

Training a Minesweeper Solver

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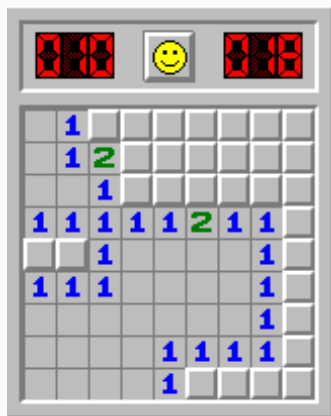
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Introduction

WHAT IS MINESWEEPER ??



- **Minesweeper**, a puzzle game introduced in 1960's requires spatial awareness and ability to work with incomplete information. Utilizing different Machine Learning approaches, we implemented solvers that makes use of **Reinforcement Learning**.
- A modified Q-learning algorithm was enhanced by function approximation, which was able to effectively generalize learning of the state space.

- Playing a game of **Minesweeper** involves uncovering tiles until the player uncovers a tile containing a mine or uncovers all of the tiles that don't contain mines.
- As the game progresses, the player is given limited information regarding the location of the mines on the board.
- Using this information the Minesweeper solver should determine which action to take at each stage of the game.

Goal

The solvers of **Minesweeper** are implemented using three algorithms :

- Simplified Q-learning
- Modified Q-learning
- Deep Q-learning

Q-Learning

- At each step s , choose the action a which maximizes the function $Q(s, a)$
- Q is the estimated utility function – it tells us how good an action is given a certain state
- $Q(s, a)$ = Immediate reward for making an action + best utility (Q) for the resulting state

- Here modified version of Q-learning is used to discover the best actions for each given board configuration.
- As we are more interested in the immediate reward rather than the end game result, we have not taken into consideration the final max optimized value.
- We estimate which tile is least likely to have a mine in a given board configuration by finding the tile with the highest Q value

Q-learning :

$$Q(s, a) = r(s, a) + \gamma \max_{a'} (Q(s', a'))$$

γ = Relative value of delayed vs.immediate rewards (0to1)

$r(s,a)$ = Immediate Reward

s' = The new state after action a

a, a' : Actions

s, s' : States

Selected action:

$$\pi(s) = \operatorname{argmax}_a (Q(s, a))$$

The original **Bellman Equation** allows the algorithm to learn not just about the direct reward of the particular action but whether the particular action is more likely to lead to reward in the long-term.

But in **Minesweeper** we are interested in the immediate reward, whether a particular move will uncover a mine on a specific board configuration.

$$Q(s, a) = r(s, a) + Q(s, a)$$

Q-Learning Algorithm

Begin probing by selecting a corner

While not game over **do**

$S \leftarrow$ current state of the board

 Array \leftarrow all tiles on frontier

For tile in Array **do**

$P(s,a) \leftarrow P(s,a)+1$

End for

 Probe *random square* in Array

End while

Modified Q-Learning Algorithm

Begin probing by selecting a corner

While not game over **do**

$S \leftarrow$ current state of the board

 Array \leftarrow all tiles on frontier

For tile in Array **do**

$P(s,a) \leftarrow P(s,a)+1$

End for

 Probe *square* in Array with the least probability of being a mine

End while

Deep Q-learning

Deep Q-Learning

Begin probing by selecting a corner

While not game over **do**

 S <- current state of the board

 Array <- all tiles on frontier

 Choose random tile t from frontier

 Append (S,t) in *experience buffer*

If experience buffer is full **do**

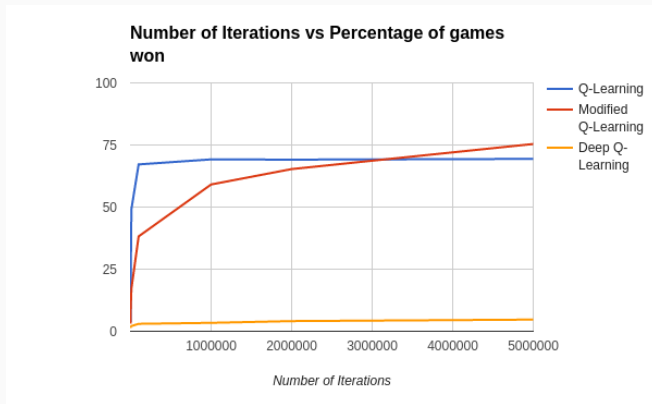
Train Network on randomly chosen samples from buffer

End If

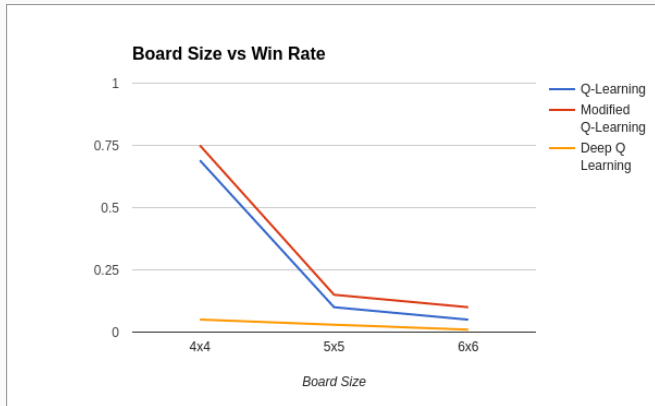
End while

Analysis

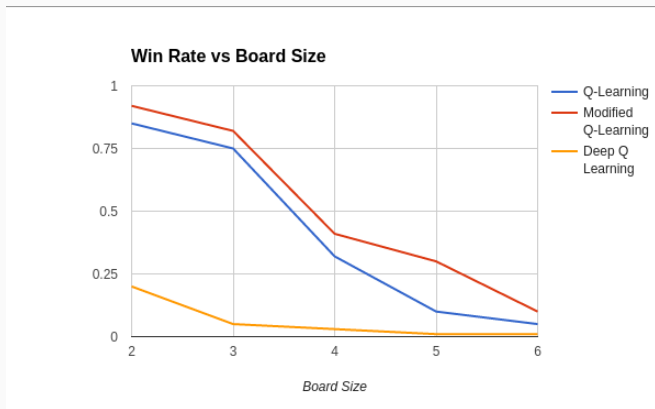
COMPARISONAL ANALYSIS



COMPARISONAL ANALYSIS



COMPARISONAL ANALYSIS



Conclusion

Conclusion




- While Q-learning is good for board sizes, it's performance decreases drastically for larger boards.
- To solve the above problem, Deep Q-learning has been implemented.
- The version of Deep Q-learning implemented is not giving satisfactory results.

Future Works

Considering a fixed state space for the Neural Network.

DEMO

References

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-  **Reinforcement Learning.**
<http://reinforcementlearning.ai-depot.com/>