# Search and the Labor Market

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# 1 Quick Review

## 1.1 Markov Chain Model of Unemployment

Recall model of unemployment:

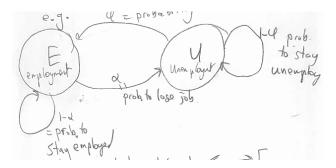


Figure 1: Lake Model with Markov Chains

where  $\phi$  probability of U to E transition should come from consumer decisions.

- This model will endogenize based on dynamic decisions of consumers looking for, and potentially rejecting, jobs.
- This, in turn, endogenizes unemployment rate.

### 1.2 Random Variables Review

• p is a non-negative random variable with pdf (probability density function) f(p) and cdf (cumulative distribution function):

$$F(p) \equiv \int_0^p f(s)ds \tag{1}$$

Assume F(0) = 0, F(B) = 1, where B is the upper bound.

• If p has continuous values on [0, B], then:

$$\underbrace{\mathbb{E}\left[p\right]}_{\text{mean of }p} = \int_{0}^{B} pf(p)dp \tag{2}$$

An alternative formula using the CDF,<sup>1</sup>

$$\mathbb{E}\left[p\right] = \int_0^B \left(1 - F(p)\right) dp \tag{3}$$

- Consider two independent random variables:  $\{p_1, p_2\}$ , both from F. Event  $\underbrace{\{p_1 < p\}}_{\text{follows } F(p)} \underbrace{\bigcap_{\text{intersection}}}_{\text{"or"}} \{p_2 < p\}$  denotes event max  $\{p_1, p_2\} < p$ .
- Since independent  $F(p)F(p) = F(p)^2$ , as probability max  $\{p_1, p_2\} < p$ . Then:

$$\mathbb{E}\left[\max\{p_1, p_2\}\right] = \int_0^B \left[1 - F(p)^2\right] dp \tag{8}$$

where  $\mathbb{E}\left[\max\left\{p_1, p_2\right\}\right] > \mathbb{E}\left[p_1\right]$ 

$$\int_{a}^{b} u dv = \underbrace{uv \mid_{a}^{b}}_{\text{evaluated at } b} - \int_{a}^{b} v du \tag{4}$$

Using our definition of expectation:

$$\int_{0}^{B} \underbrace{p}_{\substack{u=p\\ \forall u=dp\\ du=dp}} \underbrace{f(p)dp}_{\substack{dv=f(p)dp\\ \forall v=F(p)=\int_{0}^{p} f(s)ds}} = \int_{0}^{B} u dv$$

$$(5)$$

By formula,

$$= pF(p) \mid_{0}^{B} - \int_{0}^{B} F(p)dp \tag{6}$$

$$=\underbrace{BF(B)}_{B} - 0 - \int_{0}^{B} F(p)dp = \underbrace{\int_{0}^{B} (1 - F(p)) dp = \mathbb{E}[p]}_{Alternative for while for expectation}$$

$$(7)$$

<sup>&</sup>lt;sup>1</sup>To derive, Recall integration by parts,

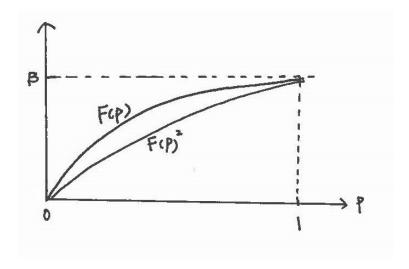


Figure 2: Expectation among  $\max(p_1, p_2)$ 

## 2 McCall Search Model

### 2.1 Setup

### • Background:

- Infinitely lived risk-neutral worker wants to maximize the expected P.D.V:

$$\mathbb{E}_0 \left[ \sum_{t=0}^{\infty} \beta^t y_t \right], \text{ with } y_t = \begin{cases} w_t \text{ if employed} \\ c \text{ if unemployed} \end{cases}$$
 (9)

where  $\beta \in (0,1)$ 

- Each period, an unemployment worker draws <u>one</u> offer to work at wage w forever, the wage is drawn from the distribution of wages in the economy F(w), where F(0) = 0, F(B) = 1
- There is <u>recall</u> of previous wages each period
- Each period, can accept current wage draw from F(w), or reject it and collect c > 0 (unemployment benefits) and draw again next period.
- If accept, the worker will work forever at w and can neither be fired nor quit.

#### • Problem:

- Find the worker's optimal strategy for accepting or rejecting draws

### • Solution:

- Let Q = optimal value of the problem for a worker about to draw a wage offer

$$Q \equiv \underbrace{\mathbb{E}\left[\sum_{t=0}^{\infty} \beta^t y_t\right]}_{\text{expectation before wage is drawn}} \tag{10}$$

- Let v(w) be the optimal value of the problem for a previously unemployed worker who has just drawn w and is about to decide what to do,

$$\underbrace{Q}_{\text{value}} = \underbrace{\int_{0}^{B} v(w) f(w) dw}_{\text{expected } \underline{\text{value}}} \text{ over draws}$$
(11)

- Key observation: write recursively as value today in terms of tomorrow

$$\frac{v(w)}{\text{value if draw } w} = \max_{\{\text{accept,reject}\}} \left\{ \underbrace{w}_{\text{Wage today}} + \underbrace{\beta v(w)}_{\text{Discounted value keeping wage}}, \underbrace{c}_{\text{Unemployment Benefits today}} + \underbrace{\beta \cdot Q}_{\text{Discounted value drawing wage}} \right\}$$
(12)

Recognizing if you accepting w once, you would keep accepting it,  $v(w|\text{accept}) = w + \beta v(w|\text{accept})$  for accepted wage. So  $v(w|\text{accept}) = \frac{w}{1-\beta}$ , and

$$\underbrace{v(w)}_{\text{value if draw } w} = \max_{\{\text{accept,reject}\}} \left\{ \underbrace{\frac{w}{1-\beta}}_{\text{PDV of benefits today}}, \underbrace{\frac{c}{\text{discounted discounted value tomorrow}}}_{\text{wage forever}} \right\}$$
(13)

Note:  $c + \beta \cdot Q$  is independent of w, since  $Q = \int v(w') f(w') dw'$ 

- Plug in for Q to find a <u>functional equation</u> in v(w),

$$v(w) = \max\left\{\frac{w}{1-\beta}, c+\beta \int_0^B v(w')f(w')dw'\right\}$$
(14)

# 2.2 Solve the Bellman Equation

**Solution**: A v(w) and accept or reject policy which fulfills (14) for  $\beta$ , f(.), c.

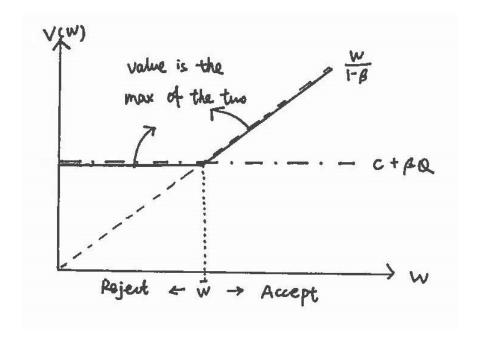


Figure 3: McCall model

• <u>Method 1</u>: Numerically iterate.

$$v_{j+1}(w) = \max\left\{\frac{w}{1-\beta}, c+\beta \int_0^B v_j(w')f(w')dw'\right\}$$
 (15)

Let  $v_0(w) = 0$ , evaluate  $v_1(w)$ , etc.

Guaranteed to converge to a stationary v(w)

• Method 2 Guess-and-verify

$$v(w) = \max\left\{\frac{w}{1-\beta}, c+\beta \int_0^B v(w')f(w')dw'\right\}$$
(16)

- Note,  $\bar{w}$  is point where indifferent between accept and reject.

$$v(\bar{w}) = \frac{\bar{w}}{1-\beta} = c + \beta \int_0^B v(w')f(w')dw', \text{ where } \begin{cases} \text{reject if } w' < \bar{w} \\ \text{accept if } w' > \bar{w} \end{cases}$$
(17)

$$= c + \beta \int_0^{\bar{w}} v(w') f(w') dw' + \beta \int_{\bar{w}}^B v(w') f(w') dw'$$
 (18)

$$= c + \beta \int_{0}^{\bar{w}} \underbrace{\frac{\bar{w}}{1 - \beta}}_{\text{reject value by}} f(w') dw' + \beta \int_{\bar{w}}^{B} \underbrace{\frac{w'}{1 - \beta}}_{\text{accept value}} f(w') dw'$$
(19)

$$=\frac{\bar{w}}{1-\beta}\tag{20}$$

- Expand LHS:

$$\int_{0}^{\bar{w}} \frac{\bar{w}}{1-\beta} f(w') dw' + \int_{\bar{w}}^{B} \frac{\bar{w}}{1-\beta} f(w') dw'$$
 (21)

$$= c + \beta \underbrace{\int_0^{\bar{w}} \frac{\bar{w}}{1 - \beta} f(w') dw'}_{\text{subtract to LHS}} + \beta \int_{\bar{w}}^B \frac{w'}{1 - \beta} f(w') dw' \tag{22}$$

 $\Rightarrow$  [After the subtract and add above]

$$(1-\beta)\int_0^{\bar{w}} \frac{\bar{w}}{1-\beta} f(w')dw' - c = \frac{1}{1-\beta} \int_{\bar{w}}^B (\beta w' - \bar{w}) f(w')dw'$$
 (23)

– By adding  $\bar{w} \int_{\bar{w}}^{B} f(w') dw'$  to both sides:

$$\bar{w} \left[ \int_{0}^{\bar{w}} f(w')dw' + \int_{\bar{w}}^{B} f(w')dw' \right] - c = \underbrace{\frac{\bar{w} - c}{\text{cost of searching one more period}}}_{\text{cost of searching one more period}} = \underbrace{\frac{\beta}{1 - \beta} \int_{\bar{w}}^{B} (w' - \bar{w}) f(w')dw'}_{\text{benefit of searching one more period}}$$
(24)

- Given  $\beta, c, f(.)$ , we can solve for  $\bar{w}$ , then use policy:  $\begin{cases} \text{reject if } w < \bar{w} \\ \text{accept if } w > \bar{w} \end{cases}$
- Then let h(w)= Benefit of search at  $w=\frac{\beta}{1-\beta}\int_w^B(w'-w)f(w')dw'$   $h(0)=\frac{\beta}{1-\beta}\mathbb{E}\left[w\right]>0$

$$h(B) = 0$$

$$h'(w) < 0, h'' > 0$$

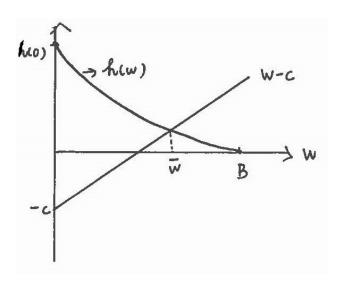


Figure 4: search model

## 2.3 The probability to accept

• Prob(accept | unemployed) = Prob( $w \ge \bar{w}$ ) =  $1 - F(\bar{w}) \equiv \pi$ 

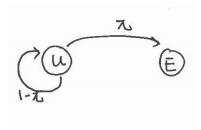


Figure 5: markov chain in search model

ullet In this basic setup, no firing or quitting, so E is absorbing. (Problem set includes firing)

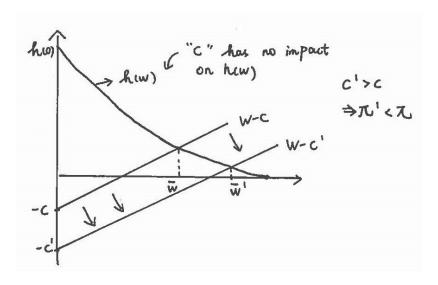


Figure 6: What is role of unemployment benefits?