

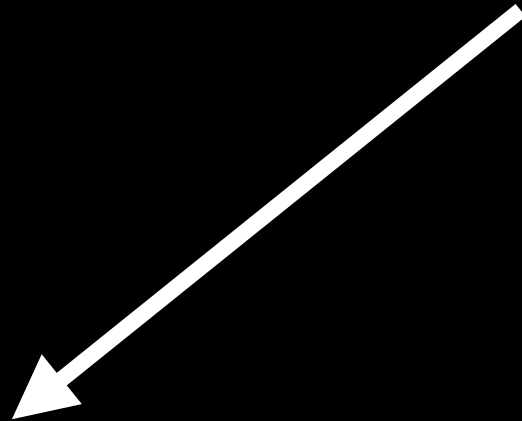
Reinforcement Learning

Kushal Shah @ Sitare University

Machine Learning

Machine Learning

Machine Learning



Supervised Learning



Unsupervised Learning

Machine Learning



```
graph TD; ML[Machine Learning] --> SL[Supervised Learning]; ML --> UL[Unsupervised Learning];
```

Supervised Learning

$$Y = f(X)$$

Unsupervised Learning

Machine Learning

```
graph TD; ML[Machine Learning] --> SL[Supervised Learning]; ML --> UL[Unsupervised Learning];
```

Supervised Learning

$$Y = f(X)$$

Handwriting Recognition

Face Detection

Speech Recognition

Unsupervised Learning

Machine Learning

Supervised Learning

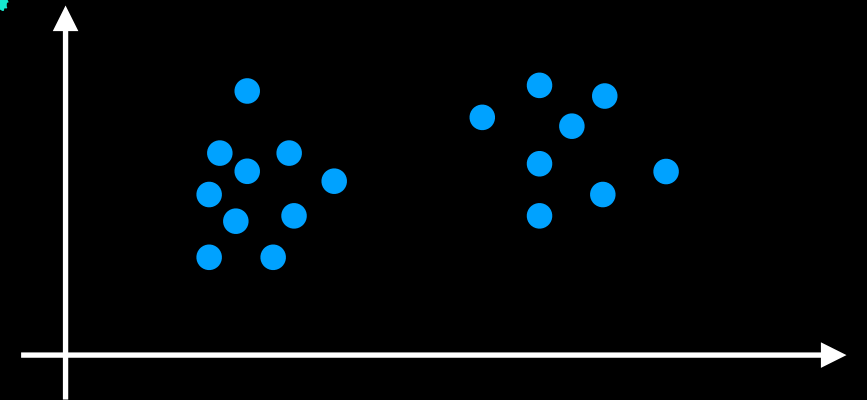
$$Y = f(X)$$

Handwriting Recognition

Face Detection

Speech Recognition

Unsupervised Learning



Machine Learning

Supervised Learning

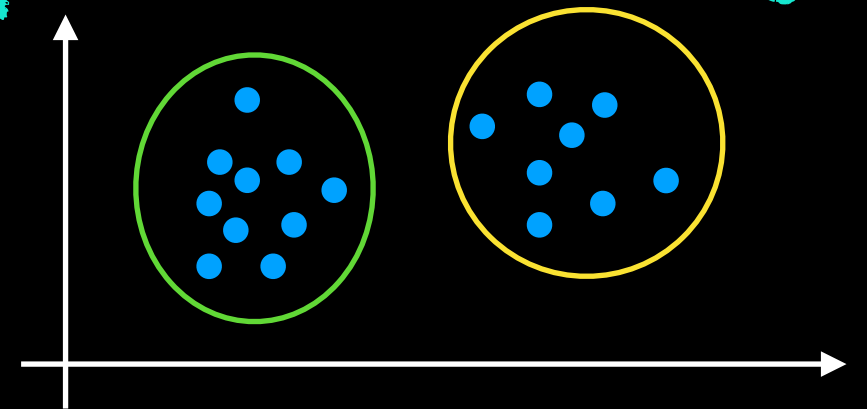
$$Y = f(X)$$

Handwriting Recognition

Face Detection

Speech Recognition

Unsupervised Learning



Machine Learning

Supervised Learning

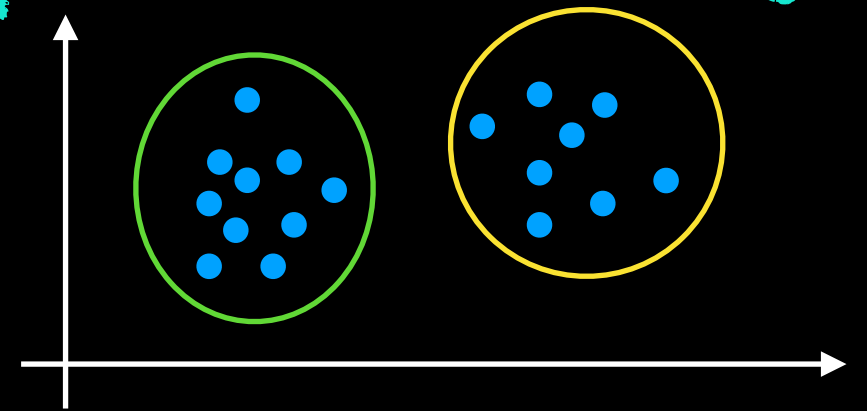
$$Y = f(X)$$

Handwriting Recognition

Face Detection

Speech Recognition

Unsupervised Learning



Reinforcement Learning

Machine Learning

Supervised Learning

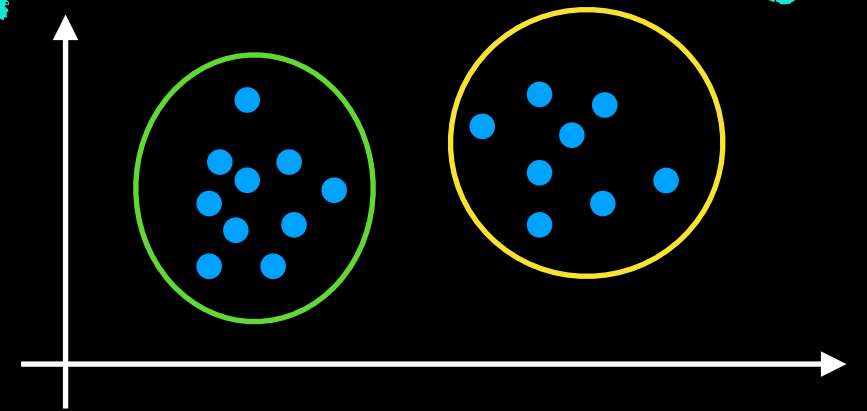
$$Y = f(X)$$

Handwriting Recognition

Face Detection

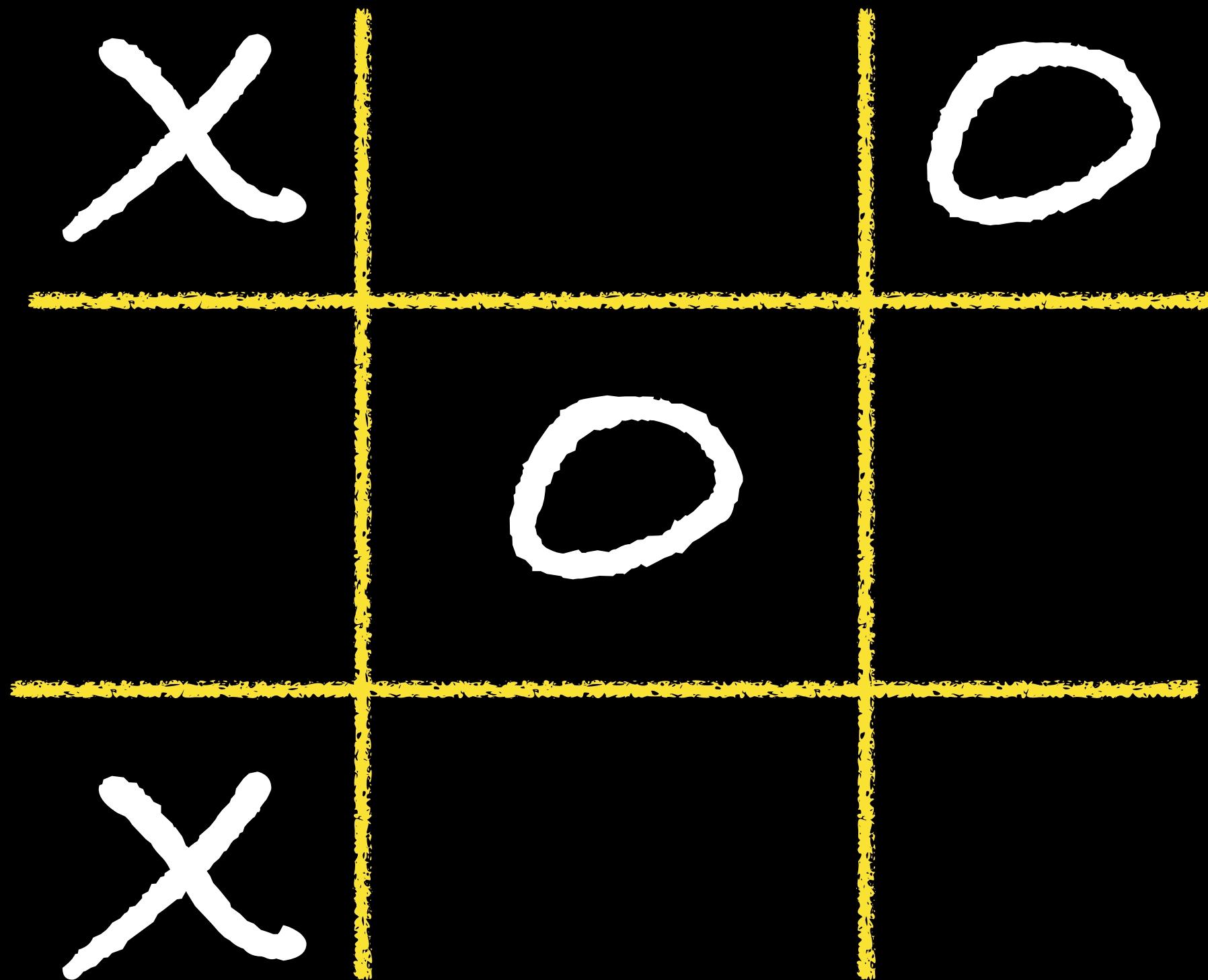
Speech Recognition

Unsupervised Learning



Reinforcement Learning

Learning to perform action in a certain environment,
closest to what biological systems do.



Environment

X

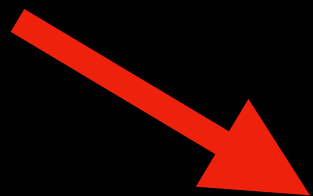
O

O

X

Agent

Environment



X

O

O

X

Agent

Environment



X

O

O

X

State/Configuration

Agent

Environment



X

O

O

X

Action

State/Configuration

Agent

Environment



X

O

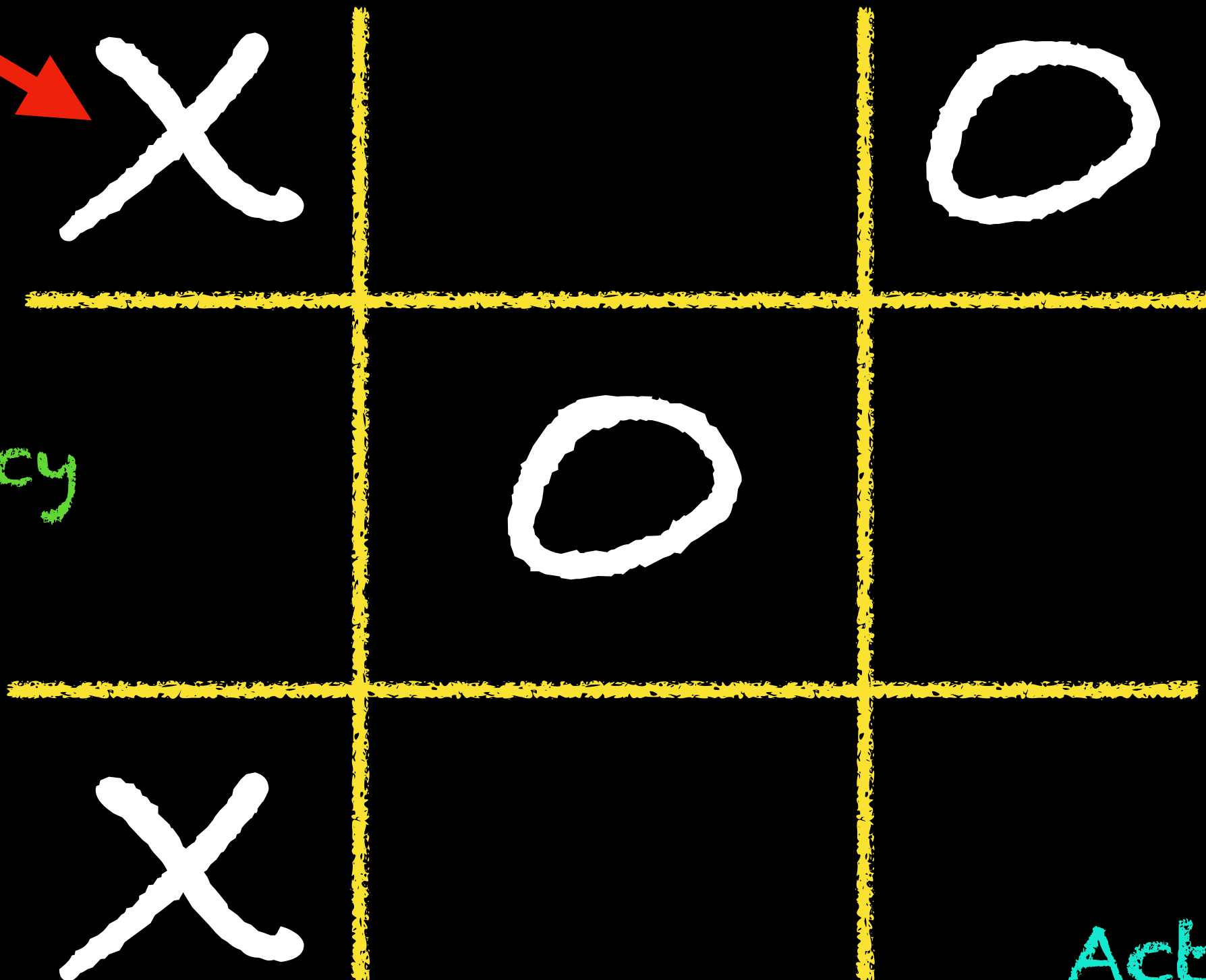
Policy

O

X

Action

State/Configuration



Agent

Environment



X

O

Policy

Reward

O

X

Action

State/Configuration

Agent

Environment



X

O

Policy

Reward

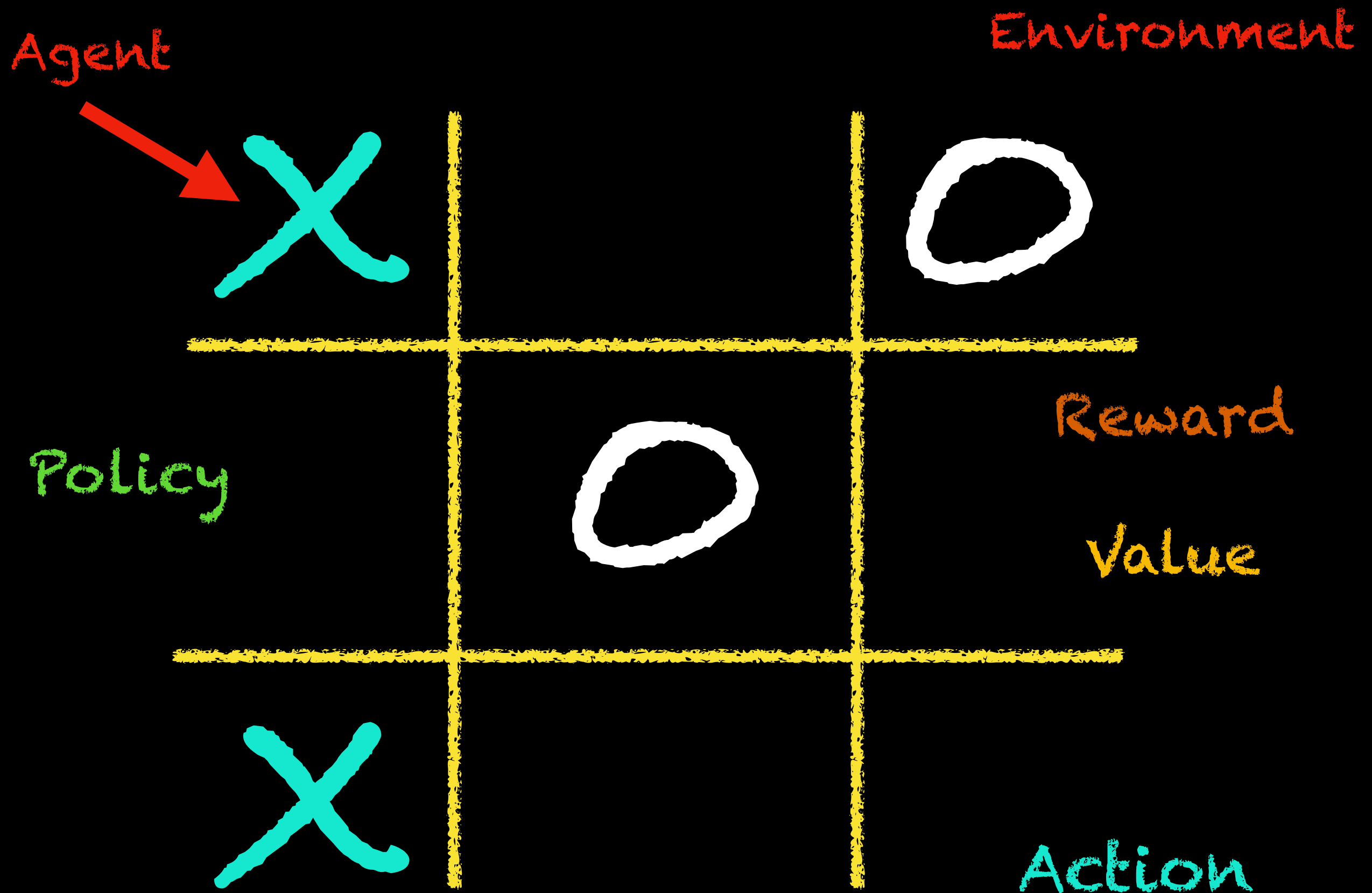
O

Value

X

Action

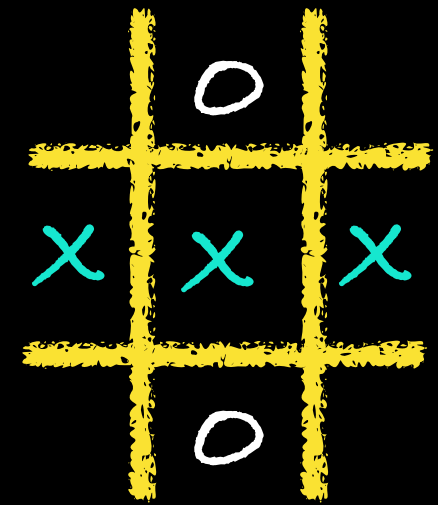
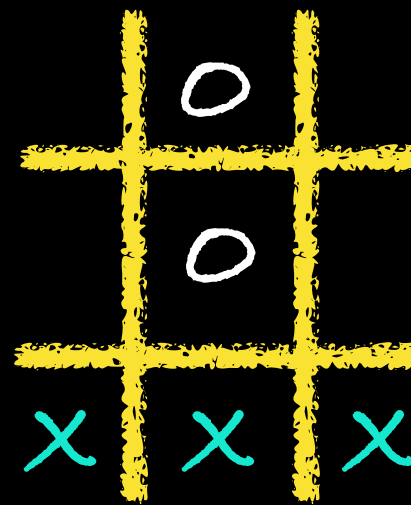
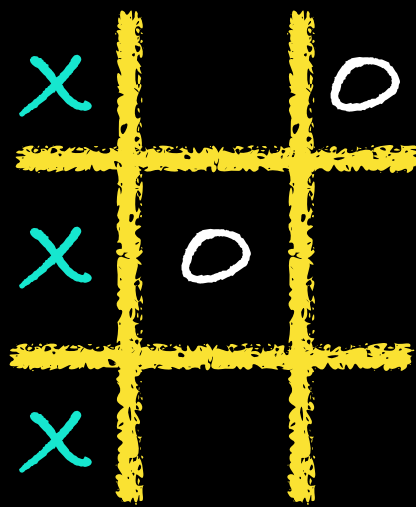
State/Configuration



Total Number of States : 3^9

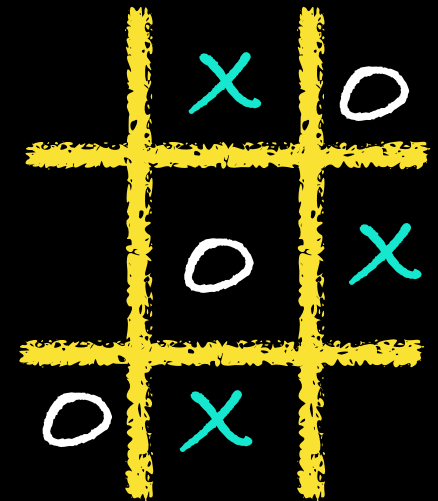
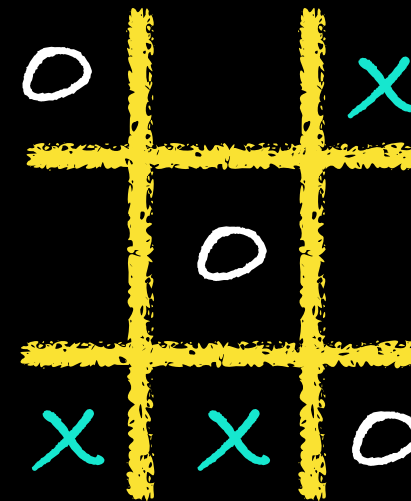
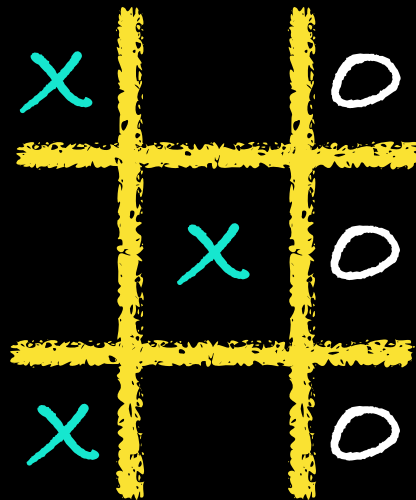
Win!

value = +1



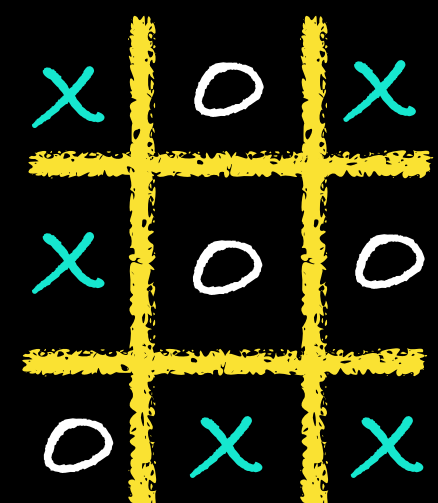
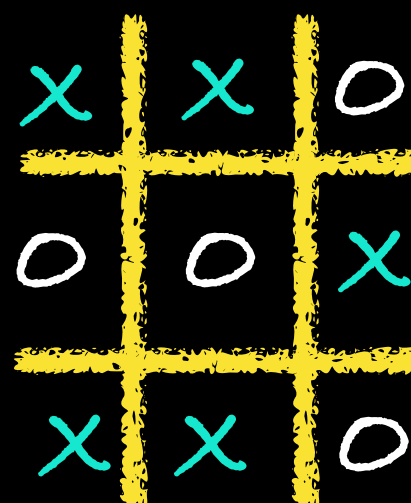
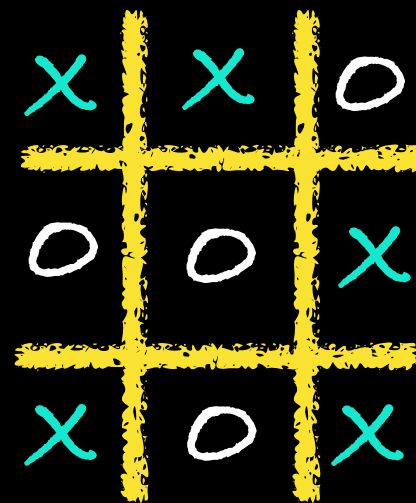
Lose!

value = -1



Draw!

value = 0



Total Number of States : 3^9

Agent



X

O

Policy

O

Reward

Value

X

Action

Total Number of States : 3^9

$$V(S_t) = V(S_t) + \alpha [V(S_{t+1}) - V(S_t)]$$

Agent



X

O

Policy

O

Reward

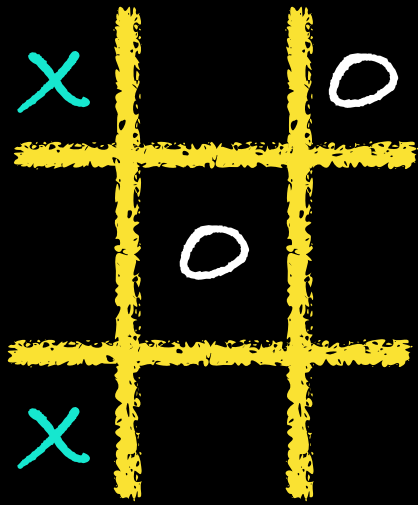
Value

X

Action

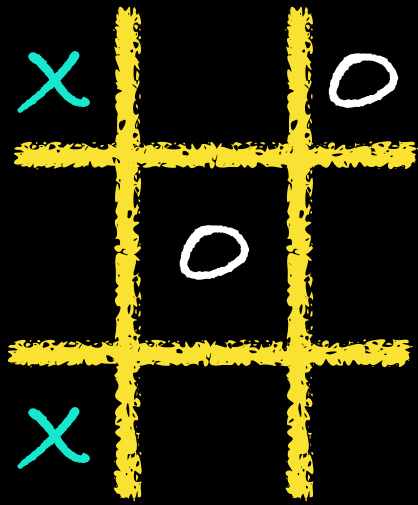
Total Number of States : 3^9

$$V(S_t) = V(S_t) + \alpha [V(S_{t+1}) - V(S_t)]$$

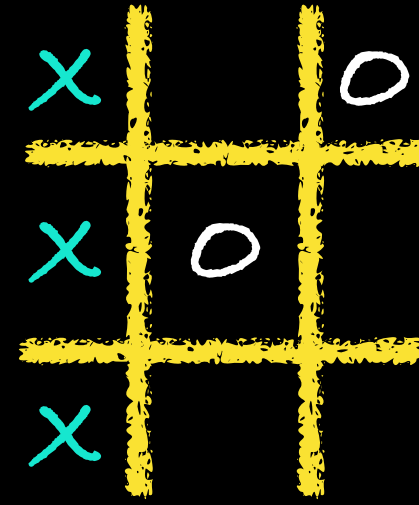


$$V(S_4) = 0$$

$$V(S_t) = V(S_t) + \alpha [V(S_{t+1}) - V(S_t)]$$

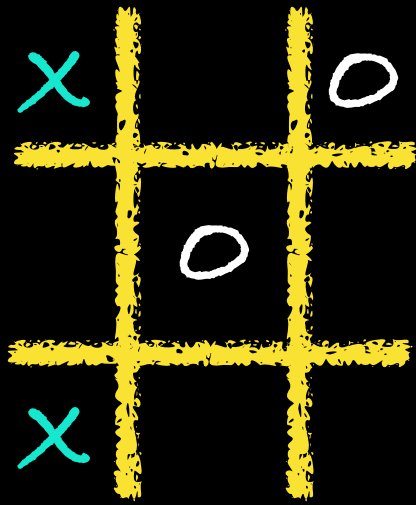


$$V(S_4) = 0$$



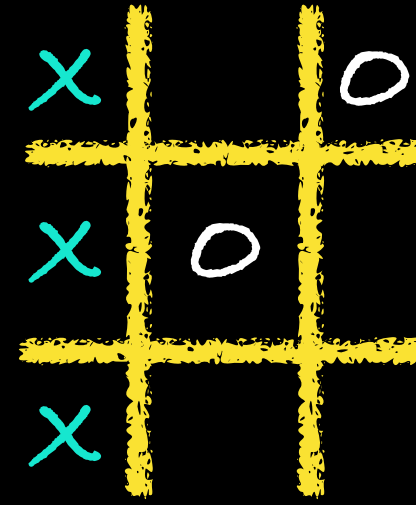
$$V(S_5) = 1$$

$$V(S_t) = V(S_t) + \alpha [V(S_{t+1}) - V(S_t)]$$



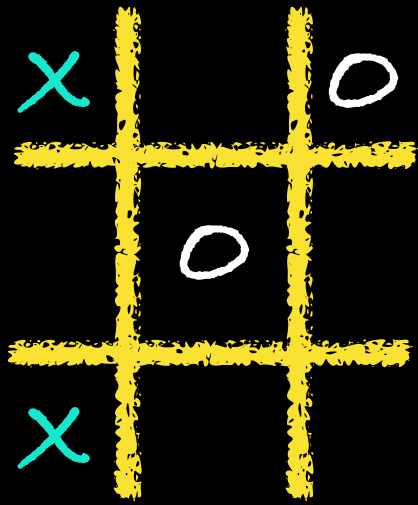
$$V(S_4) = 0$$

$$\rightarrow V(S_4) = 0 + \alpha [1 - 0] = \alpha$$



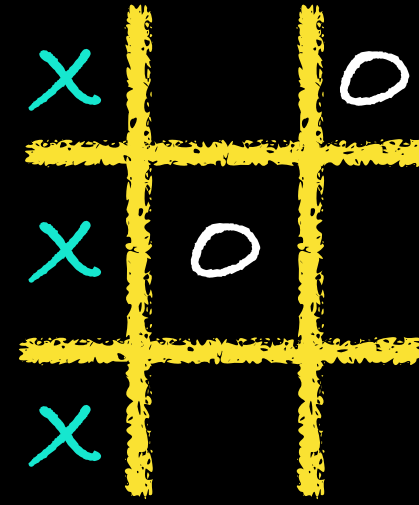
$$V(S_5) = 1$$

$$V(S_t) = V(S_t) + \alpha [V(S_{t+1}) - V(S_t)]$$

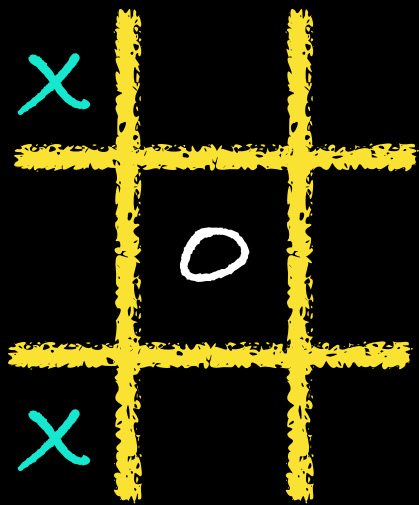


$$V(S_4) = 0$$

$$\rightarrow V(S_4) = 0 + \alpha [1 - 0] = \alpha$$

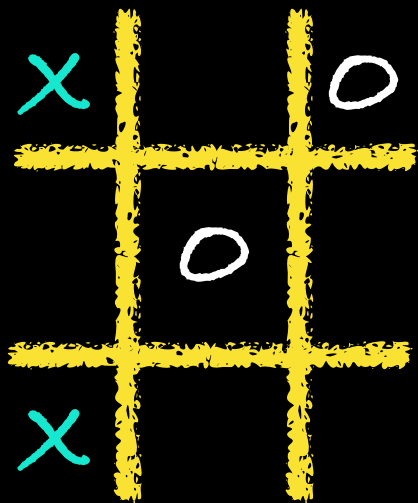


$$V(S_5) = 1$$



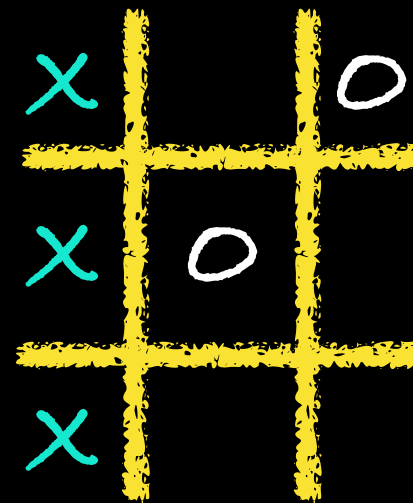
$$V(S_3) = 0$$

$$V(S_t) = V(S_t) + \alpha [V(S_{t+1}) - V(S_t)]$$

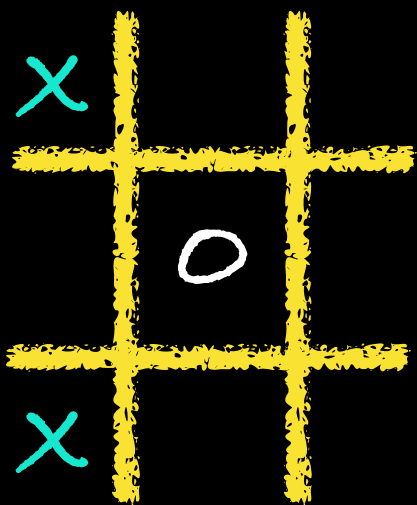


$$V(S_4) = 0$$

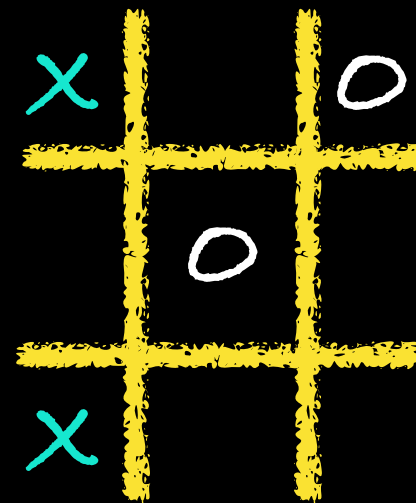
$$\rightarrow V(S_4) = 0 + \alpha [1 - 0] = \alpha$$



$$V(S_5) = 1$$

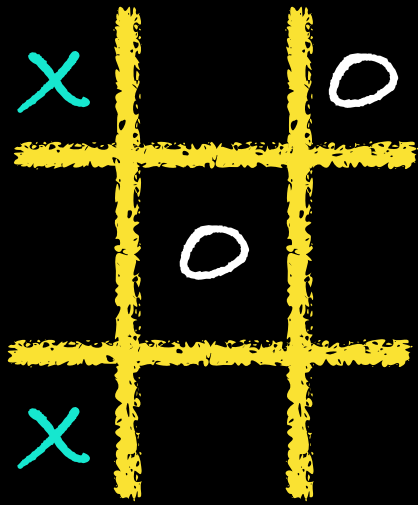


$$V(S_3) = 0$$



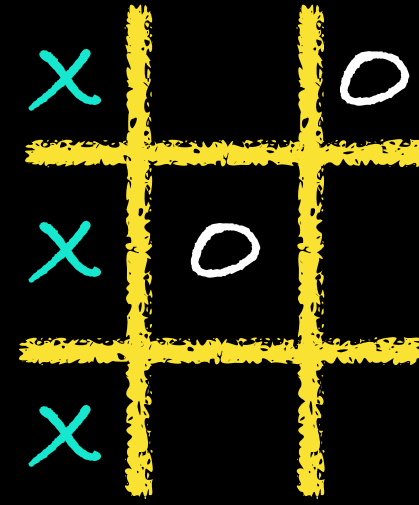
$$V(S_4) = \alpha$$

$$V(S_t) = V(S_t) + \alpha [V(S_{t+1}) - V(S_t)]$$

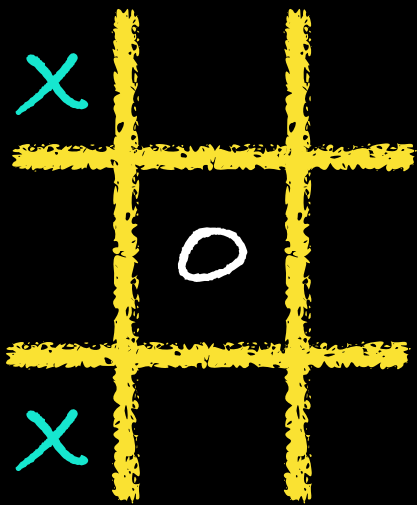


$$V(S_4) = 0$$

$$\rightarrow V(S_4) = 0 + \alpha [1 - 0] = \alpha$$

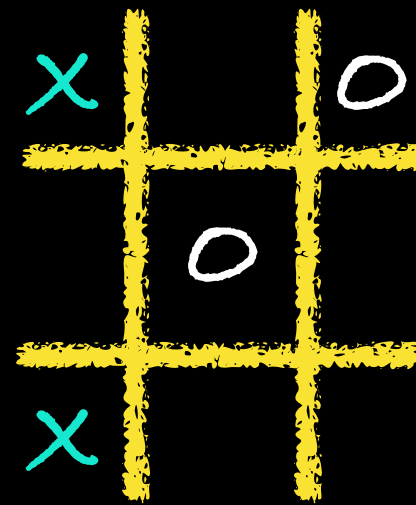


$$V(S_5) = 1$$



$$V(S_3) = 0$$

$$\rightarrow V(S_3) = 0 + \alpha [\alpha - 0] = \alpha^2$$



$$V(S_4) = \alpha$$

$$V(S_t) = V(S_t) + \alpha [V(S_{t+1}) - V(S_t)]$$



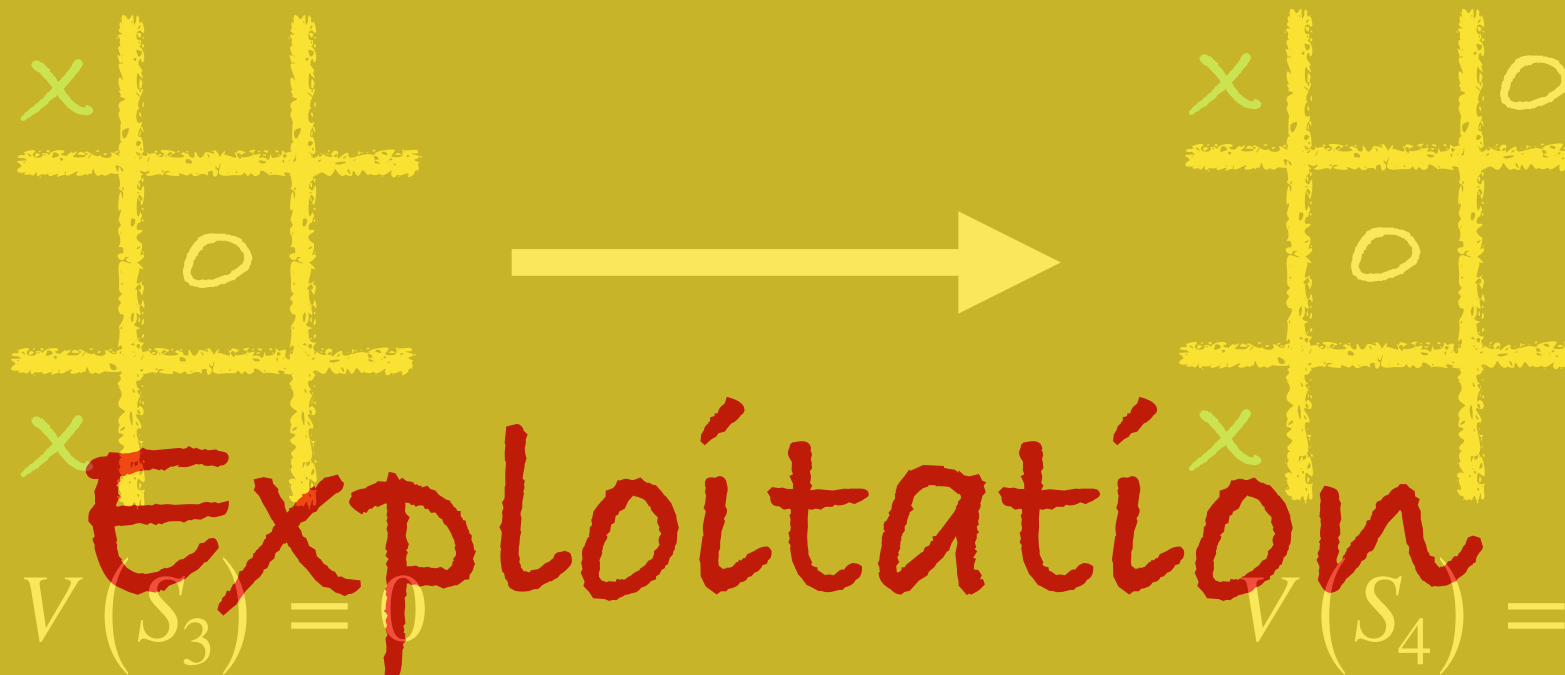
$$V(S_4) = 0$$

$$V(S_5) = 1$$

Policy

$$\rightarrow V(S_4) = 0 + \alpha [1 - 0] = \alpha$$

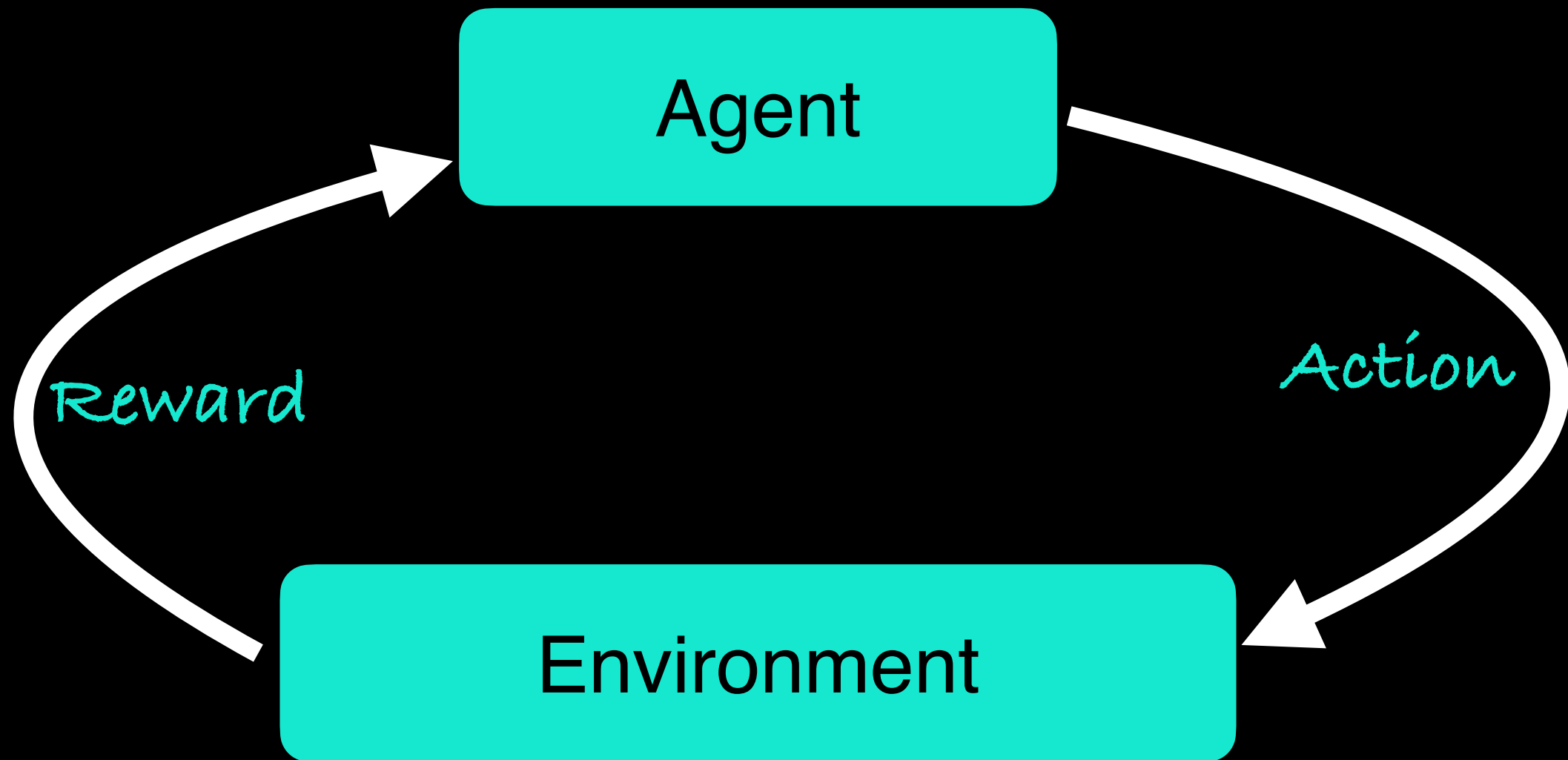
VS.

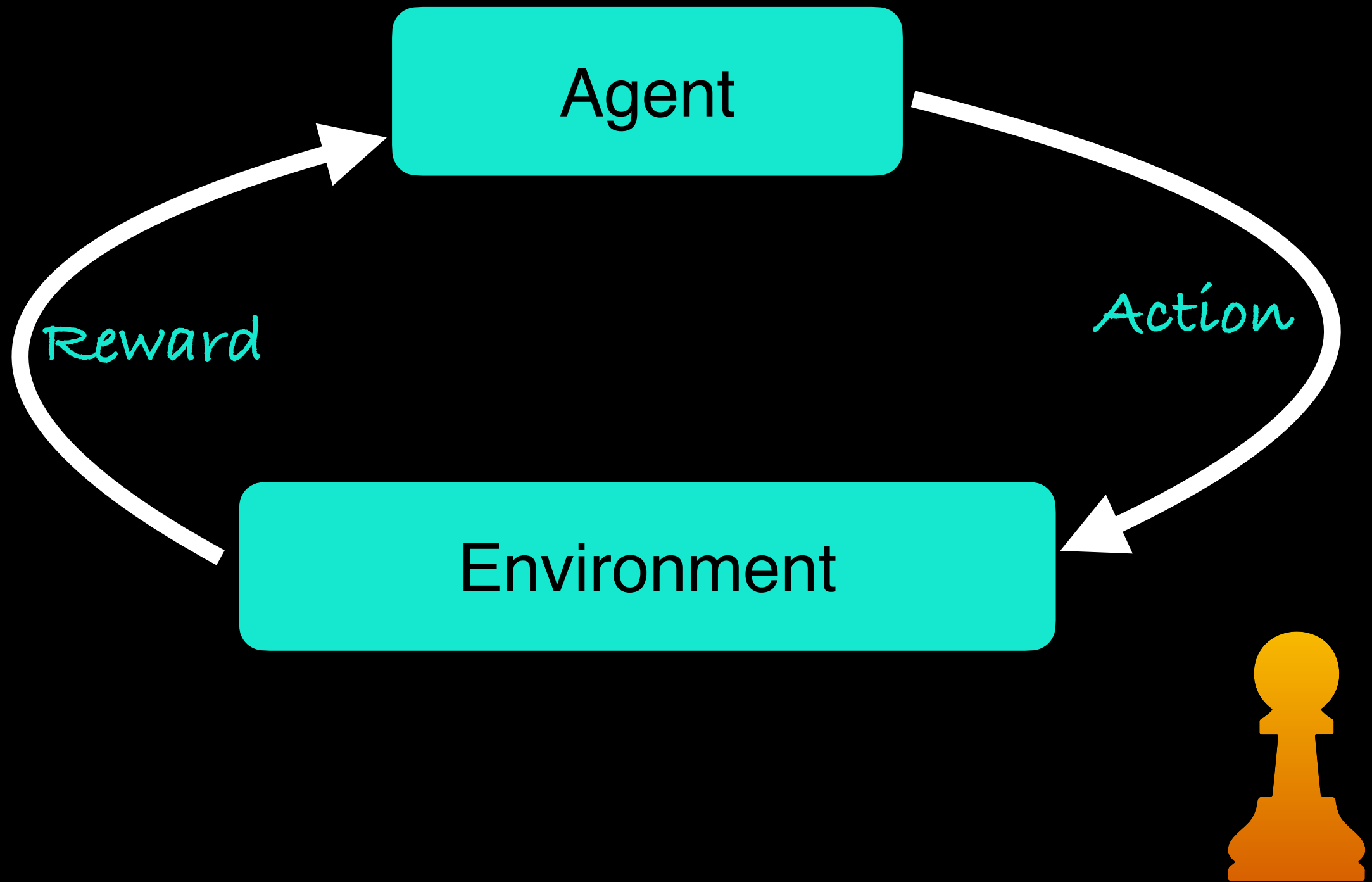


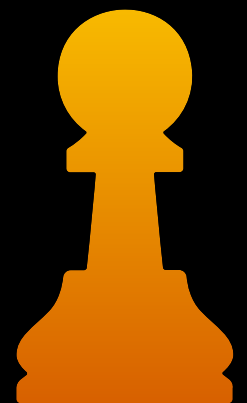
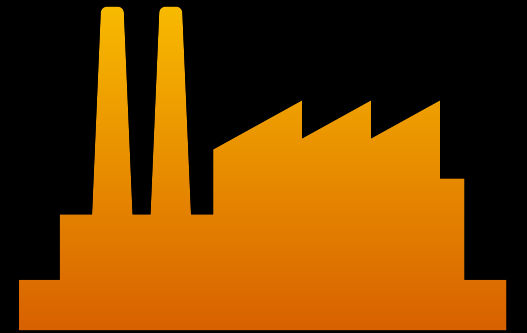
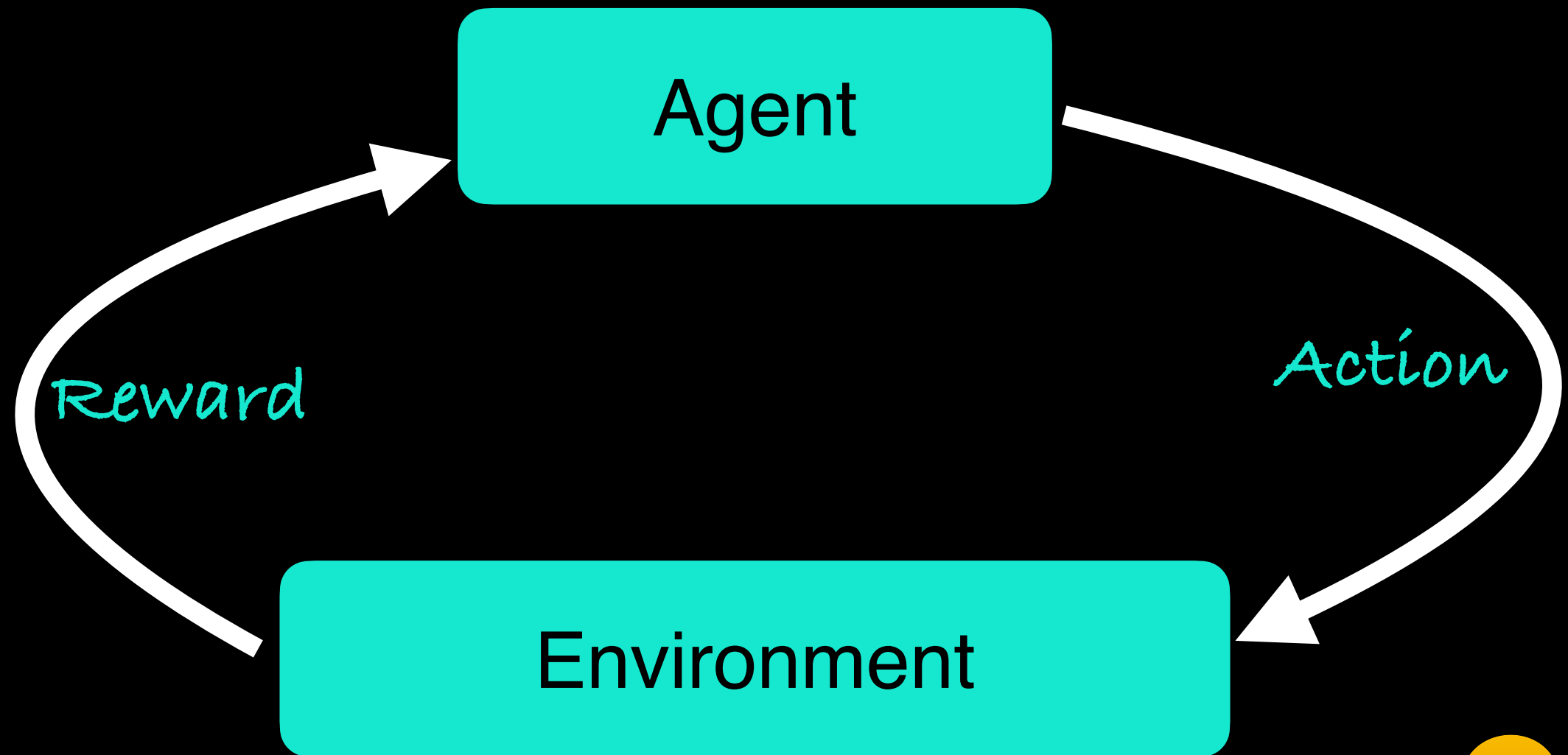
$$V(S_3) = 0$$

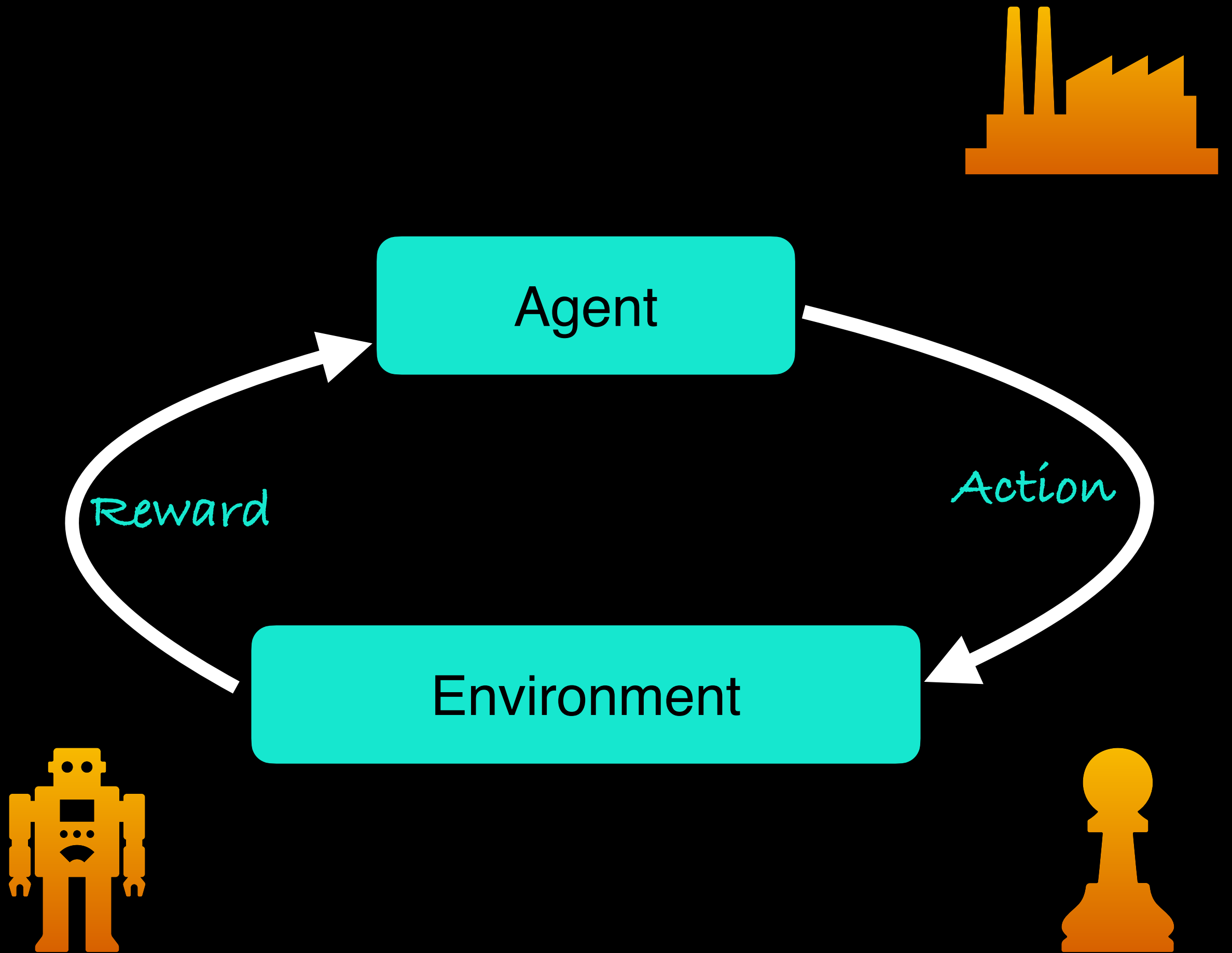
$$V(S_4) = \alpha$$

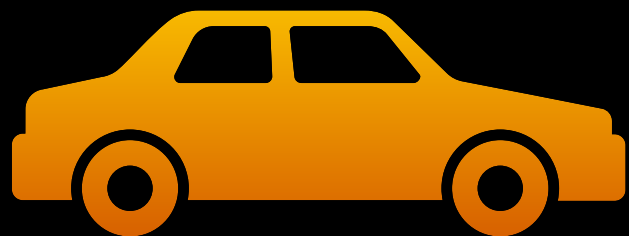
$$\rightarrow V(S_3) = 0 + \alpha [\alpha - 0] = \alpha^2$$











Agent



Reward



Action

Environment

