User Manual

PlayMaker 8

ASDD – MSDD Learning Reinforcement Learning Dynamic Programming

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1. Generating Input Data

Both Rules and Maps versions are working with: ~/InputFile.txt

It contains X States & Actions, like this:

Input File (8 entries)			
Predator Model	Paint Model		
W,E,E,W,E,E W,E,E,E,E,W W,E,E,W,E,N W,E,E,W,E,E W,E,E,E,E,W W,E,E,W,E,W	p,C,D,b,=,E p,C,d,b,-,A p,c,d,b,=,R p,c,D,b,=,R p,c,D,b,=,I p,c,D,B,=,R p,c,D,B,=,A		
W,E,E,W,E,W	P,c,D,B,=,A		

In order to obtain these « simple » Sensors, (almost) all of the *toString()* functions were copied and modified to the *translation()* functions.

1.1 Predator Model

To generate data from the Predator Model, go to:

EnvModel.PredatorModel/PredatorTester.java

At the beggining of the file, set NUM_MOVES to the number of data wanted.

I suggest 100K for a complete simulation.

Run the file, the data set will be generated in: ~/LogFilesResults.txt

Copy the whole content of this one, and paste it to ~/InputFile.txt

The old behaviour of this file can be obtained by uncommenting line 82:

paintEnvironment.testAgentRecords();

1.2 Paint Model

To generate data from the Paint Model, go to:

EnvModel.PaintModel/PaintTester.java

At the beggining of the file, set NUM_MOVES to the number of data wanted.

I suggest 100K for a complete simulation.

Run the file, the data set will be generated in: ~/LogFilesResults.txt

Copy the whole content of this one, and paste it to ~/InputFile.txt

The old behaviour of this file can be obtained by uncommenting line 100:

predatorEnvironment.testAgentRecords();

2. Learn Rules

2.1 MSDD

Go to: V_Tester/Tester.java

To Use MSDD, set MSDD (line 33) to true.

Set the *maxnodes* parameter (line 36) to 50.000 for a complete simulation. (The higher limit for MSDD and PreatorModel is around 23K)

To learn MSDD Rules, set *load_MSDD_rules* (line 34) to **false.**

New closedList & RuleSetList will be generated and exported to:

~/Rules.txt & ~/RSList.txt

2.2 ASDD

Go to: V_Tester/Tester.java

To Use ASDD, set ASDD (line 40) to true. Also check that MSDD is set to false.

To learn ASDD Rules, set load ASDD rules (line 41) to false.

New closedList & RuleSetList will be generated and exported to:

~/Rules.txt & ~/RSList.txt

3.1 MSDD

Go to: V_Tester/Tester.java

To Use MSDD, set MSDD (line 33) to true.

To load MSDD Rules, set load_MSDD_rules (line 34) to true.

The files ~/Rules.txt & ~/RSList.txt will be imported.

3.2 **ASDD**

Go to: V_Tester/Tester.java

To Use ASDD, set *ASDD* (line 40) to **true**. Also check that *MSDD* is set to **false**.

To load ASDD Rules, set load_ASDD_rules (line 41) to true.

The files ~/Rules.txt & ~/RSList.txt will be imported.

4. Using Maps

Go to: V_Tester/Tester.java

To Use Maps, set MSDD (line 33) to false & ASDD (line 40) to false.

5. Reinforcement Learning

Go to: V_Tester/Tester.java

To perform Reinforcement Learning, set Reinforcement Learning (line 53) to true.

Set the Reinforcement_Learning_steps parameter (line 54) to:

- **500.000** (or more) if you're using Maps
- 20.000 if you're using Rules.

The State Generator is selected automatically, depending on whether you use Maps or Rules.

5.1 Value Table & Action Value Table

To use Value Table, set use_Value_Table (line 58) to true.

To use Action Value Table, set the above parameter to false.

If you use Action Value Table, you can enable Dynamic Programming by setting use Dynamic Programming (line 60) to true.

Here is a reminder of what both Tables look like:

Value Table	Action Value Table
STATE 1 [E, W, W, E, A, *] Value : 1.71	STATE 1 [W, E, E, W, E, *] Actions : [N, E, S, W] Values : [0.58, 0.79, 0.75, 0.59]
STATE 2 [E, E, W, W, A, *] Value : 1.66	STATE 2 [W, E, E, E, E, *] Actions : [N, E, S, W] Values : [0.74, 0.71, 0.83, 0.72]

5.2 Decision Table

In order to get a much more quicker system when running the Predator App, we can convert Value & ActionValue Tables into Decision Tables.

To use Decision Tables, set use_decision_tables (line 55) to true.

Here is what a Decision Table looks like:

Decision Table

STATE 1 [W, E, E, E, E, *] Action: S STATE 2 [E, E, A, E, E, *] Action : S STATE 3 [E, E, E, A, E, *] Action: W STATE 4 [E, W, E, E, E, *] Action: W

This allows us to quickly « read » what the better action is, instead of « computing » an answer at every step.

6. Running the Predator App

Go to: EnvController.PredatorEnvController/PredatorEnvApp.java

Depending on what was generated in the StateTable.txt, you should uncomment one of these lines :

- getPredatorEnvironment().updateEnvironment() (line 56) is the old version. It will use the default project, including Reinforcement Learning according to the last saved Percep & Actions records.
- getPredatorEnvironment().updateEnvironmentFromStateActionValueTable() (line 60) will import the StateTable.txt file (containing an ActionValue Table), and ask at every step for an answer from the State Generator.
- getPredatorEnvironment().updateEnvironmentFromStateValueTable()
 (line 63) will import the StateTable.txt file (containing a Value Table),
 and ask at every step for an answer from the State Generator.
- getPredatorEnvironment().updateEnvironmentFromDecisionTable() (line 67) will import the StateTable.txt file (containing a Decision Table), and directly act without any other process required. This way is a lot faster, it is the last version.

These functions can be accessed in *EnvModel/TickEnvironment.java*

When Running the App, you'll see in the console that each Step has a number (in order to stop around 1.5K for instance, and that each Rewarded Step prints « REWARD ».

As a consequence, to get the performance of the Simulation, copy the console text into a .txt file, search for « Reward » (this will give you the number of rewarded States), and divide it by the number of steps.