

# Waterfall Gridworld

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Machine and Reinforcement Learning in Control Applications

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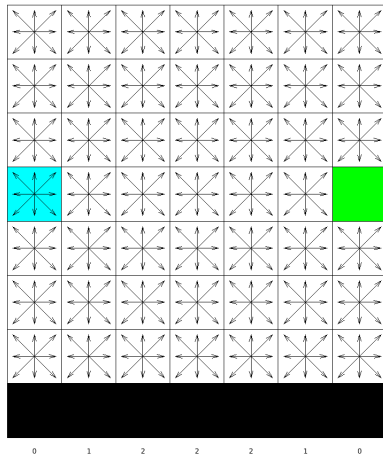
# Problem



Learn to move in an unknown map with external disturbances

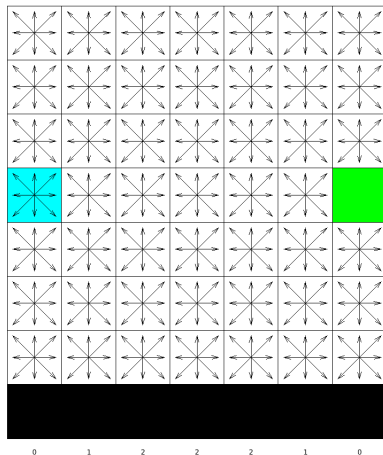
# Problem formulation

- Consider the grid world on the right;
- The goal is to reach the green box;
- A waterfall pushes the agent toward the bottom of the grid



# Problem formulation

- 8 possible directions:  
**N, S, E, W, NE, NW, SE, SW;**
- Reward:
  - $-1$  for each step



# Model

- The **state** is the position in the Gridworld
  - we have  $X \cdot Y$  states.
- The **action** is the direction of the movement
  - we have 8 actions.

# TD( $\lambda$ )

TD(0)



# TD( $\lambda$ )

TD(0)



MC / TD(1)



# TD( $\lambda$ )

TD(0)



TD( $\lambda$ )



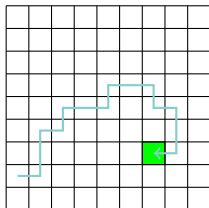
MC / TD(1)



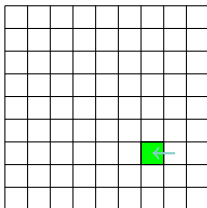


# Advantages of $TD(\lambda)$

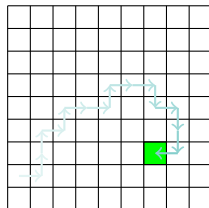
Path taken



Learnt from  
 $TD(0)$



Learnt from  
 $TD(\lambda)$



# Eligibility traces

- The eligibility traces keep memory of the visited states;
- At each step, the eligibility trace of visited state is updated:

- **Accumulating traces:**

$$E_{t+1}(s, a) = E_t(s, a) + 1$$

- **Replacing traces:**

$$E_{t+1}(s, a) = 1$$

- **Dutch traces:**

$$E_{t+1}(s, a) = (1 - \alpha)E_t(s, a) + 1$$

- At each step, the eligibility trace of non-visited states decays

$$E_{t+1}(s, a) = \gamma \lambda E_t(s, a)$$

# Estimates update

- Apply the TD( $\lambda$ ) prediction method to state–action pairs.
- The TD error for state-value prediction is

$$\delta_t = R_{t+1} + \gamma Q_t(S_{t+1}, A_{t+1}) - Q_t(S_t, A_t)$$

- The updates are

$$Q_{t+1}(s, a) = Q_t(s, a) + \alpha \delta_t E_t(s, a), \quad \forall s, a$$