# **RLSimion**

# **Project: RLSimion**

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### **API Reference**

- Actor
- ActorCritic
- BhatnagarSchedule
- CACLALearner
- ConfigNode
- DDPG
- <u>DeterministicPolicyGaussianNoise</u>
- <u>DiscreteEpsilonGreedyDeepPolicy</u>
- <u>DiscreteFeatureMap</u>
- <u>DiscreteSoftmaxDeepPolicy</u>
- <u>DoubleQLearning</u>
- DQN
- <u>DynamicModel</u>
- ETraces
- ExperienceReplay
- Experiment
- FeatureList
- FeatureMap
- FunctionSampler
- FunctionSampler2D
- FunctionSampler3D
- <u>GaussianNoise</u>
- GaussianRBFGridFeatureMap
- GreedyQPlusNoisePolicy
- IncrementalNaturalActorCritic
- InterpolatedValue
- <u>LinearStateActionVFA</u>
- <u>LinearStateVFA</u>
- LinearVFA
- <u>Logger</u>
- LQRController
- MemBlock
- OffPolicyActorCritic
- OffPolicyDeterministicActorCritic
- OrnsteinUhlenbeckNoise
- <u>PIDController</u>
- QEGreedyPolicy
- QLearningCritic
- QSoftMaxPolicy

- RegularPolicyGradientLearner
- RewardFunction
- SARSA
- SimGod
- <u>SimionApp</u>
- SimionMemBuffer
- SimionMemPool
- <u>SimpleEpisodeLinearSchedule</u>
- SingleDimensionGrid
- SinusoidalNoise
- StatsInfo
- StochasticGaussianPolicy
- TDCLambdaCritic
- TDLambdaCritic
- <u>TileCodingFeatureMap</u>
- ToleranceRegionReward
- <u>TrueOnlineTDLambdaCritic</u>
- WindTurbineBoukhezzarController
- <u>WindTurbineJonkmanController</u>
- WindTurbineVidalController
- <u>WireConnection</u>
- World

## **Actor**

## **Class Actor**

Source: actor.cpp

### Methods

double selectAction(const State s, Action a)

• Summary

Iterates over the actor's policy learners so that every one determines its output action

- Parameters
  - s: Input initial state
  - a: Output action
- Return Value

Iterates over the actor's policy learners so that every one determines its output action

void update(const State s, const Action a, const State\* s\_p, double r, double td)

Summary

Iterates over all the actor's policy learners so that every one learns from an experience tuple  $\{s,a,s\_p,r,td\}$ 

- Parameters
  - s: Initial state
  - a: Action
  - \_sp: Resultant state
  - r: Reward
  - td: Temporal-Difference error calculated by the critic

# **Actorcritic**

### **Class ActorCritic**

Source: actor-critic.cpp

### Methods

double selectAction(const State s, Action a)

Summary

Objects that implement both an actor and a critic call the actor's selectAction() method

- Parameters
  - s: Initial state
  - a: Output action
- Return Value

Objects that implement both an actor and a critic call the actor's selectAction() method

double update(const State s, const Action a, const State \*s\_p, double r, double behaviorProb)

• Summary

Encapsulates the basic Actor-Critic update: the critic calculates the TD error and the actor updates its policy accordingly

- Parameters
  - s: Initial state
  - a: Action
  - \_sp: Resultant state
  - r: Reward
  - behaviorProb: Probability by which the actor selected the action. Should be ignored
- Return Value

Encapsulates the basic Actor-Critic update: the critic calculates the TD error and the actor updates its policy accordingly

# Bhatnagarschedule

# Class BhatnagarSchedule

Source: parameters-numeric.cpp

### Methods

double get()

• Summary

Implements the schedule function proposed by Bhatnagar

## Caclalearner

## **Class CACLALearner**

Source: actor-cacla.cpp

### Methods

void update(const State s, const Action a, const State \*s\_p, double r, double td)

• Summary

Updates the policy using the CACLA update rule

- Parameters
  - s: Initial state
  - a: Action

- \_sp: Resultant state
- r: Reward
- td: Temporal-Difference error

# Confignode

# **Class ConfigNode**

Source: config.cpp

### Methods

int countChildren(const char\* name)

• Summary

Returns the number of children this node has with the given name

bool getConstBoolean(const char\* paramName, bool defaultValue)

Summary

Retrieves the value of a parameter as a boolean

- Parameters
  - o paramName: The name of the parameter
  - defaultValue: Its default value (will be used if the parameter is not found)
- Return Value

Retrieves the value of a parameter as a boolean

int getConstInteger(const char\* paramName, int defaultValue)

• Summary

Retrieves the value of a parameter as an integer

- Parameters
  - paramName: The name of the parameter
  - o defaultValue: Its default value (will be used if the parameter is not found)
- Return Value

Retrieves the value of a parameter as an integer

double getConstDouble(const char\* paramName, double defaultValue)

• Summary

Retrieves the value of a parameter as a double

- Parameters
  - paramName: The name of the parameter
  - defaultValue: Its default value (will be used if the parameter is not found)
- Return Value

Retrieves the value of a parameter as a double

void saveFile(const char\* pFilename)

• Summary

Saves all the configuration nodes below the current to a file

- Parameters
  - o pFilename: The path to the file

void saveFile(FILE\* pFile)

• Summary

Saves all the configuration nodes below the current to an already open file

Parameters

pFile: The handle to the already open file

### void clone(ConfigFile\* parameterFile)

Summary

Makes a shallow copy of the a configuration file

- Parameters
  - o parameterFile:

# **Ddpg**

### **Class DDPG**

Source: DDPG.cpp

### Methods

double selectAction(const State s, Action a)

• Summary

Implements action selection for the DDPG algorithm adding the output of the actor and exploration noise signal

- Parameters
  - s: State
  - a: Output action

double update(const State s, const Action a, const State \* s\_p, double r, double behaviorProb)

Summary

Updates the critic and actor using the DDPG algorithm

- Parameters
  - s: Initial state
  - a: Action
  - \_sp: Resultant state
  - r: Reward

# **Deterministic policy gaussiannoise**

## Class DeterministicPolicyGaussianNoise

Source: vfa-policy.cpp

### Methods

double selectAction(const State s, Action a)

• Summary

A Deterministic Policy Gaussian Noise policy uses a single function representing the deterministic output. In training episodes, the noise signal used for exploration is sampled and added to the deterministic output

- Parameters
  - s: Initial state
  - a: Output state

void getFeatures(const State state, FeatureList outFeatureList)

• Summary

Uses the policy's feature map to return the features representing the state

- Parameters
  - state: State
  - outFeatureList: Output feature list

# Discreteepsilongreedydeeppolicy

# Class DiscreteEpsilonGreedyDeepPolicy

Source: deep-vfa-policy.cpp

### Methods

int selectAction(const std::vector& values)

Summary

Deep RL version of the epsilon-greedy action selection algorithm

- Parameters
  - values: Estimated Q(s,a) for each discrete action. Size should equal the number of discrete actions
- Return Value

Deep RL version of the epsilon-greedy action selection algorithm

# **Discretefeaturemap**

# Class DiscreteFeatureMap

Source: featuremap-discrete.cpp

### Methods

void map(vector>& grids, const vector& values, FeatureList outFeatures)

Summary

Implements a feature mapping function that maps state-actions to boxes. Only one feature will be active

- Parameters
  - $\circ \;\; \textit{grids} \text{:} \; \text{Input grids for every state-variable used}$
  - values: The values of every state-variable used
  - outFeatures: The output list of features

void unmap(size\_t feature, vector& grids, vector& outValues)

Summary

Inverse of the feature mapping operation. Given a feature it returns the state-action to which it corresponds.

- Parameters
  - feature: The index of the feature
  - grids: The set of grids used to discretize each variable
  - $\circ~\mbox{\it outValues}$  . The set of output values for every state-action variable

# **Discretesoftmaxdeeppolicy**

# Class DiscreteSoftmaxDeepPolicy

Source: deep-vfa-policy.cpp

### Methods

int selectAction(const std::vector& values)

Summary

Deep-RL version of the Soft-Max action selection policy

- Parameters
  - $\circ \ \ \textit{values} \text{: Estimated Q(s,a) for each discrete action. Size should equal the number of discrete actions}$
- Return Value

Deep-RL version of the Soft-Max action selection policy

# **Doubleqlearning**

## **Class DoubleQLearning**

Source: q-learners.cpp

### Methods

double update(const State s, const Action a, const State \*s\_p, double r, double probability)

Summary

Updates the estimate of the Q-function using the Double Q-Learning update rule with tuple {s,a,s\_p,r}. The main difference with respect to Q-function is that it uses two different sets of weights for the function, updating a random set of weights toward the other set of weights. Should offer better stability than regular Q-Learning

- Parameters
  - s: Initial state
  - a: Action
  - \_sp: Resultant state
  - r: Reward
- Return Value

Updates the estimate of the Q-function using the Double Q-Learning update rule with tuple {s,a,s\_p,r}. The main difference with respect to Q-function is that it uses two different sets of weights for the function, updating a random set of weights toward the other set of weights. Should offer better stability than regular Q-Learning

# Dqn

### **Class DQN**

Source: DQN.cpp

## Methods

double selectAction(const State s, Action a)

Summary

Implements the action selection algorithm for a Q-based Deep RL algorithm

- Parameters
  - s: State
  - a: Output action

double update(const State s, const Action a, const State \* s\_p, double r, double behaviorProb)

Summary

Implements DQL algorithm update using only one Neural Network for both evaluation and update

- Parameters
  - s: Initial state
  - a: Action
  - \_sp: Resultant state
  - r: Reward

# **Dynamicmodel**

# Class DynamicModel

Source: world.cpp

### Methods

size\_t addStateVariable(const char name, const char units, double min, double max, bool bCircular)

This method must be called from the constructor of DynamicModel subclasses to register state variables. Calls are parsed by the source code parser and listed beside the parameters of the class in the class definition file (config.xml)

#### Parameters

- name: Name of the variable (i.e. "speed")
- o units: Metrical unit (i.e., "m/s")
- o min: Minimum value this variable may get. Below this, values are clamped
- o max: Maximum value this variable may get. Above this, values are clamped
- bCircular: This flag indicates whether the variable is circular (as angles)

### Return Value

This method must be called from the constructor of DynamicModel subclasses to register state variables. Calls are parsed by the source code parser and listed beside the parameters of the class in the class definition file (config.xml)

size\_t addActionVariable(const char name, const char units, double min, double max, bool bCircular)

#### Summary

This method must be called from the constructor of DynamicModel subclasses to register action variables. Calls are parsed by the source code parser and listed beside the parameters of the class in the class definition file (config.xml)

#### • Parameters

- name: Name of the variable (i.e. "speed")
- o units: Metrical unit (i.e., "m/s")
- o min: Minimum value this variable may get. Below this, values are clamped
- o max. Maximum value this variable may get. Above this, values are clamped
- o bCircular. This flag indicates whether the variable is circular (as angles)

#### Return Value

This method must be called from the constructor of DynamicModel subclasses to register action variables. Calls are parsed by the source code parser and listed beside the parameters of the class in the class definition file (config.xml)

void addConstant(const char\* name, double value)

### Summary

This method can be called from the constructor of DynamicModel subclasses to register constants. These are also parsed

### Parameters

- o name: Name of the constant
- value: Literal value (i.e. 6.5). The parser will not recognise but literal values

### int getNumConstants()

### • Summary

Returns the number of constants defined in the current DynamicModel subclass

### double getConstant(int i)

### • Summary

Returns the value of the i-th constant

### Parameters

• i: Index of the constant to be retrieved

### • Return Value

Returns the value of the i-th constant

### double getConstant(const char\* constantName)

### Summary

An alternative version of getConstant() that uses the name as input

### Parameters

o constantName: The name of the constant

Return Value

An alternative version of getConstant() that uses the name as input

double getReward(const State s, const Action a, const State \*s\_p)

Summary

This method calculates the reward associated with tuple {s,a,s\_p}

- Parameters
  - s: Initial state
  - a: Action
  - \_sp: Resultant state
- Return Value

This method calculates the reward associated with tuple  $\{s,a,s\_p\}$ 

## **Etraces**

### Class ETraces

Source: etraces.cpp

### Methods

void update(double factor)

Summary

Etraces implement a technique that updates recently visited states with the current reward. This method updates the factor of each trace, so that they decay with time according to parameter Lambda. Not compatible with Experience-Replay, which is currently favored.

- Parameters
  - factor: Update factor (depends on the learning algorithm)

void addFeatureList(FeatureList\* inList, double factor)

Summary

This method adds current state's features to the traces

- Parameters
  - inList: Features of the current state
  - factor: Factor given to these features

# **Experiencereplay**

# **Class ExperienceReplay**

Source: experience-replay.cpp

### Methods

bool bUsing()

Summary

Returns whether Experience-Replay is enabled or not

size\_t getUpdateBatchSize()

• Summary

Returns the size of each update batch

bool bHaveEnoughTuples()

• Summary

Returns whether there are enough tuples in the buffer to run a batch

void addTuple(const State s, const Action a, const State\* s\_p, double r, double probability)

Adds an experience tuple to the circular buffer used

- Parameters
  - s: Initial state
  - o a: Action
  - sp: Resultant state
  - r: Reward
  - o probability: Probability by which the action was taken

# **Experiment**

# **Class Experiment**

Source: experiment.cpp

### Methods

# double getExperimentProgress()

Summary

Returns the progress of the experiment (normalized in range [0,1])

### double getTrainingProgress()

• Summary

The normalized progress taking only into account the training episodes (normalized in range [0,1])

### double getEpisodeProgress()

• Summary

Normalized progress with respect to the current episode in range  $\left[0,1\right]$ 

# bool isEvaluationEpisode()

• Summary

Returns whether the current is an evaluation episode

# void reset()

Summary

Resets the experiment to the starting conditions

## void nextStep()

• Summary

Increments the current step

# bool isValidStep()

• Summary

Returns whether the current step is valid or we have already finished the episode

### bool isValidEpisode()

• Summary

Returns whether the current episode is valid or we have already finished the experiment

### void nextEpisode()

• Summary

Used to advance the simulation to the next episode

### bool isFirstEpisode()

• Summary

Is this the first episode?

### bool isLastEpisode()

• Summary

Is this the last episode?

void timestep(State s, Action a, State s\_p, Reward r)

Summary

Called every time-step. Controls when and what information to log, and also the timers to decide if the progress must be udated

- Parameters
  - o s: Initial state of the last tuple
  - a: Action in the last tuple
  - \_sp: Resultant state of the last tuple
  - r: Reward of the last tuple

## **Featurelist**

### **Class FeatureList**

Source: features.cpp

### Methods

void resize(size\_t newSize, bool bKeepFeatures)

Summary

Resizes the feature list, allocating more memory

- Parameters
  - o newSize: New size
  - bKeepFeatures: true if we want to preserve the features on the list

void mult(double factor)

• Summary

Multiplies all the features by a factor

- Parameters
  - o factor. The factor value

double getFactor(size\_t index)

Summary

Returns the factor of a given feature on the list

- Parameters
  - index: The index of the feature
- Return Value

Returns the factor of a given feature on the list

double innerProduct(const FeatureList \*inList)

• Summary

Implements an inner-product operation

- Parameters
  - inList: Second operand of the multiply operation
- Return Value

Implements an inner-product operation

void copyMult(double factor, const FeatureList \*inList)

• Summary

Copies a list multiplied by a factor on this feature list. Resizes the list if needed

### Parameters

- o factor. The factor to multiply by
- o inList: The feature list to copy

### void addFeatureList(const FeatureList \*inList, double factor)

### Summary

Adds feature to this list, multiplied by a factor

#### Parameters

- inList: Feature list to be added
- factor: Factor used to multiply

## void add(size\_t index, double value)

#### Summary

Adds a single feature

#### Parameters

- index: The index of the feature
- · value: The value of the feature

### void spawn(const FeatureList \*inList, size\_t indexOffset)

#### Summary

All features (indices and values) are spawned by those in inList. This means that this list contains 2 features a 5-feature space, and inList contains 3 features a 6-feature space, after this operation, this list will contain 23=6 features from a 56=30 feature space

#### Parameters

- inList: Second list used as an operand
- o indexOffset: Feature-index offset used for the second list

# void applyThreshold(double threshold)

# • Summary

Removes any feature with an activation factor under the threshold

### Parameters

o threshold: Threshold value

### void normalize()

### • Summary

Normalizes features so that the sum of all the activation factors are  $\ensuremath{\mathsf{1}}$ 

# void copy(const FeatureList\* inList)

### • Summary

Copies in this list the given one

### Parameters

o inList: Source feature list to copy

# void offsetIndices(size\_t offset)

# • Summary

Adds an offset to all the feature indices

# • Parameters

offset: Offset value

### void split(FeatureList outList1, FeatureList outList2, size\_t splitOffset)

### • Summary

Splits this feature list in two lists: features with an index below splitOffset go to the first output list, and those above go to the second output list

#### Parameters

- o outList1: Output list 1
- o outList2: Output list 2
- splitOffset: Index used to split the feature list

### void multIndices(int mult)

Summary

Multiplies all the feature indices by mult

- Parameters
  - mult: Value used to multiply

# **Featuremap**

# Class FeatureMap

Source: featuremap.cpp

### Methods

void getFeatures(const State s, const Action a, FeatureList\* outFeatures)

• Summary

Calculates the features for any given state-action

- Parameters
  - s: State
  - a: Action
  - outFeatures: Output feature list

### void getFeatureStateAction(size\_t feature, State s, Action a)

• Summary

Given a feature index, this method returns the state-action to which the feature corresponds. If the feature map uses only states, the output action is left unmodified

- Parameters
  - feature: Index of the feature
  - s: Output state
  - o a: Output action

# **Functionsampler**

## **Class FunctionSampler**

Source: function-sampler.cpp

### Methods

size\_t getNumOutputs()

• Summary

Returns the number of outputs of the sampler

# Functionsampler2d

# **Class FunctionSampler2D**

Source: function-sampler.cpp

### Methods

string getFunctionId()

Returns the name of the function

size\_t getNumSamplesX()

Summary

Returns the number of samples in X (the image's width)

size\_t getNumSamplesY()

Summary

Returns the number of samples in Y (the image's height)

# Functionsampler3d

# **Class FunctionSampler3D**

Source: function-sampler.cpp

### Methods

string getFunctionId()

• Summary

Returns the name of the function

size\_t getNumSamplesX()

• Summary

Returns the number of samples in X (the image's width)

size\_t getNumSamplesY()

• Summary

Returns the number of samples in Y (the image's height)

# Gaussiannoise

### Class GaussianNoise

Source: noise.cpp

### Methods

double getNormalDistributionSample(double mean, double sigma)

• Summary

Returns a sample from a Gaussian distribution function. Used to generate noise

- Parameters
  - mean: Mean value of the distribution
  - $\circ~$  sigma : Sigma of the distribution
- Return Value

Returns a sample from a Gaussian distribution function. Used to generate noise

double getSample()

• Summary

Returns a sample from a Gaussian distribution function. Used to generate noise

• Return Value

Returns a sample from a Gaussian distribution function. Used to generate noise

# Gaussianrbfgridfeaturemap

# Class GaussianRBFGridFeatureMap

Source: featuremap-rbfgrid.cpp

#### Methods

void map(vector>& grids, const vector& values, FeatureList outFeatures)

Summary

 $Implements\ a\ Gaussian\ Radial\mbox{-}Basis\ feature\ mapping\ function\ that\ maps\ state\mbox{-}actions\ to\ feature.$ 

- Parameters
  - o grids: Input grids for every state-variable used
  - values: The values of every state-variable used
  - outFeatures: The output list of features

void unmap(size\_t feature, vector& grids, vector& outValues)

Summary

Inverse of the feature mapping operation. Given a feature it returns the state-action to which it corresponds.

- Parameters
  - feature: The index of the feature
  - grids: The set of grids used to discretize each variable
  - o outValues: The set of output values for every state-action variable

# Greedyqplusnoisepolicy

# Class GreedyQPlusNoisePolicy

Source: q-learners.cpp

### Methods

 $\label{eq:const_decomposition} double \ \mbox{selectAction(LinearStateActionVFA pQFunction, const State s, Action* a)}$ 

Summary

Implements an action selection policy that adds noise to the greedily selected action

- Parameters
  - o pQFunction: The Q-function
  - s: Current state
  - a: Output action
- Return Value

Implements an action selection policy that adds noise to the greedily selected action

# Incrementalnaturalactorcritic

### Class IncrementalNaturalActorCritic

Source: actor-critic-inac.cpp

### Methods

double update(const State s, const Action a, const State \*s\_p, double r, double behaviorProb)

Summary

Updates the policy and the value function using the Incremental Natural Actor Critic algorithm in "Model-free Reinforcement Learning with Continuous Action in Practice" (Thomas Degris, Patrick M. Pilarski, Richard S. Sutton), 2012 American Control Conference

- Parameters
  - s: Initial state
  - a: Action

- \_sp: Resultant state
- r: Reward
- behaviorProb: Probability by which the actor selected the action

### Return Value

Updates the policy and the value function using the Incremental Natural Actor Critic algorithm in "Model-free Reinforcement Learning with Continuous Action in Practice" (Thomas Degris, Patrick M. Pilarski, Richard S. Sutton), 2012 American Control Conference

### double selectAction(const State s, Action a)

### • Summary

The actor selects an action following the policies it is learning

- Parameters
  - s: Initial state
  - o a: Action

#### Return Value

The actor selects an action following the policies it is learning

# Interpolatedvalue

# Class InterpolatedValue

Source: parameters-numeric.cpp

#### Methods

### double get()

### Summary

Returns a sample from a linear function determined by linear interpolation between (x1,y1) and (x2,y2), where x1 are x2 are given as normalized experiment progress

### Return Value

Returns a sample from a linear function determined by linear interpolation between (x1,y1) and (x2,y2), where x1 are x2 are given as normalized experiment progress

# Linearstateactionvfa

### Class LinearStateActionVFA

Source: vfa.cpp

### Methods

void getFeatures(const State s, const Action a, FeatureList\* outFeatures)

### Summary

Given a state-action pair, it calculates the features for each feature map separately (state and action) and then combines using spawn() and offsetIndices() so that the resultant features belong to the full state-action feature space

- Parameters
  - s: State
  - a: Action
  - outFeatures: Output feature list

# void getFeatureStateAction(size\_t feature, State s, Action a)

### Summary

Given a feature index, it returns in s and a the values of the variables to which the feature corresponds

### • Parameters

- feature: Index of the feature
- s: Output state

```
    a: Output action

double get(const State s, const Action a)
 • Summary
   Evaluates Q(s,a)
void argMax(const State s, Action a, bool bSolveTiesRandomly)

    Summary

   Calculates the action a that maximizes Q(s,a)

    Parameters

    s: State

     o a: Output action that maximizes Q(s,a)
     • bSolveTiesRandomly: In case of tie, this flag sets whether return a random action or the first one
double max(const State* s, bool bUseFrozenWeights)

    Summary

   Calculates the maximum value of Q(s,a) for state s
 • Parameters

 s: State

     • bUseFrozenWeights: If set and it makes sense, will use the target function
 • Return Value
   Calculates the maximum value of Q(s,a) for state s
void getActionValues(const State s,double outActionValues)

    Summary

   Returns an array with the values for each action feature

    Parameters

    s: State

     • outActionValues: Output action values, one for every feature in the action feature map
Linearstatevfa
```

## Class LinearStateVFA

Source: vfa.cpp

### Methods

```
void setInitValue(double initValue)
```

Summary

Sets the initial value of the function

Parameters

void getFeatures(const State s, FeatureList outFeatures)

Uses the state feature map to calculate the features of a state

void getFeatureState(size\_t feature, State\* s)

Given a feature, it uses the state feature map to return the state variable's value in s

double get(const State \*s)

Summary

## Linearvfa

### Class LinearVFA

Source: vfa.cpp

### Methods

double get(const FeatureList \*pFeatures,bool bUseFrozenWeights)

Summary

Returns the value of the linear Value Function Approximator for the input state-action given as a list of features.

- Parameters
  - o pFeatures: Input list of features
  - ${\color{blue} \bullet} \ \, \textit{bUseFrozenWeights} \hbox{: Flag used to determine whether to use the online or target function} \\$
- Return Value

Returns the value of the linear Value Function Approximator for the input state-action given as a list of features.

void saturateOutput(double min, double max)

• Summary

Sets the function to saturate its output in range [min,max]

void setIndexOffset(unsigned int offset)

Summary

Sets the index offset used. Handy if we want to represent f(s,a) with two different feature maps: one for the state and another one for the action

- Parameters
  - offset: Offset added to feature indices

void add(const FeatureList\* pFeatures, double alpha)

Summary

Adds a feature list (each feature has an index and a factor) to the weights in the function. Some of the indices might not belong to this function

- Parameters
  - o pFeatures: Feature list to be added
  - alpha: Gain parameter used to move current weights toward those in the feature list

void set(size\_t feature, double value)

• Summary

Sets the value of a function weight

# Logger

# **Class Logger**

Source: logger-functions.cpp

### Methods

void openFunctionLogFile(const char\* filename)

• Summary

Creates a file where functions will be logged

- Parameters
  - o filename: Path to the output file

void closeFunctionLogFile()

• Summary

Closes the file used for logging functions

### void writeFunctionLogSample()

### • Summary

Adds a sample from each function to the log file

### void setOutputFilenames()

#### Summary

Registers the output files

### bool isEpisodeTypeLogged(bool evalEpisode)

### Summary

Returns whether the given type of episode is being logged

### Parameters

• evalEpisode: true if we want to query about evaluation episodes, false otherwise

### void writeLogFileXMLDescriptor(const char\* filename)

### • Summary

Creates an XML file with the description of the log file: variables, scene file...

### Parameters

• filename:

### void firstEpisode()

#### • Summary

Must be called before the first episode begins to initialize log files, timers,...

### void firstStep()

### • Summary

Must be called before the first step to write episode headers and the initial state in the log file. It also takes a snapshot of the functions and logs them if we are logging functions

### void lastStep()

### • Summary

Must be called after the last step in an episode has finished. The episode is marked as finished in the log file and, if the current is an evaluation episode, the progress is updated

## void timestep(State s, Action a, State s\_p, Reward r)

### Summary

Logs if needed a new step {s,a,s\_p,r}, and adds a new sample to statistics

### Parameters

- s: Initial state
- a: Action
- \_sp: Resultant state
- r: Reward

# void logMessage(MessageType type, const char\* message)

### Summary

Logging function that formats and dispatches different types of messages: info/warnings/errors, progress, and evaluation. Depending on whether the app is connected via a named pipe, the message is either sent via the pipe or printed on the system console. Error log messages throw an exception to terminate the program

### Parameters

- type:
- message:

# Lqrcontroller

### Class LQRController

Source: controller.cpp

### Methods

double evaluate(const State s, const Action a, unsigned int index)

Summary

Calculates one of the outputs of the LQR controller

- Parameters
  - s: Initial state
  - a. Action
  - o index: Index of the output
- Return Value

Calculates one of the outputs of the LQR controller

# **Memblock**

## **Class MemBlock**

Source: mem-block.cpp

### Methods

void dumpToFile()

• Summary

Saves the contents of the memory block to a temporary file.

void restoreFromFile()

• Summary

Restores the contents of a memory block from file

# **Offpolicyactorcritic**

# Class OffPolicyActorCritic

Source: actor-critic-offpac.cpp

### Methods

double update(const State s, const Action a, const State \*s\_p, double r, double behaviorProb)

Summary

Updates the policy and the value function using the Incremental Natural Actor Critic algorithm in "Off-Policy Actor-Critic" (Thomas Degris, Martha White, Richard S. Sutton), Proceedings of the 29 th International Conference on Machine Learning, Edinburgh, Scotland, UK, 2012. arXiv:1205.4839v5 [cs.LG] 20 Jun 2013

- Parameters
  - s: Initial state
  - a: Action
  - \_sp: Resultant state
  - r: Reward
  - behaviorProb: Probability by which the actor selected the action
- Return Value

Updates the policy and the value function using the Incremental Natural Actor Critic algorithm in "Off-Policy Actor-Critic" (Thomas Degris, Martha White, Richard S. Sutton), Proceedings of the 29 th International Conference on Machine Learning, Edinburgh, Scotland, UK, 2012. arXiv:1205.4839v5 [cs.LG] 20

### double selectAction(const State s, Action a)

#### Summary

The actor selects an action following the policies it is learning

#### Parameters

- o s: Initial state
- a: Action

### Return Value

The actor selects an action following the policies it is learning

# Offpolicydeterministicactorcritic

# Class OffPolicyDeterministicActorCritic

Source: actor-critic-opdac.cpp

### Methods

double update(const State s, const Action a, const State \*s\_p, double r, double behaviorProb)

#### Summary

Updates the policy and the value function using the Incremental Natural Actor Critic algorithm in "Off-policy deterministic actorcritic (OPDAC)" in "Deterministic Policy Gradient Algorithms" (David Silver, Guy Lever, Nicolas Heess, Thomas Degris, Daan Wierstra, Martin Riedmiller). Proceedings of the 31 st International Conference on Machine Learning, Beijing, China, 2014. JMLR: WCP volume 32

### Parameters

- s: Initial state
- a: Action
- ∘ \_sp: Resultant state
- r: Reward
- behaviorProb: Probability by which the actor selected the action

### Return Value

Updates the policy and the value function using the Incremental Natural Actor Critic algorithm in "Off-policy deterministic actorcritic (OPDAC)" in "Deterministic Policy Gradient Algorithms" (David Silver, Guy Lever, Nicolas Heess, Thomas Degris, Daan Wierstra, Martin Riedmiller). Proceedings of the 31 st International Conference on Machine Learning, Beijing, China, 2014. JMLR: WCP volume 32

### double selectAction(const State s, Action a)

### Summary

The actor selects an action following the policies it is learning

### Parameters

- s: Initial state
- a: Action

### Return Value

The actor selects an action following the policies it is learning

### Ornsteinuhlenbecknoise

## Class OrnsteinUhlenbeckNoise

Source: noise.cpp

# Methods

### double getSample()

### Summary

 $Returns \ a \ sample \ from \ an \ Ornstein \ Uhlenbeck \ process: https://en.wikipedia.org/wiki/Ornstein\%E2\%80\%93Uhlenbeck\_process \ Used \ to \ generate \ temporally-process \ temporally-process \ used \ to \ generate \ temporally-process \ temporally-$ 

correlated noise

• Return Value

Returns a sample from an Ornstein Uhlenbeck process: https://en.wikipedia.org/wiki/Ornstein%E2%80%93Uhlenbeck\_process Used to generate temporally-correlated noise

### **Pidcontroller**

### **Class PIDController**

Source: controller.cpp

### Methods

double evaluate(const State s, const Action a, unsigned int output)

Summary

Calculates one of the outputs of the PID controller

- Parameters
  - s: Initial state
  - a: Action
  - o index: Index of the output
- Return Value

Calculates one of the outputs of the PID controller

# **Qegreedypolicy**

# Class QEGreedyPolicy

Source: q-learners.cpp

### Methods

double selectAction(LinearStateActionVFA pQFunction, const State s, Action\* a)

Summary

Implements an epsilon-greedy action selection policy, selecting the action with the maximum  $\mathsf{Q}(\mathsf{s},\mathsf{a})$  value

- Parameters
  - pQFunction: The Q-function
  - s: Current state
  - a: Output action
- Return Value

Implements an epsilon-greedy action selection policy, selecting the action with the maximum Q(s,a) value

# **Qlearningcritic**

# Class QLearningCritic

Source: q-learners.cpp

## Methods

double update(const State s, const Action a, const State \*s\_p, double r, double probability)

Summary

Updates the estimate of the Q-function using the Q-Learning update rule with tuple {s,a,s\_p,r}

- Parameters
  - s: Initial state
  - a: Action

- \_sp: Resultant state
- r: Reward
- Return Value

Updates the estimate of the Q-function using the Q-Learning update rule with tuple {s,a,s\_p,r}

# **Qsoftmaxpolicy**

# Class QSoftMaxPolicy

Source: q-learners.cpp

### Methods

double selectAction(LinearStateActionVFA pQFunction, const State s, Action\* a)

Summary

Implements a Soft-Max action selection policy controlled by temperature parameter Tau

- Parameters
  - pQFunction: The Q-function
  - s: Current state
  - a: Output action
- Return Value

Implements a Soft-Max action selection policy controlled by temperature parameter Tau

# Regularpolicygradientlearner

# Class RegularPolicyGradientLearner

Source: actor-regular.cpp

### Methods

void update(const State s, const Action a, const State \*s\_p, double r, double td)

• Summary

Updates the policies using a regular gradient-descent update rule

- Parameters
  - s: Initial state
  - a: Action
  - \_sp: Resultant state
  - r: Reward
  - behaviorProb: Probability by which the actor selected the action

## Rewardfunction

# **Class RewardFunction**

Source: reward.cpp

## Methods

void addRewardComponent(IRewardComponent\* rewardComponent)

Summary

RewardFunction can use more than one scalar reward and they are added using this method. Scalar rewards must derive from IRewardComponent

- Parameters
  - rewardComponent: The new scalar reward to be added

double getReward(const State s, const Action a, const State\* s\_p)

Calculates the total reward based on the different scalar rewards. If we only define one reward function, its value will be returned

- Parameters
  - s: Initial state
  - o a: Action
  - sp: Resultant state
- Return Value

Calculates the total reward based on the different scalar rewards. If we only define one reward function, its value will be returned

void initialize()

Summary

DynamicModel subclasses should call this initialization method after adding the reward functions

void override(double reward)

• Summary

If we want to override the final reward in some special states (i.e. a negative reward if FAST simulator crashed) we can call this method from the DynamicModel

- Parameters
  - reward: The reward we want to give the agent

## Sarsa

### Class SARSA

Source: q-learners.cpp

### Methods

double selectAction(const State s, Action a)

Summary

implements SARSA On-policy action selection algorithm

- Parameters
  - s: Initial state
  - a: Output action

 $\label{lem:const_state} \mbox{double update(const State s, const Action a, const State* s\_p, double r, double probability)}$ 

Summary

Updates the estimate of the Q-function using the SARSA update rule with tuple {s,a,s p,r}

- Parameters
  - s: Initial state
  - a: Action
  - \_sp: Resultant state
  - r: Reward
- Return Value

Updates the estimate of the Q-function using the SARSA update rule with tuple  $\{s,a,s\_p,r\}$ 

# **Simgod**

## Class SimGod

Source: simgod.cpp

### Methods

double selectAction(State s, Action a)

Iterates over all the Simions to let each of them select their actions

### Parameters

- s: Initial state
- a: Action variable where Simions write their selected actions

#### Return Value

Iterates over all the Simions to let each of them select their actions

## void update(State s, Action a, State\* s\_p, double r, double probability)

### • Summary

Iterates over all the Simions to let them learn from the last real-time experience tuple

#### Parameters

- s: Initial state
- a: Action
- \_sp: Resultant state
- r: Reward
- o probability: Probability by which the action was taken. Should be ignored

### void postUpdate()

### • Summary

If Experience-Replay is enabled, several tuples are taken from the buffer and given to the Simions to learn from them

### void deferredLoad()

### • Summary

Iterates over all the objects implementing DeferredLoad to do all the heavyweight-lifting stuff

# double getGamma()

### • Summary

Returns the value of gamma

## int getTargetFunctionUpdateFreq()

### • Summary

 $Returns \ the \ number \ of \ steps \ after \ which \ deferred \ V-Function \ updates \ are \ to \ be \ done. \ 0 \ if \ we \ don't \ use \ Freeze-V-Function$ 

### Return Value

Returns the number of steps after which deferred V-Function updates are to be done. 0 if we don't use Freeze-V-Function

## bool bUpdateFrozenWeightsNow()

### • Summary

Returns whether we need to update the "frozen" copies of any function using Freeze-Target-Function

### • Return Value

Returns whether we need to update the "frozen" copies of any function using Freeze-Target-Function

### size\_t getExperienceReplayUpdateSize()

### • Summary

Returns the number of each update batch using Experience-Replay

### • Return Value

Returns the number of each update batch using Experience-Replay

# **Simionapp**

# **Class SimionApp**

### Methods

### void printRequirements()

### Summary

This method prints the run-time requirements of an experiment instead of running it

void registerStateActionFunction(string name, StateActionFunction\* pFunction)

### • Summary

Called from function-learning objects to register an instance of a function to log (if configured to do so)

#### Parameters

- o name: Name of the function
- pFunction: Pointer to the function

### void wireRegister(string name)

### • Summary

Wires allow us to connect inputs with outputs. This method registers a wire by name

### Parameters

• name: The new wire's name

### void wireRegister(string name, double minimum, double maximum)

#### Summary

This method registers a wire by name and also sets its value range

#### Parameters

- o name:
- minimum:
- maximum:

void registerTargetPlatformInputFile(const char targetPlatform, const char filepath, const char\* rename)

# • Summary

Used to register input files to a specific platform.

### • Parameters

- targetPlatform: Target platform: Win-32, Win-64, Linux-64,...
- $\circ~$  filepath: Path to the required file (exe, dll, data file, ...)
- rename: Name given to the required file in the host machine

## void registerTargetPlatformOutputFile(const char targetPlatform, const char filepath)

### Summary

Used to register output files to a specific platform

### Parameters

- targetPlatform: Target platform: Win-32, Win-64, Linux-64,...
- $\circ~$  filepath: Path to the required file (exe, dll, data file, ...)

# void registerInputFile(const char filepath, const char rename)

## Summary

Used to register input files common to all the target platforms

### Parameters

- $\circ~$  filepath: Path to the required file (exe, dll, data file, ...)
- $\circ \;$  rename : Name given to the required file in the host machine

### void registerOutputFile(const char\* filepath)

Used to register output files common to all the target platforms

#### Parameters

- filepath: Path to the required file (exe, dll, data file, ...)
- rename: Name given to the required file in the host machine

### void run()

### • Summary

The app's main-loop that starts the simulation and runs until it finishes.

### void initRenderer(string sceneFile, State s, Action a)

#### Summary

This method is called from run() and initializes the real-time rendering window (if configured to do so)

#### Parameters

- o sceneFile: Name of the file with the definition of the scene
- s: Current state
- a: Current action

### void initFunctionSamplers(State s, Action a)

### • Summary

After adding all the StateActionFunctions to be logged/drawn, this method is called to initialize them

### Parameters

- s: State
- a: Action

### void updateScene(State s, Action a)

### • Summary

Updates the graphical objects bound in the scene file using the current value of their bounded state/action variables

### Parameters

- s:
- a:

# Simionmembuffer

# Class SimionMemBuffer

Source: mem-buffer.cpp

# Methods

### BUFFER\_SIZE getBlockSizeInBytes()

### Summary

Returns the size of a memory block in the parent memory pool

### • Return Value

Returns the size of a memory block in the parent memory pool  $% \label{eq:control_eq} % \label{eq:control_eq}$ 

# **Simionmempool**

# Class SimionMemPool

Source: mem-pool.cpp

### Methods

void initialize(MemBlock\* pBlock)

This method resets each interleaved buffer within a MemBlock to its initial value

#### Parameters

o pBlock: Memory block

### void init(BUFFER\_SIZE blockSize)

Summary

After adding all the requested buffers, this method initializes all the necessary data so that several MemBlocks are allocated

#### Parameters

• blockSize: Desired MemBlock size (element count)

### void copy(IMemBuffer pSrc, IMemBuffer pDst)

Summary

Copies data from one buffer to another. They must belong to the same MemPool

#### • Parameters

- o pSrc: Source buffer
- o pDst: Destination buffer

# Simpleepisodelinearschedule

# Class SimpleEpisodeLinearSchedule

Source: parameters-numeric.cpp

### Methods

### double get()

Summary

Returns a sample of a linear function determined by a starting value, an ending value and the current experiment progress

Return Value

Returns a sample of a linear function determined by a starting value, an ending value and the current experiment progress

# Singledimensiongrid

# Class SingleDimensionGrid

Source: single-dimension-grid.cpp

### Methods

### size\_t getClosestFeature(double value)

• Summary

Within the one-dimension grid, this method returns the index of the feature closest to the given value

- Parameters
  - value: Value of the variable to which this grid corresponds
- Return Value

Within the one-dimension grid, this method returns the index of the feature closest to the given value

## double getFeatureValue(size\_t feature)

Summary

Given a feature index, it returns the value of the variable to which this single-dimension grid corresponds

- Parameters
  - feature:
- Return Value

# Sinusoidalnoise

## Class SinusoidalNoise

Source: noise.cpp

### Methods

double getSample()

• Summary

Returns a sample from a sinusoidal signal. Used to generate noise

• Return Value

Returns a sample from a sinusoidal signal. Used to generate noise

## **Statsinfo**

## **Class StatsInfo**

Source: stats.cpp

### Methods

void reset()

Summary

Resets the stats

void addSample(double value)

• Summary

Adds a sample to the collection of values

• Parameters

· value: New sample

double getMin()

• Summary

Returns the maximum sampled value

double getMax()

• Summary

Returns the minimum sampled value

double getAvg()

• Summary

Returns the mean value of all samples

double getStdDev()

• Summary

Returns the standard deviation of all the samples

# Stochasticgaussianpolicy

# Class StochasticGaussianPolicy

Source: vfa-policy.cpp

### Methods

double selectAction(const State s, Action a)

A DeterministicPolicyGaussianNoise policy uses a function to represent the mean value of the function at each state and a second function to represent the variance of the output at each state. To calculate the policy's output, the two functions are evaluated for a state, and then the values used to sample a normal distribution, which is the actual output of the policy

- Parameters
  - s: Initial state
  - o a: Output state

void getFeatures(const State state, FeatureList outFeatureList)

Summary

Uses the policy's feature map to return the features representing the state

- Parameters
  - o state: State
  - outFeatureList: Output feature list

## **Tdclambdacritic**

## Class TDCLambdaCritic

Source: critic-tdc-lambda.cpp

### Methods

double update(const State s, const Action a, const State \*s\_p, double r, double rho)

• Summary

Updates the value function using the TDC update rule

- Parameters
  - s: Initial state
  - a: Action
  - \_sp: Resultant state
  - r: Reward
  - rho: Importance sampling
- Return Value

Updates the value function using the TDC update rule

# **Tdlambdacritic**

# Class TDLambdaCritic

Source: critic-td-lambda.cpp

### Methods

double update(const State s, const Action a, const State \*s\_p, double r, double rho)

Summary

Updates the value function using the popular TD(lambda) update rule

- Parameters
  - s: Initial state
  - a: Action
  - \_sp: Resultant state
  - r: Reward
  - rho: Importance sampling
- Return Value

# **Tilecodingfeaturemap**

# Class TileCodingFeatureMap

Source: featuremap-tilecoding.cpp

### Methods

void map(vector>& grids, const vector& values, FeatureList outFeatures)

Summary

Implements a Tile-Coding feature mapping function that maps state-actions to feature: https://www.cs.utexas.edu/~pstone/Papers/bib2htmlinks/SARA05.slides.pdf

- Parameters
  - o grids: Input grids for every state-variable used
  - o values: The values of every state-variable used
  - outFeatures: The output list of features

void unmap(size\_t feature, vector& grids, vector& outValues)

• Summary

Inverse of the feature mapping operation. Given a feature it returns the state-action to which it corresponds.

- Parameters
  - o feature: The index of the feature
  - grids: The set of grids used to discretize each variable
  - outValues: The set of output values for every state-action variable

# **Toleranceregionreward**

# Class ToleranceRegionReward

Source: reward.cpp

### Methods

double getReward(const State s, const Action a, const State \*s\_p)

Summary

This reward function returns a reward depending on the value of an error variable with respect to its tolerance region

- Parameters
  - s: Initial state
  - a: Action
  - \_sp: Resultant state
- Return Value

This reward function returns a reward depending on the value of an error variable with respect to its tolerance region

# **Trueonlinetdlambdacritic**

### Class TrueOnlineTDLambdaCritic

Source: critic-true-online-td-lambda.cpp

### Methods

double update(const State s,const Action a,const State \*s\_p, double r, double rho)

• Summary

Updates the value function using the True-Online TD(Lambda) update rule in "True Online TD(lambda)" (Harm van Seijen, Richard Sutton), Proceedings of the 31st International Conference on Machine learning

- Parameters
  - o s: Initial state
  - a: Action
  - \_sp: Resultant state
  - r: Reward
  - rho: Importance sampling
- Return Value

Updates the value function using the True-Online TD(Lambda) update rule in "True Online TD(lambda)" (Harm van Seijen, Richard Sutton), Proceedings of the 31st International Conference on Machine learning

# Windturbineboukhezzarcontroller

### Class WindTurbineBoukhezzarController

Source: controller.cpp

### Methods

double evaluate(const State s,const Action a, unsigned int output)

• Summary

Calculates one of the outputs of the Variable-Speed Wind Turbine by Boukhezzar

- Parameters
  - s: Initial state
  - o a: Action
  - index: Index of the output
- Return Value

Calculates one of the outputs of the Variable-Speed Wind Turbine by Boukhezzar

# Windturbinejonkmancontroller

## Class WindTurbineJonkmanController

Source: controller.cpp

### Methods

double evaluate(const State s,const Action a, unsigned int output)

• Summary

Calculates one of the outputs of the Variable-Speed Wind Turbine by Jonkman

- Parameters
  - s: Initial state
  - a: Action
  - o index: Index of the output
- Return Value

Calculates one of the outputs of the Variable-Speed Wind Turbine by Jonkman

# Windturbinevidalcontroller

## Class WindTurbineVidalController

Source: controller.cpp

### Methods

double evaluate(const State s, const Action a, unsigned int output)

Calculates one of the outputs of the Variable-Speed Wind Turbine by Vidal

- Parameters
  - s: Initial state
  - o a: Action
  - o index: Index of the output
- Return Value

Calculates one of the outputs of the Variable-Speed Wind Turbine by Vidal

### Wireconnection

## **Class WireConnection**

Source: parameters-numeric.cpp

### Methods

double get()

Summary

Returns the current value of a wire connection

# World

### **Class World**

Source: world.cpp

### Methods

double getDT()

• Summary

This method returns the Delta\_t used in the experiment

• Return Value

This method returns the Delta\_t used in the experiment

### void reset(State \*s)

Summary

Reset state variables to the initial state from which simulations begin (it may be random)

- Parameters
  - s: State variable that holds the initial state

double executeAction(State s, Action a, State \*s\_p)

Summary

Method called every control time-step. Internally it calculates calculates the length of the integration steps and calls several times DynamicModel::executeAction()

- Parameters
  - s: The variable with the current state values
  - a: The action to be executed
  - \_sp: The variable that will hold the resultant state
- Return Value

Method called every control time-step. Internally it calculates calculates the length of the integration steps and calls several times DynamicModel::executeAction()