1 Many-to-Many Matching with Rational Inattention

When player i of type t meets player j of type s, the match utility ε_i^j is drawn from $F_t^s(\varepsilon)$, not observable to player i, but could be learned costly through signal x.

The cost of a signal x is proportional to $\mathcal{I}(\varepsilon, x) = \text{Entropy }(\varepsilon) - \mathbb{E}_{x[\text{ Entropy }(\varepsilon|x)]}$, let the cost parameter be $\frac{1}{\lambda}$.

1.1 One-to-one matching case (K = 1):

With complete information:

Player i can observe the match utility ε_i^j , then acceptance rule is a threshold rule:

$$\begin{cases} A & \text{if } \varepsilon_i^j > x_t^s = V_{t,\emptyset} \\ R & \text{otherwise} \end{cases}$$

where $V_{t,\emptyset}$ is the continuation value of a type t.

With rational inattention:

Player i chooses a signal with distribution $f(x|\varepsilon): \mathbb{R} \to \Delta(X)$ and a strategy $\sigma: X \to \{A, R\}$ that maximizes:

$$\max_{f_{i},\sigma_{i}} E\left[u_{i}\left(\sigma_{i},\sigma_{-i},\varepsilon_{i},\varepsilon_{-i}\right) - \frac{1}{\lambda}I\left(\varepsilon,x\right)\right]$$

where $u_i(A, A, \varepsilon_i, \varepsilon_{-i}) = \varepsilon_{-i}$

and
$$u_i(A, R, \varepsilon_i, \varepsilon_{-i}) = u_i(R, A, \varepsilon_i, \varepsilon_{-i}) = u_i(R, R, \varepsilon_i, \varepsilon_{-i}) = V_{t,\emptyset}$$

This problem can be simplified to choosing a choice rule $\pi_i(\varepsilon)$ that maximizes:

$$\max_{\pi_{i}} \mathbb{E}\left[u_{i}\left(\sigma_{i}, \sigma_{-i}, \varepsilon_{i}, \varepsilon_{-i}\right) - \frac{1}{\lambda}I\left(\varepsilon_{-i}, x\right)\right]$$

where
$$\pi_i(\varepsilon) = \mathbb{P}\left\{\sigma_i(x) = A \middle| \varepsilon_{-i} = \varepsilon\right\}$$

Substituting in the utility functions and the opponent's choice rule:

$$\max_{\pi_i} \mathbb{E}_{\varepsilon \left[\pi_i(\varepsilon_{-i}) \pi_{-i(\varepsilon_i)} \varepsilon_{-i} + \left(1 - \pi_i(\varepsilon_{-i}) \pi_{-i(\varepsilon_i)} \right) V_{t,\emptyset} - c_I I(\varepsilon_{-i}, \sigma_i) \right]}$$

The solution is given by:

$$\pi_{i}\left(\varepsilon_{-i}\right) = \frac{q_{i}e^{u_{i}\left(A,\varepsilon_{-i}\right)}}{q_{i}e^{u_{i}\left(A,\varepsilon_{-i}\right)} + \left(1 - q_{i}\right)e^{u_{i}\left(R,\varepsilon_{-i}\right)}}$$

where q_i maximizes: $\mathbb{E}_{\pi_0} \left[\log \left(q_i e^{u_i(A, \varepsilon_{-i})} + (1 - q_i) e^{u_i(R, \varepsilon_{-i})} \right) \right]$

Substituting in the utility functions:

$$\pi_{i}\left(\varepsilon_{-i}\right) = \frac{q_{i}e^{\mathbb{E}_{\varepsilon_{i}}\left[\pi_{-i\left(\varepsilon_{i}\right)}\varepsilon_{-i}+\left(1-\pi_{-i\left(\varepsilon_{i}\right)}\right)V_{t,\emptyset}\right]}}{q_{i}e^{\mathbb{E}_{\varepsilon_{i}}\left[\pi_{-i\left(\varepsilon_{i}\right)}\varepsilon_{-i}+\left(1-\pi_{-i\left(\varepsilon_{i}\right)}\right)V_{t,\emptyset}\right]}+\left(1-q_{i}\right)e^{V_{t,\emptyset}}}$$

where q_i maximizes: $\mathbb{E}_{\varepsilon_{-i}} \left[\log \left(q_i e^{\pi_{-i(\varepsilon_i)}\varepsilon_{-i} + \left(1 - \pi_{-i(\varepsilon_i)}\right)V_{t,\emptyset}} + (1 - q_i) e^{V_{t,\emptyset}} \right) \right]$

The derivative has the same sign as: $e^{\pi_{-i(\varepsilon_i)}\varepsilon_{-i}+\left(1-\pi_{-i(\varepsilon_i)}\right)V_{t,\emptyset}}-e^{V_{t,\emptyset}}$

Therefore, $q_i =$

$$\begin{cases} 0 & \text{if } \pi_{-i(\varepsilon_i)} \varepsilon_{-i} + \left(1 - \pi_{-i(\varepsilon_i)}\right) V_{t,\emptyset} > V_{t,\emptyset} \\ 1 & \text{if } \pi_{-i(\varepsilon_i)} \varepsilon_{-i} + \left(1 - \pi_{-i(\varepsilon_i)}\right) V_{t,\emptyset} < V_{t,\emptyset} \end{cases}$$

or, $q_i =$

$$\begin{cases} 0 & \text{if } \varepsilon_{-i} > V_{t,\emptyset} \\ 1 & \text{if } \varepsilon_{-i} < V_{t,\emptyset} \end{cases}$$

also, $\pi_i(\varepsilon_{-i}) =$

$$\begin{cases} 0 & \text{if } \varepsilon_{-i} > V_{t,\emptyset} \\ 1 & \text{if } \varepsilon_{-i} < V_{t,\emptyset} \end{cases}$$

$$\pi_{t}^{s} = \mathbb{E}_{\varepsilon\left[\pi_{i}\left(\varepsilon_{-i}\right)\right]} = F_{t}^{s}\left(V_{t,\emptyset}\right)$$

2 Many-to-Many Matching with Learning

2.1 Complete Information case (match utility is type):

Each player has types $t \in \mathcal{T}$, meets with each other in continuous time τ

Denote the value of type s at time $\tau V_{s,\tau}$ and the value of a match between s and s' is $U_s^{s'}$

Parameters c_s is the cost of waiting and $\rho_s^{s'}$ the rate at which a type s meets with a type s'

2.2 Values

Let $\lambda_s^{s'}$ be the rate of arrival of offers (meeting and acceptance)

$$V_{s,\tau} = \mathbb{E}_{s'} \left[\lambda_s^{s'} d\tau \max \left\{ V_{s,\tau+d\tau}, U_s^{s'} \right\} + \left(1 - \lambda_s^{s'} d\tau \right) (V_{s,\tau+d\tau}) \right] - c_s d\tau$$

$$c_s d\tau = \mathbb{E}_{s'} \left[\lambda_s^{s'} d\tau \max \left\{ U_s^{s'} - V_{s,\tau+d\tau}, 0 \right\} + V_{s,\tau+d\tau} - V_{s,\tau} \right]$$

$$c_s = \mathbb{E}_{s'} \left[\lambda_s^{s'} \max \left\{ U_s^{s'} - V_{s,\tau+d\tau}, 0 \right\} \right] + \frac{V_{s,\tau+d\tau} - V_{s,\tau}}{d\tau}$$

$$c_s = \mathbb{E}_{s'} \left[\lambda_s^{s'} \max \left\{ U_s^{s'} - V_{s,\tau}, 0 \right\} \right] + \dot{V}_{s,\tau}$$

2.3 Continuous Type

With continuous types s and in equilibrium:

$$c_{s} = \int_{\underline{s}}^{\overline{s}} \lambda_{s}^{s'} \max \left\{ U_{s}^{s'} - V_{s}, 0 \right\} dF\left(s'\right)$$

Suppose each player with type s uses the threshold strategy $x^*(s)$, when meeting with a type s':

$$\begin{cases} A & \text{if } s' \ge x^{\star}(s) \\ R & \text{otherwise} \end{cases}$$

Then the values should be equal at threshold: $V_s = U_s^{x^*(s)}$

and
$$\lambda_s^{s'} = (1 - F(x^{\star}(s'))) \rho_s^{s'}$$

$$c_{s} = \int_{\underline{s}}^{\bar{s}} \rho_{s}^{s'} \left(1 - F\left(x^{\star}(s')\right) \max \left\{ U_{s}^{s'} - U_{s}^{x^{\star}(s)}, 0 \right\} dF\left(s'\right) \right.$$

$$= \left(1 - F\left(x^{\star}(s')\right) \right) \int_{x^{\star}(s)}^{\bar{s}} \rho_{s}^{s'} \left(U_{s}^{s'} - U_{s}^{x^{\star}(s)} \right) dF\left(s'\right)$$

$$= \left(1 - F\left(x^{\star}(s')\right) \right) \int_{x^{\star}(s)}^{\bar{s}} \left(1 - F\left(s'\right) \right) \frac{\partial}{\partial s'} \left(\rho_{s}^{s'} \left(U_{s}^{s'} - U_{s}^{x^{\star}(s)} \right) \right) ds'$$

This equation characterizes the set of thresholds $x^{\star}\left(s\right)$ \forall $s\in\mathcal{T}$ on both sides of the market

2.4 Discrete Types

With discrete types s:

The same threshold strategy is used but $x^{\star}(s)$ is determined by:

$$U_{s}^{x^{\star}(s)-1} \leq V_{s} \leq U_{s}^{x^{\star}(s)}$$

$$c_{s} = \sum_{s'=x^{\star}(s)}^{\bar{s}} \rho_{s}^{s'} \left(1 - F\left(x^{\star}\left(s'\right)\right)\right) \left(U_{s}^{s'} - U_{s}^{x^{\star}(s)}\right) p_{s}$$

where $p_s = dF(s) = \mathbb{P}\{S = s\}$

2.5 Two Types

With two types $s \in \{h, l\}$:

Either
$$x^{\star}(h) = h$$
 or $x^{\star}(h) = l$

Case $1: x^{*}(h) = h$:

$$c_h = \rho_h^h p_h \left(U_h^h - V_h \right) p_h$$

$$\Rightarrow V_h = U_h^h - \frac{c_h}{\rho_h^h p_h^2}$$
then $U_h^l \le V_h \le U_h^h$ implies
$$\Rightarrow U_h^l \le U_h^h - \frac{c_h}{\rho_h^h p_h^2} \le U_h^h$$

$$\Rightarrow 0 \le \frac{c_h}{\rho_h^h p_h^2} \le U_h^h - U_h^l$$

The above condition gives the equilibrium with only h-h and l-l matchings

Case $2: x^{\star}(h) = l$:

$$\begin{split} c_h &= \rho_h^h \left(U_h^h - V_h \right) p_h + \rho_h^l \left(U_h^l - V_h \right) p_l \\ &\Rightarrow V_h = \frac{\rho_h^h U_h^h p_h + \rho_h^l U_h^l p_l - c_h}{\rho_h^h p_h + \rho_h^l p_l} \\ &\Rightarrow V_h = U_h^l + \frac{\rho_h^h \left(U_h^h - U_h^l \right) p_h - c_h}{\rho_h^h p_h + \rho_h^l p_l} \end{split}$$
 then $V_h \leq U_h^l$ implies
$$\Rightarrow U_h^l + \frac{\rho_h^h \left(U_h^h - U_h^l \right) p_h - c_h}{\rho_h^h p_h + \rho_h^l p_l} \leq U_h^l$$

$$\Rightarrow \rho_h^h \left(U_h^h - U_h^l \right) p_h - c_h \leq 0$$

$$\Rightarrow U_h^h - U_h^l \leq \frac{c_h}{\rho_h^h p_h} \end{split}$$

The above condition gives the equilibrium with both h-h,h-l and l-l matchings

2.6 Complete Information case (match utility is random)

Each player meets at rate $\lambda_{s'}$. During a meeting with type s', match utility is drawn from a type specific distribution $\varepsilon_{s'} \sim F_{s'}(\varepsilon)$.

2.7 Value

$$\begin{split} V_{s,\tau} &= \sum_{s'} d\tau \lambda_{s'} \pi_{s'}^s \int_0^\infty \max \left\{ V_{s,\tau+d\tau}, \varepsilon_{s'} dF_{s'}\left(\varepsilon\right) \right\} + \left(1 - \sum_{s'} d\tau \lambda_{s'} \pi_{s'}^s \right) V_{s,\tau+d\tau} - c_s d\tau \\ c_s &= \sum_{s'} \lambda_{s'} \pi_{s'}^s \int_0^\infty \max \left\{ \varepsilon_{s'} - V_s, 0 \right\} dF_{s'}\left(\varepsilon\right) \\ c_s &= \sum_{s'} \lambda_{s'} \pi_{s'}^s \int_{x_s}^\infty \left(\varepsilon_{s'} - x_s \right) dF_{s'}\left(\varepsilon\right) \end{split}$$

Special case: $F_{s'} \sim \text{Exp }(\beta_{s'})$

$$\int_{x_s}^{\infty} (\varepsilon_{s'} - x_s) dF_{s'}(\varepsilon) = \frac{e^{x_s \beta_{s'}}}{\beta_{s'}}$$

$$\pi_{s'}^s = 1 - F_s(x_{s'}) = e^{-x_{s'} \beta_s}$$

$$\therefore c_s = \sum_{s'} \lambda_{s'} e^{-x_{s'} \beta_s} \frac{e^{x_s \beta_{s'}}}{\beta_{s'}}$$

If there is only one type:

$$c = \lambda e^{-x\beta} \frac{e^{-x\beta}}{\beta}$$
$$\Rightarrow x = \frac{1}{2\beta} \log \left(\frac{\lambda}{c\beta}\right)$$

If there are two types $s \in \{h, l\}$:

$$\begin{cases} c_h = \lambda_h e^{-x_h \beta_h} \frac{e^{-x_h \beta_h}}{\beta_h} + \lambda_l e^{-x_l \beta_h} \frac{e^{-x_h \beta_l}}{\beta_l} \\ c_l = \lambda_h e^{-x_h \beta_l} \frac{e^{-x_l \beta_h}}{\beta_h} + \lambda_l e^{-x_l \beta_l} \frac{e^{-x_l \beta_l}}{\beta_l} \end{cases}$$

$$\Rightarrow x_s = \frac{1}{2\beta_s} \log \left(\frac{2\lambda_s}{c_s \beta_s} \right) \text{ for } s \in \{h, l\}$$

2.8 Incomplete Information case (match utility is random, type is observable)

The type of a player is given by his own type s and her belief p over her types, p_s being the belief that he is type s. Summanize the types by $A = (s, p) \in (T, \Delta(T))$. Note that A can also be summarized by his history of acceptance and rejections $A = (s, A(s_1), A(s_2), ..., A(s_n))$ where s_i the type of person she met in her i-th meeting and $A(s) \in \{a, r\}$.

Each player with belief p uses the strategy
$$\begin{cases} a & \text{if } \varepsilon_{s'} > x_p \\ r & \text{otherwise} \end{cases}$$

2.9 Value

$$\begin{split} V_{s,A,\tau} &= \sum_{s'} d\tau \lambda_{s'} \sum_{s} \pi_{s'}^{s} p_{s} \int \max \left\{ \varepsilon_{s'}, V_{s,A+a(s'),\tau+d\tau} dF_{s'} \left(\varepsilon \right) \right. \\ &+ \sum_{s'} d\tau \lambda_{s'} \sum_{s} \left(1 - \pi_{s'}^{s} \right) p_{s} V_{s,A+r(s'),\tau+d\tau} \\ &+ \left(1 - \sum_{s'} d\tau \lambda_{s'} \right) V_{s,A,\tau+d\tau} - c_{s} d\tau \end{split}$$

in equilibrium,

$$c_{s} = \sum_{s'} \lambda_{s'} ((1) + (2) + (3))$$

$$(1) \left(\sum_{s} \pi_{s'}^{s} p_{s} \right) \int_{x_{p}}^{\infty} \varepsilon_{s'} - x_{p} dF_{s'} (\varepsilon)$$

$$(2) \left(\sum_{s} \pi_{s'}^{s} p_{s} \right) \left(V_{s,p+a(s')} - V_{s,p} \right)$$

$$(3) \left(1 - \sum_{s} \pi_{s'}^{s} p_{s} \right) \left(V_{s,p+r(s')} - V_{s,p} \right)$$

with belief dynamics,

$$\mathbf{p} + a\left(s'\right) = \frac{\pi_{s'}^{s} p_{s}}{\sum_{t} \pi_{s'}^{t} p_{t}}$$

$$p + r(s') = \frac{(1 - \pi_{s'}^s) p_s}{\sum_{t} (1 - \pi_{s'}^t) p_t}$$

$$\pi_{s'}^{s} = \int_{p} \left(1 - F_{s}\left(x_{p}\right)\right) \mathbb{P}\left\{p | s'\right\}$$

3 Two Period $Mo\delta$

3.1 $Mo\delta$

In the first period, people meet, observes the type and match utility of the other person s' and $\varepsilon_{s'} \sim F_{s'}$, and decide to accept and receive the match utility or reject the other person and continue to the second.

In the second period, staying single leads to a fixed utility depending on the type of the player u_s .

3.2 Strategies

Everyone starts with the same prior, so they have the same thresholds meeting a type s':

$$\begin{cases} a & \text{if } \varepsilon_{s'} \ge x_{0,s'} \\ r & \text{otherwise} \end{cases}$$

In period two, the reservation utility does not depend on the type of the other person, so:

$$x_{1,a,s'} = x_{1,a,s} \ \forall \ s,s'$$

$$x_{1,r,s'} = x_{1,r,s} \ \forall \ s,s'$$

3.3 Masses

In the first period, let the masses be:

$$m_{0,s} = p_s$$

In the second period:

type s who got an acceptance in period 1 :
$$m_{1,a,s} = \left(\sum_{s'} p_{s'} \cdot F_{s'}\left(x_{0,s'}\right)\right) \left(1 - F_s\left(x_{0,s}\right)\right)$$

type s who got a rejection in period 1: $m_{1,r,s} = F_s(x_{0,s})$

3.4 Beliefs

In the first period everyone has the same belief:

$$p_0(s) = p_s$$

In the second period:

after an acceptance
$$: p_{1,a}\left(s\right) = \frac{1 - F_s\left(x_{0,s}\right)}{\displaystyle\sum_{t} 1 - F_t\left(x_{0,t}\right)}$$
 after an rejection $: p_{1,r}\left(s\right) = \frac{F_s\left(x_{0,s}\right)}{\displaystyle\sum_{t} F_t\left(x_{0,t}\right)}$

3.5 Thresholds

In period two, threshold is equal to the value if the person is accepted but chooses to reject:

$$x_{1,a,s} = \sum_{t} p_{1,a}\left(t\right) u_{t}$$

$$x_{1,r,s} = \sum_{t} p_{1,r}\left(t\right) u_{t}$$

In period one, threshold is equal to the value if the person is accepted but chooses to reject:

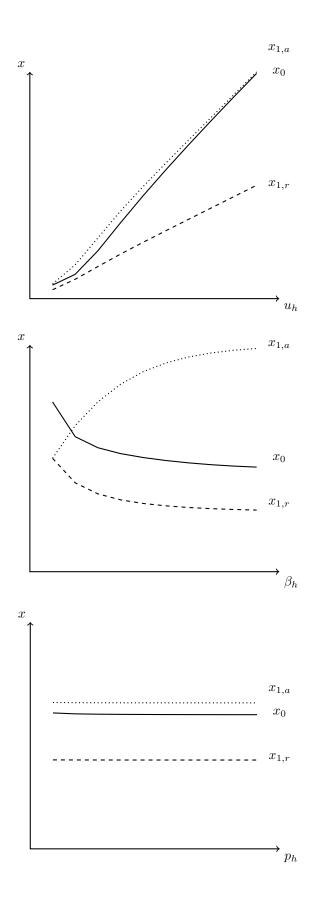
$$x_{0,s} = \sum_{t} m_{1,a,t} \left(\sum_{t'} \left(p_{1,a} \left(t' \right) \left(1 - F_{t'} \left(x_{1,a,t'} \right) \right) \right) \int_{x_{1,a,t}}^{\infty} \varepsilon_{t} dF_{t} + \sum_{t'} \left(p_{1,a} \left(t' \right) F_{t'} \left(x_{1,a,t'} \right) \right) x_{1,a,s} \right) + \sum_{t} m_{1,r,t} \left(\sum_{t'} \left(p_{1,a} \left(t' \right) \left(1 - F_{t'} \left(x_{1,r,t'} \right) \right) \right) \int_{x_{1,a,t}}^{\infty} \varepsilon_{t} dF_{t} + \sum_{t'} \left(p_{1,a} \left(t' \right) F_{t'} \left(x_{1,r,t'} \right) \right) x_{1,a,s} \right) \right)$$

Note that $x_{1,a,s}$ is independent of s, so $x_{0,s}$ should be independent of s as well.

3.6 Numerical Solutions

Consider the case with two types $s \in \{h, l\}$, F_s is exponentially distributed with rate β_{s} .

Start with $p_h = p_l = \frac{1}{2}$, $\beta_l = 1$, $\beta_h = 2$, $u_l = 0$, $u_h = 1$, the following plots changes u_h from 1 to 10, β_h from 1 to 10 and $\frac{1}{p_h}$ from 1 to 10.



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4.1 $Mo\delta$

- 1. Continum of agents, continuous time, discounting
- 2. Type $x \in \{H, L\}$ is match utility
- 3. History of opponent known (later relaxed)

4.2 Results

- $1. \ \mathrm{x}$ decrease from rejection
- 2. x increase from offer by type below reservation
- 3. Smaller change if unknown history of opponent
- 4. Larger propotion of L type in market

5 Overlapping Generations $Mo\delta$

5.1 Model 1

When people meet, they observe the age, type and match utility of the other person (t, s, ε) where $t \in \{0,1\}, s \in \{h,l\}, \varepsilon \sim F_s$.

5.2 Strategies

Let $x_{t,z}^{t',s'}$ denote the threshold of a person with age t with history z, when meeting with a person with age t' and type s'. The history $z=\emptyset$ when t=0 and $z\in\{(a,0),(a,1),(r,0),(r,1)\}$, denoting whether the person is accepted or rejected by an age 0 or 1 person.

At age 0, let the threshold be:

For each
$$(t', s'), x_0^{t', s'}$$

At age 1, reservation utility does not depend on the age and type of the other person:

For each
$$z, x_{1,z}^{t',s'} = x_{1,z}^{\tilde{t},\tilde{s}} \forall (t',s',\tilde{t},\tilde{s})$$

5.3 Masses

At each time, there are 10 groups of people, for $s \in \{h, l\}$:

$$\begin{split} m_{0,s} &= p_s \\ m_{1,a,0,s} &= \left(\sum_{s'} p_{0,s'} F_{s'} \left(x_0^{0,s'}\right)\right) \left(1 - F_s \left(x_0^{0,s}\right)\right) \\ m_{1,a,1,s} &= \left(\sum_{s'} p_{1,s'} F_{s'} \left(x_0^{1,s'}\right)\right) \left(\sum_{z'} p_{1,z'} \left(1 - F_s \left(x_{1,z'}^{0,s}\right)\right)\right) \\ m_{1,r,0,s} &= F_s \left(x_0^{0,s}\right) \\ m_{1,r,1,s} &= \left(\sum_{z'} p_{1,z'} F_s \left(x_{1,z'}^{0,s}\right)\right) \end{split}$$

5.4 Beliefs

At age 0, everyone has the same belief:

$$p_0\left(s\right) = p_s$$

At age 1, the beliefs from being rejected by age 0 and age 1 person are different:

$$p_{1,a,0}(s) = \frac{1 - F_s\left(x_0^{0,s}\right)}{\sum_{\tilde{s}} \left(1 - F_{\tilde{s}}\left(x_0^{0,\tilde{s}}\right)\right)}$$

$$p_{1,a,1}(s) = \frac{\sum_{z'} p_{1,z'}\left(1 - F_s\left(x_{1,z'}^{0,s}\right)\right)}{\sum_{\tilde{s}} \left(\sum_{z'} p_{1,z'}\left(1 - F_{\tilde{s}}\left(x_{1,z'}^{0,\tilde{s}}\right)\right)\right)}$$

$$p_{1,r,0}(s) = \frac{F_s\left(x_0^{0,s}\right)}{\sum_{\tilde{s}} F_{\tilde{s}}\left(x_0^{0,\tilde{s}}\right)}$$

$$p_{1,r,1}(s) = \frac{\sum_{z'} p_{1,z'}F_s\left(x_{1,z'}^{0,\tilde{s}}\right)}{\sum_{\tilde{s}} p_{1,z'}\left(\sum_{z'} F_{\tilde{s}}\left(x_{1,z'}^{0,\tilde{s}}\right)\right)}$$

5.5 Thresholds

At age 1,

$$x_{1,z}^{t',s'} = \sum_{s} p_{1,z}(s) u_s$$

At age 0,

$$\begin{split} x_{0}^{t',s'} &= \sum_{\tilde{s},z'} m_{1,\tilde{z'},s} \sum_{s,z} p_{1,z}\left(s\right) \left(1 - F_{s}\left(x_{1,z}^{1,s}\right)\right) \int_{x_{1,z}^{1,s}}^{\infty} \varepsilon_{\tilde{s}} dF_{\tilde{s}} + \sum_{s,z} p_{1,z}\left(s\right) F_{s}\left(x_{1,z}^{1,s}\right) x_{1,z}^{1,s} \\ &+ \sum_{\tilde{s}} m_{\tilde{0,s}} \sum_{s,z} p_{1,z}\left(s\right) \left(1 - F_{s}\left(x_{1,z}^{1,s}\right)\right) \int_{x_{1,z}^{1,s}}^{\infty} \varepsilon_{\tilde{s}} dF_{\tilde{s}} + \sum_{s,z} p_{1,z}\left(s\right) F_{s}\left(x_{1,z}^{1,s}\right) x_{1,z}^{1,s} \end{split}$$

5.6 Model 2

When people meet, they observe the age and match utility of the other person (t, ε) where $t \in \{0, 1\}$, $\varepsilon \sim F_s$, but not the type of the other person.

5.7 Strategies

Let $x_{t,z}^{t'}$ denote the threshold of a person with age t with history z, when meeting with a person with age t'. The history $z=\emptyset$ when t=0 and $z\in\{(a,0),(a,1),(r,0),(r,1)\}$, denoting whether the person is accepted or rejected by an age 0 or 1 person.

At age 0, let the threshold be:

For each
$$t', x_0^{t'}$$

At age 1, reservation utility does not depend on the age and type of the other person:

For each
$$z, x_{1,z}^{t'} = x_{1,z}^{\tilde{t}} \ \forall \ \left(t', \tilde{t}\right)$$

5.8 Masses

At each time, there are 10 groups of people, for $s \in \{h, l\}$:

$$m_{0,s} = p_s$$

$$m_{1,a,0,s} = \left(\sum_{s'} p_{0,s'} F_{s'} \left(x_0^0\right)\right) \left(1 - F_s\left(x_0^0\right)\right)$$

$$m_{1,a,1,s} = \left(\sum_{s'} p_{1,s'} F_{s'} \left(x_0^1\right)\right) \left(\sum_{z'} p_{1,z'} \left(1 - F_s\left(x_{1,z'}^0\right)\right)\right)$$

$$m_{1,r,0,s} = F_s\left(x_0^0\right)$$

$$m_{1,r,1,s} = \left(\sum_{z'} p_{1,z'} F_s\left(x_{1,z'}^0\right)\right)$$

5.9 Beliefs

At age 0, everyone has the same belief:

$$p_0(s) = p_s$$

At age 1, the beliefs from being rejected by age 0 and age 1 person are different:

$$p_{1,a,0}(s) = \frac{1 - F_s(x_0^0)}{\sum_{\tilde{s}} (1 - F_{\tilde{s}}(x_0^0))}$$

$$p_{1,a,1}(s) = \frac{\sum_{z'} p_{1,z'} (1 - F_s(x_{1,z'}^0))}{\sum_{\tilde{s}} \left(\sum_{z'} p_{1,z'} (1 - F_{\tilde{s}}(x_{1,z'}^0))\right)}$$

$$p_{1,r,0}(s) = \frac{F_s(x_0^0)}{\sum_{\tilde{s}} F_{\tilde{s}}(x_0^0)}$$

$$p_{1,r,1}(s) = \frac{\sum_{z'} p_{1,z'} F_s(x_{1,z'}^0)}{\sum_{\tilde{s}} p_{1,z'} \left(\sum_{z'} F_{\tilde{s}}(x_{1,z'}^0)\right)}$$

5.10 Thresholds

At age 1,

$$x_{1,z}^{t'} = \sum_{s} p_{1,z}\left(s\right) u_s$$

At age 0,

$$x_{0}^{t'} = \sum_{\tilde{s},z'} m_{1,\tilde{z'},s} \sum_{s,z} p_{1,z}\left(s\right) \left(1 - F_{s}\left(x_{1,z}^{1}\right)\right) \int_{x_{1,z}^{1}}^{\infty} \varepsilon_{\tilde{s}} dF_{\tilde{s}} + \sum_{s,z} p_{1,z}\left(s\right) F_{s}\left(x_{1,z}^{1}\right) x_{1,z}^{1}$$
$$+ \sum_{\tilde{s}} m_{\tilde{0},\tilde{s}} \sum_{s,z} p_{1,z}\left(s\right) \left(1 - F_{s}\left(x_{1,z}^{1}\right)\right) \int_{x_{1,z}^{1}}^{\infty} \varepsilon_{\tilde{s}} dF_{\tilde{s}} + \sum_{s,z} p_{1,z}\left(s\right) F_{s}\left(x_{1,z}^{1}\right) x_{1,z}^{1}$$

6 Many-to-Many Matching with Learning (Old Model)

6.1 Incomplete Information case:

Let the type of a player $s \in [0,1]$ be the belief of being high type,

When type s meets another player with type s', she can perfectly observe the actual type, meaning $s' \in \{0,1\}$

Let the match utilities be $U_s^{s'} = sU_h^{s'} + (1-s)U_l^{s'}$

Need $c_h = c_l = c$ so that the player cannot identify her own type

6.2 Value

The two type incomplete information case is like the continuous type complete information case:

$$c = s \left(\lambda_s^h \max\left\{U_s^h - V_s, 0\right\}\right) + (1 - s) \left(\lambda_s^l \max\left\{U_s^l - V_s, 0\right\}\right)$$
 where $\lambda_s^{s'} = s + (1 - s) F_{s'}(s^*)$

and F_t is the equilibrium distribution of beliefs of players with true type t

and $s^{\star} = \max_{s} \left\{ x^{\star} \left(s \right) = l \right\}$ is the threshold type who accepts a meeting with l type

Here, the assumption is every player accepts a meeting with h type, and only players with type $s \in [0, s^*]$ accepts a meeting with l type

6.3 Incomplete Information case (old model):

Each player has one of two types $t \in \{h, l\}$,

When player i meets player j, i observes the number of past meetings j had (or equivalently the age of j) and a sample match utility $u_{i,j}$ drawn from $F_{t_j} \in \{F_h, F_l\}$ depending on j's type,

Then each player chooses one of two actions $\sigma_i \in \{ \text{ accept }, \text{ reject } \}$:

If both players choose accept, player i obtains match utility $\mathbb{E}\left[F_{t_j}\right] = u_{t_j} \in \{u_h, u_l\}$;

Otherwise, player i continues searching.

6.4 Question:

- 1. Two types of behavior in equilibrium?
 - (a) Players sets low acceptance threshold because high type players have exited the market early Here, large proportion of players with high age has low types
 - (b) Players sets high acceptance threshold to wait for acceptance from other high type players Here, large proportion of players with high age has high types
- 2. How do acceptance rule change with initial prior?
 - (a) More skewed distribution will lead to lower acceptance threshold
 - (b) Uniform prior lead to highest acceptance threshold

6.5 Types and masses:

After $k \in \{0, 1, 2, ...\}$ meetings, player i's type is given by a pair (t_i, A_i) , where:

 $t_i \in \mathcal{T}$ is the true type of player i

 $A_i \in \{a,r\}^k$ is a vector of length k, the state (or history) of player i

For example, $A_i = \{r, r, a, r, a\}$ means player i met 5 other players, rejected the 1st, 2nd and 4th, and was rejected by 3rd and 5th player

(Note: the type need to specify the age of the other player in the meetings as well)

Then $\sum_{k} A_{i}(r)$ is the number of rejections by player i and $\sum_{k} A_{i}(a)$ is the number of rejections for player i by other players

Let m(t, A) denote the mass of players with type t and state A,

Then $m(t, k) = \sum_{A \in A_k} m(t, A)$ is the mass of players with k meetings,

6.6 Strategies and Acceptance Probabilities:

Consider the threshold strategy for player i with type (t, A) meeting player j with age k,

$$\begin{cases} \text{accept} & \text{if } u_{i,j} \ge x_A^k \\ \text{reject} & \text{otherwise} \end{cases}$$

Define acceptance probabilities, the probability of a player with type (t, A) accepting another player with age k

$$\begin{split} \pi^k_{t,A} &= \mathbb{P}\left\{ \text{ accept } |t_i = t, A_i = A, |A_j| = k \right\} \\ \pi^k_{t,k'} &= \mathbb{P}\left\{ \text{ accept } |t_i = t, |A_i| = k', |A_j| = k \right\} \\ &= \sum_{A \in \mathcal{A}_{t'}} \pi^k_{t,A} \frac{m\left(t,A\right)}{m\left(t,k'\right)} \end{split}$$

Acceptance probabilities should be consistent with the threshold strategy and the beliefs about the types.

6.7 Beliefs and Updating Process:

The posterior belief should be consistent with the equilibrium mass of type A:

$$p_{A}(t) = \mathbb{P}\left\{t_{i} = t | A_{i} = A\right\}$$

$$= \frac{m(t, A)}{\sum_{s \in T} m(s, A)}$$

$$p_{k}(t) = \mathbb{P}\left\{t_{i} = t | |A_{i}| = k\right\}$$

$$= \frac{m(t, k)}{\sum_{s \in T} m(s, k)}$$

Belief is only updated if player i chooses accept and player j with age k chooses reject:

$$p_{A+a}\left(t\right) = p_{A}\left(t\right) \frac{\sum_{B \in \mathcal{A}_{k}} F_{t}\left(x_{B}^{|A|}\right) \frac{m\left(B\right)}{m\left(k\right)}}{\sum_{s \in T} p_{A}\left(s\right) \sum_{B \in \mathcal{A}_{k}} F_{s}\left(x_{B}^{|A|}\right) \frac{m\left(B\right)}{m\left(k\right)}}$$

$$p_{A+r}\left(t\right) = p_A\left(t\right)$$

Also, these beliefs should be consistent with the equilibrium masses:

$$p_{A+a}(t) = \frac{m(t, A+a)}{\sum_{s} m(s, A+a)}$$

$$m(t, A+r) = \frac{m(t, A+r)}{\sum_{s} m(t, A+r)}$$

$$p_{A+r}(t) = \frac{m(t, A+r)}{\sum_{s} m(s, A+r)}$$

6.8 Values

Acceptance probabilities should determine the thresholds:

$$\sum_{t \in T} p_A(t) \pi_{t,A}^k = \sum_{t \in T} p_k(t) \left(1 - F_t(x_A^k)\right)$$

At the threshold $x_A^k = x$, the values from accepting and rejection should be the same

$$\sum_{t \in T} \left(\frac{f_t(x) p_k(t)}{\sum_{s \in T} f_s(x) p_k(s)} \right) \left(\pi_{t,k}^{|A|} u_t + \left(1 - \pi_{t,k}^{|A|} \right) (V_{A+a} - c) \right) = V_{A+r} - c$$

6.9 Equilibrium Flows

The flow into state A should be equal to the flow out of state A for all types t:

$$m\left(t,A\right)\pi_{t,A}^{k}\left(1-\sum_{s\in T}p_{k}\left(s\right)\pi_{s,k}^{|A|}\right))=m\left(t,A+a\right)\text{ not correct right now}$$

$$m\left(t,A\right)\left(1-\pi_{t,A}^{k}\right)=m\left(t,A+r\right)$$

(Note: need probability of meeting k)