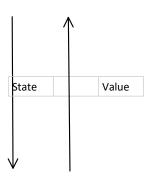
Init -> Create a dictionary (self.values)

- -> Fill out the dictionary based off of previous runs
- -> Loop over the number of iterations
- -> Loop over all the states
- -> Replace all the values in self.values with Helper[1]
- -> Create a copy of self.values (.copy())

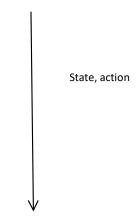


CAFV -> Return Helper()[0]



Helper -> Return a (Action, Value) (maxarg (a))

- -> Take in a state, Get all actions
- -> Loop over all the actions
- -> If state has no actions (None, 0.0)



CQVFV -> mdp.getTransitonStatesAndProbs -> R(s) + discount(sum(T(s,a,s')) * U(s')

Question 4

Init -> Create a dictionary (self.whatever = util.Counter)

Init -> Create a dictionary (self.whatever = util.Counter)

GetQValue -> return self.whatever[(state,action)]

CVFQV -> given a state, get the list of action

- -> Iterate over all the actions (edge case of no actions)
- -> Return 0.0 if no actions
- -> Call GetQValue on all (state, action) pairs and take the max

CAFQV -> given a state, list of actions

- -> Iterate over all the actions
- -> No actions = return none
- -> Call GetQValue and return the max action
- -> Potential for multiple equivalent bests actions
- -> Choose a random of those best actions (random.choice())

Update -> self.whatever[(state,action)] +=

-> self.alpha * (reward + (self.discount * self.getValue(nextState)) - self.whatever[(state,action)]

Question 5
GetAction -> return getPolicy(state) (CAFQV)
-> self.epsilon

- -> util.coinFlip(epsilon) (same as below, eval probability)
- -> random.random() < self.epsilon -> Take a random action
- -> Random action is random.choice(legalActions)
- -> What happens when no actions -> return None