# Package 'ReinforcementLearning'

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Title Model-Free Reinforcement Learning

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<b>Description</b> Performs model-free reinforcement learning in R. This implementation allows to learn an optimal policy based on sample sequences consisting of states, actions and rewards. In addition, it supplies multiple predefined reinforcement learning algorithms, such as experience replay.
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<b>Depends</b> R ( $>= 3.1.0$ )
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epsilonGreedyActionSelection

*Performs*  $\varepsilon$ *-greedy action selection* 

#### **Description**

Implements  $\varepsilon$ -greedy action selection. In this strategy, the agent explores the environment by selecting an action at random with probability  $\varepsilon$ . Alternatively, the agent exploits its current knowledge by choosing the optimal action with probability  $1-\varepsilon$ .

### Usage

```
epsilonGreedyActionSelection(Q, state, epsilon)
```

#### **Arguments**

Q State-action table of type hash.

state The current state.

epsilon Exploration rate between 0 and 1.

#### Value

Character value defining the next action.

#### References

Sutton and Barto (1998). Reinforcement Learning: An Introduction, Adaptive Computation and Machine Learning, MIT Press, Cambridge, MA.

experienceReplay

Performs experience replay

## **Description**

Performs experience replay. Experience replay allows reinforcement learning agents to remember and reuse experiences from the past. The algorithm solely requires input data in the form of sample sequences consisting of states, actions and rewards. The result of the learning process is a stateaction table Q that allows to infer the best possible action in each state.

#### Usage

```
experienceReplay(D, Q, control, ...)
```

#### **Arguments**

D	A dataframe containing the input data for reinforcement learning. Each row	
	represents a state transition tuple (s,a,r,s_new).	

Q Existing state-action table of type hash.

control Control parameters defining the behavior of the agent.

... Additional parameters passed to function.

gridworldEnvironment 3

#### Value

Returns an object of class hash that contains the learned Q-table.

#### References

Lin (1992) Self-improving reactive agents based on reinforcement learning, planning and teaching. Machine Learning.

Watkins (1992). Q-learning. Machine Learning.

#### See Also

ReinforcementLearning

gridworldEnvironment Defines an environment for a gridworld example

#### **Description**

Function defines an environment for a 2x2 gridworld example. Here, an agent is intended to navigate from an arbitrary start position to a goal position. The grid is surrounded by a wall which makes it impossible for the agent to move off the grid. In addition, the agent faces a wall between s1 and s4. If the agent reaches the goal position, it earns a reward of 10. Crossing each square of the grid leads to a negative reward of -1.

## Usage

```
gridworldEnvironment(state, action)
```

#### **Arguments**

state The current state.

Action to be executed. action

## Value

List containing the next state and the reward.

lookupActionSelection Converts a name into an action selection function

## **Description**

Input is a name for the action selection, output is the corresponding function object.

## Usage

lookupActionSelection(type)

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#### **Arguments**

type

A string denoting the type of action selection. Allowed values are epsilon-greedy or random.

#### Value

Function that implements the specific learning rule.

lookupLearningRule

Loads reinforcement learning algorithm

### **Description**

Decides upon a learning rule for reinforcement learning. Input is a name for the learning rule, output is the corresponding function object.

#### Usage

lookupLearningRule(type)

### **Arguments**

type

A string denoting the learning rule. Allowed values are experienceReplay.

#### Value

Function that implements the specific learning rule.

## See Also

ReinforcementLearning

policy

Calculates the reinforcement learning policy

### **Description**

Calculates reinforcement learning policy from a given state-action table Q. The policy is the decision making function of the agent and defines the learning agent's way of behaving at a given time.

## Usage

policy(x)

## Arguments

Х

Variable which encodes the behavior of the agent. This can be either a matrix, data.frame or a rl object.

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#### Value

Returns the learned policy.

#### See Also

ReinforcementLearning

randomActionSelection Performs random action selection

## **Description**

Performs random action selection. In this strategy, the agent always selects an action at random.

### Usage

```
randomActionSelection(Q, state, epsilon)
```

#### **Arguments**

Q State-action table of type hash.

state The current state.

epsilon Exploration rate between 0 and 1 (not used).

## Value

Character value defining the next action.

ReinforcementLearning Performs reinforcement learning

## Description

Performs model-free reinforcement learning. Requires input data in the form of sample sequences consisting of states, actions and rewards. The result of the learning process is a state-action table and an optimal policy that defines the best possible action in each state.

### Usage

```
ReinforcementLearning(data, s = "s", a = "a", r = "r", s_new = "s_new",
  learningRule = "experienceReplay", iter = 1, control = list(alpha = 0.1,
  gamma = 0.1, epsilon = 0.1), verbose = F, model = NULL, ...)
```

#### **Arguments**

data	A dataframe containing the input sequences for reinforcement learning. Each row represents a state transition tuple (s,a,r,s_new).
S	A string defining the column name of the current state in data.
a	A string defining the column name of the selected action for the current state in data.
r	A string defining the column name of the reward in the current state in data.
s_new	A string defining the column name of the next state in data.
learningRule	A string defining the selected reinforcement learning agent. The default value and only option in the current package version is experienceReplay.
iter	(optional) Iterations to be done. iter is an integer greater than 0. By default, iter is set to 1.
control	(optional) Control parameters defining the behavior of the agent. Default: alpha = $0.1$ ; gamma = $0.1$ ; epsilon = $0.1$ .
verbose	If true, progress report is shown. Default: false.
model	(optional) Existing model of class r1. Default: NULL.
	Additional parameters passed to function.

#### Value

An object of class rl with the following components:

Q Resulting state-action table.

Q\_hash Resulting state-action table in hash format.

Actions Set of actions.

States Set of states.

Policy Resulting policy defining the best possible action in each state.

 $Reward Sequence \ Rewards \ collected \ during \ each \ learning \ episode \ in \ {\tt iter}.$ 

Reward Total reward collected during the last learning iteration in iter.

#### References

Sutton and Barto (1998). Reinforcement Learning: An Introduction, Adaptive Computation and Machine Learning, MIT Press, Cambridge, MA.

## **Examples**

```
# Sampling data (1000 grid sequences)
data <- sampleGridSequence(1000)

# Setting reinforcement learning parameters
control <- list(alpha = 0.1, gamma = 0.1, epsilon = 0.1)

# Performing reinforcement learning
model <- ReinforcementLearning(data, s = "State", a = "Action", r = "Reward",
s_new = "NextState", control = control)

# Printing model
print(model)</pre>
```

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```
# Plotting learning curve
plot(model)
```

sampleExperience

Sample state transitions from an environment function

#### **Description**

Function generates sample experience in the form of state transition tuples.

## Usage

```
sampleExperience(N, env, states, actions, actionSelection = "random",
  control = list(alpha = 0.1, gamma = 0.1, epsilon = 0.1), model = NULL,
  ...)
```

#### Arguments

N Number of samples.

env An environment function.

states A character vector defining the environment states.

actions A character vector defining the available actions.

actionSelection

(optional) Defines the action selection mode of the reinforcement learning agent.

Default: random.

control (optional) Control parameters defining the behavior of the agent. Default: alpha = 0.1;

gamma = 0.1; epsilon = 0.1.

model (optional) Existing model of class rl. Default: NULL.

... Additional parameters passed to function.

## Value

An dataframe containing the experienced state transition tuples s,a,r,s\_new. The individual columns are as follows:

State The current state.

Action The selected action for the current state.

Reward The reward in the current state.

NextState The next state.

#### See Also

```
ReinforcementLearning gridworldEnvironment
```

#### **Examples**

```
# Define environment
env <- gridworldEnvironment

# Define states and actions
states <- c("s1", "s2", "s3", "s4")
actions <- c("up", "down", "left", "right")

# Sample 1000 training examples
data <- sampleExperience(N = 1000, env = env, states = states, actions = actions)</pre>
```

sampleGridSequence

Sample grid sequence

#### **Description**

Function uses an environment function to generate sample experience in the form of state transition tuples.

## Usage

```
sampleGridSequence(N, actionSelection = "random", control = list(alpha =
   0.1, gamma = 0.1, epsilon = 0.1), model = NULL, ...)
```

### **Arguments**

Number of samples.

actionSelection

(optional) Defines the action selection mode of the reinforcement learning agent.

Default: random.

 $control \qquad \qquad (optional) \ Control \ parameters \ defining \ the \ behavior \ of \ the \ agent. \ Default: \ alpha = 0.1;$ 

gamma = 0.1; epsilon = 0.1.

model (optional) Existing model of class rl. Default: NULL.

... Additional parameters passed to function.

#### Value

An dataframe containing the experienced state transition tuples s,a,r,s\_new. The individual columns are as follows:

State The current state.

Action The selected action for the current state.

Reward The reward in the current state.

NextState The next state.

#### See Also

```
gridworldEnvironment
ReinforcementLearning
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state

Creates a state representation for arbitrary objects

#### **Description**

Converts object of any class to a reinforcement learning state of type character.

#### Usage

```
state(x, ...)
```

#### **Arguments**

x An object of any class.

... Additional parameters passed to function.

#### Value

Character value defining the state representation of the given object.

tictactoe

Game states of 100,000 randomly sampled Tic Tac Toe games.

## Description

A dataset containing 406,541 games states of Tic Tac Toe. The player who succeeds in placing three of their marks in a horizontal, vertical, or diagonal row wins the game. All states are observed from the perspective of player X who is also assumed to have played first.

## Usage

tictactoe

#### **Format**

A data frame with 406,541 rows and 4 variables:

**State** The current game state, i.e. the state of the 3x3 grid.

**Action** The move of player X in the current game state.

NextState The next observed state after action selection of player X and B.

**Reward** Indicates terminal and non-terminal game states. Reward is +1 for 'win', 0 for 'draw', and -1 for 'loss'.

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