



# ECON408: Computational Methods in Macroeconomics

*Search Models of Unemployment and Dynamic Programming*

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# Overview

# Motivation

- In the previous lecture we described a model with employment and unemployment as a Markov Chain
- Central to that were arrival rates of transitions between the  $U$  and  $E$  states at some  $\lambda$  probability each period.
- But the worker didn't have a choice whether to accept the job or not. In this lecture we will investigate a simple model where workers search for jobs, and the  $\lambda$  becomes an endogenous choice
- As with the previous lecture on the [Permanent Income model](#), the benefit is that we can consider policy counterfactuals which may affect the workers choices
- Finally, we will review fixed points and connect it to Bellman Equations more formally

# Materials

- Adapted from QuantEcon lectures coauthored with John Stachurski and Thomas J. Sargent
  - Job Search I: The McCall Search Model
  - Job Search II: Search and Separation
  - A Lake Model of Employment and Unemployment

```
1 using LinearAlgebra, Statistics
2 using Distributions, LaTeXStrings
3 using Plots.PlotMeasures, NLsolve, Roots, Random, Plots
4 default(;legendfontsize=16, linewidth=2, tickfontsize=12,
5           bottom_margin=15mm)
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# The McCall Model

# Summary

- See [here](#) for a more minimal version
- The McCall model is a model of “search” in the labor market
- A worker can be in a state of unemployment or employment. at wage  $W_t$
- The key decision: if they are unemployed and receive a job offer at wage  $W_t$ , should they accept it or keep searching?
- Assume that wages come from a fixed, known distribution with discrete values  $w' \in \{w_1, \dots, w_N\}$  and probabilities  $\{p_1, \dots, p_N\}$

# Preferences

- Let  $Y_t$  be the stochastic payoffs of the consumer
  - $Y_t = W_t$  if employed at wage  $W_t$
  - $Y_t = c$ , unemployment insurance, while unemployed
- Let  $u(\cdot)$  be a standard utility function with  $u'(\cdot) > 0$  and  $u''(\cdot) \leq 0$
- Preferences over stochastic incomes  $\{Y_{t+j}\}_{j=0}^{\infty}$  given time  $t$  information are

$$\mathbb{E}_t \sum_{j=0}^{\infty} \beta^j u(Y_{t+j})$$

- We will try to rewrite this recursively just as we did when calculating PDVs



# Timeline and Decisions

- If they are employed at wage  $w$  they
  - get the wage  $w$  and enter the next period.
  - have some probability  $\alpha$  the job ends and they become unemployed before next period starts, otherwise the  $w$  does not change
- If they are unemployed they
  - get unemployment insurance,  $c$
  - have some probability,  $\gamma$  of getting a wage offer,  $w'$  for the next period.
  - Given some wage offer,  $w'$ , they can choose to accept the job and enter employment, or to reject the offer and remain unemployed
  - Cannot recall previously rejected wages (but wouldn't, in equilibrium)

# A Trade Off

- The worker faces a trade-off:
  - Waiting too long for a good offer is costly, since the future is discounted.
  - Accepting too early is costly, since better offers might arrive in the future.
- To decide optimally in the face of this trade off, we use **dynamic programming**
- The key is to start with the state:
  - employment status
  - the current wage,  $w$ , if employed
- Then determine the feasible set of actions, and how the state evolves under each action

# Value Functions

- All unemployment states are the same since nothing is changing over time and the previous wages are irrelevant
  - Hence, let  $U$  be the value of being unemployed
- The only thing that matters for the employed worker is the current wage,  $w$ 
  - Hence, let  $V(w)$  be the value of being employed at wage  $w$
- We will write recursive equations for these value functions
- Recursive equations defining the value of being in a state today in terms of the value of different states tomorrow are [Bellman Equations](#)

# Actions and Transitions if Employed

- The employed agent is passive
- While employed, at the end of the period
  - With some probability  $\alpha$  they become unemployed with value  $U$
  - Otherwise, they remain employed with value  $V(w)$
- In fancier models, they could engage in [on the job search](#), etc.

# Actions and Transitions if Unemployed

- While unemployed, at the end of the period
  - With some probability  $\gamma$  they get a wage offer  $w'$
  - If they **accept** they enter employment at wage  $w'$  next period (i.e.  $V(w')$ )
  - If they didn't get an offer, or **rejected** offer  $w'$ , they remain unemployed with values  $U$

# Summarizing Value Functions

- The value function for the unemployed worker is

$$\begin{aligned} U &= u(c) + \beta [(1 - \gamma)U + \gamma \mathbb{E}[\max\{U, V(w')\}]] \\ &= u(c) + \beta \left[ (1 - \gamma)U + \gamma \sum_{i=1}^N \max\{U, V(w_i)\} p_i \right] \end{aligned}$$

- The value function for the employed worker is

$$V(w) = u(w) + \beta [(1 - \alpha)V(w) + \alpha U]$$

- These are the **Bellman Equations** for the McCall model

# Reorganize as a Fixed Point

- Since there are only  $N$  values of  $w_i$ , we can define  $V(w_i) \equiv V_i$  and solve for  $N + 1$  values (i.e.,  $V_1, \dots, V_N, U$ )
- Let  $X \equiv [V_1 \ \dots \ V_N \ U]^\top$
- Define the Bellman Operator  $T : \mathbb{R}^{N+1} \rightarrow \mathbb{R}^{N+1}$  stacking Bellman Equations

$$T(X) \equiv \begin{bmatrix} u(w_1) + \beta [(1 - \alpha)V_1 + \alpha U] \\ \vdots \\ u(w_N) + \beta [(1 - \alpha)V_N + \alpha U] \\ u(c) + \beta \left[ (1 - \gamma)U + \gamma \sum_{i=1}^N \max\{U, V_i\} p_i \right] \end{bmatrix}$$

- Then the fixed point of  $T(\cdot)$  (i.e.,  $T(X) = X$ ) is the solution to the problem

# Alternative Formulation

- As practice with Bellman Equations, consider an alternative formulation which is qualitatively identical
- Instead of writing the choice while in the  $U$  state between periods, you can write it with a single Value Function  $V(w)$ 
  - In that case, a state of unemployment must be mapped into the framework.
  - For example, you could have a wage offer of  $0$  (although  $c$  would also work)



# Alternative Formulation Bellman Equation

- The  $V(w)$  is now the value of having a wage offer, not of choosing to work at that offer

$$V(w) = \max \left\{ u(w) + \beta [(1 - \alpha)V(w) + \alpha V(0)], \right. \\ \left. u(c) + \beta [(1 - \gamma)V(0) + \gamma \mathbb{E}(V(w'))] \right\}$$

- Let  $\bar{w}$  the minimum  $w_i$  such that

$$u(w_i) + \beta [(1 - \alpha)V(w_i) + \alpha V(0)] > u(c) + \beta [(1 - \gamma)V(0) + \gamma \mathbb{E}(V(w'))]$$

- Then we can interpret this as the **reservation wage**. If the  $w'$  were distributed continuously, then it might be an exact wage at equality
- With this, the  $\bar{w}$  will be a kink in the  $V(w)$  function at  $\bar{w}$

# Bellman Operators and Fixed Points

# Bellman Equations and Fixed Points

- Before we solve the McCall model, we need to review Fixed Points now that we have more tools
- Given a dynamic programming problem written as a Bellman equation, one common approach is to organize it as a fixed point
  - Write the Bellman equation as  $V = T(V)$  for some operator  $T$
  - Check if  $T$  is a [contraction mapping](#)
- In general, this is a fixed point in a function space
  - If the state space is continuous, you will need to discretize  $V$  and/or the state space(e.g., the [Tauchen Method](#))
  - If it is already discrete states, then the functions usually just map indices to values (i.e., can represent as a vector)

# Algorithms

- Given a Bellman Equation  $V = T(V)$ , we can solve for  $V$  through a variety of algorithms. For example,
  - Guess  $V^0$  and **iterate**  $V^{n+1} = T(V^n)$  until convergence, called **Value Function Iteration (VFI)**
  - Alternatively, solve the fixed point problem using some specialized algorithm
- One advantage of VFI is that the **Banach Fixed Point Theorem** shows uniqueness of the algorithm, even if it is not always the fastest approach
- In fact, we have already used this for solving simple **asset pricing problems**

# Repeat of Markov Asset Pricing as a Fixed Point

- Lets re-do [that exercise](#) as practice with a few additions
- Payoffs are in  $\mathbf{y} \equiv [y_L \quad y_H]^\top$ 
  - $\mathbb{P}(y_{t+1} = y_H | y_t = y_L) = \alpha$
  - $\mathbb{P}(y_{t+1} = y_L | y_t = y_H) = \gamma$
- Instead of linear utility, assume risk-averse utility  $u(y) = \frac{y^{1-\sigma}-1}{1-\sigma}$  for  $\sigma \geq 0$
- Because the process is Markov and payoffs do not depend on time, we can write this recursively

$$p(y) = u(y) + \beta \mathbb{E} [p(y') | y]$$

# Expanding out as a Fixed point

- But there are only 2 possible states, so  $p \equiv [p_L \quad p_H]^\top \in \mathbb{R}^2$
- Rewriting this as a system of equations

$$\begin{aligned} p_L &= u(y_L) + \beta \mathbb{E} [p(y') | y_L] = u(y_L) + \beta [(1 - \alpha)p_L + \alpha p_H] \\ p_H &= u(y_H) + \beta \mathbb{E} [p(y') | y_H] = u(y_H) + \beta [\gamma p_L + (1 - \gamma)p_H] \end{aligned}$$

- Stack  $p \equiv [p_L \quad p_H]^\top$  and  $u_y \equiv [u(y_L) \quad u(y_H)]^\top$

$$p = u_y + \beta \begin{bmatrix} 1 - \alpha & \alpha \\ \gamma & 1 - \gamma \end{bmatrix} p \equiv T(p)$$

- Then the fixed point of  $T(\cdot)$  (i.e.,  $T(p) = p$ ) is the solution to the problem

# Solving Numerically with a Fixed Point

```
1 y = [3.0, 5.0] #y_L, y_H
2 sigma = 0.5
3 u_y = (y.^(1-sigma) .- 1) / (1-sigma) # CRRA utility
4 beta = 0.95
5 alpha = 0.2
6 gamma = 0.5
7 iv = [0.8, 0.8]
8 A = [1-alpha alpha; gamma 1-gamma]
9 sol = fixedpoint(p -> u_y .+ beta * A * p, iv) # T(p) := u_y + beta A p
10 p_L, p_H = sol.zero # can unpack a vector
11 @show p_L, p_H, sol.iterations
12 @show (I - beta * A) \ u_y;
```

```
(p_L, p_H, sol.iterations) = (34.63941760551722, 36.04925584308621, 4)
(I - beta * A) \ u_y = [34.63941760551727, 36.04925584308624]
```

# Decisions and Valuations

- We have shown the importance of Markovian assumptions to ensure tractability
  - If decisions only depend on the current state, and not the time itself, and the state is Markovian, then we can write a recursive problem.
- This approach is especially powerful when agent's need to make a **decision** given their state - as in the McCall model
  - As always, in economics we usually implement decisions as optimization/maximization problems
  - The key is that the  $T(\cdot)$  operator can itself be complicated, and include constrained maximization/etc.



# McCall Solution

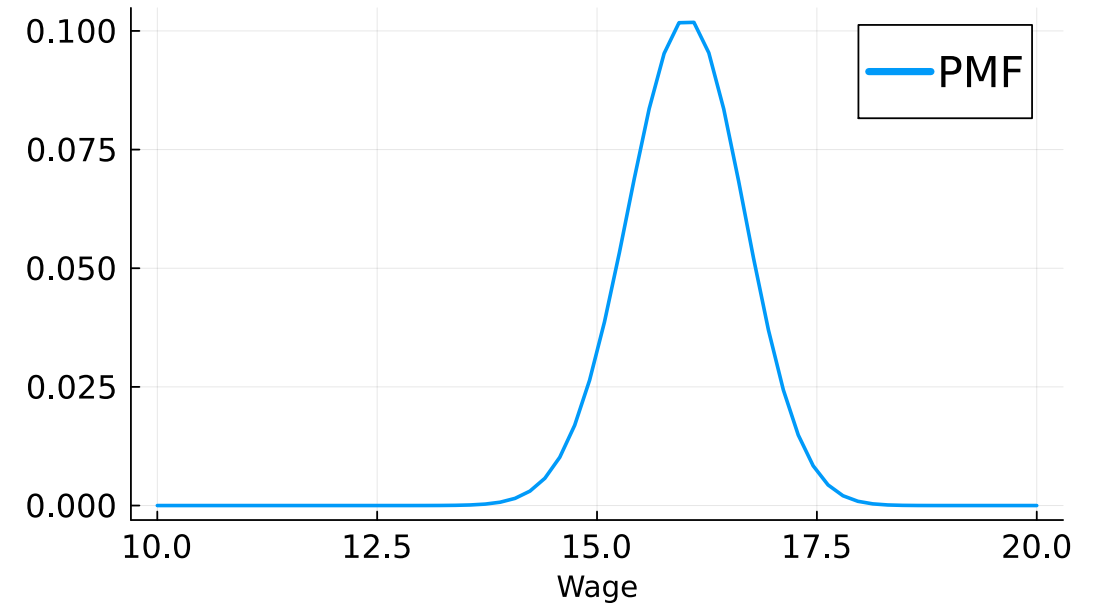
# McCall Parameters

- Choose some distribution for wages, such as [BetaBinomial](#)
- CRRA Period utility:  $u(c) = \frac{c^{1-\sigma}-1}{1-\sigma}$ . Nests the  $\lim_{\sigma \rightarrow 1} \frac{c^{1-\sigma}-1}{1-\sigma} = \log(c)$

```
1 function mccall_model(;
2     alpha = 0.2, # prob lose job
3     beta = 0.98, # discount rate
4     gamma = 0.7, # prob job offer
5     c = 6.0, # unemployment compensation
6     sigma = 2.0, # CRRA parameters
7     w = range(10, 20, length = 60), # wage values
8     p = pdf.(BetaBinomial(59, 600, 400), 0:length(w)-1)) # probs for each wage
9     # why the c <= 0 case? Makes it easier when finding equilibrium
10    u(c, sigma) = c > 0 ? (c^(1 - sigma) - 1) / (1 - sigma) : -10e-6
11    u_c = u(c, sigma)
12    u_w = u.(w, sigma)
13    return (; alpha, beta, sigma, c, gamma, w, p, u_w, u_c)
14 end
```

# Wage Distribution

```
1 mcm = mccall_model()  
2 plot(mcm.w, mcm.p; xlabel = "Wage",  
3      label = "PMF", size = (600, 400))
```



# Reminder: Value Functions

- The value function for the unemployed worker is

$$\begin{aligned} U &= u(c) + \beta [(1 - \gamma)U + \gamma \mathbb{E}[\max\{U, V(w')\}]] \\ &= u(c) + \beta \left[ (1 - \gamma)U + \gamma \sum_{i=1}^N \max\{U, V(w_i)\} p_i \right] \end{aligned}$$

- The value function for the employed worker is

$$V(w) = u(w) + \beta [(1 - \alpha)V(w) + \alpha U]$$

- These are the **Bellman Equations** for the McCall model

# Bellman Operator

- Given the stacked  $V$  and  $U$  we can implement the  $T : \mathbb{R}^{N+1} \rightarrow \mathbb{R}^{N+1}$  operator

```
1 function T(X;mcm)
2     (;alpha, beta, gamma, c, w, p, u_w, u_c) = mcm
3     V = X[1:end-1]
4     U = X[end]
5     V_p = u_w + beta * ((1 - alpha) * V .+ alpha * U)
6     # Or, expanding out with a comprehension
7     # V_p = [ u_w[i] + beta * ((1 - alpha) * V[i] + alpha * U) for i in 1:length(w)]
8     U_p = u_c + beta * (1 - gamma) * U + beta * gamma * sum(max(U, V[i]) * p[i] for i in 1:length(w))
9     return [V_p; U_p]
10 end
```

# Reservation Wage

- The reservation wage is the wage at which the worker is indifferent between accepting and rejecting the offer
- In the case of a continuous wage distribution, it may be an exact state
- However, with a discrete number of wages it will usually lie between two wages
- Define the reservation wage as the smallest wage such that the worker would accept the offer
  - In the problem, this is the smallest  $\bar{w}$  such that  $\max\{U, V(\bar{w})\} > U$
  - Given a  $V$  and  $U$  vector we find the index where  $V_i - U > 0$

# Solution

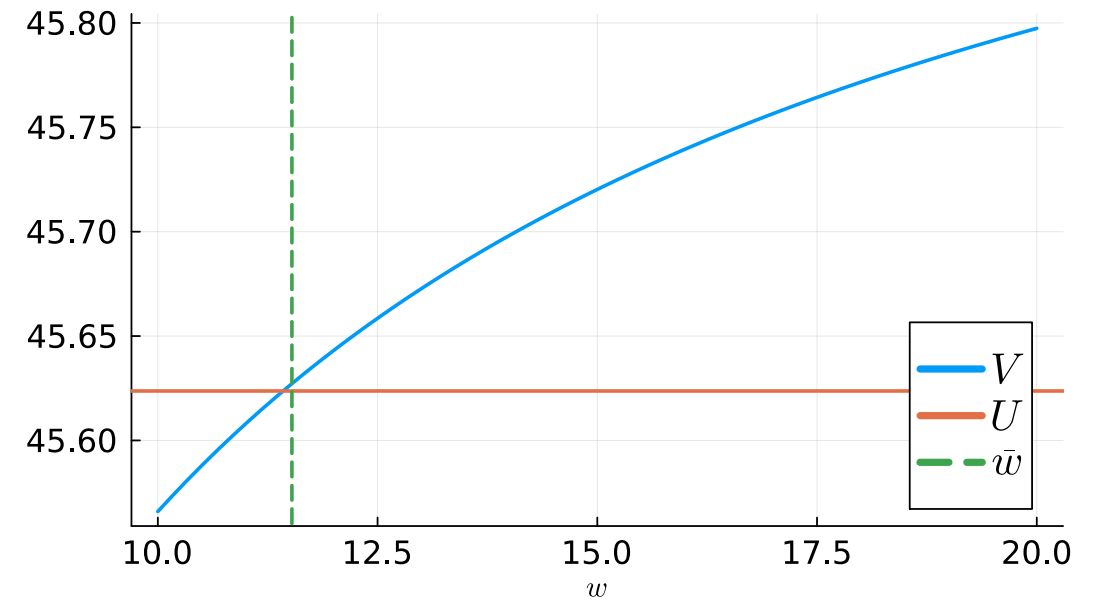
```
1 function solve_mccall_model(mcm; U_iv = 1.0, V_iv = ones(length(mcm.w)), tol = 1e-5, iter = 2_000)
2     (; alpha, beta, sigma, c, gamma, w, sigma, p) = mcm
3     x_iv = [V_iv; U_iv] # initial x val
4     xstar = fixedpoint(X -> T(X;mcm), x_iv, iterations = iter, xtol = tol, m = 0).zero
5     V = xstar[1:end-1]
6     U = xstar[end]
7
8     # compute the reservation wage
9     wbarindex = searchsortedfirst(V .- U, 0.0)
10    if wbarindex >= length(w) # if this is true, you never want to accept
11        w_bar = Inf
12    else
13        w_bar = w[wbarindex] # otherwise, return the number
14    end
15    return (;V, U, w_bar)
16 end
```

# Results

```

1 mcm = mccall_model()
2 sol = solve_mccall_model(mcm)
3 plot(mcm.w, sol.V; label = L"V",
4       xlabel=L"w", size=(600, 400))
5 hline!(mcm.w, [sol.U], label=L"U")
6 vline!([sol.w_bar]; linestyle = :dash,
7         label = L"\bar{w}")

```





# Interpretation

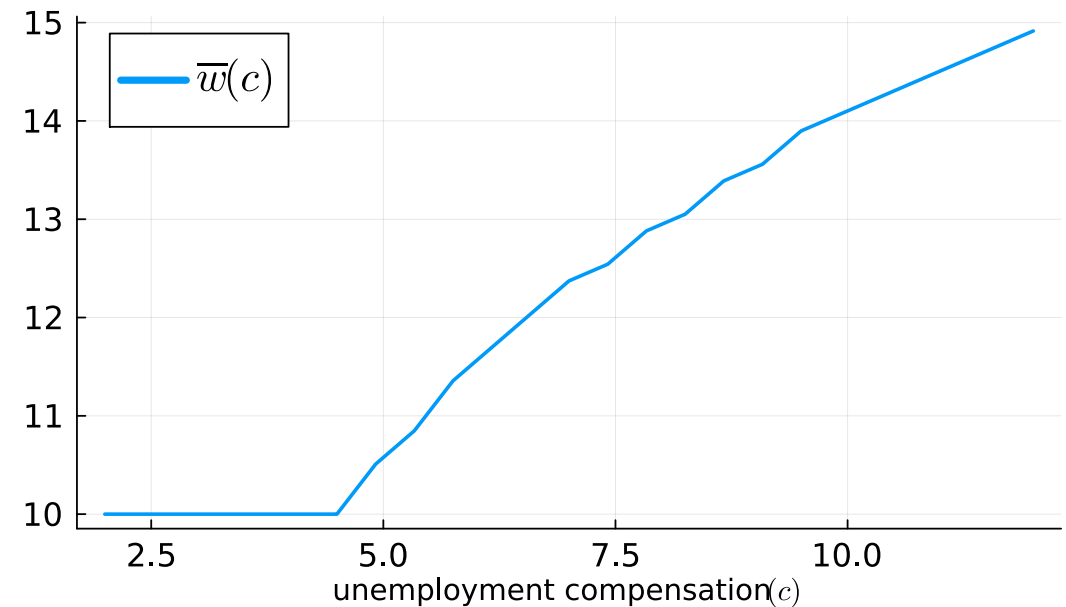
- The value of being employed is increasing in the wage, but has concavity due to the CRRA utility
- The value of unemployment is constant since the
  - wage distribution is fixed over time
  - the unemployment insurance is independent of previous wages
- While the agent can calculate the  $V(w)$  for  $w < \bar{w}$ , when thinking through decisions, they will never accept a wage offer below the reservation wage
- Lets do further analysis of the reservation wage through **comparative statics** (i.e., modifying parameters)

# Comparative Statics

# Changing Unemployment Insurance

- Change in  $c$  affect value of searching for new job
- Too high and they will never accept a job, too low and they accept every job

```
1 c_vals = range(2, 12, length = 25)
2 w_bars = [solve_mccall_model(
3             mccall_model(;c)).w_bar
4             for c in c_vals]
5 plot(c_vals, w_bars;
6       xlabel = L"unemployment compensation$(c)",
7       label = L"\overline{w}(c)",
8       size = (600, 400))
```



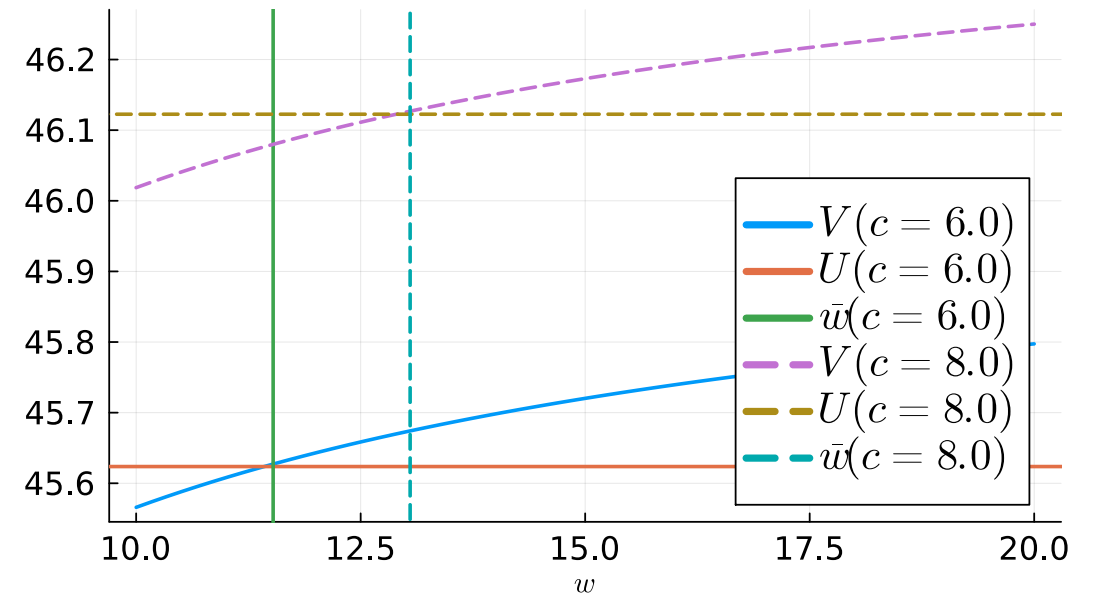
# Analyzing Value Functions

- Why does the  $V(w)$  change with  $c$ ?

```

1 mcm = mccall_model()
2 sol = solve_mccall_model(mcm)
3 plot(mcm.w, sol.V; label = L"V(c=6.0)",
4      xlabel=L"w", size=(600, 400))
5 hline!(mcm.w, [sol.U], label=L"U(c=6.0)")
6 vline!([sol.w_bar];
7        label = L"\bar{w}(c=6.0)")
8 mcm2 = mccall_model(c = 8.0)
9 sol2 = solve_mccall_model(mcm2)
10 plot!(mcm2.w, sol2.V; label = L"V(c=8.0)",
11        linestyle = :dash)
12 hline!(mcm2.w, [sol2.U], label=L"U(c=8.0)",
13        linestyle = :dash,)
14 vline!([sol2.w_bar]; linestyle = :dash,
15        label = L"\bar{w}(c=8.0)")

```

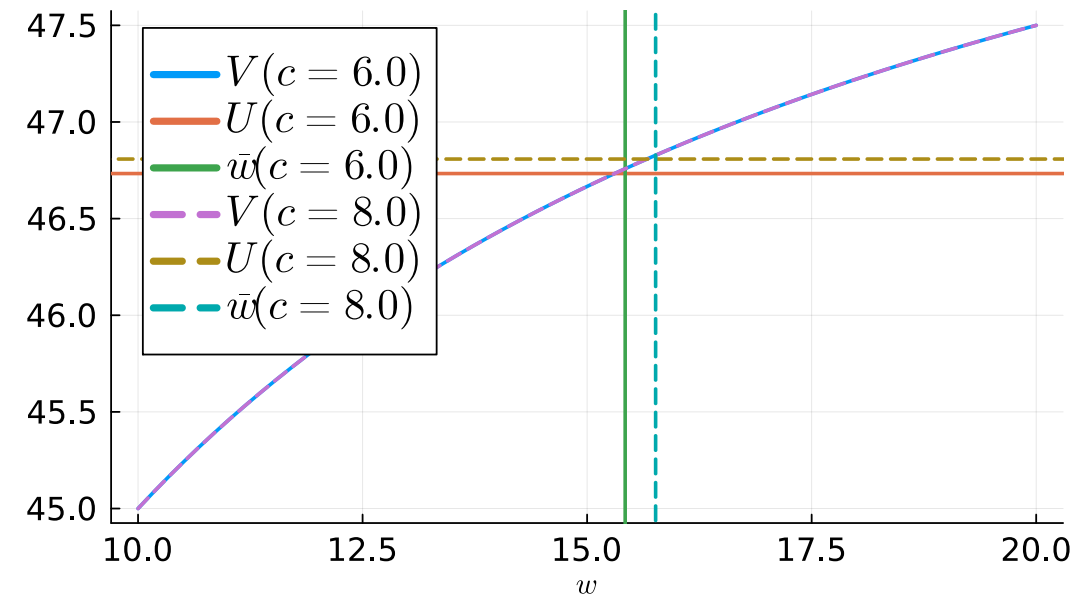


# Analyzing Value Functions ( $\alpha = 0$ )

```

1 mcm = mccall_model(;alpha = 0.0)
2 sol = solve_mccall_model(mcm)
3 plot(mcm.w, sol.V; label = L"V(c=6.0)",
4      xlabel=L"w", size=(600, 400))
5 hline!(mcm.w, [sol.U], label=L"U(c=6.0)")
6 vline!([sol.w_bar];
7        label = L"\bar{w}(c=6.0)")
8 mcm2 = mccall_model(;c = 8.0, alpha = 0.0)
9 sol2 = solve_mccall_model(mcm2)
10 plot!(mcm2.w, sol2.V; label = L"V(c=8.0)",
11        linestyle = :dash)
12 hline!(mcm2.w, [sol2.U], label=L"U(c=8.0)",
13        linestyle = :dash,)
14 vline!([sol2.w_bar]; linestyle = :dash,
15        label = L"\bar{w}(c=8.0)")

```



# Changing Job Destruction Rate

- How would  $\alpha$  affect reservation wages? Why?

```
1 alpha_vals = range(0.2, 0.5, length = 25)
2 w_bars = [solve_mccall_model(
3             mccall_model(;alpha)).w_bar
4             for alpha in alpha_vals]
5 plot(alpha_vals, w_bars;
6       xlabel = L"job destruction rate  $(\alpha)$ ",
7       label = L"\overline{w}(\alpha)",
8       size = (600, 400))
```

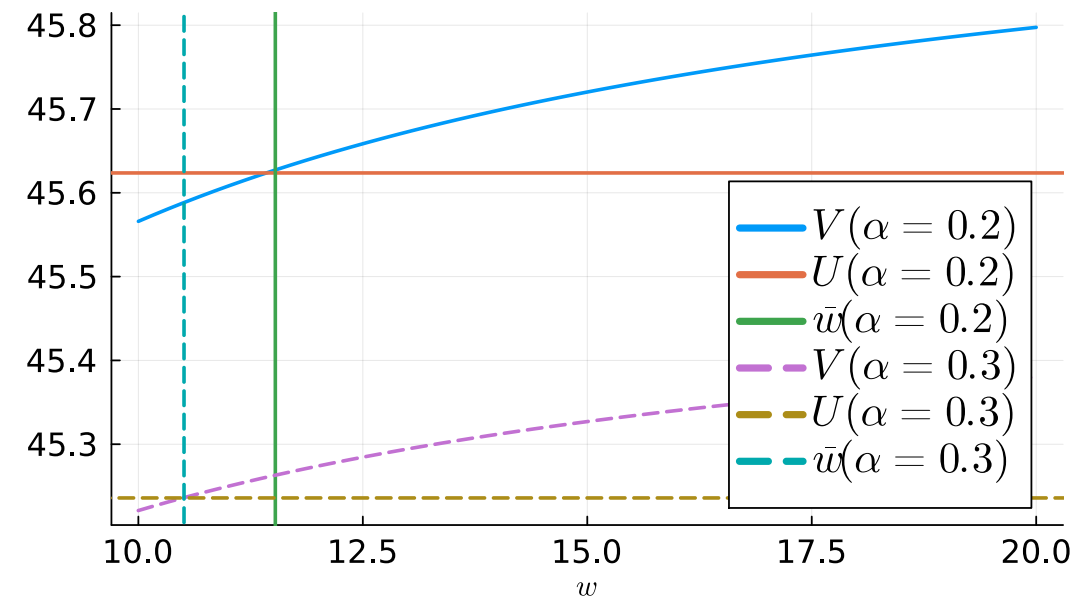


# Job Destruction Rate

```

1 mcm = mccall_model()
2 sol = solve_mccall_model(mcm)
3 plot(mcm.w, sol.V; label = L"V(\alpha=0.2)",
4       xlabel=L"w", size=(600, 400))
5 hline!(mcm.w, [sol.U], label=L"U(\alpha=0.2)")
6 vline!([sol.w_bar];
7         label = L"\bar{w}(\alpha=0.2)")
8 mcm2 = mccall_model(;alpha = 0.3)
9 sol2 = solve_mccall_model(mcm2)
10 plot!(mcm2.w, sol2.V; label=L"V(\alpha=0.3)",
11        linestyle = :dash)
12 hline!(mcm2.w, [sol2.U], label=L"U(\alpha=0.3)",
13        linestyle = :dash,)
14 vline!([sol2.w_bar]; linestyle = :dash,
15         label = L"\bar{w}(\alpha=0.3)")

```



# Connecting $\bar{w}$ to the Unemployment Rate

- Going back to our [Lake Model](#)
- Recall that the probability to transition from  $E$  to  $U$  is  $\lambda$
- Our model of search leads to an endogenous choice of  $\lambda$ . In particular, while in  $U$ ,
  - With probability  $\gamma$  they get a wage offer
  - Conditional on a wage offer, the probability to accept a wage is the probability that the wage is above the reservation wage

$$\lambda = \gamma \mathbb{P}(w > \bar{w}) = \gamma \sum_{i=1}^N \mathbb{1}(w_i > \bar{w}) p_i$$



# Tax Policy

# Government Budget and Fiscal Policy

- Unemployment insurance is a transfer from the government to the unemployed
- We will assume that:
  - The government finances the unemployment insurance through a lump-sum tax  $\tau$  on all wages.
  - We can subtract it from  $c$  as well for simplicity
  - Must balance the budget at the steady-state level
- If there are a normalized measure  $\mathbf{1}$  in the economy, then the government budget constraint is

$$\tau = \bar{u}c$$

# Wage Conditional on Employment

- Given that the consumer values their wage or unemployment with  $V(w)$  and  $U$ , a benevolent planner would use the same criteria
- Let  $\bar{e}$  and  $\bar{u}$  be the steady state fraction of employed and unemployed individuals
- Since consumers would never accept a  $w < \bar{w}$  we know the wage distribution **conditional on being employed** is

$$\mathbb{P}(w_i | w_i > \bar{w}) = \frac{p_i}{\sum_{j=i}^N p_j}$$

# Aggregate Welfare

- The aggregate welfare, given  $U$  and  $V(w)$  depends on the steady state fraction of unemployed and employed individuals

$$W \equiv \bar{u}U + \bar{e}\mathbb{E}[V(w)|w > \bar{w}]$$

# Reminder: Lake Model

```
1 function lake_model(; lambda = 0.283, alpha = 0.013, b = 0.0124, d = 0.00822)
2     g = b - d
3     A = [(1 - lambda) * (1 - d) + b (1 - d) * alpha + b
4           (1 - d) * lambda (1 - d) * (1 - alpha)]
5     A_hat = A ./ (1 + g)
6     x_0 = ones(size(A_hat, 1)) / size(A_hat, 1)
7     sol = fixedpoint(x -> A_hat * x, x_0)
8     converged(sol) || error("Failed to converge in $(sol.iterations) iter")
9     x_bar = sol.zero
10    return (; lambda, alpha, b, d, A, A_hat, x_bar)
11 end
```

# Computing Steady State Quantities with Possibly Unbalanced Budget

```
1 function compute_optimal_quantities(c_pretax, tau; p, sigma, gamma, beta, alpha, w_pretax, b, d)
2     c = c_pretax - tau
3     w = w_pretax .- tau
4     mcm = mccall_model(; alpha, beta, gamma, sigma, p, c, w)
5     (; V, U, w_bar) = solve_mccall_model(mcm)
6     accept_wage = w .> w_bar # indices of accepted wages
7     prop_accept = dot(p, accept_wage) # proportion of accepted wages
8     lambda = gamma * prop_accept
9     lm = lake_model(; lambda, alpha, b, d)
10    u_bar, e_bar = lm.x_bar
11    V_wealth = (dot(p, V .* accept_wage)) / dot(p, accept_wage)
12    welfare = e_bar .* V_wealth + u_bar .* U
13    return (;w_bar, lambda, V, U, u_bar, e_bar, welfare)
14 end
```

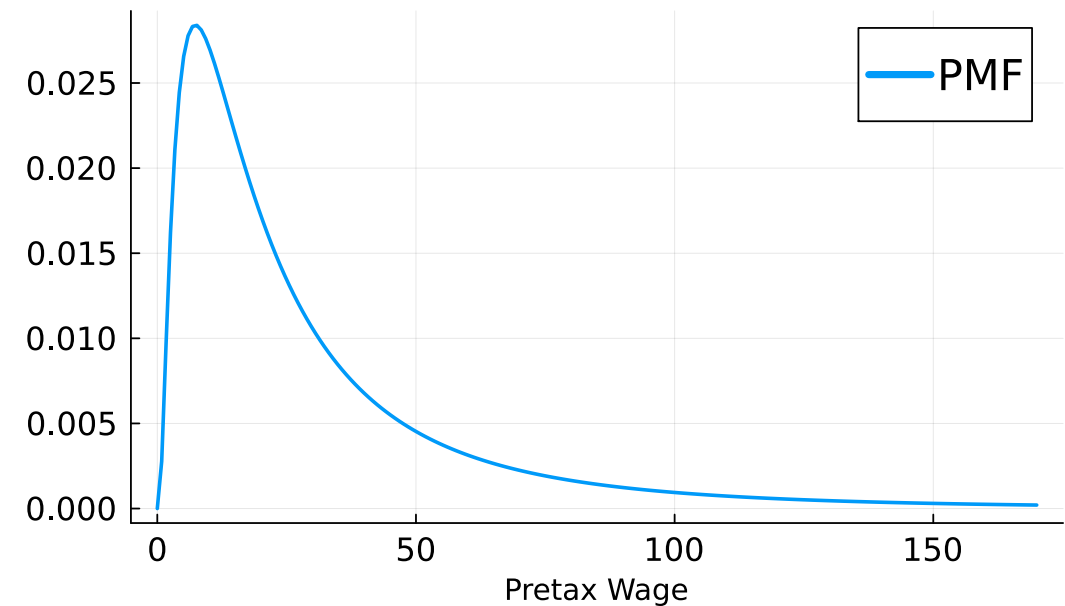
# Balancing the Budget

- Fixing  $c$ , modify  $\tau$  until  $\tau - \bar{u}c \approx 0$

```
1 function find_balanced_budget_tax(c_pretax; p, sigma, gamma, beta, alpha, w_pretax, b, d)
2     function steady_state_budget(tau)
3         (;u_bar, e_bar) = compute_optimal_quantities(c_pretax, tau; p, sigma, gamma, beta,
4                                                     alpha, w_pretax, b, d)
5         return tau - u_bar * c_pretax
6     end
7     # Find root
8     tau = find_zero(steady_state_budget, (0.0, 0.9 * c_pretax))
9     return tau
10 end
```

# Example Parameters

```
1 alpha = (1 - (1 - 0.013)^3)
2 b = 0.0124
3 d = 0.00822
4 beta = 0.98
5 gamma = 1.0
6 sigma = 2.0
7 # Discretized log normal
8 log_wage_mean = 20
9 wage_grid_size = 200
10 w_pretax = range(1e-3, 170,
11                  length = wage_grid_size + 1)
12 wage_dist = LogNormal(log(log_wage_mean), 1)
13 p = pdf.(wage_dist, w_pretax)
14 p = p ./ sum(p)
15 plot(w_pretax, p; xlabel = "Pretax Wage",
16      label = "PMF", size = (600, 400))
```





# Solving for various $c$

```
1 function calculate_equilibriums(c_pretax; p, sigma, gamma, beta, alpha, w_pretax, b, d)
2     tau_vec = similar(c_pretax)
3     u_vec = similar(c_pretax)
4     e_vec = similar(c_pretax)
5     welfare_vec = similar(c_pretax)
6     for (i, c_pre) in enumerate(c_pretax)
7         tau = find_balanced_budget_tax(c_pre; p, sigma, gamma, beta, alpha, w_pretax, b, d)
8         (;u_bar, e_bar, welfare) = compute_optimal_quantities(c_pre, tau; p, sigma, gamma, beta, alpha,
9                                                                w_pretax, b, d)
10        tau_vec[i] = tau
11        u_vec[i] = u_bar
12        e_vec[i] = e_bar
13        welfare_vec[i] = welfare
14    end
15    return tau_vec, u_vec, e_vec, welfare_vec
16 end
```

# Results with Various $c$

```
1 # levels of unemployment insurance we wish to study
2 c_pretax = range(5, 140, length = 60)
3 tau_vec, u_vec, e_vec, welfare_vec = calculate_equilibriums(c_pretax; p, sigma, gamma, beta, alpha,
4                                                             w_pretax, b, d)
5
6 # plots
7 plt_unemp = plot(title = "Unemployment", c_pretax, u_vec, color = :blue, xlabel = "c", label = "")
8 plt_tax = plot(title = "Tax", c_pretax, tau_vec, color = :blue, xlabel = "c", label = "")
9 plt_emp = plot(title = "Employment", c_pretax, e_vec, color = :blue, xlabel = "c", label = "")
10 plt_welf = plot(title = "Welfare", c_pretax, welfare_vec, color = :blue, xlabel = "c", label = "")
11
12 plot(plt_unemp, plt_emp, plt_tax, plt_welf, layout = (2, 2))
```

# Results with Various $c$

