

# Minesweeper AI

Using Logic Inference, CSP and CNN Approaches

Chenyang Zhang, Chunxu Guo, Jiahao Huang, Qianyu Liu, Tianyi Zhang

Team 4

# Problem Statement

## Minesweeper AI

- **Two perspectives**
  - Deterministic: know where is a mine for sure—Inference
  - Non-deterministic: need to guess where is a mine—Probability
- **NP-Complete problem**
- **Goal**
  - Higher winning rate
  - Computational efficiency



# Related Works

# State of the Art

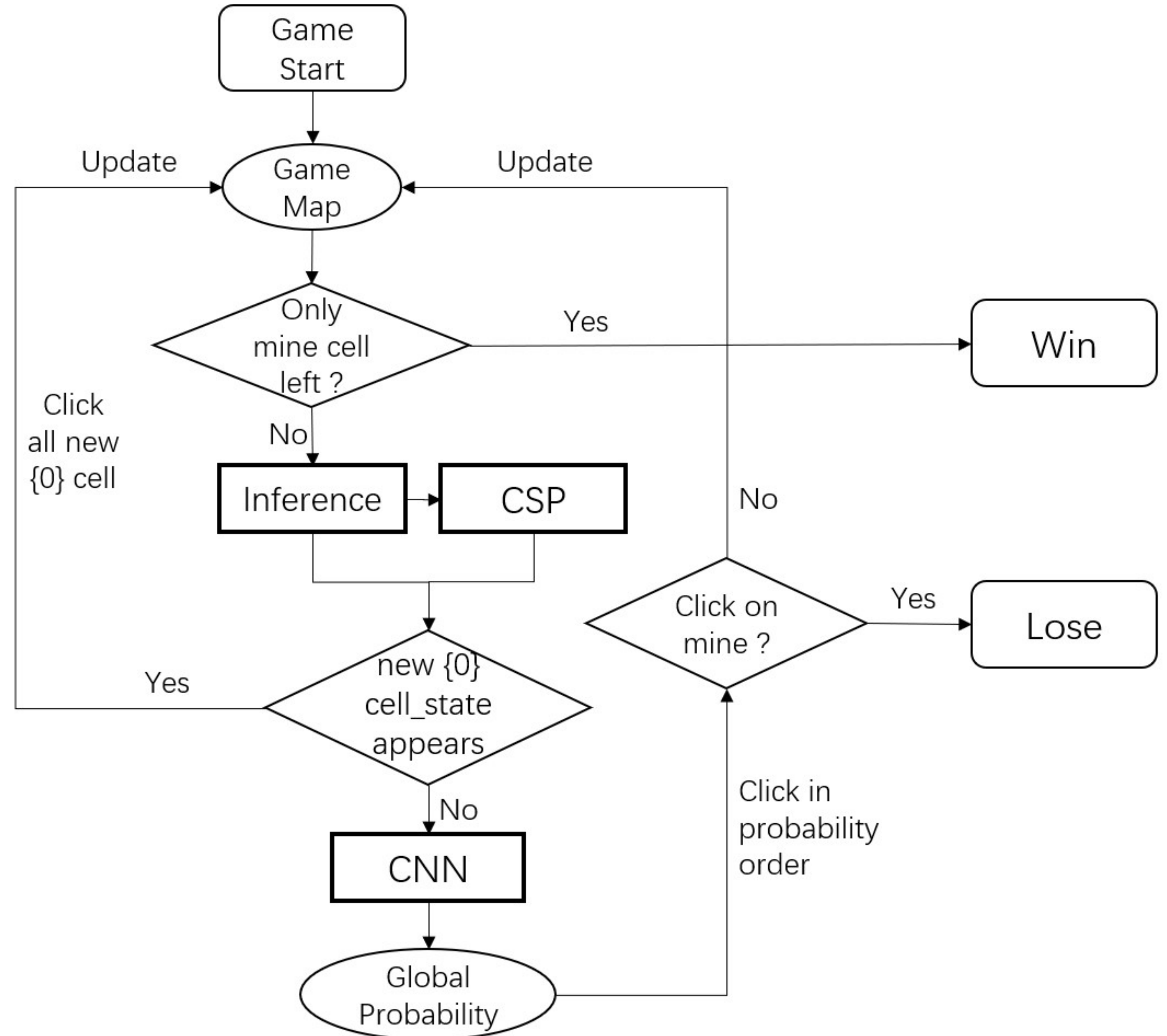
## Related Works

- **Logic:** SAT solver
  - Inference process is slow
- **Search:** CSP
- **RL:** Q Learning
  - Hard to define the state: too many states if use the whole map
  - Poor performance
- **Neural Network:** CNN

# Approach

# Pipeline Approach

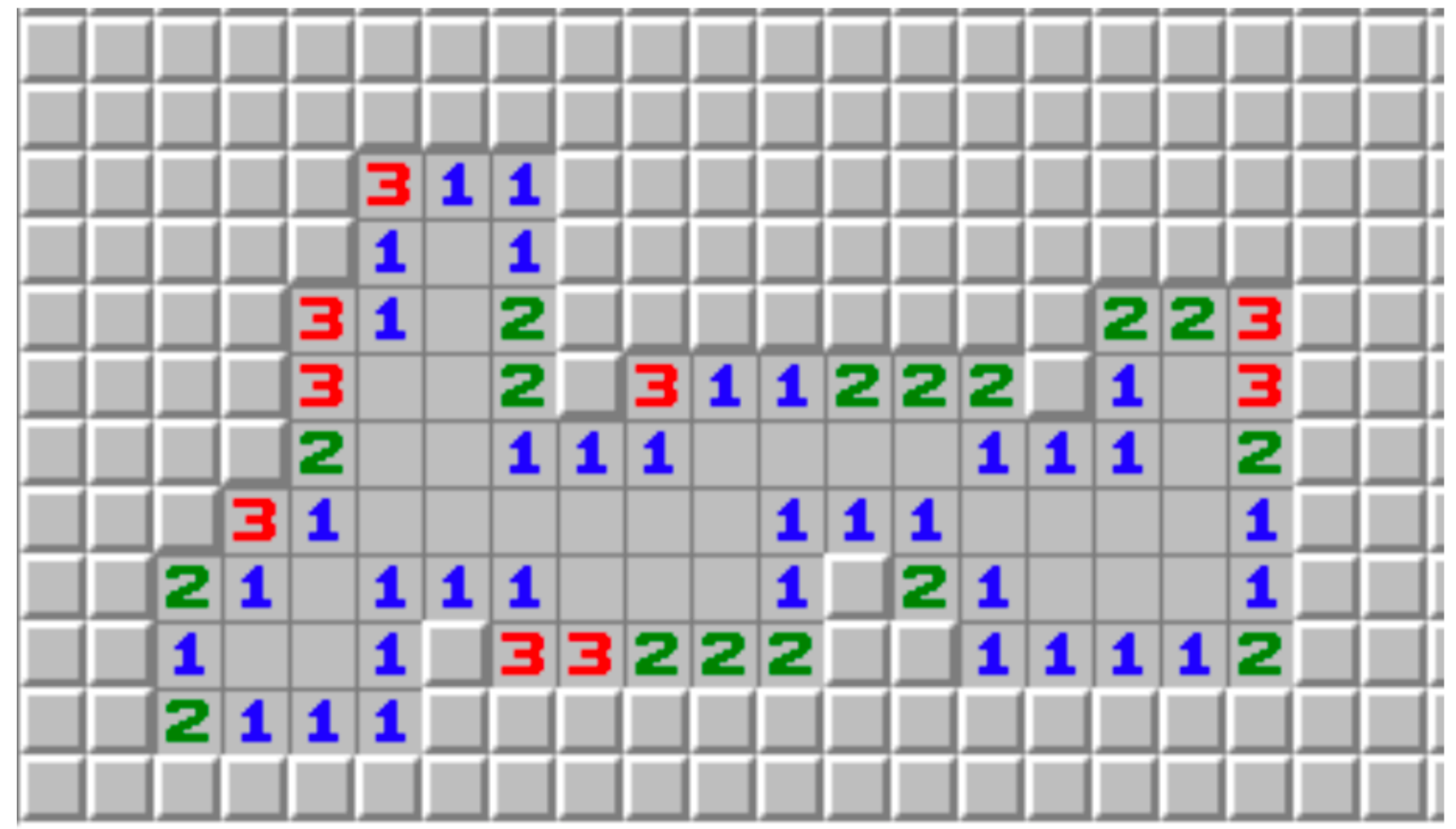
- **Logic Inference**
  - Fast & Accurate
- **CSP**
  - Complex Inference
- **CNN**
  - Global Probability



# Stage 1: Logic Inference


## Approach

- **Goal:** Fast and Accuracy check for deterministic case
- **Define 4 rules**
  - Flag
  - Revealing
  - Elimination
  - Confirmation



# Approach



- **Goal:** Fast and Accuracy check for deterministic case
  - **Define 4 rules**
    - Flag
    - Revealing
    - Elimination
    - Confirmation
- 
- |  |  |   |   |   |   |   |   |   |  |
|--|--|---|---|---|---|---|---|---|--|
|  |  |   |   |   |   |   |   |   |  |
|  |  |   |   |   |   |   |   |   |  |
|  |  |   |   |   | 3 | 1 | 1 |   |  |
|  |  |   |   | 1 | 1 |   | 1 |   |  |
|  |  |   | 1 | 3 | 1 |   | 2 | 1 |  |
|  |  |   | 1 | 3 |   |   | 2 | 1 |  |
|  |  | 1 | 1 | 2 |   |   | 1 | 1 |  |
|  |  | 1 | 3 | 1 |   |   |   |   |  |
|  |  | 2 | 1 |   | 1 | 1 | 1 |   |  |





# Stage 2: CSP

## Approach

- **Goal:** find hidden deterministic cells
- **Method:**
  - Define *variables* and *constraints*
    - *variables*: for every cell,  $v = \{0\}$  (safe) or  $\{1\}$  (mine) or  $\{0, 1\}$  (both possible)
    - *constraints*: for every number,  $\text{number} = \text{SUM}(\text{neighboring true values})$
  - Basic solution: backtracking search
  - All variables start from state  $\{0, 1\}$ , and any state changes means new deterministic information



# Stage 3: CNN

## Approach

- **Goal:** handle non-deterministic case—Probability
- **Inspiration:**
  - 3x3 grid is of great importance in Minesweeper
  - Combining many convolutional layers with 3x3 kernels gives larger receptive field
  - We may find more global information using CNN

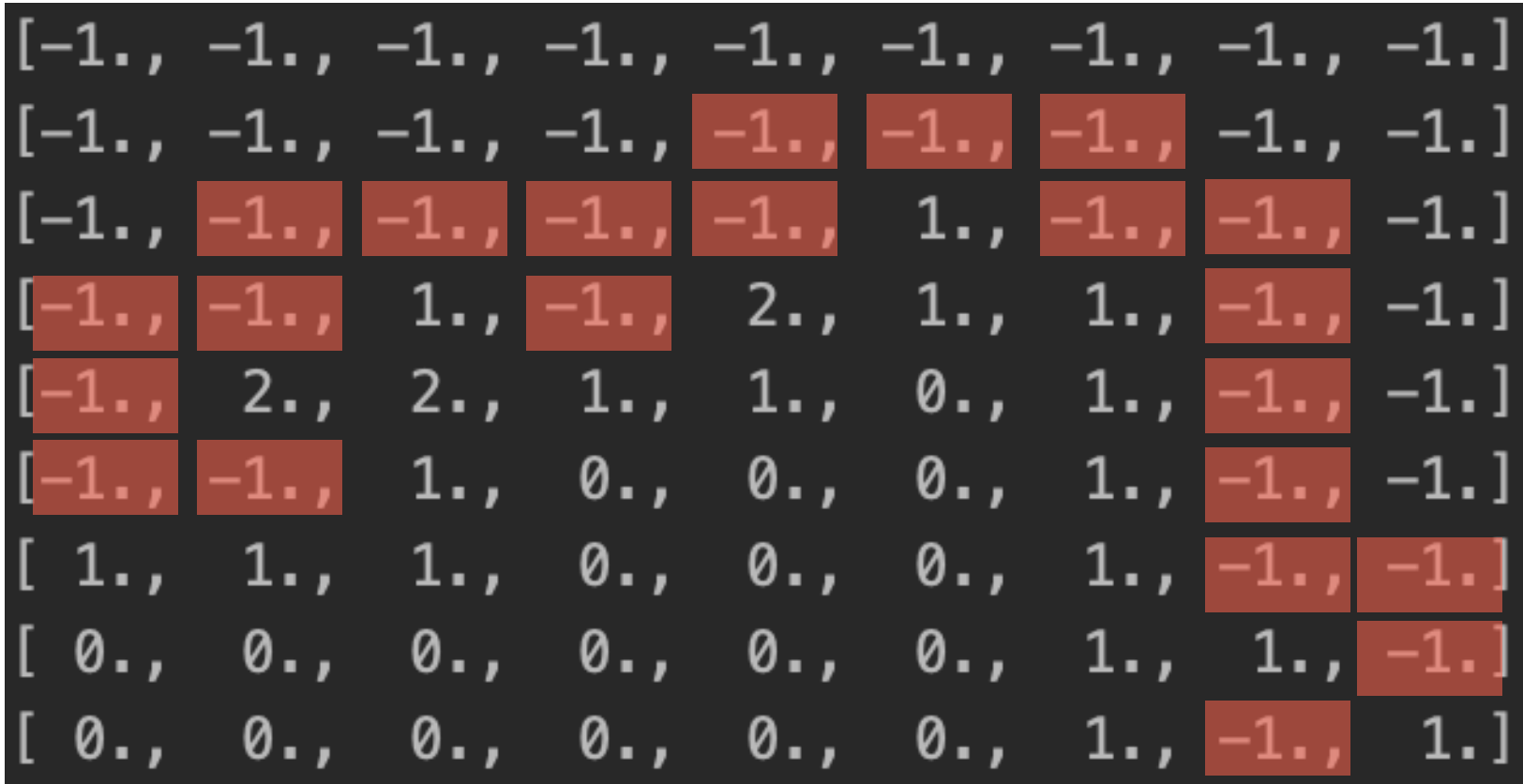
# Stage 3: CNN

## Approach

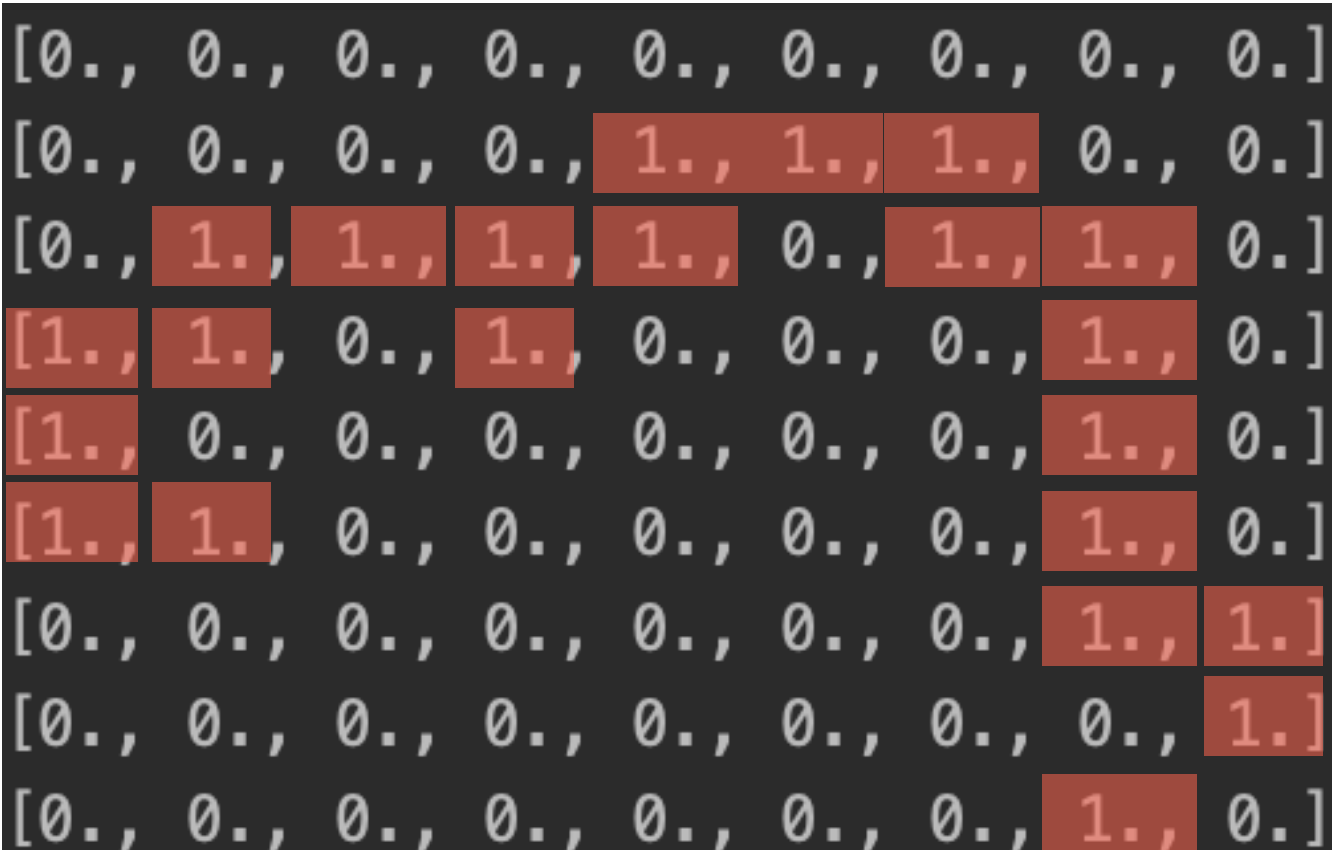
- **Net Architecture:**
  - 5 convolutional layers with 3x3 kernels
  - 3 fully connected layers
- **Training:**
  - Input: game state map, Output: mine probability map
  - Easy to train—not resource-consuming
  - Loss: only focus on the ‘edge’

# Stage 3: CNN

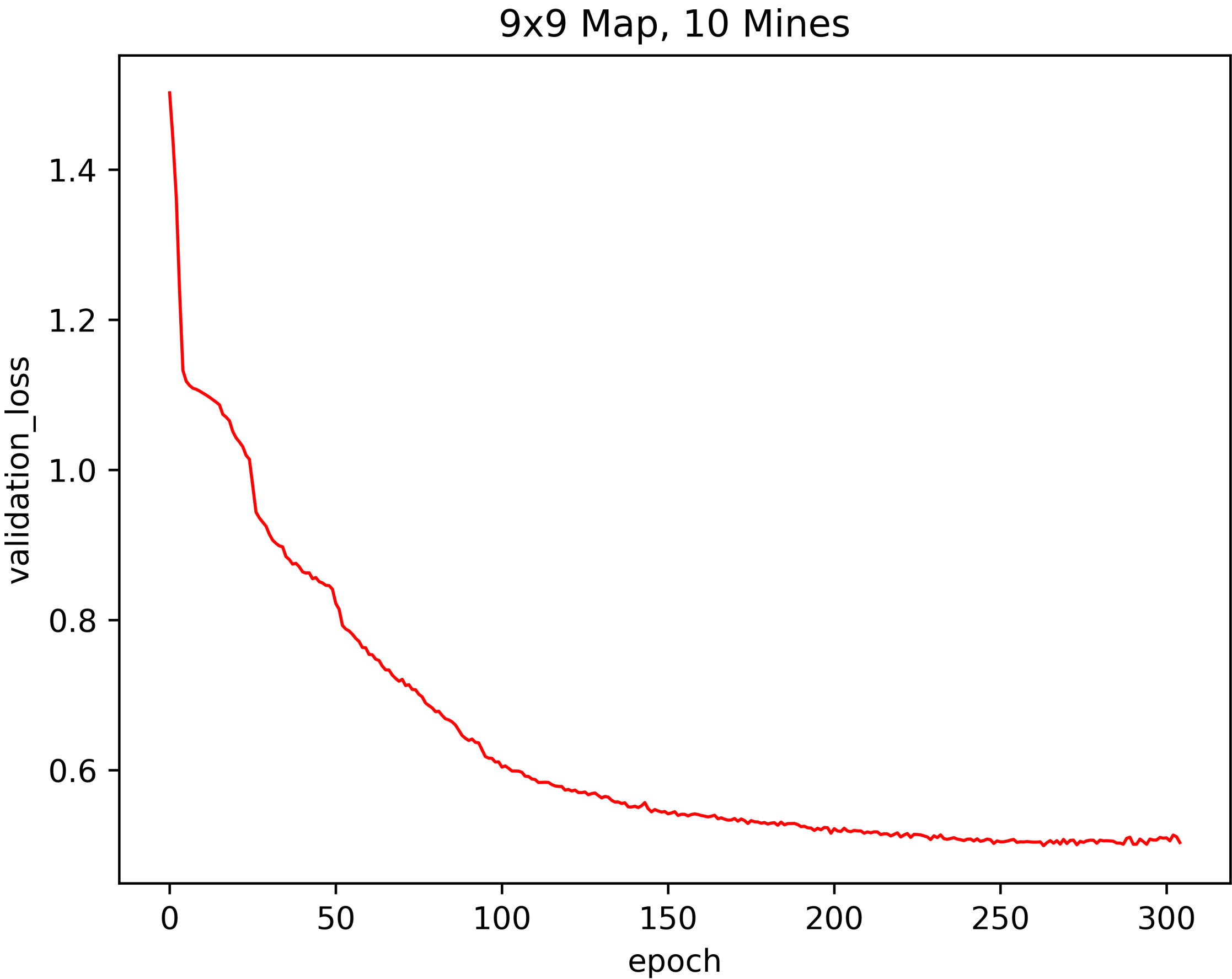
## Approach



Game State Map

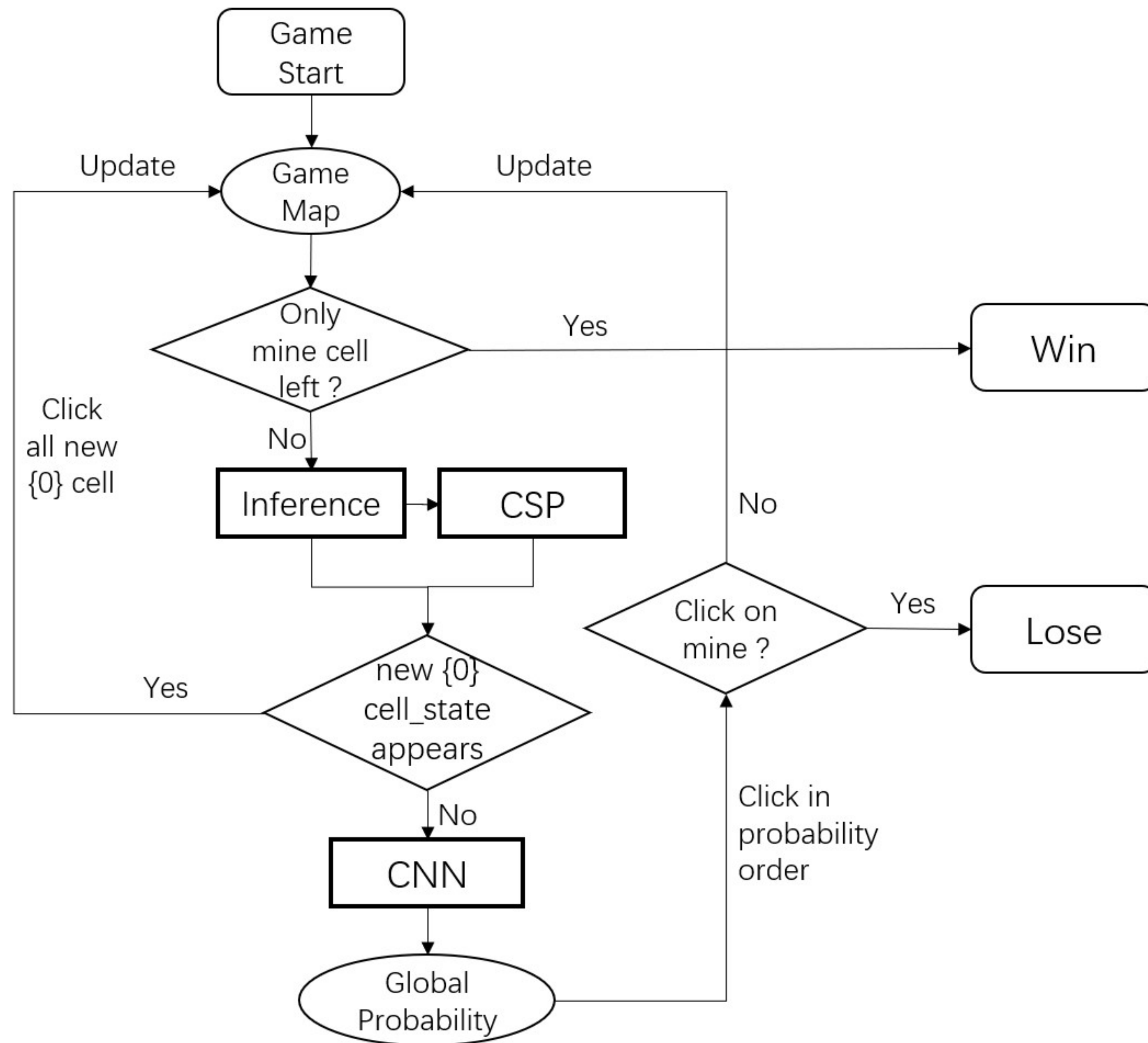


Edge Map



# Pipeline Approach

- **Logic Inference**
  - Fast & Accurate
- **CSP**
  - Complex Inference
- **CNN**
  - Global Probability



# Experiments and Results

# Winning Rate Improvement

## Experiments and Results

Approach	9x9 Map, 10 Mines	16x16 Map, 40 Mines
Random pick	0%	0%
Logic	70.7%	31.4%
CSP	45%	1%
Net	32.8%	0%
Logic + CSP	84.5%	48.7%
CSP + Net	77%	7%
Logic + CSP + Net	95%	66.1%

Testing: Play 1000 rounds random generated game.

# Conclusion



# Conclusion

## Minesweeper AI

- **Contribution:**
  - Combining deterministic (logic inference, CSP) and non-deterministic (CNN) models
  - Good performance & Computational efficiency
- **Future Works:**
  - Do filtering optimization (Forward Checking, Arc Consistency) with CSP
  - Expert mode (16x30, 100 mines) optimization