Assignment

Write a generic Markov process solver.

The program should take 4 flags (which will be understood better later) and an input file.

- -df: a float discount factor [0, 1] to use on future rewards, defaults to 1.0 if not set
- -min: minimize values as costs, defaults to false which maximizes values as rewards
- -tol: a float tolerance for exiting value iteration, defaults to 0.01
- -iter: an integer that indicates a cutoff for value iteration, defaults to 100

e.g.

mdp -df .9 -tol 0.0001 some-input.txt

Grading

- out of 100: 90 points for correct behavior
 - This means numbers to align within a power of 10 of the tolerance.
 - e.g. if -tol=0.01 then your answers should be <= 0.1 from mine
- 10 points for well written code: clear data structures, methods, control flow, etc.

Markov process solver

You will implement the algorithm described in class:

```
\pi = initial policy (arbitrary) 
V = initial values (perhaps using rewards) 
for { 
V = ValueIteration(\pi) // computes V using stationery P 
\pi' = GreedyPolicyComputation(V) // computes new P using latest V 
if \pi == \pi' then return \pi, V 
\pi = \pi' }
```

Where:

- ValueIteration computes a transition matrix using a fixed policy, then iterates by recomputing values for each node using the previous values until either:
 - no value changes by more than the 'tol' flag,
 - or -iter iterations have taken place.
- GreedyPolicyComputation uses the current set of values to compute a new policy. If -min
 is not set, the policy is chosen to maximize rewards; if -min is set, the policy is chosen to
 minimize costs.

The value of an individual state is computed using the Bellman equation for a Markov property

$$v(s) = r(s) + df * P * v$$

Where:

- v on the RHS is the previous values for each state
- df is the -df discount factor flag applied to future rewards
- P is the transition probability matrix computed using the policy and the type of node this is (see below)
- r(s) is the reward/cost for being in this particular state.

Input file and State types

Node/state names should be alphanumeric

The input file consists of 4 types of input lines:

- Comment lines that start with # and are ignored (as are blanklines)
- Rewards/costs lines of the form 'name = value' where value is an integer
- Edges of the form 'name : [e1, e2, e2]' where each e# is the name of an out edge from name
- Probabilities of the form 'name % p1 p2 p3' (more below)

These lines may occur in any order and do not have to be grouped, so this is valid:

```
A = 7
B % .9
C : [B, A]
C=-1
A : [B, A]
A % .2 .8
B : [A, C]
```

Probability entries:

A node with the same number of probabilities as edges is a *chance node*, with synchronized positions.

```
e.g.
A:[B, C, D]
A% 0.1 0.2 0.7
```

Indicates that from node A there is a 10% chance of transitioning to node B, 20% to

node C and 70% to node D.

Note: these probabilities should sum to 1 or your program should indicate an error.

A node with a single probability 'name: p' is a *decision node*, with the given success rate. As in the maze example, failures are split evenly amongst the other edges. e.g.

```
F: [C, E, G]
F%.8
```

•

Given a policy of F -> E, transition probabilities would be {C=.1 E=.8 G=.1}, noting that this changes with the policy.

Note: with p < 1 this is Q-learning where alpha=1-p

Nodes:

- If a node has edges but no probability entry, it is assumed to be a decision node with p=1
- If a node has edges but no reward entry, it is assumed to have a reward of 0
- If a node has no edges it is terminal. A probability entry for such a node is an error.
- If a node has a single edge it always transitions there. (this is useful for capturing some reward on the way)
- A node referenced as an edge must separately have one of the three entries to be valid
- Therefore to create a 0 value terminal node you must do 'name = 0'

Output

Your program should output the optimal policy (if applicable, a problem with only chance nodes has no policy), and the values of each node/state (under that policy if applicable).

See examples for the format.

Examples

- 1. Maze from slides produces solution.
- 2. Publishing Decision Tree produces solution.
- 3. Restaurant Decision Tree run with '-min' produces solution.
- 4. Student Markov Process produces solution.
- 5. Student MDP produces solution.
- 6. CMU Teacher example with -df=.9 produces solution.

Submission

- On Brightspace, a zip file containing: Code for the program
- Makefile/BUILD/pom.xml/etc needed to build your code
- README indicating how to compile and run your code
- In your README include any classmates you consulted with

•	 As previously indicated, the code should build on department linux machines 			