HW3: Multi-Agent Search

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Part I. Implementation

Part 1: Minimax Search

```
lastAgent = gameState.getNumAgents()-1 # record the number of the last agent
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               The recursion stop when the depth equals to the given depth, or the the game is either a winning
state or losing state, and return the score calculated by self.evaluationFunction.
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               3. Run for loop to do recursion to evaluate the score of each legal action and store the results
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                4. After getting all the evaluated scores, determine if the 'agentIdx' represents ghosts, return
               zero depth, and if it is in the zero depth, then return the action which has the highest score
               def minimax(state, depth, agentIdx):
    if depth == self.depth or state.isWin() or state.isLose():
                        return self.evaluationFunction(state)
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                    actions = state.getLegalActions(agentIdx)
                        actions.remove(Directions.STOP) # we don't want the pacman to stop in some cases,
                    for action in actions:
                        nextState = state.getNextState(agentIdx, action)
                              resVal.append(minimax(nextState, depth, agentIdx+1))
                            resVal.append(minimax(nextState, depth+1, 0))
                    if agentIdx:
                        return min(resVal)
                        return max(resVal) if depth else actions[resVal.index(max(resVal))]
               return minimax(gameState, 0, 0)
```

Part 2: Alpha-Beta Pruning

```
# Begin your code (Part 2)

lastAgent = gameState.getNumAgents()-1 # record the number of the last agent

"""

Alpha-Beta Pruning

1. Initialize 'initVal' to positive infinity if the agentIdx represents ghosts, or to negative infinity if the agentIdx represents pacman.

2. The recursion stop when the depth equals to the given depth, or the the game is either a winning state or losing state, and return the score calculated by self.evaluationFunction.

3. If the depth and agentIdx both aren't zero, run for loop to do recursion to evaluate the score of each legal action. If agentIdx is zero, 'initVal' equals to the maximum value between 'initVal' and the returned score, and update the 'alpha' if 'initVal' is larger than 'alpha'; else, 'initVal' equals to the minimum value between 'initVal' and the returned score, and update the 'beta' if 'initVal' is less than 'beta'.

4. If 'alpha' is larger than 'beta', break the for loop and return the 'initVal'; if not, do the next for loop and return 'initVal' after finishing.

5. If the depth and agentIdx are both zero, initialize an empty list 'resVal' to record the evaluated score, and run for loop to get scores and store in the 'resVal' and update the 'initVal' and 'alpha'.

6. If 'alpha' is larger than 'beta', break the for loop; if not, do the next for loop. After finishing, return the action which value is equal to 'alpha'.

"""
```

```
def alphabeta(state, depth, agentIdx, alpha=float('-inf'), beta=float('inf')):
    if depth == self.depth or state.isWin() or state.isLose():
       return self.evaluationFunction(state)
    initVal = float('inf') if agentIdx else float('-inf')
   actions = state.getLegalActions(agentIdx)
   if Directions.STOP in actions:
       actions.remove(Directions.STOP) # we don't want the pacman to stop in some cases.
   if depth or agentIdx:
       for action in actions:
           nextState = state.getNextState(agentIdx, action)
           if agentIdx == 0:
               initVal = max(initVal, alphabeta(nextState, depth, 1, alpha, beta))
                alpha = max(alpha, initVal)
               if agentIdx == lastAgent:
                   initVal = min(initVal, alphabeta(nextState, depth+1, 0, alpha, beta))
                   initVal = min(initVal, alphabeta(
                       nextState, depth, agentIdx+1, alpha, beta))
                beta = min(beta, initVal)
           if alpha > beta:
               break
       return initVal
       resVal = []
        for action in actions:
           resVal.append(alphabeta(state.getNextState(0, action), depth, 1, alpha, beta))
           initVal = max(initVal, resVal[-1])
           alpha = max(alpha, initVal)
            if alpha > beta:
               break
       return actions[resVal.index(alpha)]
return alphabeta(gameState, 0, 0)
```

Part 3: Expectimax Search

```
lastAgent = gameState.getNumAgents()-1 # record the number of the last agent
Expectimax Search
1. The steps are the same as Minimax Search, but return the mean value of 'resVal' rather than minimal
def expectimax(state, depth, agentIdx):
   if depth == self.depth or state.isWin() or state.isLose():
       return self.evaluationFunction(state)
   resVal = []
   actions = state.getLegalActions(agentIdx)
   if Directions.STOP in actions:
       actions.remove(Directions.STOP) # we don't want the pacman to stop in some cases.
    for action in actions:
       nextState = state.getNextState(agentIdx, action)
        if agentIdx != lastAgent:
           resVal.append(expectimax(nextState, depth, agentIdx+1))
           resVal.append(expectimax(nextState, depth+1, 0))
    if agentIdx:
        return sum(resVal)/len(resVal)
        if depth:
           return max(resVal)
            return actions[resVal.index(max(resVal))]
return expectimax(gameState, 0, 0)
```

Part 4: Evaluation Function

```
# Begin your code (Part 4)
Initialize variables and get the current game state we want.
score = currentGameState.getScore()
pos = currentGameState.getPacmanPosition()
foodList = currentGameState.getFood().asList()
capsuleList = currentGameState.getCapsules()
ghostStates = currentGameState.getGhostStates()
minFoodDist = float('inf')
minCapsuleDist = float('inf')
scaredGhostDist = float('inf')
Calculate the minimal position of food, capsule, and scred ghosts.
for food in foodList:
    minFoodDist = min(minFoodDist, manhattanDistance(pos, food))
for capsule in capsuleList:
    minCapsuleDist = min(minCapsuleDist, manhattanDistance(pos, capsule))
for ghost in ghostStates:
    if ghost.scaredTimer > 0:
        scaredGhostDist = min(scaredGhostDist, manhattanDistance(pos, ghost.getPosition()))
minimal scared ghost distance.
return score+(10/(minFoodDist))+(20/(minCapsuleDist))+(200/(scaredGhostDist))
```

Part II. Results & Analysis

Question Part1

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Question Part2

Question Part3

Question Part4

Final



Discussion

In pacman game, we can get 10 pts by eating food, get 50 pts by eating capsules, and get 200 pts by eating scared ghosts. Thus, I assume that the formula is

$$CurrentScore + (\frac{x}{minFoodDist}) + (\frac{y}{minCapsuleDist}) + (\frac{z}{scaredGhostDist})$$

In the beginning, I make x, y, and z be 10, 50, 200, respectively. However, I discovered that I will get different results if I change the parameter x, y, or z, and interestingly, if I only raise the value z, the result score will become higher, but sometimes the higher value I use, the worst score I will get.

Therefore, I've tried many reasonable values of x, y, and z, and finally, I consider

$$CurrentScore + (\frac{10}{minFoodDist}) + (\frac{20}{minCapsuleDist}) + (\frac{200}{scaredGhostDist})$$

is the most suitable formula for the evaluation function.