COMP3702 - Assignment 3

Roy Portas - 43560846

October 12, 2016

1 Problem Definition

1.1 State Space

The state space will be a set of tuples of integers from 0 to the maximum number of items the shop can stock. The size of the tuple is the number of items a shop can stock. Thus for a tiny store it will be $\{(0,0),(0,1),(1,0),\ldots,(3,3)\}$.

1.2 Action Space

The actions space will be the set of all actions that can be performed by the customers. Thus for a tiny store, it will look like the following:

```
buy 0 of item 1,
buy 1 of item 1,
...,
buy 2 of item 2,
buy 3 of item 2
}
```

1.3 Transition Function

The transitions of items are indepedant from each other, meaning that buying one item type does not influence the probability of buying another item type. Thus the transition function for a tiny store can be represented as follows:

$$T((i_1, i_2), (t_1, t_2), i'_1, i'_2) = T_{i_1}(i_1, t_1, i_1) \cdot T_{i_2}(i_2, t_2, i_2)$$

The transition function will be a matrix for each item type where the rows and columns are the number of items the store stocks.

Item 1	(0, 0)	(0, 1)	(1, 0)
(0, 0)	0.2	0.3	0.1
(0, 1)	0.2	0.3	0.1

In this table, the rows represent the current state and the columns represent the next state.

1.4 Reward Function

The reward function will be the net profit made by the store after performing an action. As with the transition function, the reward for one item type is independent to the reward of other item types.

1.5 Discount Factor

The discount factor is given in the input file.

1.6 Value Function

A value function associates a state with the value, or 'worth' of being in that state.

$$V_{\pi}(s) = \sum_{s'} T(s, \pi(s), s') [R(s, \pi(s)s') + dV_{\pi}(s')]$$

However since the reward is independent from the next state, the reward function can be represented as R(s, a), yielding:

$$V_{\pi}(s) = R(s, \pi(s)) + d\sum_{s'} T(s, \pi s, s') V_{\pi}(s')$$

1.6.1 Calculating the value function

$$(I - dP)v = r$$
$$v = (I - dP)^{-1}r$$

Where:

I: Identity matrix

d: Discount factor

P: Probability matrix

v: Value matrix, what we want to find

r: Reward matrix

2 Algorithm Description

- 2.1 Policy
- 2.2 Online Method
- 2.3 Offline Method