# Managing congestion in dynamic matching markets

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### Introduction

SEARCH FRICTIONS IN DYNAMIC MARKETS

### Search frictions

#### Search literature:

Diamond (1982), Mortensen (1982), Pissarides (1984) Hosios (1990), Mortensen & Pissarides (1994), Moen (1997) Acemoglu & Shimer (2000), Kennes, King & Julien (2001) Guerrieri (2008), Kircher (2009), Lewis & Wang (2013)

### Take-home messages:

- Costly search, coordination failure cause inefficiency
- Generally, as search costs vanish, efficiency is recovered

# Dynamic online matching markets



These sites provide value by *facilitating search*, allowing agents who can profitably transact to find each other.

### Consequences of low frictions

- Finding and contacting potential partners
   is easier than ever, but
   many inquiries may not get a response.
- If response rates are low,
   people may spend lots of time applying,
   even if each one is quick.
- This can lower response rates even further; a vicious cycle.

#### Anecdotal evidence

### oDesk worker:

"I applied to more than 100 jobs with no response... I have 4.8 star feedback, and I meet all the qualifications needed for the job."

#### oDesk website:

When clients send out invitations and freelancers don't reply, it's a frustrating experience that makes those clients less likely to hire anyone.

#### Our work

Model a two-sided market in which:

- Matching is dynamic and asynchronous
- Learning preferences (screening) is costly
- Agents may be unavailable (this is unobservable)

Online labor markets, with employers and applicants



### Key conclusions

When application costs are low:

- 1. Applicants contact many employers.
- 2. Employers screen applicants who are no longer available.
- 3. This can drive employer welfare to zero

Lower frictions may not be universally good.

Intervening by limiting applications can:

- generate Pareto improvements, and
- boost employer welfare to a natural upper-bound

# A dynamic matching market

THE GAME THEORETIC MODEL

# A dynamic matching market

#### Two-sided:

Applicants and employers

#### Asynchronous:

Agents arrive and depart over time

#### One-to-one:

Agents match to at most one other agent

#### Homogeneous:

• Ex ante, the agents look the same to each other

#### **Overview**

- 1. Employers arrive, and live for unit lifetime
- 2. Applicants arrive, and apply to a subset of employers in the system
- 3. Upon leaving, employers screen and make offers to applicants
- 4. Upon receiving offers, applicants can accept or reject

# **Employer utility**

Agents are utility maximizers.

#### Employers:

- Want a *compatible* (i.e., qualified) applicant
- **P**(s compatible with b) =  $\beta$  (i.i.d.)
- Value from a compatible match: V
- Must screen before hiring
   Screening cost per applicant: C<sub>s</sub>

#### Overall utility:

 $V \times 1$ [match to a compatible applicant]  $-C_S \times \#$  of screened applicants

# **Applicant utility**

Agents are utility maximizers.

#### Applicant:

- Cost per application: C<sub>A</sub>
- Value from any match: W

#### Overall utility:

$$W \times 1$$
[matched to a employer]  
-  $C_A \times \#$  of applications sent

Note: As a result, "accept the first offer" is a (weak) dominant strategy.

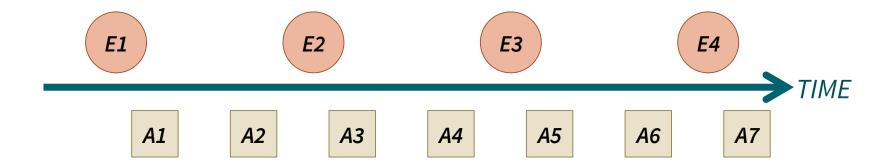
# **Dynamics: Employer arrivals**

Employers arrive at rate *n*.



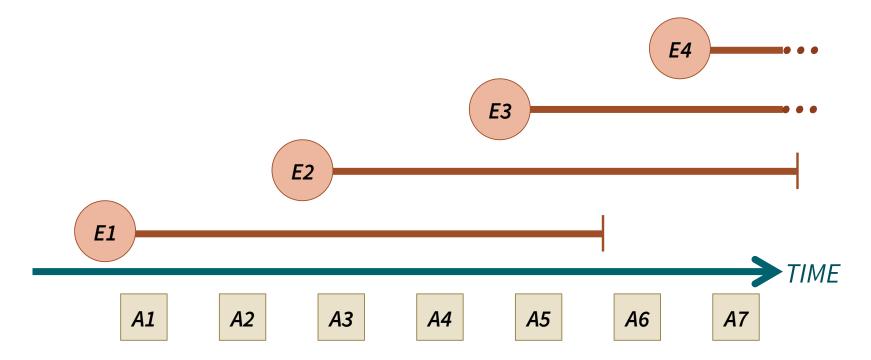
# **Dynamics: Applicant arrivals**

Applicants arrive at rate Rn. [R = market imbalance]



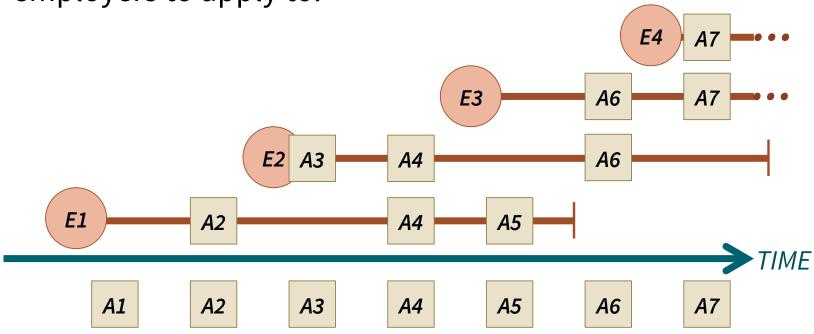
# **Dynamics: Employer lifetimes**

Employers live for unit lifetime.



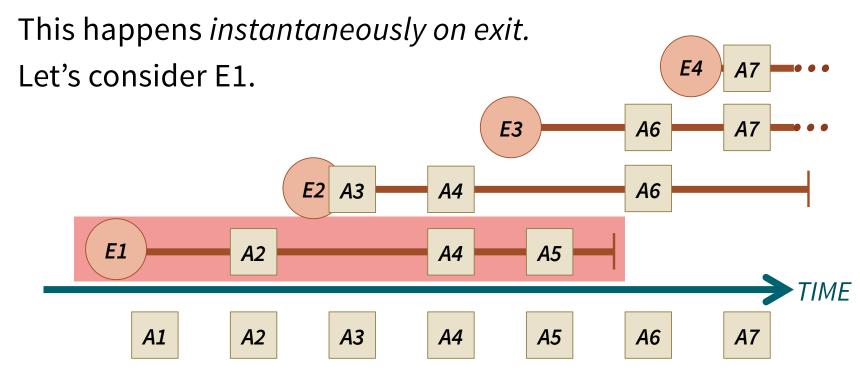
# **Dynamics: Applications**

On arrival, applicants choose a subset of employers to apply to.



# Dynamics: Employer screening and offers

On exit, employers can screen for compatibility, and make offers to compatible applicants.



# Dynamics: Employer screening and offers

Let's consider E1.

Suppose she screens A2 and A5, and both are not compatible.

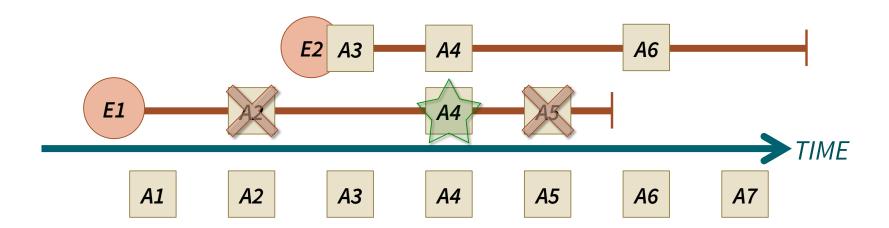
But A4 is compatible, so E1 makes her an offer.

A4 is available, and it is a weak dominant strategy to accept, so she accepts.



# **Dynamics: Matching**

So E1 and A4 are matched.

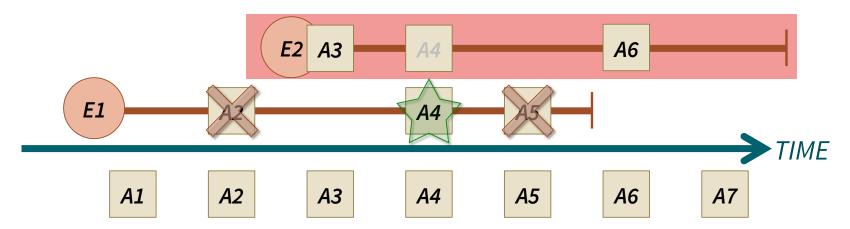


### Dynamics: A lack of availability information

So E1 and A4 are matched.

Now A4 is no longer available... but E2 doesn't know that when she exits!

If E2 screens A4, this is *wasted effort* – even if A4 turned out to be compatible.



### Dynamics: A lack of availability information

So E1 and A4 are matched.

Now A4 is no longer available... but E2 doesn't know that when she exits!

If E2 screens A4, this is *wasted effort* – even if A4 turns out to be compatible.



"Maybe that person had

a limited number of hours to give away,

multiple interviews set up, and

somebody else just snatched him/her before you."

### The challenge

Each employer chooses a screening and offer strategy. Each applicant chooses an application strategy.

Agents may hold complex beliefs, and play complex strategies as a result.

Characterizing PBE is challenging.

### A mean field approach

APPROXIMATING EQUILIBRIA VIA ASYMPTOTICS

### A mean field model

We simplify analysis by considering a mean field model:

Seems plausible that in markets with "many" agents:

- 1. For each applicant, whether an application yields an offer becomes *i.i.d* (say probability *p*)
- 2. For each employer, at time of exit, whether each applicant is available becomes *i.i.d.* (say probability *q*)

Do these assumptions help analyze the game? Do they hold as market size grows?

Answer to both is YES!

# Optimal responses

Under previous assumptions, given p and q, agents have simple optimal strategies.

Applicant optimal response:

- Expected number of applications *m* 

Employer optimal response: A (possibly) mixed strategy

- With probability a

Simple sequential screening: screen, then make offer if compatible; repeat if needed

- With probability 1 – a

Exit without screening anyone

### Mean field equilibrium

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A mean field equilibrium is
a pair of strategies (m, a) and
a pair (p, q)
such that
agents play optimally, given their beliefs, and
agent beliefs are consistent with agent strategies
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#### Theorem:

MFE exist and are unique.

### Mean field approximation

#### Theorem:

As  $n \to \infty$ , the initial mean field assumptions hold.

In particular, any MFE is an approximate equilibrium for large enough  $n\,.$ 

(Proof via a stochastic contraction argument.)

### **Model conclusions**

THE VALUE OF SIMPLE INTERVENTIONS

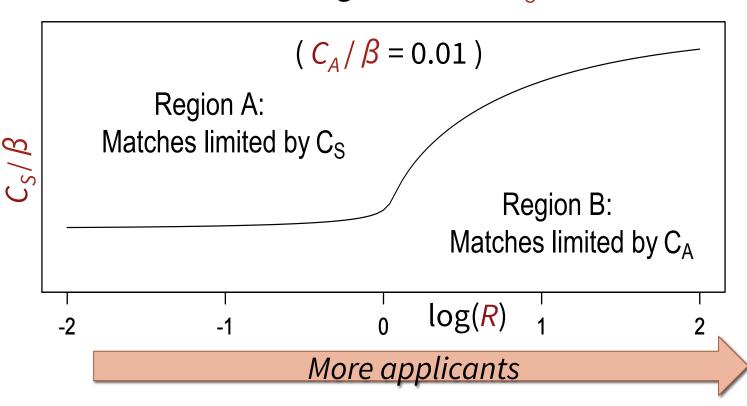
Higher cost

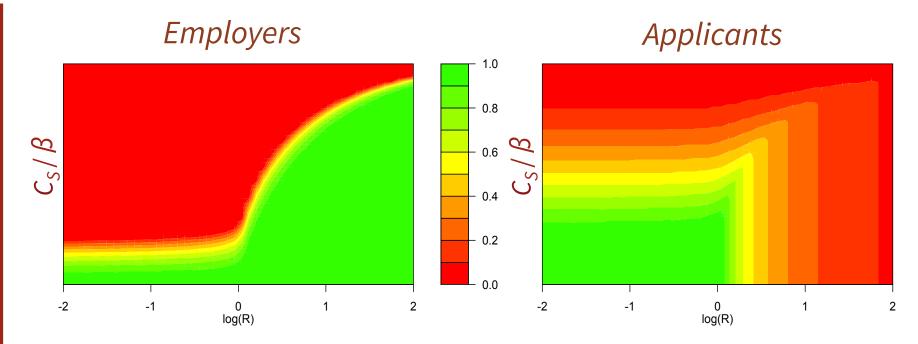
Market performance depends on three parameters:

1. # of applicants / # of employers: R

2. Normalized application cost:  $C_A/\beta$ 

3. Normalized screening cost:  $C_S/\beta$ 



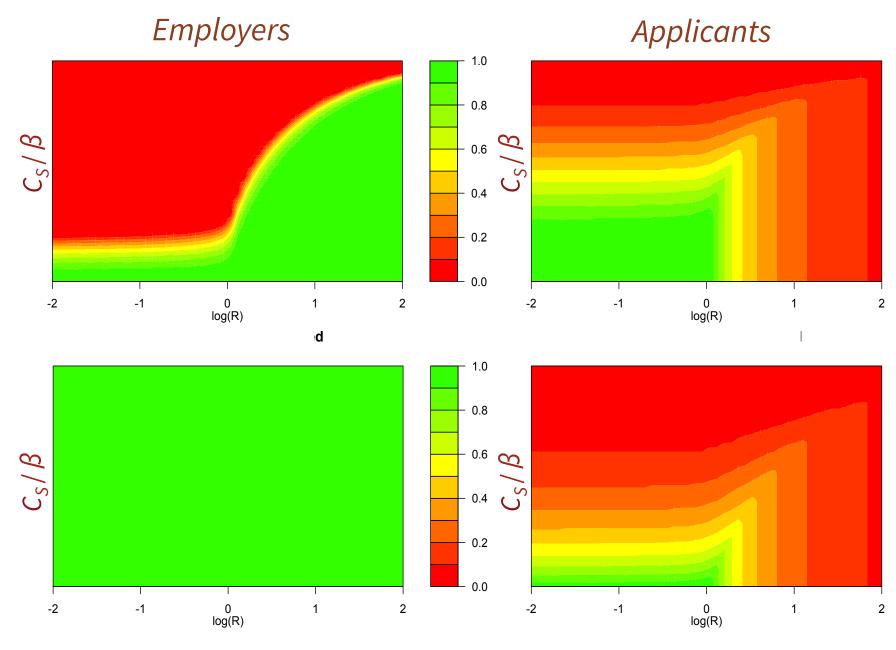


### Employers (Region A)

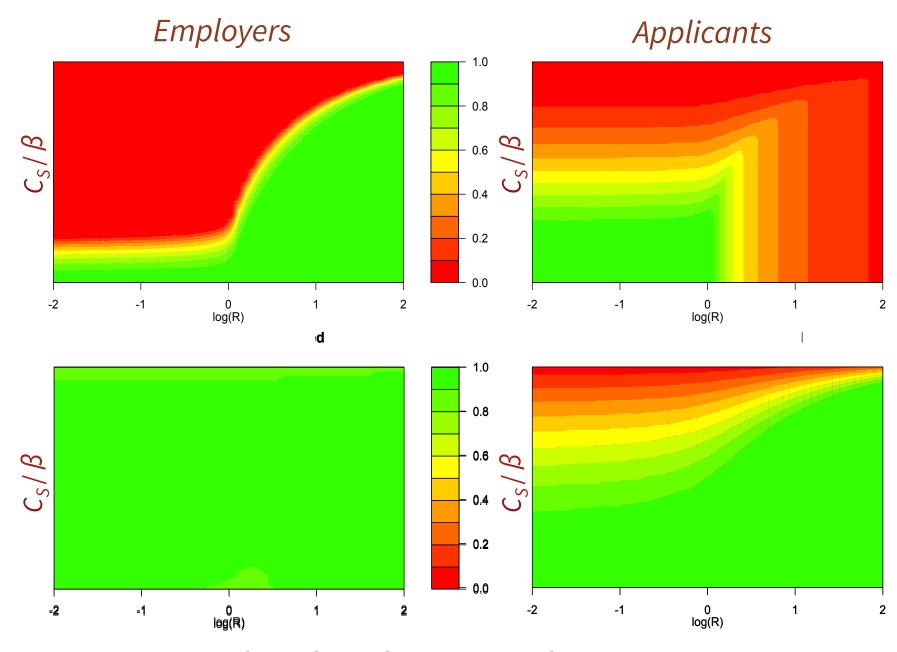
- > Applicants are unlikely to be available when screened.
- Employers get zero surplus, and some leave the market.
- Lowering  $C_A$  expands Region A.  $\triangleright$  Losses persist as  $C_A \rightarrow 0$ .

#### **Applicants:**

- > Face "tragedy of the commons."
- Losses greatest when many applicants go unmatched.



What about raising  $C_A$ ?



What about limiting applications?

# Helping applicants

Define **A** as the *maximum applicant welfare* over all choices of *m* and *a*.

#### **Theorem:**

Suppose we restrict applicants to choose  $m \le L$ .

There always exists an *L* that can strictly improve applicant welfare.

In Region B, the right choice of *L* improves applicant welfare all the way to **A.** 

#### **Theorem:**

Raising  $C_A$  only reduces applicant welfare.

# Helping employers

Define **E** as the *maximum employer welfare*, over all choices of *m* and *a*.

#### Theorem:

Suppose we restrict applicants to choose  $m \le L$ .

In Region A, the right choice of *L* improves employer welfare *from zero* all the way to **E**.

#### **Theorem:**

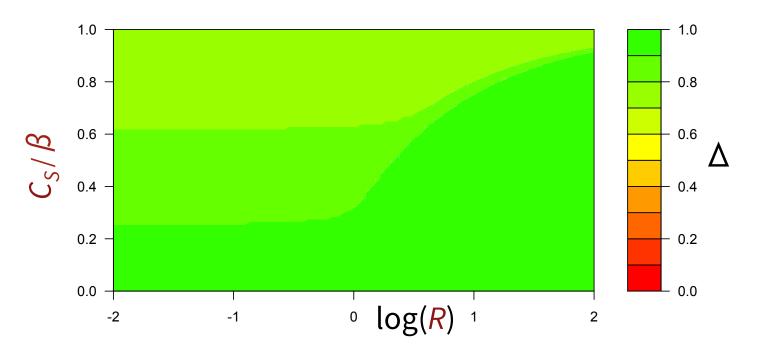
Raising  $C_A$  has the same effect.

### **Fairness**

Let **A**<sup>LIMIT</sup> (resp., **E**<sup>LIMIT</sup>) be the maximum welfare for applicants (resp., employers) via an application limit.

Let  $\Delta$  be the *largest fraction* such that we can simultaneously achieve  $\Delta \mathbf{A}^{\text{LIMIT}}$  and  $\Delta \mathbf{E}^{\text{LIMIT}}$ .

How large is  $\Delta$ ?



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#### Theorem:

 $\Delta \geq 3/4$ .

### Conclusion

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#### When:

- application costs are low,
- screening is costly,
- and availability is unknown,

there can be a significant negative externality from applicants on employers.

Simple, *low-knowledge* interventions can benefit both sides.