# GRL: a generic C++ reinforcement learning library

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#### 1 Introduction

GRL is a C++ reinforcement learning library that aims to easily allow evaluating different algorithms through a declarative configuration interface.

## 2 Directory structure

```
|-- base
                             Base library
   |-- include
                             Header files
    `-- src
                             Source files
        |-- agents
                             Agents (fixed, black box, td)
        |-- discretizers
                             Action discretizers
        |-- environments
                             Environments (pendulum, cart-pole)
       |-- experiments
                             Experiments (online, batch)
        |-- policies
                             Control policies (PID, Q-based)
        |-- predictors
                             Value function predictors (SARSA, AC)
        |-- projectors
                             State projectors (tile coding, fourier)
       |-- representations Representations (linear, ann)
       |-- samplers
                             Action samplers (greedy, e-greedy)
                             MDP solvers (VI, rollout-based)
       |-- solvers
       I-- traces
                             Elibility traces (accumulating, replacing)
       `-- visualizations
                             Visualizations (value function, policy)
|-- addons
                             Optional modules
   |-- cma
                             CMA-ES black-box optimizer
    |-- gl
                             OpenGL-based visualizations
   |-- glut
                             GLUT-based visualizer
   |-- 11r
                             Locally linear regression representation
   |-- lqr
                             Linear Quadratic Regulator solver
   |-- matlab
                             Matlab interoperability
   |-- muscod
                             Muscod interoperability
    |-- odesim
                             Open Dynamics Engine environment
```

Rigid Body Dynamics Library dynamics |-- rbdl `-- ros ROS interoperability |-- bin Python binaries (configurator) Imported external library code |-- externals |-- cfg Sample configurations |-- share Misc files `-- taskmaster Taskmaster parameter study example |-- tests Unit tests |-- CMakeLists.txt CMake instructions to build everything `-- grl.cmake CMake helper functions

# 3 Prerequisites

GRL requires some libraries in order to compile. Which ones exactly depends on which agents and environments you would like to build, but the full list is

- Git
- GCC (including g++)
- Eigen
- GLUT
- ZLIB
- QT4 (including the OpenGL bindings)
- TinyXML
- MuParser
- ODE, the Open Dynamics Engine
- Python (including Tkinter and the yaml reader)
- Lua

On Ubuntu 16.04, these may be installed with the following command:

# 4 Building

GRL may be built with or without ROS's catkin. When building with, simply merge grl.rosinstall with your catkin workspace

```
wcaarls@vbox:~$ mkdir indigo_ws
wcaarls@vbox:~$ cd indigo_ws
wcaarls@vbox:~/indigo_ws$ rosws init src /opt/ros/indigo
wcaarls@vbox:~/indigo_ws$ cd src
wcaarls@vbox:~/indigo_ws/src$ rosws merge /path/to/grl.rosinstall
wcaarls@vbox:~/indigo_ws/src$ rosws up
wcaarls@vbox:~/indigo_ws/src$ cd ...
wcaarls@vbox:~/indigo_ws$ catkin_make
   Otherwise, follow the standard CMake steps of (in the grl directory)
wcaarls@vbox:~/src/grl$ mkdir build
wcaarls@vbox:~/src/grl$ cd build
wcaarls@vbox:~/src/grl/build$ cmake ..
-- The C compiler identification is GNU 4.8.2
wcaarls@vbox:~/src/grl/build$ make
Scanning dependencies of target yaml-cpp
. . .
```

# 5 Running

The most important executables in grl are the deployer (grld) and configurator (grlc). The configurator allows you to generate configuration files easily. To see an example, run

```
wcaarls@vbox:~/src/grl/bin$ ./grlc ../cfg/pendulum/sarsa_tc.yaml
```

More information on the configurator can be found in Section 8. Once you have configured your experiment, you can either run it directly from the configurator, or save it and run it using the deployer. For example:

```
wcaarls@vbox:~/src/grl/build$ ./grld ../cfg/pendulum/sarsa_tc.yaml
```

#### 6 Build environment

The whole grl system is built as a single package, with the exception of mprl\_msgs. This is done to facilitate building inside and outside catkin. There is one CMakeLists.txt that is used in both cases. The ROS interoperability is selectively built based on whether cmake was invoked by catkin\_make or not.

Modules are built by calling their respective build.cmake scripts, which is done by grl\_build\_library. The include directory is set automatically, as is an SRC variable pointing to the library's source directory.

The build system has a simplistic dependency management scheme through grl\_link\_libraries. This calls the link.cmake files of the libraries on which the current library depends. Typically they will add some target\_link\_libraries and add upstream dependencies. grl\_link\_libraries also automatically adds the upstream library's include directory.

#### 7 Class structure

Most classes in grl derive from Configurable, a base class that standardizes configuration such that the object hierarchy may be constructed declaratively in a configuration file. Directly beneath Configurable are the abstract base classes defining the operation of various parts of the reinforcement learning environment, being:

Agent RL-GLUE<sup>1</sup> style agent interface, receiving observations in an episodic manner and returning actions.

Discretizer Provides a list of discrete points spanning a continuous space.

Environment RL-GLUE style environment interface, receiving actions and returning observations.

Experiment Top-level interface, which typically calls the agent and environment in the correct manner, but may in general implement any experiment.

Optimizer Black-box optimization of control policies, suggesting policies and acting on their cumulative reward.

Policy Basic control policy that implements the state-action mapping.

Predictor Basic reinforcement learning interface that uses transitions to predict a value function or model.

Projector Projects an observation onto a feature vector, represented as a Projection.

Representation Basic supervised learning interface that uses samples to approximate a function. As such, it generally supports reading, writing and updating of any vector-to-vector mapping.

Sampler (Stochastically) chooses an item from a vector of (generally unnormalized) values.

Trace Stores a trace of projections with associated eligibilities that can be iterated over.

<sup>1</sup>http://http://glue.rl-community.org

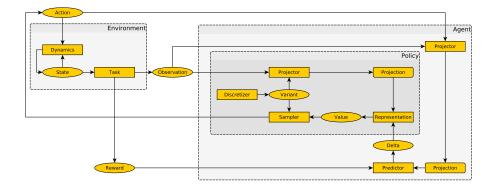


Figure 1: Information flow diagram for regular TD control. Rectangles (and dashed rectangles) are Configurable objects, while the others are the data passed between them.

Visualization Draws on the screen to visualize some aspect of the learning process.

Visualizer Keeps track of visualizations and provides the interface to the graphics subsystem.

Each abstract base class is generally implemented in various concrete classes, with or without additional hierarchy. A list can be requested by running

```
wcaarls@vbox:~/src/grl/bin$ ./grlq
```

and is also available in the appendices of this document.

A typical example of the information flow between the various classes can be seen in Figure 1, which depicts the standard TD control setting.

#### 7.1 Configuration

Each Configurable subclass must define its type and a short description using the TYPEINFO macro:

```
class OnlineLearningExperiment : public Experiment
{
   public:
     TYPEINFO("experiment/online_learning", "Interactive learning experiment")
   /* ... */
};
```

This textual description of the type is used to facilitate user configuration by limiting the selection of parameter values, as well as enforcing the type hierarchy.

In general, the textual description should follow the C++ class hierarchy, but this is not obligatory.

The basic Configurable interface has three important functions:

#### 7.1.1 request

```
virtual void request(ConfigurationRequest *config);
```

request is called by the configurator to find out which parameters the object requires to be set, and which parameters it exports for other objects to use. To do this, it should extend the given ConfigurationRequest by pushing configuration request parameters (CRPs). A basic CRP has the following signature:

```
CRP(string name, string desc, TYPE value)
```

where TYPE is one of int, double, Vector, or string. For example:

```
config->push_back(CRP("steps", "Number of steps per learning run", steps_));
config->push_back(CRP("output", "Output base filename", output_));
```

The value argument is used both to determine the type of the parameter and the default value suggested by the configurator. request may also be called while the program is running, in which case it is expected to return the current value of all parameters.

To use other Configurable objects as parameters, use

```
CRP(string name, string type, string desc, Configurable *value)
```

The extra type field restricts which Configurable objects may be used to configure this parameter. Only objects whose TYPEINFO starts with the given type are eligible. For example:

```
config->push_back(CRP("policy", "policy/parameterized",
```

```
"Control policy prototype", policy_));
```

restricts the "policy" parameter to classes derived from ParameterizedPolicy. Note that this extra type hierarchy is related to, but not derived from the actual class hierarchy. Care must therefore be taken in the correct usage of TYPEINFO.

Some parameters are not requested, but rather *provided* by an object. In that case. These have the following signature:

```
CRP(string name, string type, string desc, CRP::Provided)
```

Examples of provided parameters are the number of observation dimensions (provided by Tasks) or the current system state (provided by some Environments).

#### 7.1.2 configure

```
virtual void configure(Configuration *config);
```

configure is called after all parameters (including other Configurable objects) have been initialized. The parameter values may be accessed using mapping syntax (config["parameter"]). Note that Configurable objects are passed as void pointers and must still be cast to their actual class:

```
steps_ = config["steps"];
output_ = config["output"].str();
policy_ = (ParameterizedPolicy*)config["policy"].ptr();
```

Note the use of .str() and .ptr() for strings and objects, respectively. Provided parameters should be written to the configuration instead of read, like so:

```
config.set("state", state_);
```

#### 7.1.3 reconfigure

```
virtual void reconfigure(const Configuration *config);
```

Some parameters may be defined as reconfigurable by appending CRP::Online to the respective CRP signature. In the case of a reconfiguration, reconfigure will be called with the new values of those parameters in config. reconfigure may also be used for general messaging, equivalent to RL-GLUE's message calls. In that case, it is often helpful to reconfigure all objects in the object hierarchy, which can be done using

```
void Configurable::walk(const Configuration &config);
```

Examples are resetting the hierarchy for a new run (config["action"] = "reset") or saving the current state of all memories (config["action"] = "save"). In the latter case, Configurable::path() may be used to determine an object's location in the object hierarchy.

#### 7.2 Roles

While using the configurator, the user often has to select previously defined objects as the value of certain parameters. If all such previously defined objects are presented as possibilities, the list would quickly grow very large. To make setting these parameters easier, a class may have various *roles* while providing the same interface. In that case, only previously defined objects with a role that starts with the requested role are valid choices.

An example is a Representation, which may represent a state-value function, action-value function, control policy or model. Each has a different number of inputs and outputs, and chosing the wrong representation will result in mismatches. An object requesting a Representation may therefore request a certain role. For example:

requests any representation that represents action-values. A newly defined representation will do, of course, but from the previously defined ones only the ones with the right role are eligible.

The same strategy is used for basic types, for example:

make sure the only suggested previously defined values for the "outputs" parameter are ones with the "action\_dims" role. As an added convenience, if the parameter is defined as a *system parameter* (CRP::System), meaning that the choice is not free but rather defined by the structure of the configuration, and only a single value was previously defined, that value is automatically used.

The role that needs to be requested may depend on the role of the requesting object itself. In that case, the following signature for request should be used:

virtual void request(const std::string &role, ConfigurationRequest \*config);

# 8 Configurator

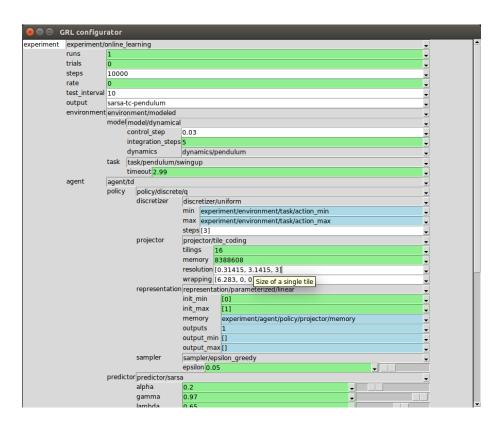


Figure 2: Python configurator user interface

#### 9 Matlab interface

If Matlab is installed (and can be found on the path), a MEX interfaces for the agents and environments is built. If you want to use these, make sure that you're building with a compatible compiler, both by setting the CC and CXX variables in your call to cmake and by correctly configuring mex.

#### 9.1 Environments

To initialize an environment, call

```
>> spec = grl_env('cfg/matlab/pendulum_swingup.yaml');
```

Where the argument specifies a configuration file that has a top-level 'environment' tag. spec gives some information about the environment, such as number of dimensions, minimum and maximum values, etc. Next, retrieve the first observation of an episode with

```
>> o = grl_env('start');
```

where o is the observation from the environment. All following steps should be called using

```
>> [o, r, t, d] = grl_env('step', a);
```

where a is the action suggested by the agent, r is the reward given by the environment, t signals termination of the episode and txtd is the length of the step. If t is 2, the episode ended in an absorbing state. When all episodes are done, exit cleanly with

```
>> grl_env('fini');
```

#### 9.2 Agents

To initialize the agent, use

```
>> grl_agent('init', 'cfg/matlab/sarsa.yaml');
```

Where the argument specifies a configuration file that has a top-level 'agent' tag. Next, give the first observation of an episode with

```
>> a = grl_agent('start', o);
```

where o is the observation from the environment and a is the action suggested by the agent. All following steps should be called using

```
>> a = grl_agent('step', d, r, o);
```

where  $\mathbf{r}$  is the reward given by the environment and txtd is the length of the step. To signal the end of an episode (absorbing state), use

```
>> a = grl_agent('end', d, r);
```

To end an episode without an absorbing state, simply start a new one. To exit cleanly after all epsiodes are finished (which also allows you to reinitialize the agent with different options), call

```
>> grl_agent('fini');
```

# A Agents

#### A.1 agent/black\_box

Agent that learns from the cumulative reward of complete rollouts

episodes int Number of episodes to evaluate policy optimizer optimizer Policy optimizer

#### A.2 agent/communicator

Communicator agent which connects GRL to a remote agent

communicatorcommunicatorComunicator which exchanges messages with an actual/virtual envobservation\_dimsint.observation\_dimsNumber of observation dimensionsaction\_dimsint.action\_dimsNumber of action dimensionsaction\_minvector.action\_minLower limit of actionaction\_maxvector.action\_maxUpper limit of action

test int.test Selection of a learning/testing agent role

#### A.3 agent/delayed\_td

Agent that learns from observed state transitions assuming non-integer values of control delay

policy mapping/policy Control policy

predictor predictor Value function predictor

control\_delay double Relative control delay: 0 (no delay) - 1 (one timestep delay)

#### A.4 agent/dyna

Agent that learns from both observed and predicted state transitions

planning\_steps int Number of planning steps per control step

planning\_horizon int Planning episode length

threads int Threads used for planning (0 = synchronous planning, 0 requires re)

policy mapping/policy Control policy

predictor Value function predictor

model observation\_model Observation model used for planning

model\_predictor predictor/model Model predictor

model\_agent agent Agent used for planning episodes

Provided parameters

state state Current observed state of planning

#### A.5 agent/filtering

Agent that filters incoming observations and outgoing actions

observation\_idx vector Index vector for downstream observation (-1=pad)

action\_idx vector Index vector for upstream action (-1=pad) action\_dims int Number of downstream action dimensions

agent agent Downstream agent

#### A.6 agent/fixed

Fixed-policy agent

policy mapping/policy Control policy

#### A.7 agent/leo/fixed

Leo fixed agent

policy mapping/policy Control policy

pub\_transition\_type signal/vector Publisher of the transition type

#### A.8 agent/leo/sma

State-machine agent for Leo

agent\_prepareagentPrepare agentagent\_standupagentSafe standup agentagent\_starteragentStarting agentagent\_mainagentMain agent

upright\_trigger trigger trigger which finishes stand-up phase and triggers preparation agent

fc\_trigger trigger trigger which checks for foot contact to ensure that robot is prepared to wa

starter\_trigger trigger Trigger which initiates a preprogrammed walking at the beginning

sub\_ic\_signal signal/vector Subscriber to the contact signal

#### A.9 agent/leo/sym\_wrapper

Leo agent that symmetrically wraps angles and controls

agent agent with reduced state-action space due to symmetry

sub\_ic\_signal signal/vector Publisher of the initialization and contact signal

#### A.10 agent/leo/td

Leo agent that learns from observed state transitions

policy mapping/policy Control policy

predictor value function predictor

pub\_transition\_type signal/vector Publisher of the transition type

#### A.11 agent/leo\_preprogrammed

Leo preprogrammed agent

rand\_gen random\_generator Random generator for action pertubation

epsilon double Exploration rate
output\_min vector.action\_min Lower limit on outputs
output\_max vector.action\_max Upper limit on outputs

#### A.12 agent/master/exclusive

Master agent that selects one sub-agent to execute

gamma double Discount rate control\_step double.control\_step Characteristic step time on which gamma is defined

predictor predictor Optional (model) predictor

agent1 agent/sub First subagent agent2 agent/sub Second subagent

#### A.13 agent/master/predicated

Master agent in which execution is predicated on preceding agent confidence

gamma double Discount rate

control\_step double.control\_step Characteristic step time on which gamma is defined

predictor predictor Optional (model) predictor

agent1 agent/sub First subagent agent2 agent/sub Second subagent

#### A.14 agent/master/random

Master agent that chooses sub-agents randomly

gamma double Discount rate

control\_step double.control\_step Characteristic step time on which gamma is defined

predictor predictor Optional (model) predictor

agent1 agent/sub First subagent
agent2 agent/sub Second subagent

#### A.15 agent/master/sequential

Master agent that executes sub-agents sequentially

predictor predictor Optional (model) predictor

agent1 agent First subagent, providing the suggested action agent2 agent Second subagent, providing the final action

exporter exporter Optional exporter for transition log (supports time, state, observation, action, reward,

# A.16 agent/master/sequential/additive

Additive master agent that executes sub-agents sequentially and adds their outputs

predictor predictor Optional (model) predictor

agent1 agent First subagent, providing the suggested action agent2 agent Second subagent, providing the final action

exporter exporter Optional exporter for transition log (supports time, state, observation, act

output\_min vector.action\_min Lower limit on outputs output\_max vector.action\_max Upper limit on outputs

#### A.17 agent/solver

Agent that successively solves learned models of the environment

interval int Episodes between successive solutions (0=asynchronous)

policy mapping/policy Control policy

predictor predictor Optional (model) predictor

solver solver Model-based solver

#### A.18 agent/sub/compartmentalized

Sub agent that is valid in a fixed state-space region

min vector.observation\_min Minimum of compartment bounding box max vector.observation\_max Maximum of compartment bounding box

agent agent Sub agent

#### A.19 agent/sub/filtering

Subagent that filters incoming observations and outgoing actions

observation\_idx vector Index vector for downstream observation (-1=pad)

action\_idx vector Index vector for upstream action (-1=pad) action\_dims int Number of downstream action dimensions

agent agent/sub Downstream subagent

## A.20 agent/sub/voluntary

Sub agent that has confidence as part of the action

dim int Action dimension that indicates confidence

agent agent Sub agent

#### A.21 agent/td

Agent that learns from observed state transitions

policy mapping/policy Control policy

predictor predictor Value function predictor

#### B Behaviors

#### B.1 behavior/leo\_squat\_sym

Leo squatting behavior with symmetrical switchers of observations

#### B.2 behavior/leo\_walk

Leo walking behavior without symmetrical switchers of observations

#### B.3 behavior/leo\_walk\_sym

Leo walking behavior with symmetrical switchers of observations

#### C Communicators

#### C.1 communicator/zeromq/pub\_sub

Zeromq class to establish a link by sending messages asynchronously (publisher/subscriber)

```
role string Role of the zeromq (Pub/Sub, Request/Reply)
sync string Syncronization ip address
pub string Publisher address
sub string subscriber address
```

# C.2 communicator/zeromq/request\_reply

Zeromq class to establish a link by sending messages synchronously (request/reply)

```
role string Role of the zeromq (Pub/Sub, Request/Reply)
sync string Syncronization ip address
addr string Address
```

#### D Converters

#### D.1 converter/state\_action\_converter

Configurable which is capable of remapping states and actions

```
state_in string Comma-separated list of state elements in the input vector state_out string Comma-separated list of state elements in the output vector action_in string Comma-separated list of action elements observed in the input vector action_out string Comma-separated list of action elements provided in the output vector
```

#### E Discretizers

#### E.1 discretizer/peaked

Peaked discretizer, with more resolution around center

```
min vector Lower limit
max vector Upper limit
steps vector Discretization steps per dimension
peaking vector Extra resolution factor around center (offset by 1/factor at edges)
```

# E.2 discretizer/policy

Returns the action suggested by a policy

```
policy mapping/policy Policy whose action to return
```

#### E.3 discretizer/split

Compound discretizer

```
identify int Identify active discretizer before (-1) or after (1) value discretizer1 discretizer. First discretizer discretizer discretizer. Second discretizer
```

#### E.4 discretizer/uniform

Uniform discretizer

```
min vector Lower limit
max vector Upper limit
steps vector Discretization steps per dimension
```

# F Dynamics

#### F.1 dynamics/acrobot

Acrobot dynamics

#### F.2 dynamics/cart\_double\_pole

Cart-double-pole dynamics from Zhong and Rock

#### F.3 dynamics/cart\_pole

Cart-pole dynamics from Barto et al.

# F.4 dynamics/flyer2d

2D flyer dynamics

#### F.5 dynamics/mountain

Mountain world dynamics

mass double Car mass gravity double Gravitational acceleration

friction double Coefficient of viscous friction between car and ground

stiffness double Spring constant of walls

map mapping/puddle Height map

#### F.6 dynamics/pendulum

Pendulum dynamics based on the DCSC MOPS

#### F.7 dynamics/rbdl

RBDL rigid body dynamics

file string RBDL Lua model file

options string Lua string to execute when loading model

points string Points

auxiliary string Model mass(mm), Center of mass (com), Center of mass velocity (comv), Angular momer

# F.8 dynamics/swimmer

Coulom's swimmer dynamics

segments double.swimmer/segments Number of swimmer segments

#### F.9 dynamics/tlm

Two-link manipulator dynamics

#### G Environments

#### G.1 environment/communicator

Communicator environment which interects with a real environment by sending and receiving messages

converter converter Convert states and actions if needed

communicator communicator Comunicator which exchanges messages with an actual/virtual environment

target\_obs\_dims int Observation dimension of a target target\_action\_dims int Action dimension of a target benchmark\_delays int Observation dimension of a target

# G.2 environment/leo2

LEO/2 environment

port string Device ID of FTDI usb-to-serial converter

bps int Bit rate

Provided parameters

state state Current state of the robot

#### G.3 environment/leo\_squat

Leo squatting environment

behavior behavior Behavior type

xmlstringXML configuration filenametarget\_envenvironmentInteraction environment

observe string Comma-separated list of state elements observed by an agent actuate string Comma-separated list of action elements provided by an agent

exporter exporter Comma-separated list of action elements provided by an agent exporter optional exporter for transition log (supports time, state, observation, a

sub\_transition\_type signal/vector Subscriber to the transition type

pub\_ic\_signal signal/vector Publisher of the initialization and contact signal

measurement\_noise double Additive measurement noise

Provided parameters

observation\_dims int.observation\_dims observation\_min observation\_max action\_dims int.action\_dims int.observation\_max int.action\_dims Number of observation dimensions Upper limit on observations Number of action dimensions

action\_min vector.action\_min Lower limit on actions action\_max vector.action\_max Upper limit on actions

#### G.4 environment/leo\_walk

Leo walking environment

behavior behavior Behavior type

xml string XML configuration filename target\_env environment Interaction environment

observe string Comma-separated list of state elements observed by an agent

actuate string Comma-separated list of action elements provided by an agent exporter optional exporter for transition log (supports time, state, observation, a

sub\_transition\_type signal/vector Subscriber to the transition type

pub\_ic\_signal signal/vector Publisher of the initialization and contact signal

measurement\_noise double Additive measurement noise

Provided parameters

 $observation\_dims$  $int.observation\_dims$ Number of observation dimensions observation\_min vector.observation\_min Lower limit on observations  $observation\_max$ vector.observation\_max Upper limit on observations  $action\_dims$ int.action\_dims Number of action dimensions  $action\_min$ vector.action\_min Lower limit on actions vector.action\_max  $action_{-}max$ Upper limit on actions

#### G.5 environment/modeled

Environment that uses a state transition model internally

model model Environment model

task task Task to perform in the environment (should match model)

exporter exporter Optional exporter for transition log (supports time, state, observation, action, reward,

Provided parameters

state signal/vector Current state of the model

#### G.6 environment/ode

Open Dynamics Engine simulation environment

xml string XML configuration filename randomize int Randomize initial state

visualize int Whether to display 3D visualization

Provided parameters

Number of observation dimensions observation\_dims int.observation\_dims  $observation\_min$ vector.observation\_min Lower limit on observations  $observation\_max$ vector.observation\_max Upper limit on observations action\_dims int.action\_dims Number of action dimensions  $action\_min$ vector.action\_min Lower limit on actions  $action_{-}max$ vector.action\_max Upper limit on actions  $reward_min$ double.reward\_min Lower limit on immediate reward  $reward_max$ double.reward\_max Upper limit on immediate reward

#### G.7environment/pre/noise

Injects noise into an environment

environment environment Environment to inject noise into sensor\_noise vector Additive sensor noise standard deviation  $actuator\_noise$ Additive actuator noise standard deviation vector

#### G.8environment/pre/shaping

Adds reward shaping to an environment

Environment to inject noise into environment environment shaping\_function mapping Potential function over states

double gamma Discount factor

#### G.9environment/sandbox

Non-Markov environment

model  $sandbox\_model$ Environment model

Task to perform in the environment (should match model) task

exporter exporter Optional exporter for transition log (supports time, state, observation, action, re

Provided parameters

signal/vector Current state of the model state

#### $\mathbf{H}$ Experiments

#### H.1 experiment/approx\_test

Approximator test experiment (supervised learning)

train\_samples int Number of training samples test\_samples int Number of test samples file string Output file (csv format)

input\_min vector Lower limit for drawing samples input\_max vector Upper limit for drawing samples

projector projector Projector (should match representation)

representation representation Learned representation mapping mapping Function to learn

#### H.2 experiment/batch\_learning

Batch learning experiment using randomly sampled experience

runs int Number of separate learning runs to perform

batches int Number of batches per learning run batch\_size int Number of transitions per batch rate int Test trial control step frequency in Hz

output string Output base filename

model model Model in which the task is set

task task Task to be solved

predictor predictor Learner

test\_agent agent Agent to use in test trials after each batch

observation\_minvector.observation\_minLower limit for observationsobservation\_maxvector.observation\_maxUpper limit for observationsaction\_minvector.action\_minLower limit for actionsaction\_maxvector.action\_maxUpper limit for actions

Provided parameters

state signal/vector Current observed state of the environment

#### H.3 experiment/multi

Run multiple experiments in parallel

instances int Number of experiments to run in parallel

experiment experiment Experiment to run

# H.4 experiment/online\_learning

Interactive learning experiment

runs int Number of separate learning runs to perform trials int Number of episodes per learning run

steps int Number of steps per learning run rate int Control step frequency in Hz

test\_interval int Number of episodes in between test trials

output string Output base filename

environment environment Environment in which the agent acts

agent agent Agent

test\_agent agent Agent to use in test trials load\_file string Load policy filename

save\_every string Save policy to 'output' at the end of event

Provided parameters

state signal/vector Current observed state of the environment action signal/vector Current action applied to the environment

curve signal/vector Learning curve

#### H.5 experiment/rpc/environment

Environment RPC server

port int Listen port

environment environment Environment to interface

# I Exporters

#### I.1 exporter/csv

Comma-separated values exporter

file string Output base filename

fields string Comma-separated list of fields to write

style string Header style variant string Variant to export

enabled int Enable writing to output file

# J Importers

#### m J.1 - importer/csv

Comma-separated values importer

file string Input base filename

fields string Comma-separated list of fields to read

# **K** Mappings

#### K.1 mapping/displacement

Mapping that returns the state displacement effected by a policy

policy mapping/policy Policy for which displacement is calculated model observation\_model Observation model on which policy acts

#### K.2 mapping/multisine

Sum of sines mapping

inputs int Number of input dimensions outputs int Number of output dimensions

sines int Number of sines

#### K.3 mapping/policy/action

Policy based on a direct action representation

sigma vector Standard deviation of exploration distribution

output\_min vector.action\_min Lower limit on outputs output\_max vector.action\_max Upper limit on outputs

projector projector.observation Projects observations onto representation space

representation representation.action Action representation

#### K.4 mapping/policy/action\_probability

Policy based on an action-probability representation

discretizer discretizer Action discretizer

projector projector Projects observation-action pairs onto representation space

representation representation Action-probability representation

#### K.5 mapping/policy/discrete/random

Policy that chooses discrete random actions

discretizer discretizer.action Action discretizer

#### K.6 mapping/policy/feed\_forward

Feed-forward policy

controls mapping Maps time to controls

# K.7 mapping/policy/mcts

Monte-Carlo Tree Search policy

model	$observation\_model$	Observation model used for planning
discretizer	discretizer.action	Action discretizer
gamma	double	Discount rate
epsilon	double	Exploration rate
horizon	int	Planning horizon
budget	double	Computational budget

# K.8 mapping/policy/parameterized/action

Parameterized policy based on a direct action representation

sigma	vector	Standard deviation of exploration distribution
$output\_min$	vector.action_min	Lower limit on outputs
$output\_max$	vector.action_max	Upper limit on outputs
projector	projector.observation	Projects observations onto representation space
representation	representation/parameterized.action	Action representation

# K.9 mapping/policy/parameterized/pid

Parameterized policy based on a proportional-integral-derivative controller

$\operatorname{setpoint}$	vector	Setpoint
outputs	$int.action\_dims$	Number of outputs
p	vector	P gains ([out1_in1,, out1_inN,, outN_in1,, outN_inN])
i	vector	I gains
d	vector	D gains (use P gain on velocity instead, if available)
il	vector	Integration limits
action_min	$vector.action\_min$	Lower limit on actions
action_max	vector.action_max	Upper limit on actions

# K.10 mapping/policy/parameterized/pidt

Parameterized policy based on a proportional-integral-derivative controller for trajectory tracking

trajectory	mapping	Maps time to setpoints
inputs	$int.observation\_dims$	Number of inputs
outputs	$int.action\_dims$	Number of outputs
p	vector	P gains ([out1_in1,, out1_inN,, outN_in1,, outN_inN])
i	vector	I gains
d	vector	D gains (use P gain on velocity instead, if available)
il	vector	Integration limits
$action\_min$	vector.action_min	Lower limit on actions
$action\_max$	$vector.action\_max$	Upper limit on actions

#### K.11 mapping/policy/parameterized/state\_feedback

Parameterized policy based on a state feedback controller

operating\_state vector Operating state around which gains are defined operating\_action vector Operating action around which gains are defined

gains vector Gains ([in1\_out1, ..., in1\_outN, ..., inN\_out1, ..., inN\_outN])

output\_min vector.action\_min Lower action limit output\_max vector.action\_max Upper action limit

#### K.12 mapping/policy/post/noise

Postprocesses policy output by injecting noise

sigma vector Standard deviation of Gaussian exploration distribution theta vector Ornstein-Uhlenbeck friction term (1=pure Gaussian noise)

policy mapping/policy Policy to inject noise into

#### K.13 mapping/policy/random

Policy that chooses continuous random actions

output\_min vector.action\_min Lower action limit output\_max vector.action\_max Upper action limit

#### K.14 mapping/policy/sample\_feedback

Policy based on state feedback controller defined over samples

output\_min vector.action\_min Lower action limit output\_max vector.action\_max Upper action limit

#### K.15 mapping/policy/uct

Monte-Carlo Tree Search policy using UCB1 action selection

model observation\_model Observation model used for planning discretizer.action Action discretizer discretizer gamma double Discount rate epsilon double Exploration rate horizon Planning horizon int budget double Computational budget

#### K.16 mapping/policy/value/q

Q-value based policy

discretizer discretizer.action Action discretizer

projector projector.pair Projects observation-action pairs onto representation space

representation representation.value/action Action-value representation

sampler sampler Samples actions from action-values

#### K.17 mapping/policy/value/q/bounded

Q-value based policy with bounded action deltas

bound vector Maximum action delta discretizer discretizer.action Action discretizer

projector projector.pair Projects observation-action pairs onto representation space

representation representation.value/action Action-value representation

sampler sampler Samples actions from action-values

## K.18 mapping/policy/value/q/ucb

UCB1 policy

discretizer discretizer.action Action discretizer

projector projector.pair Projects observation-action pairs onto representation space

representation representation.value/action Q-value representation visit\_representation representation.value/action Visit count representation uc\_p double UCB1 exploration term

#### K.19 mapping/policy/value/v

State-value based policy

gamma double Discount rate discretizer discretizer.action Action discretizer model observation\_model Observation model

projector projector.observation Projects observations onto representation space

representation representation.value/state State-value representation

sampler sampler Samples actions from state-values

#### K.20 mapping/puddle

Random 2D puddles

seed int World seed

smoothing double Standard deviation of Gaussian filter steepness double Parameter of sigmoid stretching

# K.21 mapping/represented

A mapping that internally uses a representation

projector projector Projects inputs onto representation space

representation representation Representation

#### K.22 mapping/timeline

Imported timeline mapping

importer importer.dynamic Importer with time as the first column

#### K.23 mapping/value

Mapping that returns the expected value of a value-based policy

policy mapping/policy/value Value based policy

#### L Models

## L.1 model/compass\_walker

Simplest walker model from Garcia et al.

 ${\it control\_step} \qquad {\it double.control\_step} \quad {\it Control step time}$ 

integration\_steps int Number of integration steps per control step

slope\_angle double.slope\_angle Inclination of the slope

#### L.2 model/dynamical

State transition model that integrates equations of motion

control\_step double.control\_step Control step time

integration\_steps int Number of integration steps per control step

dynamics dynamics Equations of motion

#### L.3 model/pinball

Model of a ball on a plate

control\_step double.control\_step Control step time

integration\_steps int Number of integration steps per control step

restitution double Coefficient of restitution

radius double Ball radius maze int Maze ID

#### L.4 model/puddle

Puddle world model

drag double Velocity multiplier for puddles

map mapping/puddle Puddle map

#### L.5 model/windy

Sutton & Barto's windy gridworld model

#### M Observation\_models

#### M.1 observation\_model/approximated

Observation model based on observed transitions

jacobian\_step double Step size for Jacobian estimation

control\_step double.control\_step Control step time (0 = estimate using SMDP approximator)

differential vector.differential State dimensions for which to predict deltas

wrapping vector.wrapping Wrapping boundaries
observation\_min vector.observation\_min Lower limit on observations

observation\_max vector.observation\_max Upper limit on observations stddev\_limit double Maximum standard deviation

stddev\_limit double Maximum standard deviation of acceptable predictions, as frac projector projector.pair Projector for transition model (—S—+—A— dimensions) representation representation transition representation for transition model (—S—+2 dimensions)

#### M.2 observation\_model/fixed

Observation model based on known state transition model

jacobian\_step double Step size for Jacobian estimation

model model Environment model

task task Task to perform in the environment (should match model)

#### M.3 observation\_model/fixed\_reward

Observation model based on observed transitions but known task

jacobian\_step double Step size for Jacobian estimation

control\_step double.control\_step Control step time (0 = estimate using SMDP approximator)

differential vector.differential State dimensions for which to predict deltas

wrapping vector.wrapping Wrapping boundaries
observation\_min vector.observation\_min Lower limit on observations

observation\_max vector.observation\_max Upper limit on observations stddev\_limit double Maximum standard deviation of acceptable predictions, as frac

 $projector \quad projector.pair \quad Projector \ for \ transition \ model \ (-S-+-A-dimensions)$ 

 $representation \qquad representation. transition \qquad Representation for transition \ model \ (--S-+2 \ dimensions)$ 

task task Task to perform in the environment

# N Optimizers

#### N.1 optimizer/cma

Coverance matrix adaptation black-box optimizer

population int Population size

sigma vector Initial standard deviation (a single-element vector will be repli

policy mapping/policy/parameterized Control policy prototype

#### N.2 optimizer/rwa

Reward weighted averaging black-box optimizer

mu int Parent population size lambda int Offspring population size

sigma vector Standard deviation of exploration

policy mapping/policy/parameterized Control policy prototype

#### O Predictors

#### O.1 predictor/ac/action

Actor-critic predictor for direct action policies

importer importer.static Optional importer for pre-training

exporter exporter Optional exporter for transition log (supports observation

alpha double Critic learning rate double Actor learning rate beta double Discount rate gamma lambda double Trace decay rate update\_method Actor update method string step\_limit vector Actor exploration step limit

critic\_projector projector.observation Projects observations onto critic representation space

critic\_representation representation.value/state Value function representation

critic\_trace trace Trace of critic projections

actor\_projector projector.observation Projects observations onto actor representation space

actor\_representation representation.action Action representation
actor\_trace trace Trace of actor projections

#### O.2 predictor/ac/probability

Actor-critic predictor for action-probability policies

importer importer.static Optional importer for pre-training

exporter exporter Optional exporter for transition log (supports observation

alpha double Critic learning rate
beta double Actor learning rate
gamma double Discount rate
lambda double Trace decay rate

critic\_projector projector.observation Projects observations onto critic representation space

critic\_representation representation.value/state Value function representation critic\_trace trace Value function representation Trace of critic projections

critic\_trace trace Trace of critic projections
actor\_projector projector.pair Projects observation-action pairs onto actor representation

actor\_representation representation.value/action Action-probability representation

actor\_trace trace Trace of actor projections

discretizer discretizer.action Action discretizer

#### O.3 predictor/ac/q

Actor-critic predictor for direct action policies with a Q-value based critic

importer importer.static Optional importer for pre-training

exporter exporter Optional exporter for transition log (supports observation alpha double Critic learning rate beta double Actor learning rate gamma double Discount rate

lambdadoubleTrace decay ratekappadoubleAdvantage scaling factorupdate\_methodstringActor update methodstep\_limitvectorActor exploration step limittargetmappingTarget value at next state

critic\_projector projector.pair Projects observations onto critic representation space

critic\_representation representation.value/action Value function representation critic\_trace trace Trace of critic projections

actor\_projector projector.observation Projects observations onto actor representation space

actor\_representation representation.action Action representation
actor\_trace trace Trace of actor projections

#### O.4 predictor/ac/qv

Actor-critic predictor for direct action policies with a Q critic storing advantages over a V critic

importer importer.static Optional importer for pre-training exporter exporter Optional exporter for transition log (supports observat alpha double Critic Q learning rate double Critic V learning rate beta\_v beta\_a double Actor learning rate gamma double Discount rate lambda double Trace decay rate update\_method string Actor update method Actor exploration step limit step\_limit vector critic\_q\_projector projector.pair Projects observations onto critic Q representation space  $critic\_q\_representation$ representation.value/action Q Value function representation critic\_v\_projector projector.observation Projects observations onto critic V representation space  $critic\_v\_representation$ representation.value/state V Value function representation  $critic\_v\_trace$ Trace of critic V projections trace projector.observation Projects observations onto actor representation space  $actor\_projector$ actor\_representation representation.action Action representation Trace of actor projections actor\_trace trace

#### O.5 predictor/advantage

Advantage learning off-policy value function predictor

importer	importer.static	Optional importer for pre-training
exporter	exporter	Optional exporter for transition log (supports observation, acti
alpha	double	Learning rate
gamma	double	Discount rate
lambda	double	Trace decay rate
kappa	double	Advantage scaling factor
discretizer	discretizer.action	Action discretizer
projector	projector.pair	Projects observation-action pairs onto representation space
representation	representation.value/action	A-value representation
trace	trace	Trace of projections

#### O.6 predictor/dpg

Deterministic policy gradient predictor

importer importer.static Optional importer for pre-training

exporter exporter Optional exporter for transition log (supports observe

alpha double Advantage model learning rate

beta\_v double Critic learning rate
beta\_a double Actor learning rate
gamma double Discount rate
lambda double Trace decay rate

projector projector.state Projects observations onto representation spaces

critic\_representation representation.value/state State value function representation

critic\_trace trace Trace of critic projections

advantage\_representation representation Local advantage model representation (one output pe

actor\_representation representation Action representation

#### O.7 predictor/expected\_sarsa

Expected SARSA low-variance on-policy value function predictor

importer importer.static Optional importer for pre-training

exporter exporter Optional exporter for transition log (supports observation, acti

alpha double Learning rate gamma double Discount rate lambda double Trace decay rate

projector projector.pair Projects observation-action pairs onto representation space

representation representation.value/action Q-value representation policy mapping/policy/value Value based target policy trace trace Trace of projections

#### O.8 predictor/fqi

Fitted Q-iteration predictor

importer importer.static Optional importer for pre-training

exporter exporter Optional exporter for transition log (supports observation, a

gamma double Discount rate

transitions int Maximum number of transitions to store

iterations int Number of policy improvement rounds per episode

reset\_strategy string At which point to reset the representation

macro\_batch\_size int Number of episodes/batches after which prediction is rebuilt

discretizer discretizer.action Action discretizer

projector projector.pair Projects observations onto critic representation space representation representation.value/action Value function representation

## O.9 predictor/full/qi

Deterministic model-based action-value function predictor

importer importer.static Optional importer for pre-training

exporter exporter Optional exporter for transition log (supports observation, acti

gamma double Discount rate

model observation\_model Observation model used for planning

discretizer discretizer.action Action discretizer

projector projector.pair Projects observation-action pairs onto representation space

representation representation.value/action Action-value function representation

#### O.10 predictor/full/vi

Deterministic model-based state-value function predictor

importer importer.static Optional importer for pre-training

exporter exporter Optional exporter for transition log (supports observation, actio

gamma double Discount rate

model observation\_model Observation model used for planning

discretizer discretizer.action Action discretizer

projector projector.observation Projects observations onto representation space

representation representation.value/state State-value function representation

#### O.11 predictor/ggq

Greedy-GQ off-policy value function predictor

importer importer.static Optional importer for pre-training

exporter exporter Optional exporter for transition log (supports observation, acti

alpha double Learning rate

eta double Relative secondary learning rate (actual is alpha\*eta)

gamma double Discount rate

projector projector.pair Projects observation-action pairs onto representation space

representation representation.value/action (Q, w) representation policy mapping/policy/value Greedy target policy

#### O.12 predictor/mbfqi

Minibatch FQI predictor

importer importer.static Optional importer for pre-training

exporter exporter Optional exporter for transition log (supports observation, ac

gamma double Discount rate

transitions int Maximum number of transitions to store
minibatch\_size int Number of transitions to average gradient over.
update\_interval int Number of minibatches between target updates.

discretizer discretizer Action discretizer

projector projector.pair Projects observations onto critic representation space

representation representation.value/action Value function representation

#### O.13 predictor/model

Observation model predictor

importer importer.static Optional importer for pre-training

exporter exporter Optional exporter for transition log (supports observation, action

differential vector.differential State dimensions for which to predict deltas

wrapping vector.wrapping Wrapping boundaries

 $\begin{array}{lll} \text{projector} & \text{projector.pair} & \text{Projector for transition model } (-S-+-A-\text{dimensions}) \\ \text{representation} & \text{representation.transition} & \text{Representation for transition model } (-S-+2\text{dimensions}) \\ \end{array}$ 

#### O.14 predictor/multi

Updates multiple predictors

predictor1 predictor First downstream predictor predictor2 predictor Second downstream predictor

#### O.15 predictor/qv

QV on-policy value function predictor

importer importer.static Optional importer for pre-training

exporter exporter Optional exporter for transition log (supports observation, ac

alpha double State-action value learning rate

beta double State value learning rate

gamma double Discount rate lambda double Trace decay rate

q\_projector projector.pair Projects observation-action pairs onto representation space

q\_representation representation.value/action State-action value representation (Q)

v\_projector projector.observation Projects observations onto representation space

v\_representation representation.value/state State value representation (V)

trace trace Trace of projections

#### O.16 predictor/sarsa

SARSA on-policy value function predictor

importer importer.static Optional importer for pre-training

exporter exporter Optional exporter for transition log (supports observation, acti

alpha double Learning rate gamma double Discount rate lambda double Trace decay rate

projector projector.pair Projects observation-action pairs onto representation space

representation representation.value/action Q-value representation trace Trace of projections

#### O.17 predictor/td

TD value function predictor

importer importer.static Optional importer for pre-training

exporter exporter Optional exporter for transition log (supports observation, actio

alpha double Learning rate
gamma double Discount rate
lambda double Trace decay rate

projector projector.observation Projects observations onto representation space

representation representation.value/state State value representation trace Trace of projections

# P Projectors

#### P.1 projector/fourier

Fourier basis function projector

input\_min vector Lower input dimension limit (for scaling)
input\_max vector Upper input dimension limit (for scaling)
order int Order of approximation (bases per dimension)

parity string Whether to use odd or even bases

Provided parameters

memory int.memory Feature vector size

#### P.2 projector/grid/index

Discretizes continuous input to a linear grid index

discretizer discretizer Discretizer

Provided parameters

memory int.memory Grid size

#### P.3 projector/grid/position

Discretizes continuous input to a grid center position

discretizer discretizer Discretizer

Provided parameters

memory int.memory Grid size

#### P.4 projector/identity

Simply returns the input vector

## P.5 projector/monomial

Monomial basis function projector

operating\_input vector Origin

degree int Maximum degree of monomials

Provided parameters

memory int.memory Feature vector size

#### P.6 projector/multi

Combines multiple projections

dim int Indicator dimension (-1=union)
projector1 projector. First downstream projector
projector2 projector. Second downstream projector
memories vector.memory Memory of downstream projectors

Provided parameters

memory int.memory Feature vector size

#### P.7 projector/pre/normalizing

Preprocesses projection onto a normalized [0, 1] vector

input\_min vector Lower input dimension limit (for scaling) input\_max vector Upper input dimension limit (for scaling)

projector projector. Downstream projector

#### P.8 projector/pre/peaked

Preprocesses projection for more resolution around center

peaking vector Extra resolution factor around center (offset by 1/factor at edges)

input\_min vector Lower input dimension limit (for scaling) input\_max vector Upper input dimension limit (for scaling)

projector projector. Downstream projector

#### P.9 projector/pre/scaling

Preprocesses projection onto a scaled vector

scaling vector Scaling vector

projector projector. Downstream projector

#### P.10 projector/rbf

Projection on a grid of triangular radial basis functions

input\_min vector Lower input dimension limit input\_max vector Upper input dimension limit steps vector Basis functions per dimension

Provided parameters

memory int.memory RBF size

#### P.11 projector/sample/ann

Projects onto samples found through approximate nearest-neighbor search

Maximum number of samples to store samples int neighbors int Number of neighbors to return locality Locality of weighing function double interval Samples to accumulate before rebuilding kd-tree int incremental int Search samples that haven't been indexed yet bucket\_size int error\_bound double Number of input dimensions inputs int

# P.12 projector/sample/ertree

Projects onto samples found through the Extra-trees algorithm by Geurts et al.

samples Maximum number of samples to store int Number of trees in the forest  ${\rm trees}$ int splits Number of candidate splits Maximum number of samples in a leaf leaf\_size int inputs Number of input dimensions int Number of output dimensions outputs int

#### P.13 projector/split

Splits a feature vector into distinct sets

index vector Binary vector that specifies which dimensions to use as index discretizer Determines the distinct set based on the index dimensions projector projector projector int.memory Memory of downstream projector

Provided parameters

memory int.memory Resulting feature vector size

## P.14 projector/tile\_coding

Hashed tile coding projector

tilings int Number of tilings memory int.memory Hash table size

safe int Collision detection (0=off, 1=claim on write, 2=claim always)

resolution vector Size of a single tile

wrapping vector.wrapping Wrapping boundaries (must be multiple of resolution)

# Q Random\_generators

## Q.1 random\_generator/normal

Normal Random generator

mu double Mean

sigma double Standart deviation

#### Q.2 random\_generator/ornstein\_uhlenbeck

Ornstein-Uhlenbeck Random generator

center double Attraction point

theta double Theta sigma double Sigma

#### Q.3 random\_generator/uniform

Uniform Random generator

lower double Lower bound of an interval upper double Upper bound of an interval

## Q.4 random\_generator/uniform\_integer

Uniform Integer Random generator

ma int Upper bound of an interval [0, ma)

## R Representations

#### R.1 representation/additive

Linear combination of two representations

representation representation. First representation representation. Second representation

learning int Which representation to learn (0=both)

#### R.2 representation/communicator

Interface to an out-of-process representation

inputs int Number of input dimensions outputs int Number of output dimensions

communicator communicator Communicator which exchanges messages with the out-of-process representa

#### R.3 representation/dictionary

Stores examples as key-value pairs in a dictionary

## R.4 representation/iterative

Representation that iteratively trains a sub-representation

epochs int Learning epochs

cumulative int Add to training set instead of replacing it

representation representation. Downstream representation

#### R.5 representation/llr

Performs locally linear regression through samples

ridge double Ridge regression (Tikhonov) factor

order int Order of regression model

input\_nominals vector Vector indicating which input dimensions are nominal output\_nominals vector Vector indicating which output dimensions are nominal

outputs int Number of output dimensions

 $\begin{array}{cccc} output\_min & vector & Lower output limit \\ output\_max & vector & Upper output limit \\ \end{array}$ 

projector projector/sample Projector used to generate input for this representation

#### R.6 representation/parameterized/ann

Parameterized artificial neural network representation

inputs int Number of input dimensions outputs int Number of output dimensions hiddens vector Number of hidden nodes per layer

eta double Learning rate (0=RPROP, i0=RMSPROP)

## R.7 representation/parameterized/linear

Linear-in-parameters representation

init\_min vector Lower initial value limit Upper initial value limit init\_max vector memory Feature vector size int.memory outputs Number of outputs output\_min vector Lower output limit output\_max vector Upper output limit

# S Samplers

rand\_max

#### S.1 sampler/ac\_ornstein\_ohlenbeck

int

 $\label{lem:correlated} Action-correlated \ maximum \ search \ with \ an \ Ornstein-Uhlenbeck \ random \ chance \ of \ non-maximums$ 

discretizer discretizer.action Action discretizer theta vector Theta parameter of Ornstein-Uhlenbeck sigma Sigma parameter of Ornstein-Uhlenbeck vector center vector Centering parameter of Ornstein-Uhlenbeck signal/vector Publisher and subscriber to the value of noise (or action in the ACOU pub\_sub\_ou\_state epsilon double Exploration rate

In case of multiple maximum values select a random index among to

#### S.2 sampler/epsilon\_greedy

Maximum search with a uniform random chance of non-maximums

rand\_max int In case of multiple maximumum values select a random index among them epsilon vector Exploration rate (can be defined per action)

#### S.3 sampler/epsilon\_ornstein\_ohlenbeck

Exploitations are done by greedy action selection without constraints, as in egreedy. Explorations are done with time-correlated noise, as it is in OU. rand\_max int In case of multiple maximumum values select a random index among t

discretizer discretizer.action Action discretizer

theta vector Theta parameter of Ornstein-Uhlenbeck sigma vector Sigma parameter of Ornstein-Uhlenbeck center vector Centering parameter of Ornstein-Uhlenbeck

pub\_sub\_ou\_state signal/vector Publisher and subscriber to the value of noise (or action in the ACOU

epsilon double Exploration rate

#### S.4 sampler/epsilon\_pada

exploitations are done by greedy action selection without constraints, as in egreedy. Explorations are done with constrained set of actions, as it is in pada.

rand\_max int In case of multiple maximumum values select a random index amon

epsilon vector Exploration rate (can be defined per action)

discretizer discretizer.action Action discretizer delta vector Delta of PADA

pub\_sub\_pada\_state signal/vector Publisher and subscriber to the value of action of the PADA familiy

#### S.5 sampler/greedy

Maximum search

rand\_max int In case of multiple maximumum values select a random index among them

#### S.6 sampler/leo/action

Wrapper for an action sampler for Leo (can modify memory of samplers with memory at contact events)

sampler sampler Samples actions from action-values

sub\_ic\_signal signal/vector Subscrider to the initialization and contact signal

pub\_sub\_sampler\_state signal/vector Publisher and subscriber of the sampler state with memory such as p

#### S.7 sampler/ornstein\_ohlenbeck

Maximum search with an Ornstein-Uhlenbeck random chance of non-maximums

rand\_max int In case of multiple maximumum values select a random index among t

discretizer discretizer.action Action discretizer

theta vector Theta parameter of Ornstein-Uhlenbeck sigma vector Sigma parameter of Ornstein-Uhlenbeck center vector Centering parameter of Ornstein-Uhlenbeck

pub\_sub\_ou\_state signal/vector Publisher and subscriber to the value of noise (or action in the ACOU

## S.8 sampler/pada

Maximum search with a PADA random chance of non-maximums

rand\_max int In case of multiple maximumum values select a random index amon

epsilon vector Exploration rate (can be defined per action)

discretizer discretizer.action Action discretizer delta vector Delta of PADA

pub\_sub\_pada\_state signal/vector Publisher and subscriber to the value of action of the PADA familiy

#### S.9 sampler/pada\_ornstein\_ohlenbeck

Explorations and exploitations are same as OU, but action is selected from a constrained set, as in PADA.

rand\_max int In case of multiple maximumum values select a random index among t

discretizer discretizer.action Action discretizer

theta vector Theta parameter of Ornstein-Uhlenbeck sigma vector Sigma parameter of Ornstein-Uhlenbeck center vector Centering parameter of Ornstein-Uhlenbeck

pub\_sub\_ou\_state signal/vector Publisher and subscriber to the value of noise (or action in the ACOU

pada sampler Pada sampler

pub\_new\_action signal/vector Publisher of the signal with noise

#### S.10 sampler/softmax

Softmax (Gibbs/Boltzmann) sampler

tau double Temperature of Boltzmann distribution

#### T Sandbox\_models

#### T.1 sandbox\_model/compass\_walker

Simplest walker model from Garcia et al. with a sequential evaluation

control\_step double.control\_step Control step time

integration\_steps int Number of integration steps per control step

slope\_angle double.slope\_angle Inclination of the slope

exporter exporter Optional exporter for transition log (supports time, state, observation)

use\_avg\_velocity int Velocity type

#### T.2 sandbox\_model/leo\_squatting

State transition model that integrates equations of motion and augments state vector with additional elements

control\_step double.control\_step Control step time

integration\_steps int Number of integration steps per control step

dynamics dynamics/rbdl Equations of motion

target\_dof int.target\_dof Number of degrees of freedom of the target model

animation string Save current state or full animation

target\_env environment Interaction environment

lower\_height double.lower\_height Lower bound of root height to switch direction upper\_height Upper bound of root height to switch direction

# U Signals

## U.1 signal/matrix

Matrix-based signal (trajectory, etc.)

## U.2 signal/vector

Vector-based signal (state, observation, etc.)

#### V Solvers

#### V.1 solver/agent

Solver that uses a simulated agent

steps int Number of planning steps before solution is returned

horizon int Planning episode length start vector Starting state for planning

model observation\_model Observation model used for planning agent Agent used for planning episodes

Provided parameters

state state Current observed state of planning

#### V.2 solver/ilqg

Iterative Linear Quadratic Gaussian trajectory optimizer

horizon int Horizon

iterations int Maximum number of iterations

stddev vector Standard deviation of initial random action sequence

regularization string Regularization method model observation\_model Observation model

policy mapping/policy/sample\_feedback Sample feedback policy to adjust

#### Provided parameters

signal/matrix Predicted trajectory

#### V.3 solver/lqr

Linear Quadratic Regulator solver

operating\_state Operating state around which to linearize vector  $operating\_action$ vector Operating action around which to linearize

Observation model model  $observation\_model$ 

policy mapping/policy/parameterized/state\_feedback State feedback policy to adjust

#### V.4 solver/vi

Value iteration solver

sweeps int Number of planning sweeps before solution is returned

parallel Perform backups in parallel (requires reentrant representation)

discretizer discretizer.observation State space discretizer predictor predictor/full Predictor to iterate

#### $\mathbf{W}$ Tasks

## task/acrobot/balancing

Acrobot balancing task

Provided parameters

 $observation\_dims$  $int.observation\_dims$ Number of observation dimensions observation\_min vector.observation\_min Lower limit on observations  $observation\_max$ vector.observation\_max Upper limit on observations action\_dims int.action\_dims Number of action dimensions action\_min vector.action\_min Lower limit on actions  $action_max$ vector.action\_max Upper limit on actions Lower limit on immediate reward  $reward\_min$  $double.reward\_min$ 

 $reward_max$ double.reward\_max Upper limit on immediate reward

#### W.2task/acrobot/regulator

Acrobot regulator task

start vector Starting state goal vector Goal state

 $\begin{array}{cccc} stddev & vector & Starting \ state \ standard \ deviation \\ q & vector & Q \ (state \ cost) \ matrix \ diagonal \\ r & vector & R \ (action \ cost) \ matrix \ diagonal \end{array}$ 

function string Cost function style

smoothing double Cost function smoothing parameter

#### Provided parameters

$observation\_dims$	int.observation_dims	Number of observation dimensions
$observation\_min$	vector.observation_min	Lower limit on observations
$observation\_max$	$vector.observation\_max$	Upper limit on observations
$action\_dims$	$int.action\_dims$	Number of action dimensions
action_min	vector.action_min	Lower limit on actions
$action\_max$	vector.action_max	Upper limit on actions
$\operatorname{reward\_min}$	$double.reward\_min$	Lower limit on immediate reward
$reward\_max$	$double.reward\_max$	Upper limit on immediate reward

# $W.3 \quad task/cart\_double\_pole/balancing$

Cart-double-pole balancing task

timeout double Episode timeout

#### Provided parameters

$observation\_dims$	$int.observation\_dims$	Number of observation dimensions
$observation\_min$	$vector.observation\_min$	Lower limit on observations
$observation\_max$	$vector.observation\_max$	Upper limit on observations
$action\_dims$	$int.action\_dims$	Number of action dimensions
$action\_min$	$vector.action\_min$	Lower limit on actions
$action\_max$	$vector.action\_max$	Upper limit on actions
$\operatorname{reward\_min}$	double.reward_min	Lower limit on immediate reward
$reward\_max$	$double.reward\_max$	Upper limit on immediate reward

# $W.4 \quad task/cart\_double\_pole/regulator$

Cart-double-pole regulator task

 $\operatorname{start}$ Starting state vector goal vector Goal state

stddevvectorStarting state standard deviation Q (state cost) matrix diagonal vector q R (action cost) matrix diagonal vector

Cost function style function string

smoothing Cost function smoothing parameter double

timeout double Episode timeout

#### Provided parameters

$observation\_dims$	$int.observation\_dims$	Number of observation dimensions
$observation\_min$	vector.observation_min	Lower limit on observations
$observation\_max$	$vector.observation\_max$	Upper limit on observations
$action\_dims$	$int.action\_dims$	Number of action dimensions
action_min	vector.action_min	Lower limit on actions
$action\_max$	$vector.action\_max$	Upper limit on actions
$\operatorname{reward\_min}$	$double.reward\_min$	Lower limit on immediate reward
$reward_max$	$double.reward\_max$	Upper limit on immediate reward

#### W.5task/cart\_double\_pole/swingup

Cart-double-pole swing-up task

timeout double Episode timeout

#### Provided parameters

$observation\_dims$	$int.observation\_dims$	Number of observation dimensions
$observation\_min$	$vector.observation\_min$	Lower limit on observations
$observation\_max$	$vector.observation\_max$	Upper limit on observations
$\operatorname{action\_dims}$	$int.action\_dims$	Number of action dimensions
$action\_min$	vector.action_min	Lower limit on actions
$action\_max$	$vector.action\_max$	Upper limit on actions
$\operatorname{reward\_min}$	double.reward_min	Lower limit on immediate reward
$reward_max$	$double.reward\_max$	Upper limit on immediate reward

#### W.6task/cart\_pole/balancing

Cart-pole balancing task

timeout double Episode timeout

 $observation\_dims$  $int.observation\_dims$ Number of observation dimensions  $observation\_min$  $vector.observation\_min$ Lower limit on observations  $observation\_max$  $vector.observation\_max$ Upper limit on observations  $action\_dims$  $int.action\_dims$ Number of action dimensions Lower limit on actions  $action\_min$ vector.action\_min Upper limit on actions action\_max vector.action\_max  $double.reward\_min$ Lower limit on immediate reward  $reward_min$  $reward_max$  $double.reward\_max$ Upper limit on immediate reward

## W.7 task/cart\_pole/regulator

#### Cart-pole regulator task

start	vector	Starting state
goal	vector	Goal state
stddev	vector	Starting state standard deviation
q	vector	Q (state cost) matrix diagonal
r	vector	R (action cost) matrix diagonal
function	string	Cost function style
smoothing	double	Cost function smoothing parameter
timeout	double	Episode timeout

#### Provided parameters

observation_dims	$int.observation\_dims$	Number of observation dimensions
observation_min	vector.observation_min	Lower limit on observations
observation_max	$vector.observation\_max$	Upper limit on observations
action_dims	$int.action\_dims$	Number of action dimensions
action_min	vector.action_min	Lower limit on actions
action_max	$vector.action\_max$	Upper limit on actions
reward_min	$double.reward\_min$	Lower limit on immediate reward
$reward_max$	$double.reward\_max$	Upper limit on immediate reward

#### W.8 task/cart\_pole/swingup

#### Cart-pole swing-up task

timeout double Episode timeout
randomization int Start state randomization
shaping int Whether to use reward shaping
gamma double Discount rate for reward shaping

end\_stop\_penalty int Terminate episode with penalty when end stop is reached

Number of observation dimensions observation\_dims int.observation\_dims  $observation\_min$ vector.observation\_min Lower limit on observations  $observation\_max$ vector.observation\_max Upper limit on observations action\_dims int.action\_dims Number of action dimensions  $action\_min$ vector.action\_min Lower limit on actions  $action_{-}max$ vector.action\_max Upper limit on actions reward\_min double.reward\_min Lower limit on immediate reward  $reward_max$ double.reward\_max Upper limit on immediate reward

#### W.9 task/compass\_walker/vref

Compass walker tracking velocity task

timeout double Learning episode timeout  $initial\_state\_variation$ double Variation of initial state double.slope\_angle Inclination of the slope slope\_angle negative\_reward double Negative reward observe vector State elements observed by an agent steps int number of steps after which task is terminated reference\_velocity double Reference velocity If set, give reward per every step per\_step\_reward int

Provided parameters

 $int.observation\_dims$ Number of observation dimensions  $observation\_dims$ Lower limit on observations observation\_min vector.observation\_min  $observation\_max$ vector.observation\_max Upper limit on observations action\_dims int.action\_dims Number of action dimensions  $action\_min$ vector.action\_min Lower limit on actions  $action_max$ vector.action\_max Upper limit on actions  $double.reward\_min$ Lower limit on immediate reward reward\_min double.reward\_max Upper limit on immediate reward reward\_max

#### W.10 task/compass\_walker/vrefu

Compass walker tracking velocity task with controls minimization

timeout double Learning episode timeout initial\_state\_variation double Variation of initial state slope\_angle double.slope\_angle Inclination of the slope negative\_reward double Negative reward

observe vector State elements observed by an agent

steps int number of steps after which task is terminated

reference\_velocity double Reference velocity

per\_step\_reward int If set, give reward per every step

 $observation\_dims$ Number of observation dimensions int.observation\_dims  $observation\_min$ vector.observation\_min Lower limit on observations  $observation\_max$ vector.observation\_max Upper limit on observations Number of action dimensions action\_dims int.action\_dims  $action\_min$ vector.action\_min Lower limit on actions  $action_{-}max$ vector.action\_max Upper limit on actions reward\_min double.reward\_min Lower limit on immediate reward  $reward_max$ double.reward\_max Upper limit on immediate reward

#### W.11 task/compass\_walker/walk

Compass walker walking task

timeout double Learning episode timeout initial\_state\_variation double Variation of initial state slope\_angle double.slope\_angle Inclination of the slope negative\_reward double Negative reward

observe vector State elements observed by an agent

steps int number of steps after which task is terminated

Provided parameters

observation\_dims int.observation\_dims Number of observation dimensions  $observation\_min$ vector.observation\_min Lower limit on observations  $observation\_max$ vector.observation\_max Upper limit on observations action\_dims int.action\_dims Number of action dimensions action\_min vector.action\_min Lower limit on actions action\_max vector.action\_max Upper limit on actions  $reward\_min$ double.reward\_min Lower limit on immediate reward  $reward_max$ double.reward\_max Upper limit on immediate reward

#### W.12 task/flyer2d/regulator

#### 2D flyer regulator task

start vector Starting state goal vector Goal state

function string Cost function style

smoothing double Cost function smoothing parameter

action\_range double Range of allowed actions

timeout double Episode timeout

Number of observation dimensions observation\_dims int.observation\_dims  $observation\_min$ vector.observation\_min Lower limit on observations  $observation\_max$ vector.observation\_max Upper limit on observations action\_dims int.action\_dims Number of action dimensions  $action\_min$ vector.action\_min Lower limit on actions  $action_{-}max$ vector.action\_max Upper limit on actions reward\_min double.reward\_min Lower limit on immediate reward  $reward_max$ double.reward\_max Upper limit on immediate reward

#### W.13 task/leo\_squatting

Task specification for Leo squatting with a fixed arm

timeout double.timeout Task timeout

rand\_init int.rand\_init Initialization from a random pose

Provided parameters

Number of observation dimensions observation\_dims int.observation\_dims observation\_min vector.observation\_min Lower limit on observations observation\_max vector.observation\_max Upper limit on observations action\_dims int.action\_dims Number of action dimensions  $action\_min$ vector.action\_min Lower limit on actions  $action\_max$ Upper limit on actions vector.action\_max reward\_min double.reward\_min Lower limit on immediate reward  $reward_max$ double.reward\_max Upper limit on immediate reward

#### W.14 task/lua

User-provided task specification in LUA

file string Lua task file

options string Lua string to execute when loading task

Provided parameters

Number of observation dimensions  $observation\_dims$  $int.observation\_dims$ observation\_min vector.observation\_min Lower limit on observations  $observation\_max$ vector.observation max Upper limit on observations action\_dims int.action\_dims Number of action dimensions  $action\_min$ vector.action\_min Lower limit on actions  $action_max$ vector.action\_max Upper limit on actions  $reward\_min$  $double.reward\_min$ Lower limit on immediate reward reward\_max double.reward\_max Upper limit on immediate reward

# W.15 task/mountain/regulator

Mountain world regulator task

start	vector	Starting state
goal	vector	Goal state
stddev	vector	Starting state standard deviation
q	vector	Q (state cost) matrix diagonal
r	vector	R (action cost) matrix diagonal
function	string	Cost function style
smoothing	double	Cost function smoothing parameter

## Provided parameters

$observation\_dims$	$int.observation\_dims$	Number of observation dimensions
$observation\_min$	$vector.observation\_min$	Lower limit on observations
$observation\_max$	$vector.observation\_max$	Upper limit on observations
$action\_dims$	$int.action\_dims$	Number of action dimensions
action_min	$vector.action\_min$	Lower limit on actions
$action\_max$	$vector.action\_max$	Upper limit on actions
$\operatorname{reward\_min}$	double.reward_min	Lower limit on immediate reward
$reward_max$	$double.reward\_max$	Upper limit on immediate reward

# W.16 task/pendulum/regulator

 $Pendulum\ regulator\ task$ 

start	vector	Starting state
goal	vector	Goal state
stddev	vector	Starting state standard deviation
q	vector	Q (state cost) matrix diagonal
r	vector	R (action cost) matrix diagonal
function	string	Cost function style
smoothing	double	Cost function smoothing parameter

$observation\_dims$	$int.observation\_dims$	Number of observation dimensions
$observation\_min$	$vector.observation\_min$	Lower limit on observations
$observation\_max$	$vector.observation\_max$	Upper limit on observations
$\operatorname{action\_dims}$	$int.action\_dims$	Number of action dimensions
$action\_min$	$vector.action\_min$	Lower limit on actions
$action\_max$	$vector.action\_max$	Upper limit on actions
$reward\_min$	double.reward_min	Lower limit on immediate reward
$reward_max$	$double.reward\_max$	Upper limit on immediate reward

#### W.17task/pendulum/swingup

Pendulum swing-up task

Episode timeout timeout double

randomization double Level of start state randomization

Provided parameters

observation\_dims int.observation\_dims observation\_min vector.observation\_min observation\_max vector.observation\_max  $action\_dims$ int.action\_dims  $action\_min$ vector.action\_min action\_max vector.action\_max reward\_min double.reward\_min reward\_max double.reward\_max

Number of observation dimensions Lower limit on observations Upper limit on observations Number of action dimensions Lower limit on actions Upper limit on actions Lower limit on immediate reward

Upper limit on immediate reward

#### W.18task/pinball/movement

Pinball movement task

tolerance double Goal tolerance

Provided parameters

observation\_dims int.observation\_dims observation\_min vector.observation\_min  $observation\_max$ vector.observation\_max  $action_dims$ int.action\_dims action\_min vector.action\_min action max vector.action\_max reward\_min double.reward\_min  $reward_max$ double.reward\_max

Number of observation dimensions Lower limit on observations Upper limit on observations Number of action dimensions Lower limit on actions Upper limit on actions Lower limit on immediate reward

Upper limit on immediate reward

#### W.19task/pinball/regulator

Pinball regulator task

Starting state start vector goal vector Goal state

stddev Starting state standard deviation vector Q (state cost) matrix diagonal vector q vector R (action cost) matrix diagonal

Cost function style function string

smoothing double Cost function smoothing parameter

#### Provided parameters

 $int.observation\_dims$  $observation\_dims$  $observation\_min$ vector.observation\_min  $observation\_max$ vector.observation\_max action\_dims int.action\_dims action\_min vector.action\_min  $action_{-max}$ vector.action\_max reward\_min double.reward\_min  $reward_max$  $double.reward_max$ 

Number of observation dimensions Lower limit on observations Upper limit on observations Number of action dimensions Lower limit on actions Upper limit on actions

Lower limit on immediate reward Upper limit on immediate reward

## W.20 task/puddle/regulator

Puddle world regulator task

start Starting state vector Goal state goal vector stddev Starting state standard deviation vector vector Q (state cost) matrix diagonal q R (action cost) matrix diagonal vector function string Cost function style Cost function smoothing parameter smoothing double double Penalty multiplier for puddles penalty mapping/puddle Puddle map map

#### Provided parameters

Number of observation dimensions observation\_dims int.observation\_dims observation\_min vector.observation\_min Lower limit on observations  $observation\_max$ vector.observation\_max Upper limit on observations  $action_dims$ int.action\_dims Number of action dimensions  $action\_min$ vector.action\_min Lower limit on actions action\_max vector.action\_max Upper limit on actions reward\_min double.reward\_min Lower limit on immediate reward  $reward_max$ double.reward\_max Upper limit on immediate reward

#### W.21 task/swimmer/reaching

Swimmer reaching task

timeout double Episode timeout

randomization double Level of start state randomization segments Number of swimmer segments

$observation\_dims$	$int.observation\_dims$	Number
observation_min	$vector.observation\_min$	Lower lin
$observation\_max$	$vector.observation\_max$	Upper lii
$action\_dims$	$int.action\_dims$	Number
$action\_min$	$vector.action\_min$	Lower lin
$action\_max$	$vector.action\_max$	Upper lii
reward_min	$double.reward\_min$	Lower lin
$reward_max$	$double.reward\_max$	Upper lii

# Number of observation dimensions Lower limit on observations Upper limit on observations Number of action dimensions Lower limit on actions Upper limit on actions Lower limit on immediate reward Upper limit on immediate reward

# $W.22 ext{ } ext{task/tlm/balancing}$

Two-link manipulator balancing task

#### Provided parameters

observation_dims	int.observation_dims	Number of observation dimensions
observation_min	vector.observation_min	Lower limit on observations
observation_max	vector.observation_max	Upper limit on observations
$action\_dims$	$int.action\_dims$	Number of action dimensions
$action\_min$	vector.action_min	Lower limit on actions
$action\_max$	$vector.action\_max$	Upper limit on actions
$\operatorname{reward\_min}$	$double.reward\_min$	Lower limit on immediate reward
$reward_max$	$double.reward\_max$	Upper limit on immediate reward

## W.23 task/windy/movement

Windy gridworld movement task

#### Provided parameters

$observation\_dims$	$int.observation\_dims$	Number of observation dimensions
$observation\_min$	vector.observation_min	Lower limit on observations
$observation\_max$	$vector.observation\_max$	Upper limit on observations
$action\_dims$	$int.action\_dims$	Number of action dimensions
$action\_min$	vector.action_min	Lower limit on actions
$action\_max$	vector.action_max	Upper limit on actions
$\operatorname{reward\_min}$	$double.reward\_min$	Lower limit on immediate reward
$reward\_max$	double.reward_max	Upper limit on immediate reward

# X Traces

## X.1 trace/enumerated/accumulating

Accumulating eligibility trace using a queue of projections

## X.2 trace/enumerated/replacing

Replacing eligibility trace using a queue of projections

## Y Trigges

#### Y.1 trigger

Event trigger

min vector.observation\_min Minimum of compartment bounding box
max vector.observation\_max Maximum of compartment bounding box
delay double Settlement delay for which conditions are continuously fullfilled

#### Z Visualizations

#### Z.1 visualization/acrobot

Acrobot visualization

state signal/vector Acrobot state to visualize

## Z.2 visualization/cart\_double\_pole

Cart-double-pole visualization

state signal/vector Cart-double-pole state to visualize

## Z.3 visualization/cart\_pole

Cart-pole visualization

state signal/vector Cart-pole state to visualize

#### Z.4 visualization/compass\_walker

Compass walker visualization

state signal/vector Compass walker state to visualize

#### Z.5 visualization/field/mapping

Visualizes a mapping over a field of states

field\_dims vector Dimensions to visualize
input\_min vector Lower input dimension limit
input\_max vector Upper input dimension limit
points int Number of points to evaluate

savepoints int Number of points to evaluate when saving to file ('s')

state signal/vector Optional current state to overlay

projection string Method of projecting values onto 2d space

mapping mapping Mapping

output\_dim int Output dimension to visualize

## Z.6 visualization/field/policy/value

Visualizes the value of a policy over a field of states

field\_dims vector Dimensions to visualize
input\_min vector Lower input dimension limit
input\_max vector Upper input dimension limit
points int Number of points to evaluate

savepoints int Number of points to evaluate when saving to file ('s')

state signal/vector Optional current state to overlay

projection string Method of projecting values onto 2d space

policy mapping/policy/value Value based control policy

## Z.7 visualization/field/value

Visualizes an approximation over a field of states

field\_dims vector Dimensions to visualize
input\_min vector Lower input dimension limit
input\_max vector Upper input dimension limit
points int Number of points to evaluate

savepoints int Number of points to evaluate when saving to file ('s')

state signal/vector Optional current state to overlay

projection string Method of projecting values onto 2d space

output\_dim int Output dimension to visualize

projector projector Projects inputs onto representation space

representation representation Value representation

# Z.8 visualization/flyer2d

2D flyer visualization

state signal/vector 2D flyer state to visualize

#### $\mathbf{Z.9}$ visualization/pendulum

Pendulum visualization

signal/vector Pendulum state to visualize state

#### visualization/pinball $\mathbf{Z}.10$

Pinball visualization

state signal/vector Pinball state to visualize

#### Z.11visualization/sample

Visualizes a sample-based approximation

field\_dims Dimensions to visualize vector field\_min vector Lower visualization dimension limit  $field_{max}$ vector Upper visualization dimension limit output\_dim Output dimension to visualize int points Texture size int

Sample projector whose store to visualize projector projector/sample

#### Z.12visualization/sample/random

Visualizes an approximation over randomly sampled states

 $field\_dims$ Dimensions to visualize vector  $input_min$ vector Lower input dimension limit  $input_max$ vector Upper input dimension limit output\_dim Output dimension to visualize int points Texture size int

projector Projects inputs onto representation space projector

Value representation representation representation

#### Z.13visualization/slice

Visualizes a slice from a mapping

field\_dims vector Dimensions to visualize  $input\_min$ vector Lower input dimension limit input\_max vector Upper input dimension limit

operating\_point vector Fixed values for non-visualized dimensions

output\_dim int Output dimension to visualize Number of points to evaluate points int state signal/vector Optional current state to overlay signal/vector Optional current action to overlay action

mapping mapping Mapping to visualize

## Z.14 visualization/state

Plots state values

$input\_dims$	vector	Input dimensions to visualize
$input\_min$	vector	Lower input dimension limit
$input\_max$	vector	Upper input dimension limit
memory	int	Number of data points to draw

state signal/vector State to visualize

#### Z.15 visualization/swimmer

Swimmer visualization

state signal/vector Swimmer state to visualize

## Z.16 visualization/tlm

Two-link manipulator visualization

state signal/vector Two-link manipulator state to visualize

# Z.17 visualization/trajectory

Plots trajectories

$input\_dims$	vector	Input dimensions to visualize
$input\_min$	vector	Lower input dimension limit
$input\_max$	vector	Upper input dimension limit
trajectory	signal/matrix	Trajectory to visualize

## Z.18 visualization/windy

Windy gridworld visualization

state signal/vector Windy gridworld state to visualize

#### AA Visualizers

## AA.1 visualizer/glut

Visualizer based on the GLUT library