Implementation

```
# Begin your code (Part 1)
def minimum(state, agentIndex, depth):
    if state.isLose() or state.isWin(): # If the game is finished(win or lose)
        return self.evaluationFunction(state) # go to the evaluation function
    if agentIndex == state.getNumAgents() - 1: # If it is the last ghost
       return min(maximum(state.getNextState(agentIndex, action), depth) for action in state.getLegalActions(agentIndex))
        # Take the minimum value of the maximum function with the depth
        return min(minimum(state.getNextState(agentIndex, action), agentIndex + 1, depth) for action in
                  state.getLegalActions(agentIndex))
def maximum(state, depth):
   if state.isLose() or state.isWin() or depth == self.depth: # If the game is finished(win or lose or no more depth)
       return self.evaluationFunction(state) # go to the evaluation function
    return max(minimum(state.getNextState(0, action), 1, depth + 1) for action in state.getLegalActions(0))
best = max(gameState.getLegalActions(0),
          key=lambda action: minimum(gameState.getNextState(0, action), 1, 1))
return best
```

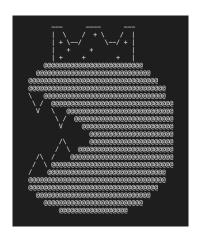
```
# Begin your code (Part 3)
def expectation(state, agentIndex, depth):
    if state.isLose() or state.isWin(): # If the game is finished(win or lose)
        return self.evaluationFunction(state) # go to the evaluation function
    probability = 1.0 / len(state.getLegalActions(agentIndex)) # the probability is one out of total num of actions
    for action in state.getLegalActions(agentIndex):
        if agentIndex == state.getNumAgents() - 1: # If it is the last ghost
           val += maximum(state.getNextState(agentIndex,
                                            action), depth) * probability # Add the maximum function times probability
           val += expectation(state.getNextState(agentIndex, action), agentIndex +
                              1, depth) * probability # If not, add the expectation with the next agent
    return val
def maximum(state, depth):
    if state.isLose() or state.isWin() or depth == self.depth: # If the game is finished(win or lose or no more depth)
        return self.evaluationFunction(state) # go to the evaluation function
    val = max(expectation(state.getNextState(0, action), 1, depth + 1)
             for action in state.getLegalActions(0)) # Take the maximum value of expectation on the next depth
    return val
best = max(gameState.getLegalActions(), key=lambda action: expectation(
   gameState.getNextState(0, action), 1, 1)) #
return best # The best action is the maximum sorted by action in expectation function
```

```
# Begin your code (Part 2)
def minimum(state, agentIndex, depth, a, b):
    if state.isLose() or state.isWin(): # If the game is finished(win or lose)
        return self.evaluationFunction(state) # go to the evaluation function
    v = float("inf")
    for action in state.getLegalActions(agentIndex):
        if agentIndex == state.getNumAgents() - 1: # If it is the last ghost
            newV = maximum(state.getNextState(
                agentIndex, action), depth, a, b) # record the return value of maximum function
        else: # If not
            newV = minimum(state.getNextState(
                agentIndex, action), agentIndex + 1, depth, a, b) # record the value of minimum function with the next agent
        v = min(v, newV) # update current value with minimum between current value and the new value
        if v < a: # If the value is smaller than the alpha</pre>
            return v
        b = min(b, v) # Save beta as the minimum between beta and current value
    return v
def maximum(state, depth, a, b):
    if state.isLose() or state.isWin() or depth == self.depth: # If the game is finished(win or lose or no more depth)
        return self.evaluationFunction(state) # go to the evaluation function
    v = float("-inf")
    if depth == 0: # If at the first layer
        best = state.getLegalActions(0) # Take the pacman's state
    for action in state.getLegalActions(0): # For the pacman's actions
        newV = minimum(state.getNextState(
            0, action), 1, depth + 1, a, b) # record the value of minimum function on the pacman and the next depth
        if newV > v: # Update the value if the new one is greater than current
            v = newV
            if depth == 0: # And if at the first layer
                 best = action # Take this action
        a = max(a, v) # Save alpha as the maximum between alpha and current value
    if depth == 0: # If at the first layer
        return best
best = maximum(gameState, 0, float("-inf"), float("inf")) # Start by taking maximum with layer 0, large alpha beta
return best
# Begin your code (Part 4)
pacman = currentGameState.getPacmanPosition() # Get the pacman's position
foods = currentGameState.getFood().asList() # Store the position of the foods in a list
ghosts = currentGameState.getGhostStates() # Get the position of the ghosts
closestFoodDis = min(manhattanDistance(pacman, food)
                    for food in foods) if foods else 5 # Get the distance of the closest food
closestGhostDis = min(manhattanDistance(pacman, ghost.getPosition())
                     for ghost in ghosts) # Get the distance of the closest ghost
score = currentGameState.getScore() # Get the score
evaluation = 1.0 / closestFoodDis + score - closestGhostDis
2. add up the original score
3. Cut off the closest ghost's distance
return evaluation
```

Results & Analysis

```
Question part3
*** PASS: test_cases/part3/0-eval-function-lose-states-1.test
*** PASS: test_cases/part3/0-eval-function-lose-states-2.test
*** PASS: test_cases/part3/0-eval-function-win-states-1.test
*** PASS: test_cases/part3/0-eval-function-win-states-2.test
*** PASS: test_cases/part3/0-expectimax1.test
*** PASS: test_cases/part3/1-expectimax2.test
*** PASS: test_cases/part3/2-one-ghost-3level.test
*** PASS: test_cases/part3/3-one-ghost-4level.test
*** PASS: test_cases/part3/4-two-ghosts-3level.test
*** PASS: test_cases/part3/5-two-ghosts-4level.test
*** PASS: test_cases/part3/6-1a-check-depth-one-ghost.test
*** PASS: test_cases/part3/6-1b-check-depth-one-ghost.test
*** PASS: test_cases/part3/6-1c-check-depth-one-ghost.test
*** PASS: test_cases/part3/6-2a-check-depth-two-ghosts.test
*** PASS: test_cases/part3/6-2b-check-depth-two-ghosts.test
*** PASS: test_cases/part3/6-2c-check-depth-two-ghosts.test
*** Running ExpectimaxAgent on smallClassic 1 time(s).
Pacman died! Score: 84
Average Score: 84.0
                84.0
Scores:
Win Rate:
                0/1 (0.00)
Record:
                Loss
*** Finished running ExpectimaxAgent on smallClassic after 0 seconds.
*** Won 0 out of 1 games. Average score: 84.000000 ***
*** PASS: test_cases/part3/7-pacman-game.test
### Question part3: 25/25 ###
```

```
Question part4
Pacman emerges victorious! Score: 1098
Pacman emerges victorious! Score: 1152
Pacman emerges victorious! Score: 903
Pacman emerges victorious! Score: 1098
Pacman emerges victorious! Score: 1331
Pacman emerges victorious! Score: 969
Pacman emerges victorious! Score: 1325
Pacman emerges victorious! Score: 1160
Pacman emerges victorious! Score: 1132
Pacman emerges victorious! Score: 947
Average Score: 1111.5
                1098.0, 1152.0, 903.0, 1098.0, 1331.0, 969.0, 1325.0, 1160.0, 1132.0, 947.0 10/10 (1.00)
Scores:
Win Rate:
*** EXTRA CREDIT: 2 points
***
         1111.5 average score (4 of 4 points)
             Grading scheme:
***
***
              < 500: 0 points
        >= 500: 2 points
>= 1000: 4 points
10 games not timed out (2 of 2 points)
***
***
***
             Grading scheme:
***
             < 0: fail
>= 0: 0 points
>= 5: 1 points
>= 10: 2 points
***
***
***
***
         10 wins (4 of 4 points)
             Grading scheme:
***
             < 1: fail
>= 1: 1 points
***
             >= 4: 2 points
>= 7: 3 points
***
***
             >= 10: 4 points
### Question part4: 10/10 ###
```



For the evaluation function, when the position of the ghosts are not considered, I can get greater scores than the previously showed scores, I think this is because of the way I calculate it, I tried subtraction and fractions and subtraction worked better, but I think it is important to take this into consideration because we not only have to check the position of the food so I implemented it anyway.