 U D2 R D' R' F U' L2 D L2 R2 F2 D2 B2 U' R2 F2 U L2 F2 B2 Parallel IDA* is solving the cube...

 $==$ Using sequential IDA* implementation $==$ 0.03704s F2 L B2 R U' B' L2 D' F L2 U' R2 U L2 B2 L2 B2 D L2 U F2 U' F2 Parallel IDA* is solving the cube...

 $==$ Using sequential IDA* implementation $==$

=== Using sequential IDA* implementation === 0.01452s U2 L2 B2 D B' U' R F' B R' D2 F2 U L2 B2 U' B2 R2 U' F2 L2 D' Parallel IDA* is solving the cube...

=== Using sequential IDA* implementation ===

 $==$ Using sequential IDA* implementation $==$ 0.00682s L U F' L R2 D B' U2 D L U B2 U' B2 R2 U' F2 D2 L2 U F2 R2 D' L2 Parallel IDA* is solving the cube...

 $==$ Using sequential IDA* implementation $==$

 $==$ Using sequential IDA* implementation $==$ 0.25002s L U' B R2 U D2 R D' R' F U' L2 D L2 R2 F2 D2 B2 U' R2 F2 U L2 F2 B2 Parallel IDA* is solving the cube...