

## Problem 1:

a)

Describe and illustrate your method of representing the KenKen puzzles, including the board and constraints.

Board:

The KenKen board is represented by a square  $n$ -by- $n$  grid of cells. The grid may contain between 1 and  $n$  boxes (cages) represented by a heavily outlined perimeter. Each cage will contain in superscript: the target digit value for the cage followed by a mathematical operator. Each cell may contain any one of the digits: 1 through  $n$  inclusive. Please refer to Figure 1a for an example KenKen board representation.

<i>a)</i>	<i>b)</i>	<i>c)</i>																											
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3	1	2																											
2	3	1																											
1	2	3																											

Figure 1: Sample KenKen puzzle board of size 3-by-3. [1]

Constraints:

$N := n^2$  “number of cells in grid”

$X = \{0, \dots, N-1\}$

$D = \{1, \dots, n\}$

$C = \{$

[for Row  $r$  in AllRows {AllDiff(AllCellsInRow( $r$ ))}],

[for Column  $c$  in AllColumns {AllDiff(AllCellsInColumn( $c$ ))}]

[for Cage  $a$  in AllCages {TargetNumberForCage( $a$ ) == ApplyCageCageOperator(OperatorForCage( $a$ ), AllCellsInCage( $a$ ))}]

}

It is implied that the *ApplyCageCageOperator* function from above takes into account the non-commutative properties of subtraction and division: subtraction and division operations are performed on only cages of size 2 and only in the order that results in a positive whole number.

Additionally, cages of size 1 and containing no explicit operator are meant to contain only the target value.

Please refer to Figure 1c for a constraint satisfied solution to Figure 1a.

**b)**

*Why would a naïve search approach be bad for this problem and why would a back-tracking approach be good?*

Naïve search approaches ignore the commutativity property of this CSP (and all others CSPs): the order of applying values to each cell does not matter.

Additionally, pieces of the KenKen puzzle can be quickly solved by inference; by performing a slow brute-force naïve search approach, we would be ignoring following properties of this type of CSP:

- 1) Node consistencies of cells given their cage's target value and operator.
- 2) Arc consistencies between cells of known value and their other corresponding row and column cells.
- 3) Path consistencies of cells outside of a cage but in the same row or column of 2 or more cage cells containing a more constrained domain than the exterior cell.
  - a. For example: Cell (1,0) in Figure 1b is constrained to 2 only because the two cells (1,1) and (1,2) can only each contain either 3 or 1, removing the possibility of a 3 or a 1 anywhere else in row 1.

Backtracking is able to find the solution more efficiently by cutting off potentially deep branches that will ultimately lead to failure by checking value assignments with domain consistencies (at the very least).

Backtracking can be made even more efficient by incorporating inference checks and informed variable and value ordering techniques such as the Most Constraining Value (MCV) heuristic.

**Problem 2:**

*a) How many solutions are there to the puzzle show in Figure 2a?*

There is 1 solution, as shown in Figure 2b.

<i>a)</i>	<i>b)</i>								
<table><tr><td>1</td><td>2÷</td></tr><tr><td>2</td><td></td></tr></table>	1	2÷	2		<table><tr><td>1</td><td>2</td></tr><tr><td>2</td><td>1</td></tr></table>	1	2	2	1
1	2÷								
2									
1	2								
2	1								

Figure 2. Sample KenKen puzzle and solution.

*b) & c) Show the complete trace of puzzle boards that were considered by the backtracking procedure, starting from the root node. Point out the steps where a branch of the search tree was abandoned.*

Please see Figure 3 on the next page.

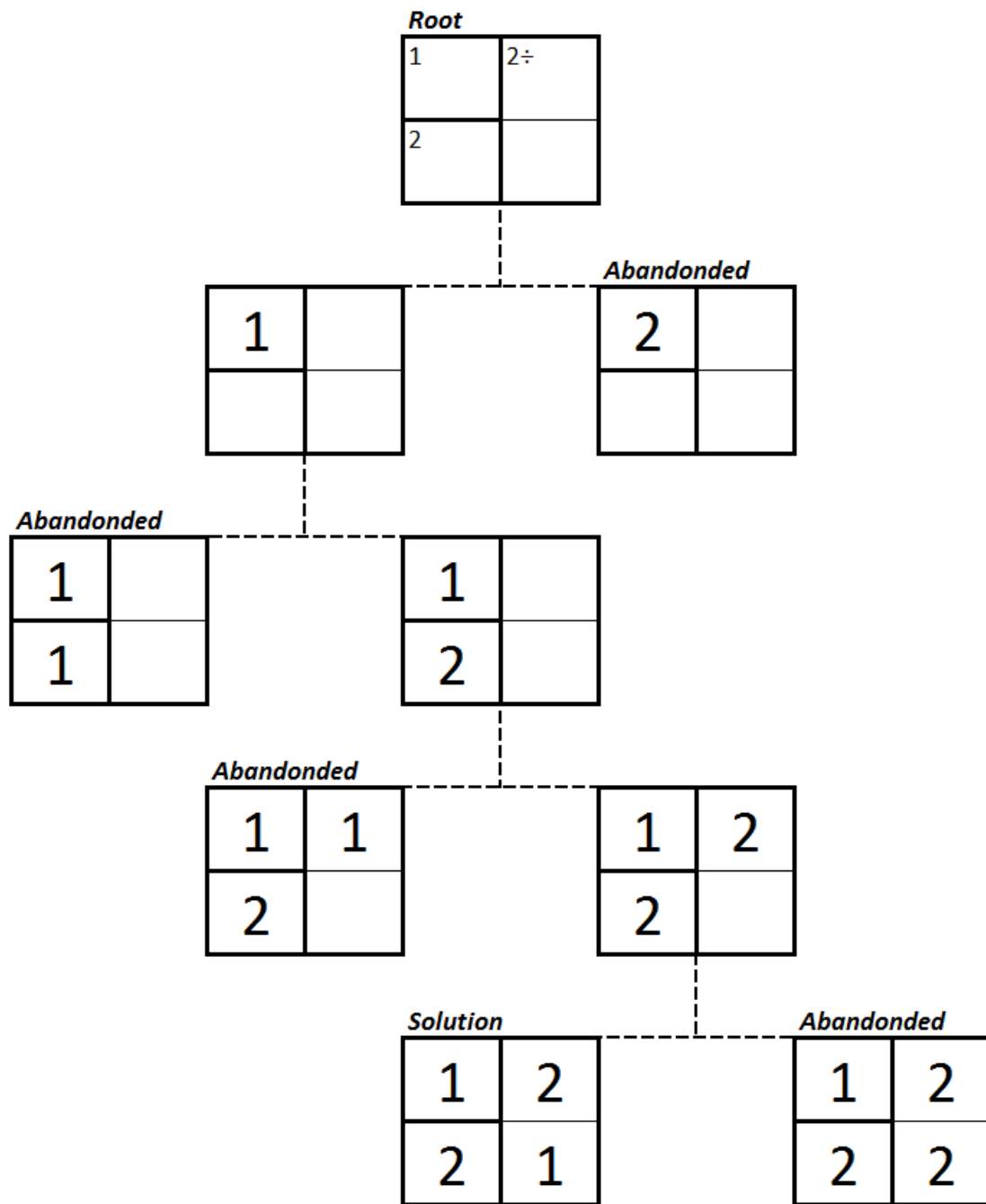


Figure 3. Backtracking procedure KenKen puzzle board trace

**Problem 5:**

*Explain whether the MCV or Least Constraining Value (LCV) heuristics can be useful for KenKen puzzles.*

The MCV heuristic is crucial to an effective backtracking solution to the KenKen CSP as it provides for the mechanism to assign simple very constrained cells (Ex: single-cell-cages) first so as to create more row and column constraints when assigning the less constrained cells. For example, in Figure 1b, assigning the highly constrained cells (0,2) and (2,0) with 2 and 1 respectively cuts the domain for cell (0,0) down from  $\{3,2,1\}$  to  $\{3\}$ , which results in a similarly highly constrained cell which imposes even more constrain other neighboring cells.

If the MCV approach leaves remaining variable assignments the LCV approach can be used to efficiently search for a correct assignment by exploring values from least used in neighboring cells to most. For example, if we wanted to start the assignment process of Figure 1b beginning at cell (2,1), it makes sense to start out with 2 because choosing either 3 or 1 would limit our future assignment options for 3 other cells as opposed to just one for the 2 assignment.

**Extra Credit Question 1:**

*Are there any other domain-specific heuristics that might further limit the number of puzzles considered by backtracking?*

After the assignment of a variable it might be beneficial to apply forward checking inference rules so as to constrain the domains of all other variables in the same row and column as the one that was just assigned.

### Extra Credit Question 2:

I implemented both the Most Constraining Variable and Least Constraining Value heuristics from the book. I found that for input\_example3.txt, use of the two heuristics resulted in a decrease of about 50 variable assignments with respect to the run without any heuristics, before the 1<sup>st</sup> solution (I) was found, as shown in Figure 4. However, the reciprocal of both heuristics (*Least* Constraining Variable and *Most* Constraining Value) resulted in a significant increase in variable assignments before the first solution was found (See LCVar/MCVal in Figure 4).

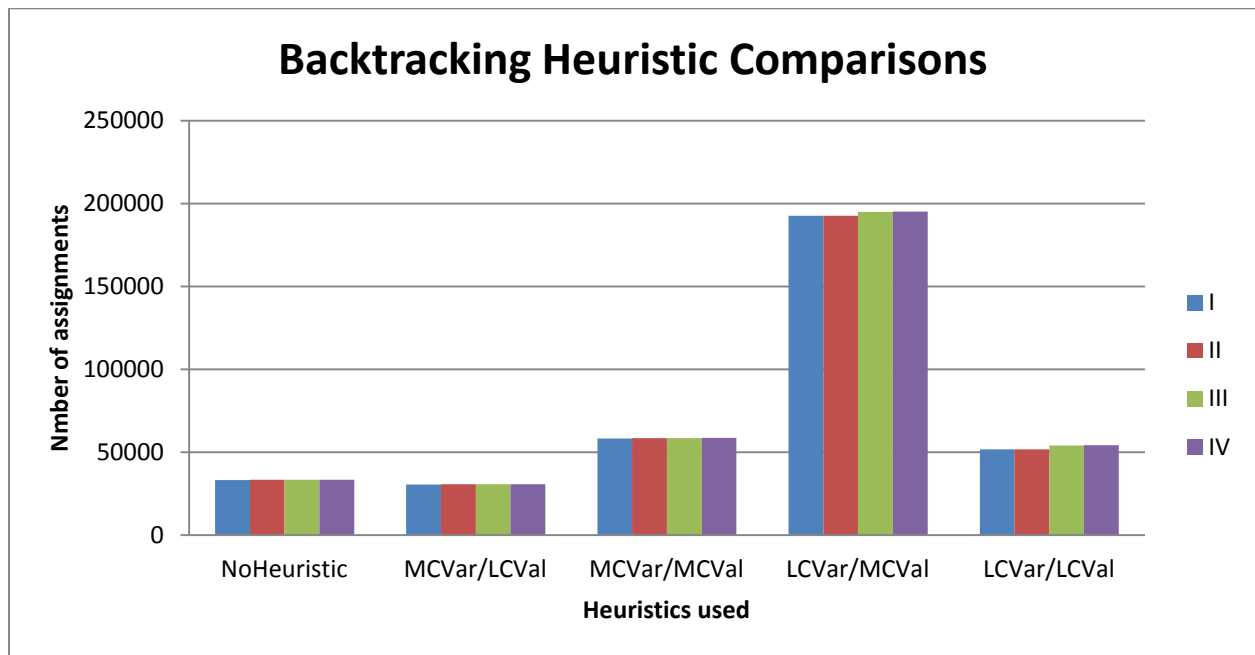


Figure 4. Number of variable assignments before each solution (I, II, III, IV) was found given each heuristic combination.

The same experiment was performed on input file: input\_q3.txt as shown in Figure 5. The MCVar/LCVal heuristic combination resulted in fewer assignments required to find the first 4 solutions than the no-heuristic test. However, the MCVar/LCVal heuristic combination performed worse than the no-heuristic when finding the last 4 solutions.

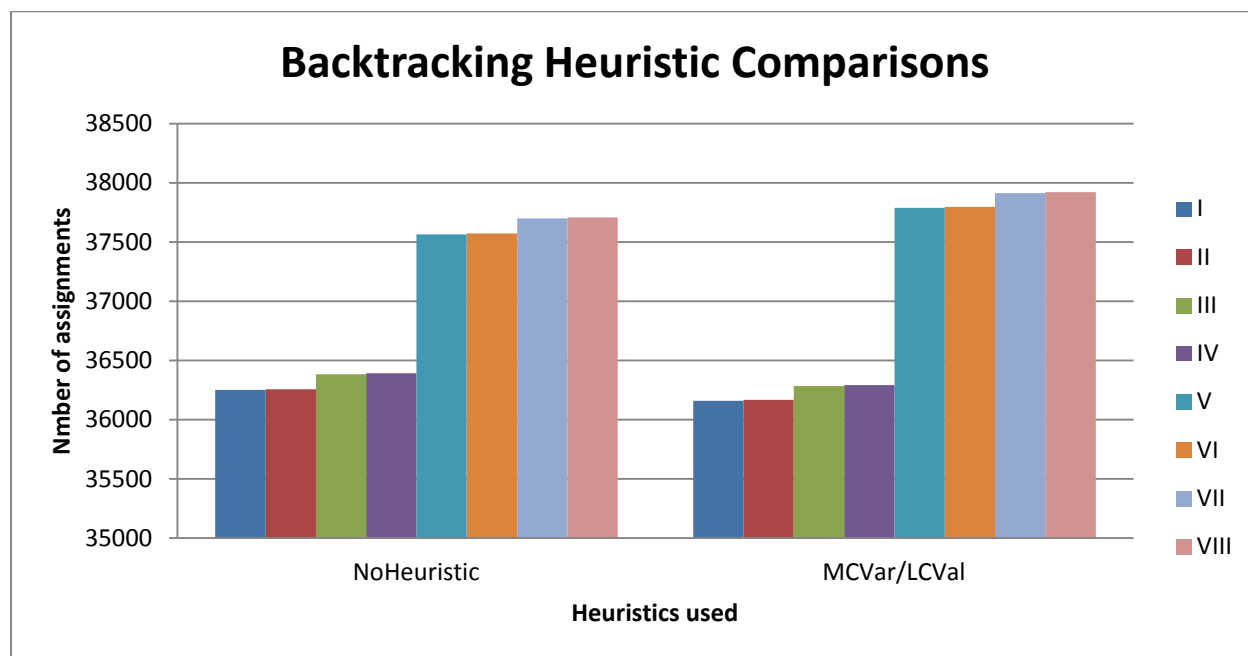


Figure 5. Number of variable assignments before each solution for input\_q3.txt was found, given each heuristic combination.

## References:

- [1] [http://www.kenken.com/2011-2012\\_TeacherPuzzles/KK\\_08282011\\_TeacherSet.pdf](http://www.kenken.com/2011-2012_TeacherPuzzles/KK_08282011_TeacherSet.pdf)