

# Labor Economics Term Paper

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Due date: 2/1

## Q1

(1) The value functions of the worker are:

$$\begin{cases} rW = w + \lambda(\bar{u} + f_e - W) & \text{for employment} \\ r\bar{u} = z + \theta q(\theta)(W - \bar{u}) & \text{for unemployment} \end{cases}$$

$W$  and  $\bar{u}$  are the values of employment and unemployment.  $w$  and  $z$  are the instantaneous utility of employment and unemployment, respectively.  $r$  is the discount factor, and  $q(\theta)$  is the matching function of  $\theta = \frac{v}{u}$ .  $v$  and  $u$  is the vacancy rate and the unemployment rate of the workers, satisfying  $q'(\theta) < 0$ . When a worker was dismissed, they receive  $f_e$  as the severance payment.

(2) The value functions of the firm are:

$$\begin{cases} rJ = (p - w) + \lambda(V - f_e - f_a J) & \text{for having a vacancy filled} \\ rV = -c + q(\theta)(J - W) & \text{for vacancy} \end{cases}$$

$J$  and  $V$  are the values of vacancy filled and vacancy.  $p$  is the the instantaneous productivity of the worker and  $w$  is the instantaneous wage.  $c$  is the expected hiring cost. When a firm fire a worker, then it have to pay the severance payment  $f_e$  to the worker, and  $f_a$  as the related cost for firing.

(3) By the free-entry/exit condition,  $V = 0$  holds. Then, the value function of vacancy is rewritten as follows:

$$J = \frac{c}{q(\theta)}$$

Then, substituting this into the value function of having vacancy filled,

$$\begin{aligned}
rJ &= p - w + \lambda(V - f_e - f_a - J) \\
p - w &= (r + \lambda) \left( \frac{c}{q(\theta)} \right) - \lambda(f_e + f_a) \\
p - w &= \frac{(r + \lambda)c - \lambda(f_e + f_a)}{q(\theta)},
\end{aligned}$$

which corresponds to the job creation condition.

(4) First, we derive the equilibrium conditions: wage condition and steady state condition. First, we derive the wage curve.

Given the value functions in (1) and (2), the wage is determined by Nash bargaining rule:

$$w = \arg \max_w (W - \bar{u})^\beta (J - V)^{1-\beta},$$

note that  $\beta$  stands for the bargaining power of the worker.

## Q2

(1) "Hazard Function" is an instantaneous rate of an individual moving from a certain state to another. In the settings of job searching, it corresponds to that of leaving from unemployment pool.

Let  $T \geq 0$  denote the duration and the cumulative distribution function  $F(\cdot)$  is defined as follows:

$$F(t) = \Pr(T \leq t).$$

Then, the hazard function  $\lambda(\cdot)$  is:

$$\lambda(t) \equiv \lim_{h \rightarrow 0} \frac{\Pr(t \leq T \leq t + h | T \geq t)}{h},$$

where  $\Pr(t \leq T \leq t + h | T \geq t)$  is given as:

$$\Pr(t \leq T \leq t + h | T \geq t) = \frac{F(t + h) - F(t)}{1 - F(t)}.$$

Then, we can rewrite the hazard function by  $F(t)$  and  $f(t) = F'(t)$ ,

$$\lambda(t) = \frac{f(t)}{1 - F(t)}.$$

Now suppose the duration is distribution according to the Weibull distribution. Then, the cumulative distribution  $F(t)$  is given as follows.

$$F(t) = 1 - \exp(-\gamma t^\alpha)$$

$\gamma$  and  $\alpha$  are parameters. Then the density function is obtained as

$$\begin{aligned}
f(t) = F'(t) &= \frac{d(1 - \exp(-\gamma t^\alpha))}{dt} \\
&= \frac{d(1 - \exp(-\gamma t^\alpha))}{d(-\gamma t^\alpha)} \cdot \frac{d(-\gamma t^\alpha)}{dt} \\
&= \gamma \alpha t^{\alpha-1}
\end{aligned}$$

Substituting this into  $\gamma(t)$ , we obtain the hazard function.

$$\lambda(t) = \gamma \alpha t^{\alpha-1}$$

Positive/Negative dependence of the hazard function on the duration is found by checking the first-order differential of the hazard function: if  $\lambda'(\cdot) > 0$ , then it has positive dependence, and negative if  $\lambda'(\cdot) < 0$ . When the duration has Weibull distribution, by  $\lambda'(t) = (\alpha - 1)\gamma \alpha t^{\alpha-2}$ . Since  $t > 0$  and  $\gamma \geq 0$  is assumed, it depends on the value of  $\alpha$ . If  $\alpha > 1$ , then the hazard ratio has the positive dependence on the duration, and  $\alpha < 1$  implies negative dependence.

(2) Suppose observation of the worker who started job-searching after time 0. We observe unemployed workers who started job-searching until time  $b$ , and we stop observation at the certain time: after then, we do not observe the worker, even though they have not finished searching. We define the time a worker  $i$  enters the initial state (start searching)  $a_i$ , and time left for her/him until we stop the observation  $c_i$ .

Then, we cannot observe the true duration  $t_i$  if the worker have not finished searching at the censoring time. “Right-censoring” stands for possible bias caused by this problem.

On the other hand, when we start observation of the workers who are searching at the time  $b$ : If s/he have finished searching at time  $b$ , then we cannot observe her/him, even if s/he started job searching after time 0. Thus, obtained sample can be selected, that is, “left-censoring” problem occurs.

(3) Denote the “observed” duration to be  $t^*$ . Assume  $t^*$  is distributed independent of  $a_i$  and  $c_i$ , then the conditional cumulative distribution function  $F(\cdot)$  on the vector of observed covariates  $X_i$  is defined as:

$$F(t^*|X_i, a_i, c_i|\theta) = F(t_i|X_i;\theta).$$

$\theta$  is the vector of parameters.

If there is no censoring problem discussed in (2), then the conditional density function of duration is  $f(t_i^*|X_i;\theta)$ . When an observation is completed, we can obtain the likelihood function of this.

Then, we argue the observation with incomplete duration. First, we consider the right-censoring problem. The probability of right-censored observation is given by  $1 - F(c_i|X_i;\theta)$ . Then, the conditional likelihood function is obtained as follows:

$$f(t_i^*|X_i;\theta)^{d_i} [1 - F(c_i|X_i;\theta)]^{1-d_i}$$

Second, we consider the left-censoring. All the observed samples satisfy that the individual is still unemployed at time  $b$ . That is,  $t_i^* \geq b_i - a_i$ . The probability of this is expressed by  $\Pr(t^* \geq b - a_i | X_i, a_i, c_i) = 1 - F(b - a_i | X_i)$ .

By the Bayes Rule, the likelihood function including right and left-censoring is:

$$L(\theta) = \prod_i^N \frac{f(t_i^* | X_i; \theta)^{d_i} [1 - F(c_i | X_i, \theta)]^{1-d_i}}{1 - F(b - a_i | X_i; \theta)}$$

Maximizing this (or log-likelihood function) with  $\theta$ , we obtain MLE: the consistent asymptotical normal estimated parameter  $\hat{\theta}$ .

(4) Holmås. 2002. Keeping nurses at work: a duration analysis *Health Economics*. 11: 493 – 503 (2002)

### Q3

(1)

(2)

(3)

(4)