$q[a] = p4*(0 + \gamma*V[a+s+1]) + (1-p4)*(0 + \gamma*V[s-a+1])$ end V[s+1], greedy\_action = findmax(q)  $\pi[s+1] = greedy_action$  $\delta = \max(\delta, \text{ abs}(v - V[s+1]))$ end if  $\delta < \theta$ break end end return V,  $\pi$ # output the optimal value function and a deterministic policy end pi =zeros(101) V = zeros(length(S)) # will hold the value function V[101] = 1V, pi= value\_iteration(V, pi) println("capital 51: ", pi[52])#at capital 51 println("capital 50: ", pi[51]) #at capital 50 println("capital 49: ", pi[50]) #at capital 49 capital 51: 1.0 capital 50: 50.0 capital 49: 1.0 Two optimal policies π1 and π2 can lead to the same expected return (a.k.a state value function) even though they take different actions in a specific state s. Hence, even though the optimal policy that you derived takes a different action compared to the text book, the sequence of actions eventually leads to the same expected return (a.k.a state value function), hence both policies are optimal. In [2]: using Plots  $\alpha = range(1,101)$  $plot(\alpha, pi[\alpha], xlabel="Capital", ylabel="Final policy")$ Out[2]: 50 - y1 40 Final policy 10 25 50 75 100 Capital In [3]:  $plot(\alpha, V[\alpha], xlabel="Capital", ylabel="Value estimates")$ Out[3]: 1.00 0.75 Value estimates 0.25 0.00 50 75 25 100 Capital In [4]: # this is on-policy every-visit MC control because we do not check for # 1st visits to states; however, for the game of blackjack, it's not # possible to visit the same state twice in an episode. you could have # an ace being counted as 11, and then later being counted as 1, but # the indicator for a usable ace is part of the state. also, note # that we maintain exploration of nonoptimal actions in the function # blackjack(). # solving the black jack game described in example 5.1 using Random, StatsBase # the face values for a suit # ace through nine, ten, jack, queen, king # note that 1=ace suit = vcat(1, collect(2:9), 10, 10, 10, 10)# a deck of cards consists of four suits: diamonds, clubs, hearts, spades deck = repeat(suit,4) shuffle!(deck) # a random permutation # to simulate an infinite deck, we can sample with replacement deal\_cards(n) = sample(deck, n, replace=true) # note that it's quite possible (in fact it's common) to have more than # one ace in a hand, but it's not possible to have two "usable" aces. # 1/2 = no/yesusable\_ace(hand) = (any(hand .== 1) && sum(hand) <= 11) ? 2 : 1score(hand) = sum(hand) + (usable\_ace(hand)==2 ? 10 : 0) # simulate an episode of blackjack according to policy  $\pi$ function blackjack( $\pi$ ,  $\in$  = 0.05) player = deal\_cards(2) dealer = deal\_cards(2) action = [] while score(player) < 12</pre> append!(player, deal\_cards(1)) end states = [[score(player), dealer[1] ,usable\_ace(player)]] if states[1] == 21 && states[3] == 2 push!(action, 2) if score(dealer) == 21 && usable\_ace(dealer) == 2 return states, action, 0 else return states, action, 1 end end while true  $a = \pi[states[length(states)][1]-11, states[length(states)][2], states[length(states)][3]] # determine the action a according to policy$ if rand()  $< \epsilon$ **if** a == 1 a=2 else a = 1 end end **if** a == 2 push!(action,a) break else push!(action, a) append!(player, deal\_cards(1)) if score(player) > 21 return states, action, -1.0 else push!(states,[ score(player), dealer[1] , usable\_ace(player)]) end end while score(dealer)<17</pre> append!(dealer, deal\_cards(1)) end if score(dealer) > 21 | states[length(states)][1]> score(dealer) elseif states[length(states)][1] < score(dealer)</pre> r = -1.0else r = 0.0end return states, action, r end # this is on-policy every-visit MC control because we do not check for # 1st visits to states; however, for the game of blackjack, it's not # possible to visit the same state twice in an episode. you could have # an ace being counted as 11, and then later being counted as 1, but # the indicator for a usable ace is part of the state. also, note # that we maintain exploration of nonoptimal actions in the function # blackjack(). function MC!  $(q, qn, \pi, S)$ **for** i = 1:10e6 i % 100000 == 0 && println("episode ", i) states, actions,  $r = blackjack(\pi)$ @assert(length(states) == length(actions)) @assert( r in [-1.0, 0.0, 1.0]) T = length(states) **for** t = 1:T s = states[t]a= actions[t] qn[s[1]-11, s[2], s[3],a] += 1 #updating qnq[s[1]-11, s[2], s[3], a] += r #updating qS[s[1]-11, s[2], s[3]], maxaction = findmax( q[s[1]-11, s[2], s[3],:]./ qn[s[1]-11, s[2], s[3],:])# find the action that maxmizes q for state s $\pi[s[1]-11, s[2], s[3]] = maxaction # update the policy greedily$ end end end # the state space consists of # the player's sum 12:21, # the dealer's showing card 1:10, # and indicator for usable ace no/yes=1/2 # for a total of 200 possible states. # the action is hit/stick=1/2 # the initial policy is to stick when the player's sum is 20 or 21, otherwise hit  $\pi = \text{fill}(1, (10, 10, 2))$  $\pi[9:10,:,:] = 2$ # stick when sum is 20 or 21 q = zeros(10, 10, 2, 2) # q(state=(player, dealer, usable), action)qn = zeros(10, 10, 2, 2) # to hold the number of observations S = zeros(10, 10, 2) $MC!(q, qn, \pi, S)$ episode 100000.0 episode 200000.0 episode 300000.0 episode 400000.0 episode 500000.0 episode 600000.0 episode 700000.0 episode 800000.0 episode 900000.0 episode 1.0e6 episode 1.1e6 episode 1.2e6 episode 1.3e6 episode 1.4e6 episode 1.5e6 episode 1.6e6 episode 1.7e6 episode 1.8e6 episode 1.9e6 episode 2.0e6 episode 2.1e6 episode 2.2e6 episode 2.3e6 episode 2.4e6 episode 2.5e6 episode 2.6e6 episode 2.7e6 episode 2.8e6 episode 2.9e6 episode 3.0e6 episode 3.1e6 episode 3.2e6 episode 3.3e6 episode 3.4e6 episode 3.5e6 episode 3.6e6 episode 3.7e6 episode 3.8e6 episode 3.9e6 episode 4.0e6 episode 4.1e6 episode 4.2e6 episode 4.3e6 episode 4.4e6 episode 4.5e6 episode 4.6e6 episode 4.7e6 episode 4.8e6 episode 4.9e6 episode 5.0e6 episode 5.1e6 episode 5.2e6 episode 5.3e6 episode 5.4e6 episode 5.5e6 episode 5.6e6 episode 5.7e6 episode 5.8e6 episode 5.9e6 episode 6.0e6 episode 6.1e6 episode 6.2e6 episode 6.3e6 episode 6.4e6 episode 6.5e6 episode 6.6e6 episode 6.7e6 episode 6.8e6 episode 6.9e6 episode 7.0e6 episode 7.1e6 episode 7.2e6 episode 7.3e6 episode 7.4e6 episode 7.5e6 episode 7.6e6 episode 7.7e6 episode 7.8e6 episode 7.9e6 episode 8.0e6 episode 8.1e6 episode 8.2e6 episode 8.3e6 episode 8.4e6 episode 8.5e6 episode 8.6e6 episode 8.7e6 episode 8.8e6 episode 8.9e6 episode 9.0e6 episode 9.1e6 episode 9.2e6 episode 9.3e6 episode 9.4e6 episode 9.5e6 episode 9.6e6 episode 9.7e6 episode 9.8e6 episode 9.9e6 episode 1.0e7 In [20]: import Plots x=range(1,11) y = range(1,11)Plots.heatmap(x , y.+11,  $\pi$ [:, :,2]) #heatmap of usable ace x axis = dealer showing, yaxis = player sum 12 to 21 Out[20]: 22.5 **-2.0** -1.9 -1.8 20.0 -1.7 -1.6 17.5 -1.5 -1.4 15.0 -1.3 -1.2 -1.112.5 6 8 10 2 In [17]: x = range(1,11)y = range(1,11)Plots heatmap (x, y.+ 11,  $\pi$ [:, :,1]) #heatmap of nonusable ace Out[17]: 22.5 **-2.0** -1.9 -1.8 20.0 -1.7 -1.6 17.5 -1.5 -1.4 15.0 -1.3 -1.2 -1.112.5 2 6 10 4 8 In [7]: # some helpful things in Julia? using Plots # two functions named f exist. they have different signatures y = range(1,10) #for the player x= range(1,10) #for the dealer Plots.surface(x, y.+11, S[y,x,1]) #non usable ace consider x = current value showing + 11 because x is the player sum Out[7]: -0.75 -0.50 0.75 -0.25 0.50 0.25 0.00 -0.25-0.25-0.50--0.5C 10 12 In [8]: Plots.surface(x, y.+11, S[y,x,2]) #surface plot of the usable ace Out[8]: -0.75 -0.50 0.75 0.50 -0.25 0.25 0.00 -0.25

--0.25

In [ ]:

In [1]: using DelimitedFiles

using Main

p4 = .4

using StatsBase

using Distributions
using LinearAlgebra

while true

 $\delta = 0.0$ 

**function** value\_iteration(V,  $\pi$ )  $\gamma$  = 1 # discount factor

S = collect(1:101)# 15 is the terminal state

 $\theta$  = 1e-3 # tolerance for convergence

q = zeros(length(A))

for a in A

v = V[s+1] # old value

for s = 1:99 # loop through the nonterminal states

A = collect(1:findmin([s,100-s])[1])