# Search and Matching - Ph.D. Training Course Lecture 4: Search and Sorting 

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## Search and Sorting

- Big focus in labor: unemployment
- Less focus: "unsuitable" employment
- Examples:
- Dentist working at a fast-food restaurant
- Ph.D. economist working as taxi driver
- Why is this hard: observational problems (output hard to observe)
- Need more theory to understand this
- Frictions: induce mismatch (but other things do as well).


## Sorting and Search Frictions: The Basics

We keep the basic elements of the framework before, but

- Each worker has a type $x$; distr. $H_{w}$
- Each job has a type $y$; distr. $H_{m}$
- The output is $f(x, y)$ [same as $V(m, w)$ with men and women]
- Matching through matching function (directed or random).
- succesful: firm gets $f(x, y)-w$ and worker gets $w$ (risk-neutrality).
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- Some prob $s \geq 0$ that job survives to next period.
- unsuccessful: workers unemployment payoff $b \geq 0$, firms get 0 .
- Potentially try next period again (discount $\delta \in[0,1)$ ).


## Sorting

How does sorting work now? Who get's matched with whom? Why?
Recall from frictionless matching: PAM if $f_{x y}>0$.
Things change with frictions:

- It is not only important which partner one gets,
- But it is also important whether one gets a partner at all.
- The second part tends to favor NAM, because the highest types have most to loose and are most likely to accept lower matches if that helps them getting matched.
- Most easily explained in directed search.


## Sorting in Directed Search

Sorting in Directed Search. (based on Eeckhout-Kircher ECTR. See also Shi 01, Shimer 05)
Assume bilateral meetings. (otherwise auctions, see Eeckhout-Kircher JET)
Firm y posts $(w, x)$ combination to maximize:

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\begin{aligned}
& \max _{x, w} m(\lambda(x, w))[f(x, y)-w] \text { s.t. } n(\lambda(x, w)) w=U(x) . \\
\Leftrightarrow & \max _{x, \lambda} m(\lambda) f(x, y)-\lambda U(x)
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FOC at optimal $\lambda=\Lambda(y)$ and $x=\mu(y)$ :

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SOC according to Hessian:

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\left(\begin{array}{cc}
m^{\prime \prime}(\Lambda) f(\mu, y) & m^{\prime}(\Lambda) f_{x}(\mu, y)-U^{\prime}(\mu) \\
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Can be done. Real complication: deal with possible non-differentiabilities,

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Remarkable symmetry. Stronger than $f_{x y}>0$. (Use graph...)

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## Sorting: Random Search

Sorting with Random Search:

- Downside for theory: much harder (illustrate matching bands)
- Applied upside: breaks perfect matching (feature of data)
- Canonical Model: Shimer-Smith ECTR
- Considitions for increasing matching bands (PAM):
- $f$ sm, $f_{x} \log$-sm, $f_{x y} \log$-sm,... (implies $f \log -s m$ )
- More interesting for applied work:
- Can we identify the production function from observed data?
- Can we say whether sorting is positive, negative, etc?
- Can we say how much value is lost from mismatch?
- How much could the market improve (increase $b$, not done yet)?


# Identificaiton of Sorting under Random Search Identification of Sorting with Random Search: <br> Fixed search costs $c>0$ (Atakan ECTR, Eeckhout-Kircher REStud, Gautier-Teulings) Surplus from $x$ matching with $y$ : 

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- Worker's type:


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- With $s \delta=1$ : firm type cannot be identified
- With $s \delta<1$ : firm type is identified by excess payments (what workers get beyond their reservation wage Hagedorn-Law-Manovskii)


## Other Identification Strategies

Other ways of identification:

- Hight and width of wage function (use picture)
(Gautier-Teulings: mismatch costs $\approx$ unemployment costs)
- Similar types of co-workers (de Melo)
- Speed of sorting with search intensity (Lentz...)

Problematic:

- Correlation of worker and firm fixed effects (reason: non-monotonicity of wage function)


## Different reason for mismatch: shocks or learning

Open questions about sorting:

- How to handle on-the-job search (important for wage dispersion, recently introduced by Lise-Robin, Hagedorn-Law-Manovskii, Gautier-Teulings...)
- How to handle ideosyncratic and aggregate shocks (Lise-Robin)
- To use it for sensible policy questions:
- What is the effect of higher unemployment insurance
- What is the effect of job protection....

Different way to think about mismatch:

- Shocks to types or learning
- Long literature going back to Waldmann...
- Short exposition based on my own work
- Message:
- Combining search and shocks might be important
- Small improvements on any of these can be a great dissertation
- Keep relevance in mind
- Keep data in mind

