

Artificial Intelligence



American
University
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Minesweeper

Team:

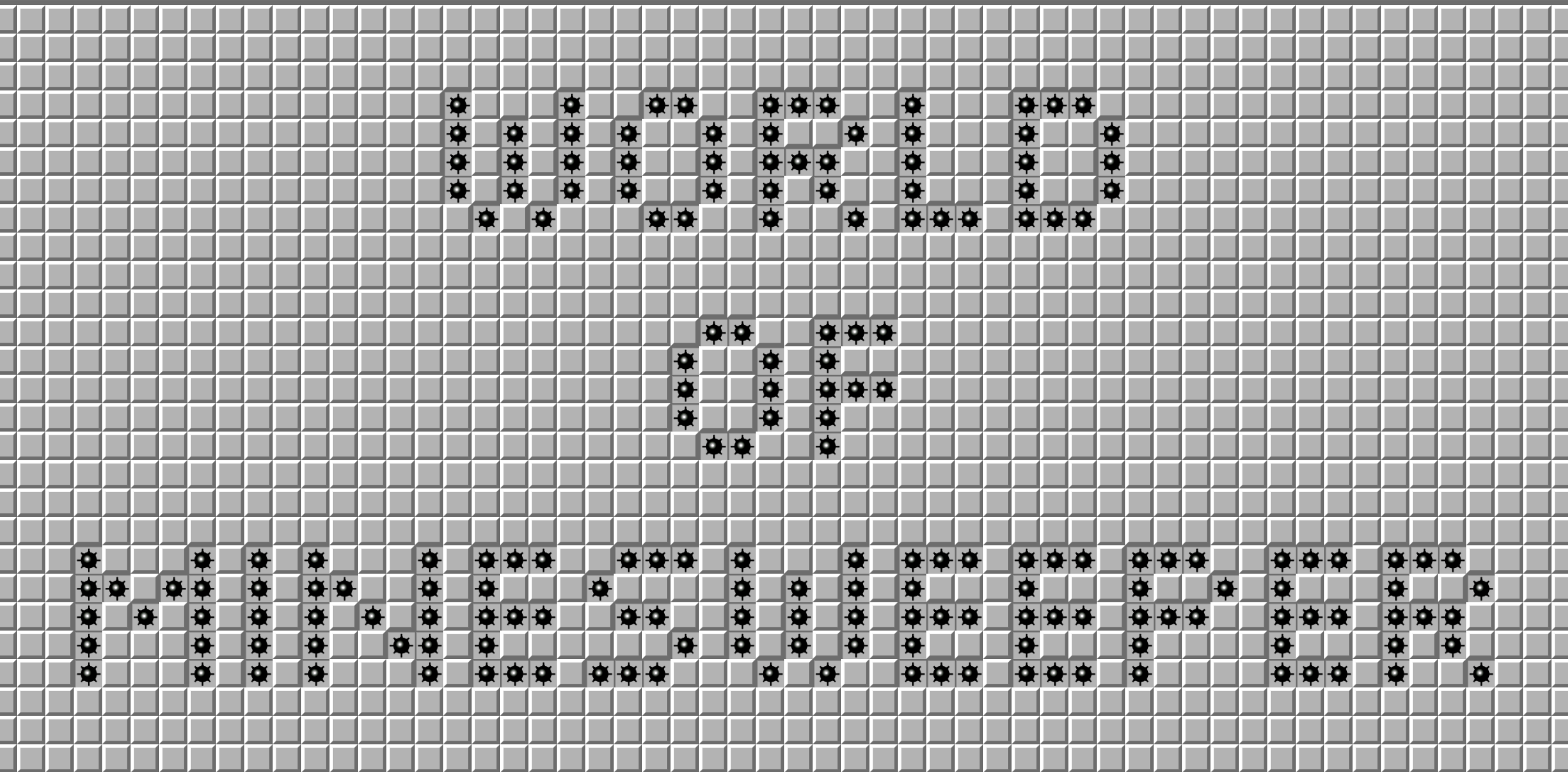
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LITERATURE REVIEW

Existing approaches to Minesweeper game

- RANDOM SEARCH (RS)
- HILL CLIMBING (HC)
- SIMULATED ANNEALING (SA)
- GENETIC ALGORITHM (GA)

Comparison of Algorithms

RANDOM SEARCH	<p>Randomly generates successor configurations and picks the best of them.</p> <p>Brings Minesweeper into an Optimization problem with a penalty function (sum of the squared differences between opened mines and the indicators).</p>	<ul style="list-style-type: none">• Outputs a good solution locally but not for the whole board.• When encounters a bad configuration, discards it and starts a new iteration(missing the potential of that bad move to yield a good solution for the whole board).
HILL CLIMBING	<p>1st step-randomly generates a solution.</p> <p>Makes small local changes with the hope of optimizing the penalty function.</p>	<ul style="list-style-type: none">• uses a penalty function• randomly places mines to generate new states and picks whatever is the best move locally• fast compared to RS
SIMULATED ANNEALING	<p>Starts with a random configuration, maintains temperature t, assigns probability values to states $-c/t$.</p>	<ul style="list-style-type: none">• keeps potentially good moves that seem to be bad locally• when t is high, similar to RS• when t is lower, similar to HC
GENETIC ALGORITHM	<p>Starts with a population of potential solutions, assigns fitness value(inverse of penalty)</p> <p>for two fittest configurations A and B,takes one part of A, and then another part of B, plus a potential mutation = new successor</p>	<ul style="list-style-type: none">• uses random mutations and partitions• penalty function decreases as it moves always to fitter states

Constraint Satisfaction Problem

CSP is a factored representation of a mathematical problem, consisting of states whose solutions must satisfy a set of constraints

CSP consists of :

- set of variables
- set of domains
- set of constraints that specify allowable combinations of values

Minesweeper as a CSP

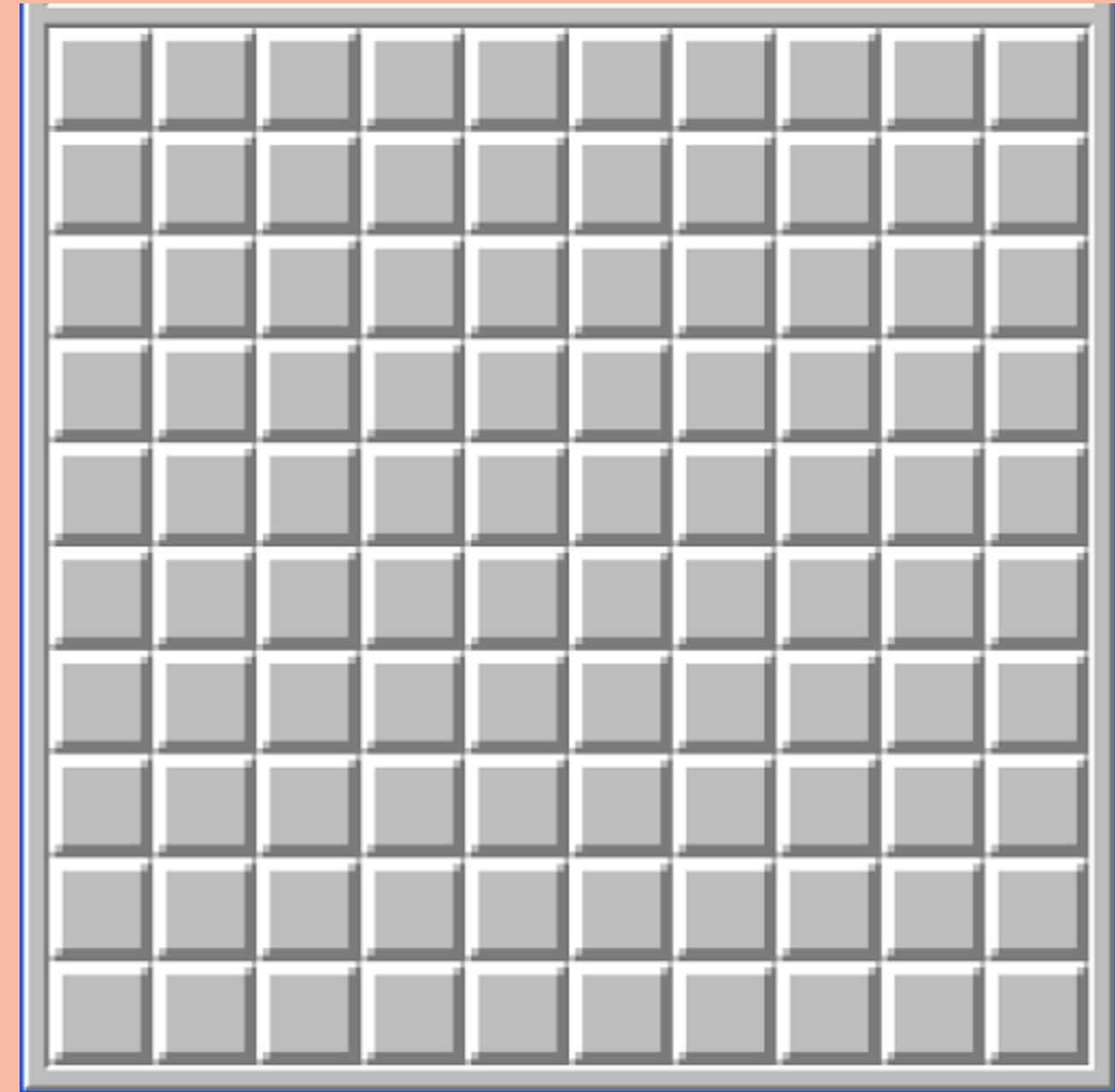
- the variables are squares of our 2D board
 - the values of the domains are -1(mine) and 0(safe)
 - each cell is in a constraint with its neighbours
- such that the sum of abs. values of the neighbors is equal to the indicator assigned to that constraint

OUR APPROACH TO MINESWEEPER



Algorithm

- **board** is **10 x 10** with **10** randomly placed mines
- **variables**- 100 squares
- **Domains** are **$\{-1, 0\}$** : "mine" = -1, "safe" = 0
- each square has an **indicator**(from **$\{0, \dots, 8\}$**)
- multiple components
- possible to encounter a mine at the first click
- Start at (1,1) (if indicator $\neq 0 \Rightarrow$ click(n,n), again
indicat. $\neq 0 \Rightarrow$ uncover corners of the board
 - **Component 1**(till 12th iter.)- uncover corners when not having insight about "safe" squares
 - **Component 2**- if indicat. $= 0 \Rightarrow$ uncover all neighbors
 - **Component 3**- if indicat. $= \#$ of covered neighb. \Rightarrow Flag them

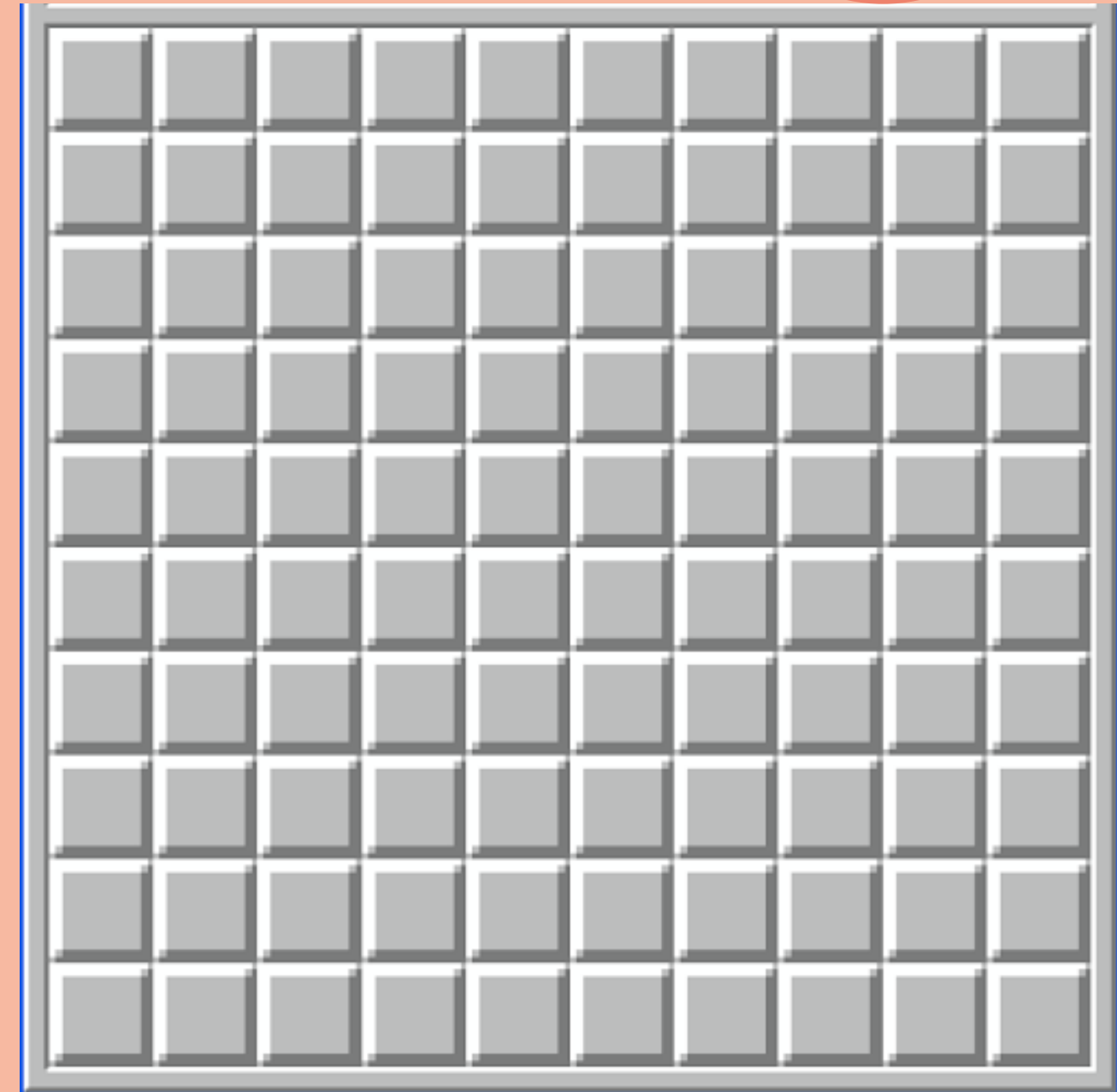


OUR APPROACH TO MINESWEEPER



Algorithm

- **Component 4** (for optimal and fast search)-
find supersets-subsets among the constraints
(of squares with Manh. dist. at most 2),
new constr' = superset-subset,
 $\text{indic.}(\text{superset}') = \text{indic}(\text{superset}) - \text{indic}(\text{subset})$
remove subset from superset
- **Component 5** Exhaustive search of the
remaining constraints. All possible universes
(permutations) are accounted for, the square with
least # of universes where it contains a mine is
selected



NOW SHOWING

Demonstration of the game:
<https://youtu.be/MN8mYPZc9Wg>

Time Complexity

$O(2^r)$ because of the exhaustive search, where r is the #of covered squares left

$O(n^4)$ where n is the length of the board

Accuracy

10/10 generated game instances were solved correctly

Reference list

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**Thank
you!**