

Computational Methods in Economics

Algorithms

January 14, 2020

Computing and Algorithms

- Algorithm: well-defined computational procedure that takes some value as **input** and produces some value as **output**.
- Designing an algorithm is about finding the most **efficient** way to create an input/output relationship.

Stable Matching (1)

- Gale and Shapley (1962): Can we design a college admissions process or a job recruiting process that was self-enforcing?
- Consider the marriage problem.
- A set of men and women desire to form partnerships.
- Each woman has a preference ordering over men.
- Each man has a preference ordering over women.
- Based on those preferences, women make offers to men, and marriages happen.

Stable Matching (2)

Initially all m in M and w in W are free.

While there is a man m who is free and hasn't proposed to every woman w

Choose such a man m

Let w be the highest-ranked woman in m 's preference list to which m has not yet proposed

If w is free then

(m , w) become engaged

Else w is currently engaged to m'

If w prefers m' to m then

m remains free

Else w prefers m to m'

(m , w) become engaged

m' becomes free

Endif

Endif

Endwhile

Constraints

- Physical constraints (RAM, storage).
- Worst-case and average-case analysis. (Benchmark)
- Order of growth (scale).

Structure

- Initialization.
- Maintenance.
- Termination.

Design

- Incremental approach
- Divide and Conquer

Divide-and-conquer approach

- **Divide** the problem into a number of subproblems that are smaller instances of the same problem.
- **Conquer** the subproblems by solving them recursively.
- **Combine** the solutions to the subproblems into the solution for the original problem.

Examples

- Search and Matching
- EM algorithm

Theory
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Example 1
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Example 2
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Computation
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Theory

Example 1

Example 2

Computation

Searching, Matching and Migrating: Space and population

- Geography
 - **Discrete** set \mathcal{J} of J cities j where people both live and work
 - Exogenous sites summarized by the distance matrix $d = (d_{jl})$
 - Endogenous **population** size L_j and **number of firms** N_j
 - Some cities may disappear in equilibrium
- Land market
 - Discrete choice: **one unit** of land per individual
 - Competition between **commercial** and residential real estate
 - Firms, vacant or active, consume ψ units

$$R_j = R(L_j + \psi N_j)$$

- All individuals are **workers**
 - Fixed total population (normalized to 1)
 - Endowed with individual **ability** $h \sim \ell(\cdot) = U([0, 1])$
 - Die at rate Ξ (perpetual youth, exit value set to zero)

Searching, Matching and Migrating: Workers

- Set-up
 - Maximize total discounted (at rate r) lifetime income
 - **Unemployed** $i = u$ are in measure $u_j(h)$ and earn b (fixed)
 - **Employed** $i = e$ are in measure $\ell_j(h) - u_j(h)$ and earn w
 - All workers are endowed with bargaining power β
 - Mobility cost $\phi_{jl} = \phi(d_{jl})$ between two cities
- Initial location problem
 - New urban workers, in measure $\omega_j(h)$, start off as unemployed
 - **Reshuffling** is irrespective of skills (no skill dynasties):

$$\omega_j(h) = \Xi \ell(h) L_j$$

- Assumption: workers may only migrate upon finding a job
 - Means that mobility costs (or housing market frictions) are very high for unemployed workers

Searching, Matching and Migrating: Firms

- Set-up
 - Distribution of latent **productivity** $p \sim g(\cdot) = U([0, 1])$
 - Firms equate **jobs**: not a theory of firm size
 - Value of a type- p vacancy: $\Pi_j^v(p)$
 - If matched, **location-invariant** production technology $f(h, p)$
- Vacancies
 - Local measure of type- p vacancy $v_j(p)$
 - Total number of vacancies $V_j = \int v_j(p) dp$
 - **Selection**: lower bound \underline{p}_j s.t. $\Pi_j^v(\underline{p}_j) = 0$
- Matches
 - Joint match distribution $m_j(h, p)$
 - Exogenous destruction at rate δ
 - Local accounting relationship $L_j - U_j \equiv N_j - V_j$

Searching, Matching and Migrating: Search

- Set-up
 - Continuous time, infinite horizon, **random** search
 - Both off-the-job and on-the-job & both within and between cities
- Jobseekers
 - Intensity-adjusted measure E_j of jobseekers in city j
 - $\varrho_{lj} = \varrho(d_{jl}) \in [0, 1]$ **connectedness** of workers in l to jobs in j
 - ζ relative search efficiency of employed workers

$$E_j = U_j + \zeta(L_j - U_j) + \sum_{k \neq j} \varrho_{kj} [U_k + \zeta(L_k - U_k)]$$

- Matching
 - $\mathcal{M}(E_j, V_j)$ number of matches in j and $\theta_{lj} = \frac{\varrho_{lj} \mathcal{M}(E_j, V_j)}{E_j \times V_j}$

Meeting rates

	Worker $(i, j) \rightarrow$ Vacancy in l (independent of h)	Vacancy in $j \rightarrow$ Worker (i, l) (independent of p)
$i = u$	$\theta_{lj} v_l(p)$	$\theta_{lj} u_l(h)$
$i = e$	$\zeta \theta_{lj} v_l(p)$	$\zeta \theta_{lj} m_l(h, p')$

Searching, Matching and Migrating: Value functions and surplus

- Agents' value functions
 - The value for a type- p firm in city j when matched with a type- h worker paid at wage w is $\Pi_j^f(h, w, p)$
 - The corresponding value for the worker is $\mathcal{V}_j^e(h, w, p)$
 - The value of a type- h unemployed worker in city j is $\mathcal{V}_j^u(h)$
- Match product (**transferability**)

$$\forall w, \quad \mathbb{P}_j(h, p) \equiv \Pi_j^f(h, w, p) + \mathcal{V}_j^e(h, w, p)$$

- Surplus

$$\mathcal{S}_{jl}^*(h, p) = \mathbb{P}_l(h, p) - \mathcal{V}_j^u(h) - \Pi_l^v(p)$$

Searching, Matching and Migrating: Wage determination

- Bargaining à-la Lise et al. (2016) over gross match surplus
 - But match occurrence depends on net surplus
- Reservation-wage strategy $\psi_{jl}(h, p)$:

$$\forall p \in \mathcal{P}_{jl}^0(h) \equiv \{y : \mathcal{S}_{jl}(h, y) \geq 0\},$$

$$\mathcal{V}_l^e(h, \psi_{jl}(h, p), p) - \mathcal{V}_j^u(h) = \beta \mathcal{S}_{jl}^*(h, p)$$

$$\forall h \in \mathcal{H}_{jl}^0(p) \equiv \{x : \mathcal{S}_{jl}(x, p) \geq 0\},$$

$$\Pi_l^f(h, \psi_{jl}(h, p), p) - \Pi_l^v(p) = (1 - \beta) \mathcal{S}_{jl}^*(h, p)$$

- Firm-switching wage-change strategy $\psi_{jl}(h, p, p')$:

$$\forall p' \in \mathcal{P}_{jl}^c(h, p) \equiv \{y : \mathcal{S}_{jj}(h, p) \leq \mathcal{S}_{jl}(h, y)\},$$

$$\mathcal{V}_l^e(h, \psi_{jl}(h, p, p'), p') - \mathcal{V}_j^u(h) = \mathcal{S}_{jj}^*(h, p) + \beta[\mathcal{S}_{jl}^*(h, p') - \mathcal{S}_{jj}^*(h, p)]$$

$$\forall p' \in \mathcal{P}_{jl}^f(h, p) \equiv \{y : \mathcal{S}_{jj}(h, y) \leq \mathcal{S}_{jl}(h, p)\},$$

$$\Pi_l^f(h, \psi_{jl}(h, p', p), p) - \Pi_l^v(p) = (1 - \beta)[\mathcal{S}_{jl}^*(h, p) - \mathcal{S}_{jj}^*(h, p')]$$

Searching, Matching and Migrating: Bellman equations

- Lifetime discounted **income** of a type- h unemployed worker in city j :

$$(r + \Xi)\mathcal{V}_j^u(h) = b - R_j + \sum_{k \in \mathcal{J}} \theta_{jk} \int_{\mathcal{P}_{jk}^0(h)} [\mathcal{V}_k^e(h, \psi_{jk}(h, y), y) - \mathcal{V}_j^u(h)] v_k(y) dy$$

- Expected discounted **profit** of a type- p vacancy in city j :

$$\begin{aligned} r\Pi_j^v(p) = & -\psi R_j + \sum_{k \in \mathcal{J}} \theta_{kj} \int_{\mathcal{H}_{kj}^0(p)} [\Pi_j^f(x, \psi_{kj}(x, p), p) - \Pi_j^v(p)] u_k(x) dx \\ & + \zeta \sum_{k \in \mathcal{J}} \theta_{kj} \int_{\mathcal{H}_{kj}^0(p)} \int_{\mathcal{P}_{kj}^f(x, p)} [\Pi_j^f(x, \psi_{kj}(x, y, p), p) - (1 - \beta)\phi_{jk} - \Pi_j^v(p)] m_k(x, y) dy dx \end{aligned}$$

- **Product** of a type- (h, p) match in city j :

$$\begin{aligned} r\mathbb{P}_j(h, p) = & f(h, p) - (1 + \psi)R_j - \delta [\mathbb{P}_j(h, p) - (\mathcal{V}_j^u(h) + \Pi_j^v(p))] - \Xi [\mathbb{P}_j(h, p) - \Pi_j^v(p)] \\ & - \zeta \sum_{k \in \mathcal{J}} \theta_{jk} \int_{\mathcal{P}_{jk}^e(h, p)} [\mathbb{P}_j(h, p) - (\mathcal{V}_k^e(h, \psi_{jk}(h, p, y), y) + \Pi_j^v(p))] v_k(y) dy \end{aligned}$$

Searching, Matching and Migrating: Steady state

- Destruction rate for a type- (h, p) match in city j :

$$\delta_j(h, p) = \delta + \Xi + \zeta \sum_{k \in \mathcal{J}} \theta_{jk} \int_{\mathcal{P}_{jk}^{\zeta}(h, p)} v_k(y) dy$$

- Motion laws on the labor market:

$$u_j(h) \left(\Xi + \sum_{k \in \mathcal{J}} \theta_{jk} \int_{\mathcal{P}_{jk}^0(h)} v_k(y) dy \right) = \delta \int m_j(h, y) dy + \omega_j(h)$$

$$m_j(h, p) \delta_j(h, p) = v_j(p) \sum_{k \in \mathcal{J}} \theta_{kj} \left(1_{h \in \mathcal{H}_{kj}^0(p)} u_k(h) + \zeta \int_{\mathcal{P}_{kj}^f(h, p)} m_k(h, y) dy \right)$$

$$v_j(p) \sum_{k \in \mathcal{J}} \theta_{kj} \int_{\mathcal{H}_{kj}^0(p)} \left(u_k(x) + \zeta \int_{\mathcal{P}_{kj}^f(x, p)} m_k(x, y) dy \right) dx = \int \delta_j(x, p) m_j(x, p) dx$$

Searching, Matching and Migrating: “Closed” forms

- Conditional on surplus, closed forms for unmatched agents:

$$\begin{aligned}
 (r + \Xi)\mathcal{V}_j^u(h) &= b - R_j + \beta \sum_{k \in \mathcal{J}} \theta_{jk} \int_{\mathcal{P}_{jk}^0(h)} \mathcal{S}_{jk}(h, y) v_k(y) dy \\
 r\Pi_j^v(p) &= -\psi R_j + (1 - \beta) \sum_{k \in \mathcal{J}} \theta_{kj} \int_{\mathcal{H}_{kj}^0(p)} \mathcal{S}_{kj}(x, p) u_k(x) dx \\
 &\quad + (1 - \beta)\zeta \sum_{k \in \mathcal{J}} \theta_{kj} \int_{\mathcal{H}_{kj}^0(p)} \int_{\mathcal{P}_{kj}^f(x, p)} [\mathcal{S}_{kj}(x, p) - \mathcal{S}_{kk}(x, y)] m_k(x, y) dy dx
 \end{aligned}$$

- This allows us to recover a closed-form expression for match product:

$$\begin{aligned}
 r\mathbb{P}_j(h, p) &= f(h, p) - (1 + \psi)R_j - (\delta + \Xi)\mathcal{S}_{jj}(h, p) - \Xi\mathcal{V}_j^u(h) \\
 &\quad + \zeta \sum_{k \in \mathcal{J}} \theta_{jk} \int_{\mathcal{P}_{jk}^c(h, p)} \beta [\mathcal{S}_{jk}(h, y) - \mathcal{S}_{jj}(h, p)] v_k(y) dy
 \end{aligned}$$

Searching, Matching and Migrating: Operationalization

Model	Parameterization
R_j	$R_0 + R_1(L_j + \psi N_j)^{R_2}$
$f(h, p)$	$[h^{(\rho-1)/\rho} + p^{(\rho-1)/\rho}]^{\rho/(\rho-1)}$
ϱ_{jl}	$\exp(-\varrho d_{jl})$
$\mathcal{M}(E_j, V_j)$	$m_0 E_j^{m_1} V_j^{1-m_1}$
ϕ_{jl}	$1_{j \neq l} \times \phi$

Algorithm

For a given $\Theta = (\underbrace{\Xi, r}_{\text{Life}}, \underbrace{\psi, R_0, R_1, R_2}_{\text{Land}}, \underbrace{b, \beta, \delta, \zeta, \rho, m_0, m_1}_{\text{Labor}}, \underbrace{\varrho, \phi}_{\text{Space}})$,

1. Assume $\mathcal{S}_{jl}(h, p)$, $u_j(h)$, $m_j(h, p)$ and θ_{jl} known
2. Compute $\ell_j(h)$, L_j , U_j , V_j (from θ_{jl}), N_j and R_j
3. Construct $\Pi_j^v(p)$
4. Find \underline{p}_j from the zero-profit cutoff condition
5. Recover $v_j(p)$
6. Update $u_j(h)$ and $m_j(h, p)$ using steady state conditions
7. Construct $\mathcal{V}_j^u(h)$ and $\mathbb{P}_j(h, p)$ and update $\mathcal{S}_{jl}(h, p)$ from its definition
8. Repeat steps 3 to 7 until convergence.

Theory
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Example 1
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Example 2
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Computation
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Theory

Example 1

Example 2

Computation

Discrete Mixing Distributions (1)

- Consider a discrete choice model
- An agent i obtains utility U_{ijt} from product j at time t .

$$U_{ijt} = \beta_i x_{ijt} + \epsilon_{ijt}$$

with

- ϵ_{ijt} is iid and extreme value distributed.
- $\beta_i \sim f(\beta \mid \theta)$ where θ represents the parameters of the distribution
- Let y_{it} denote the alternative that agent i chooses in situation t .

Mixing Distributions (2)

- Conditional on β , the probability of y_i is

$$K_i(\beta) = \prod_t \frac{\exp(x_{iy_i t})}{\sum_j \exp(\beta x_{ijt})} \quad (1)$$

- The choice probability is

$$P_i(\theta) = \int K_i(\beta) f(\beta \mid \theta) d\beta$$

Mixing Distributions (3)

- For discrete distribution with support at β_c , $c = 1, \dots, C$, the choice probability is

$$P_i(\theta) = \sum_c s_c K_i(\beta_c) \quad (2)$$

where $s_c = f(\beta_c \mid \theta)$ is the share of the population that has coefficients β_c .

Mixing Distributions (4): EM

- Select initial values β_c^0 and $s_c^0 \forall c$. Probably start with equal shares and estimates from a logit for partitions of the sample into C subsamples.
- Calculate the weights as

$$h_{ic}^0(\theta^0) = \frac{s_c^0 K_n(\beta_c^0)}{\sum_m s_m^0 K_n(\beta_m^0)} \quad (3)$$

- Update the shares as

$$s_c^1 = \frac{\sum_n h_{ic}^0}{\sum_m \sum_i h^0} \quad (4)$$

- Run C standard logits on the data using weights h_{nc}^0 in the c_{th} run, and derive β_c^1

Practice

- Number of iterations.
- Size of the update step.
- Convergence criteria.

Theory

Example 1

Example 2

Computation

Fundamentals

- ?solve or help(solve).
- R is case sensitive.
- “|-” is an assignment operator.
- Avoid loops in R.

Benchmarking

- microbenchmark
- benchmarkme