



# People and the Labour Market:

## *Incorporating Behavioural Heterogeneity to Model Unequal Outcomes*

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# Overview

**Motivation**

**Theoretical Embedding: Behaviour & the Job Search Process**

**Model: Incorporating Behavioural Heterogeneity**

**Preliminary Application to US Data**

**Discussion**



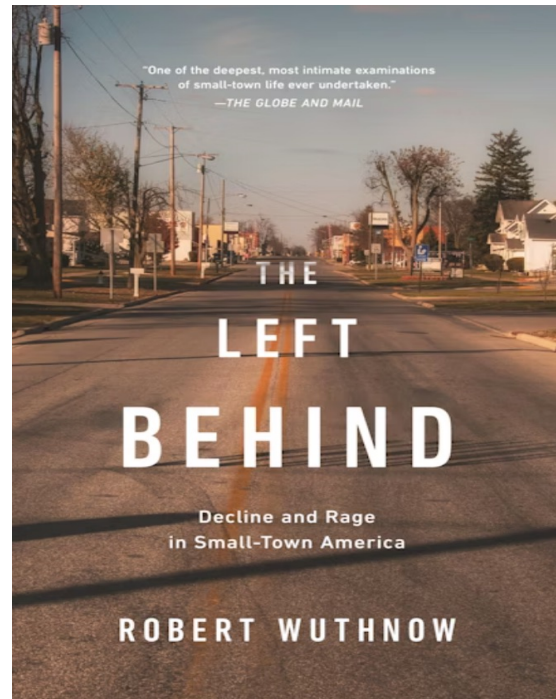
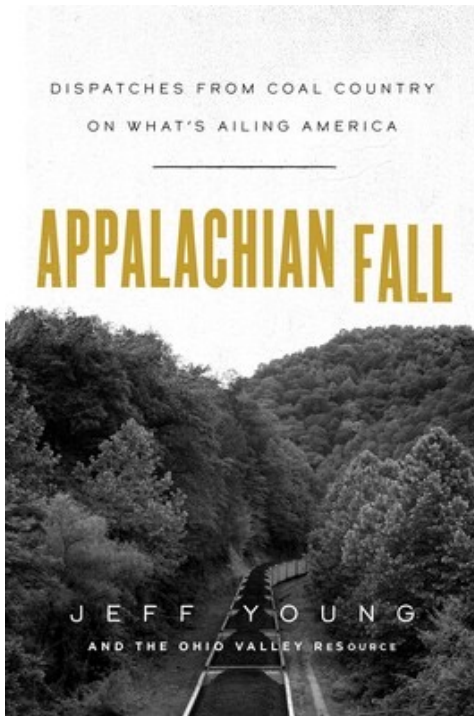
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# Motivation

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## Policy Challenge

Business cycles and structural transformation processes are often accompanied by heterogeneous labour market frictions.



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## Method

Agent-based models are a useful tool for modelling or simulating complex labour market dynamics.

# Motivation

## Some agent-based models of labour market dynamics:

Structural reform policy and unemployment and income inequality (Dosi et al. 2018)

Social networks on upskilling and job market opportunities (Gemkow and Neugart 2011)

Effect of migration in Austria (Poledna 2024)

Employment protection legislation and unemployment (Martin & Neugart 2009)

Green transition in Brazil (Berryman et al. 2023)

Decarbonisation of the US power sector (Bücker et al.)

Automation in the US labour market (del Rio-Chanona et al. 2021)

# Motivation

## Policy Challenge

Business cycles and structural transformation processes are often accompanied by heterogeneous labour market frictions.

## Method

Agent-based models are a useful tool for modelling or simulating complex labour market dynamics.

## Gap

Most existing agent-based labour market models neglect the role of behaviour in job search processes, despite their key role in determining heterogeneous outcomes.

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# Motivation

Across agent-based labour market models, **behaviour is often an afterthought despite its core role in job search processes.**

There is a wealth of evidence on *behavioural heterogeneity in job searches* from behavioural labour economics.

Some notable examples:

- **Gendered behaviour in reservation wage setting, search effort, aversion to relocation** has been found to contribute to wage gaps across countries.
- Search effort is influenced by **negative or positive affect towards the job search process itself**. The dynamics of this relationship changes over the search period.
- **Confidence in one's abilities or the job market** defines search effort and/or strategy.



# Motivation

## Policy Challenge

Business cycles and structural transformation processes are often accompanied by heterogeneous labour market frictions.

## Method

Agent-based models are a useful tool for modelling or simulating complex labour market dynamics.

## Gap

Most existing agent-based labour market models neglect the role of behaviour in job search processes, despite their key role in determining heterogeneous outcomes.

## Relevance

Job search behaviour not only underlies the labour market adjustment process itself but also contributes to labour market inefficiencies such as gender wage gaps and long-term unemployment.

# Ultimate Goal

**Behaviourally micro-founded agent-based labour market model**



Research Question(s)

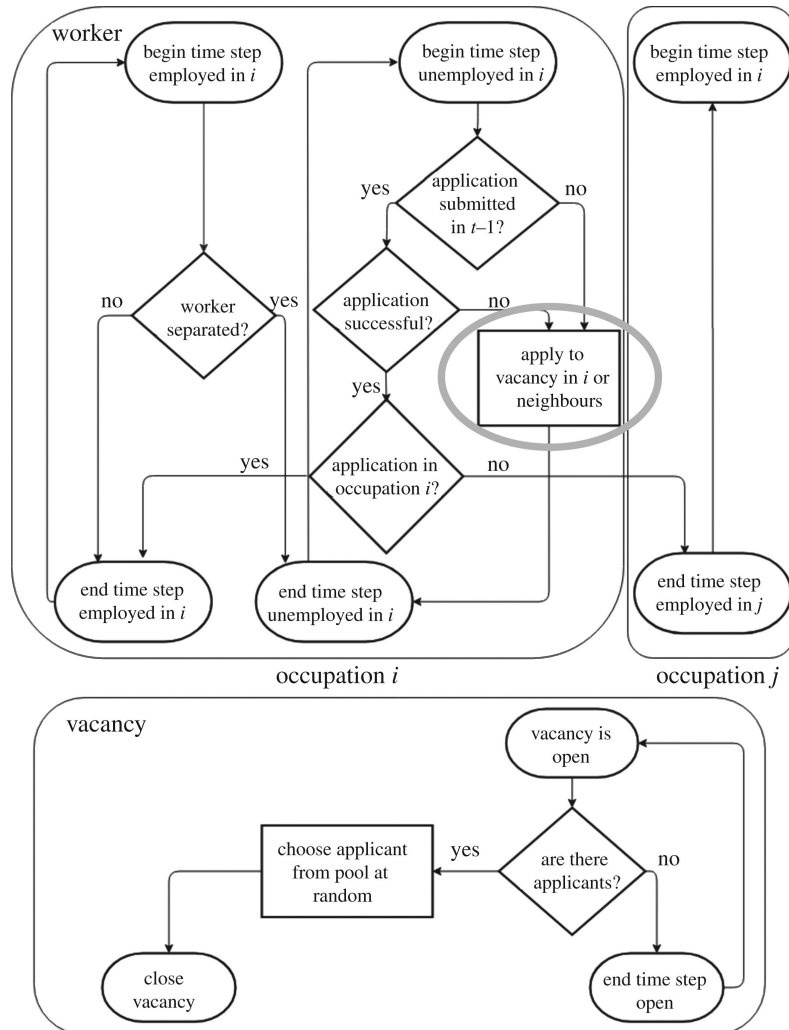
**What is the use of incorporating behavioural heterogeneity into agent-based labour market models?  
(Forthcoming) How will net-zero structural transformation interact with existing labour market inequalities?**

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# Building the Model (1)

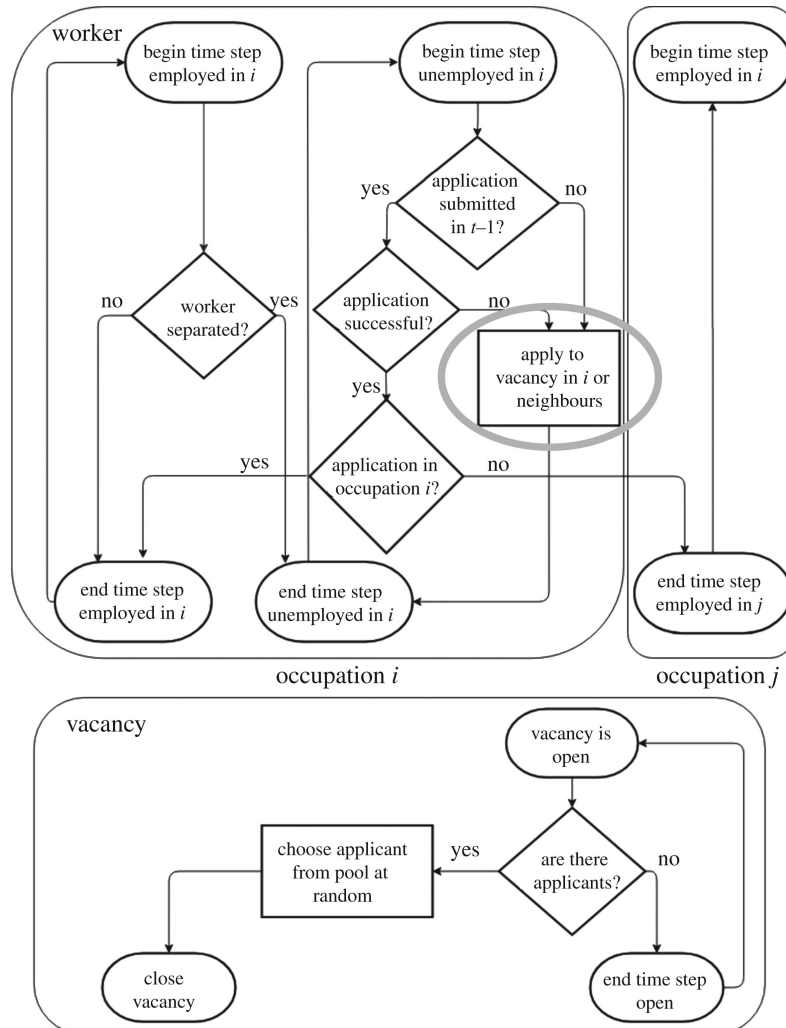
*Starting Point*

# Starting point: del Rio-Chanona et al. 2021



*del Rio-Chanona et al. 2021*

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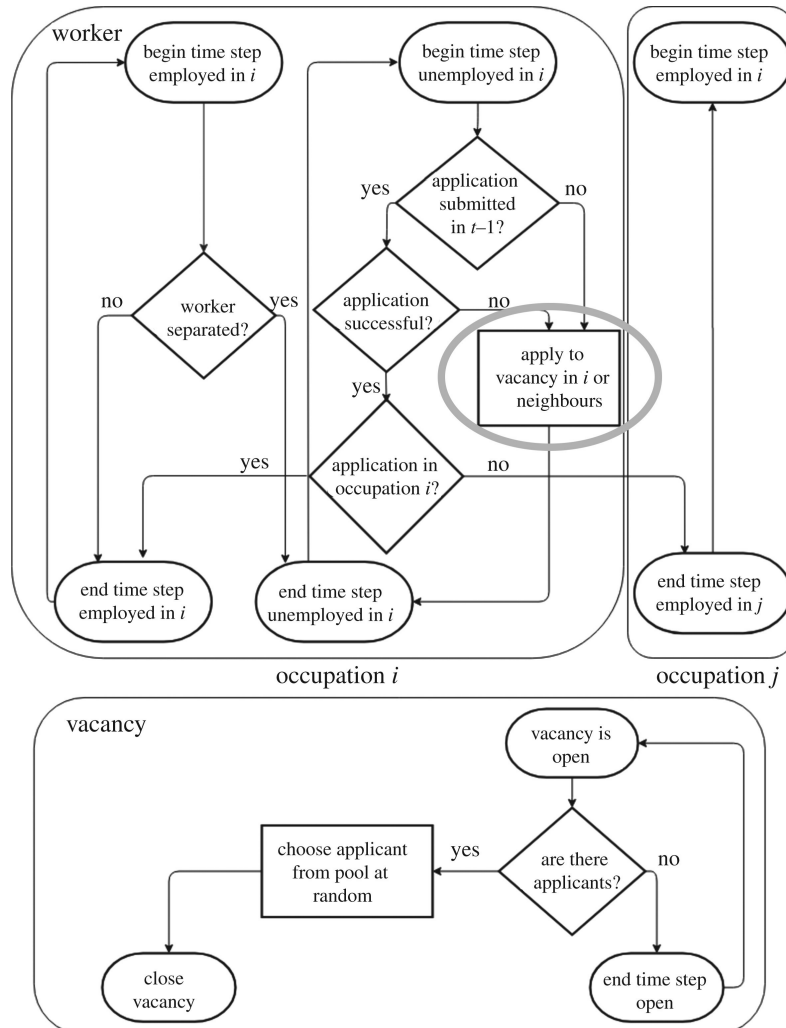
## Occupational mobility network:

**Nodes:** occupations ( $n = 464$ )

**Edges:** directed probability of transitioning between occupations

**Data:** US Current Population Survey

# Starting point: del Rio-Chanona et al. 2021



del Rio-Chanona et al. 2021

## Occupational mobility network:

**Nodes:** occupations ( $n = 464$ )

**Edges:** directed probability of transitioning between occupations

**Data:** US Current Population Survey

**Spontaneous separation and vacancy creation processes** which create movement in the model

**State-dependent separation and vacancy creation processes** which follow fluctuations in labour demand

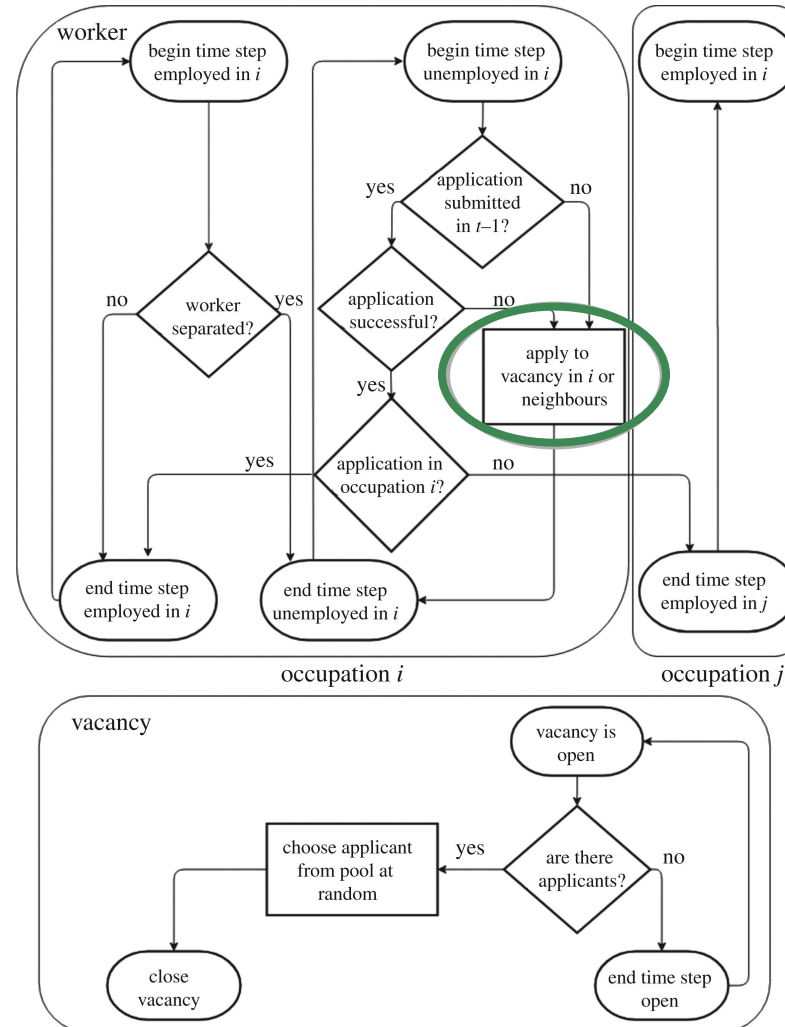
**Labour market adjustment parameter**

**Data:** Calibrated (prelim.) to most recent full and non-recession business cycle periods pre-Covid using JOLTS and BLS time series

# Building the Model (2)

*Incorporating Behavioural Heterogeneity*

# Behavioural heterogeneity in job search





# Surveying job search behaviour

## **Behaviours**

*Choice of search channel (online, newspaper, network, employment agency)*

*Delay search*

**Reservation wage**

*Other non-wage reservation preferences*

**Job search effort** *(hours spent searching, applications sent)*

*Job search intensity (planning, strategy)*

**“Aiming high or low”**

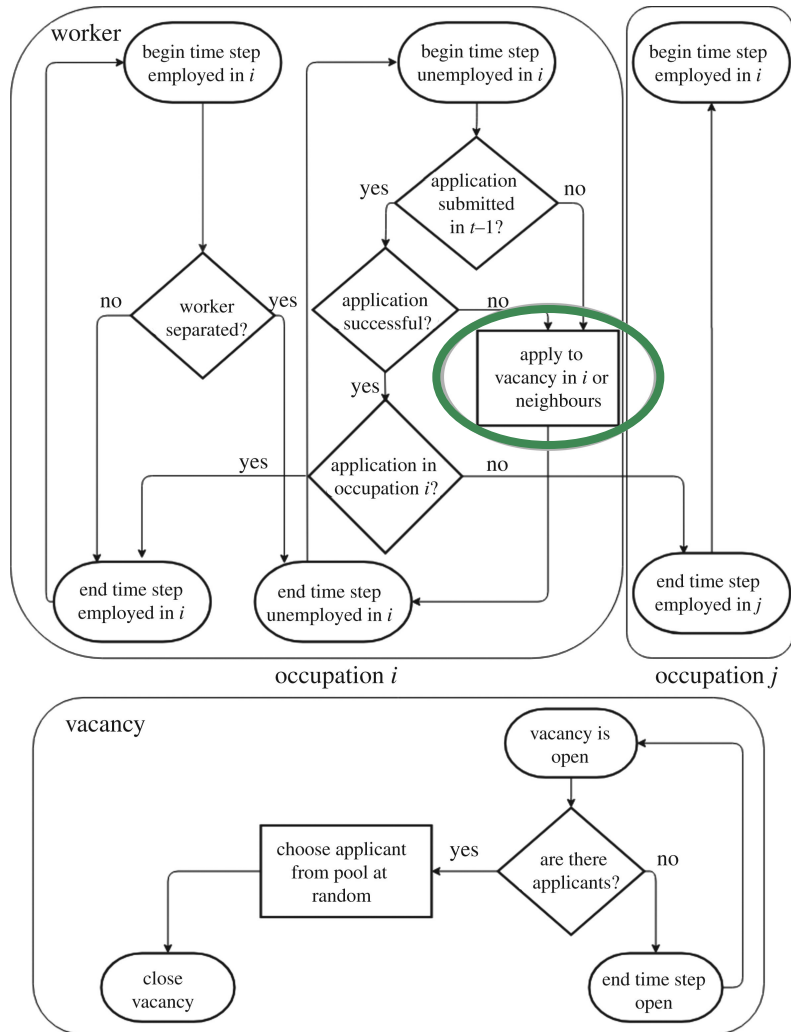
*Apply only for newer job postings*

**Non-linear search effort**

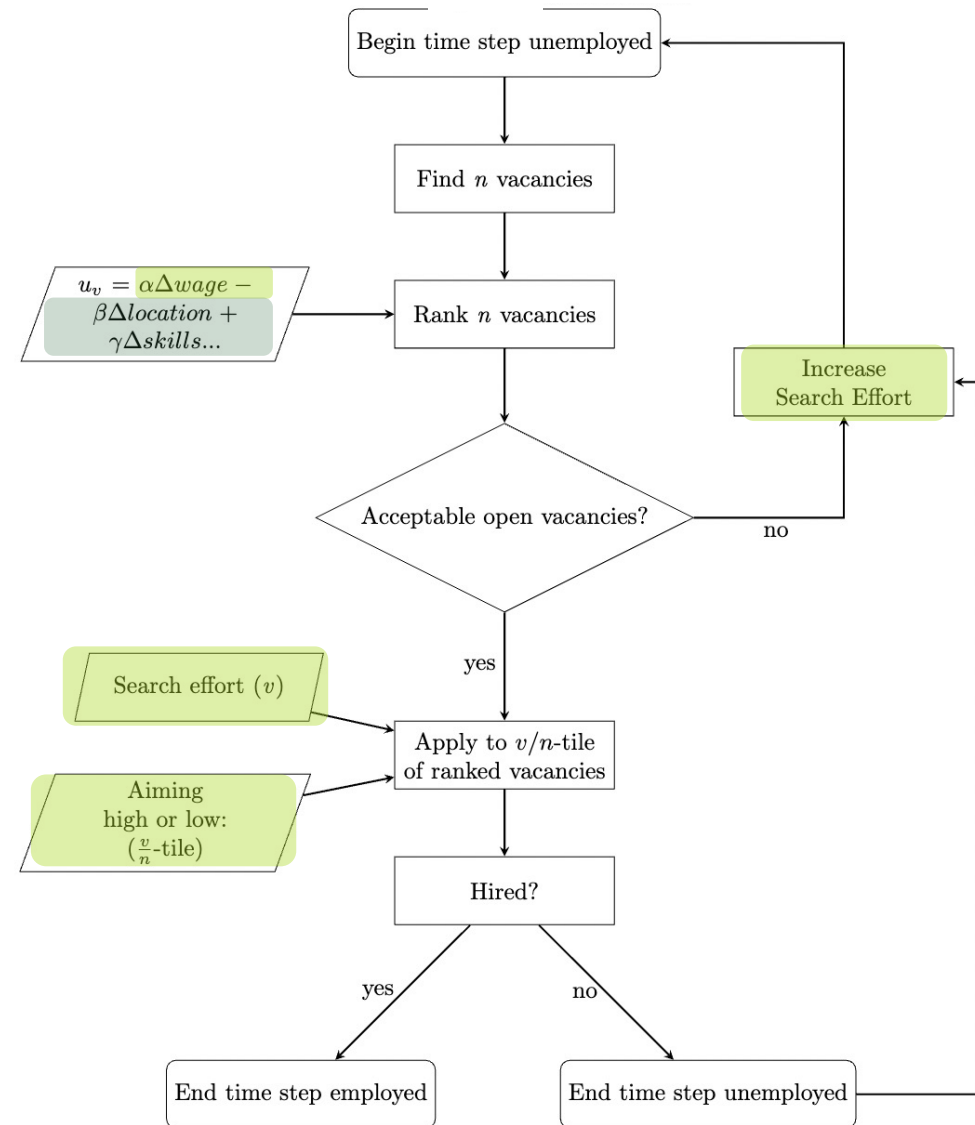
*Dynamism/Oscillating Search Effort*

*Cyclicalities of job search in tight labour markets*

# Incorporating behavioural heterogeneity

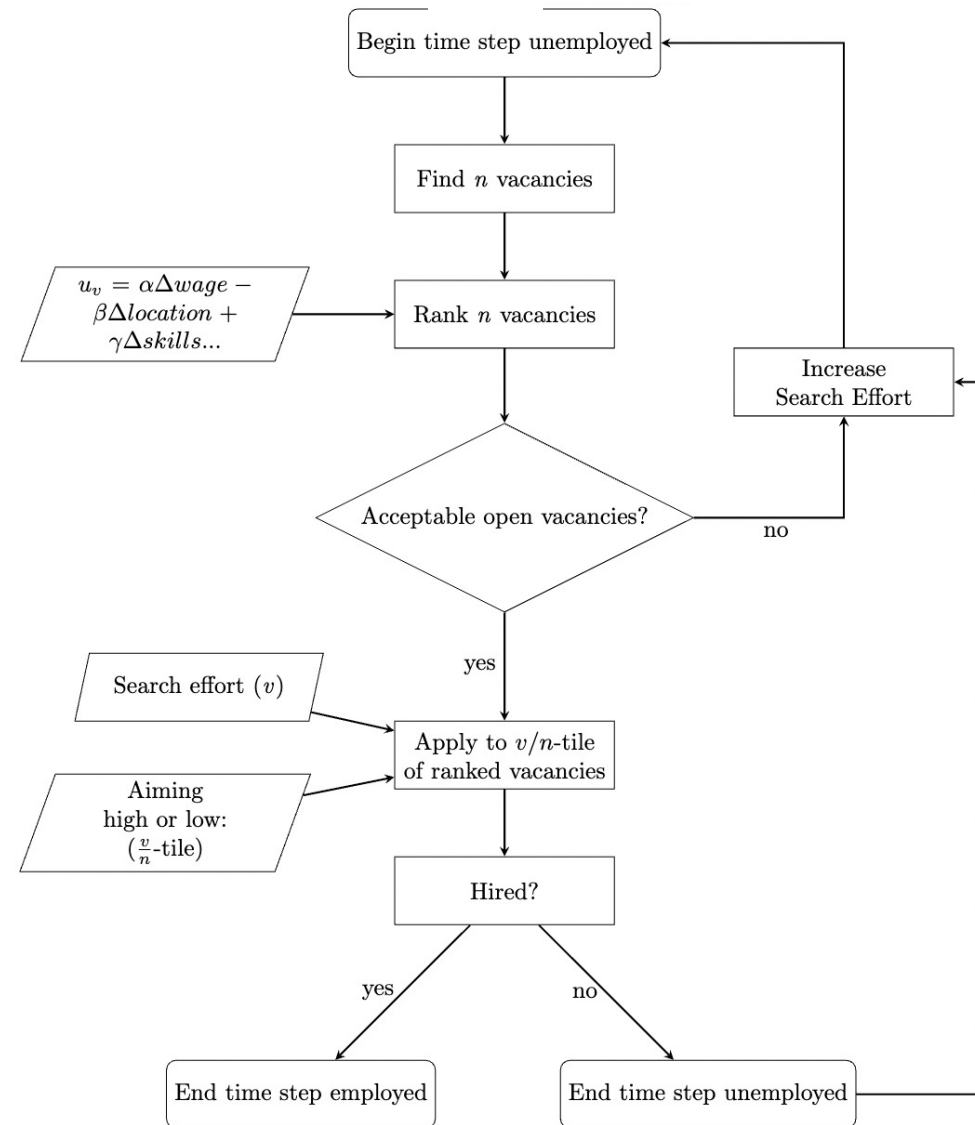


*del Rio-Chanona et al. 2021*



# Incorporating behavioural heterogeneity

Vacancies:

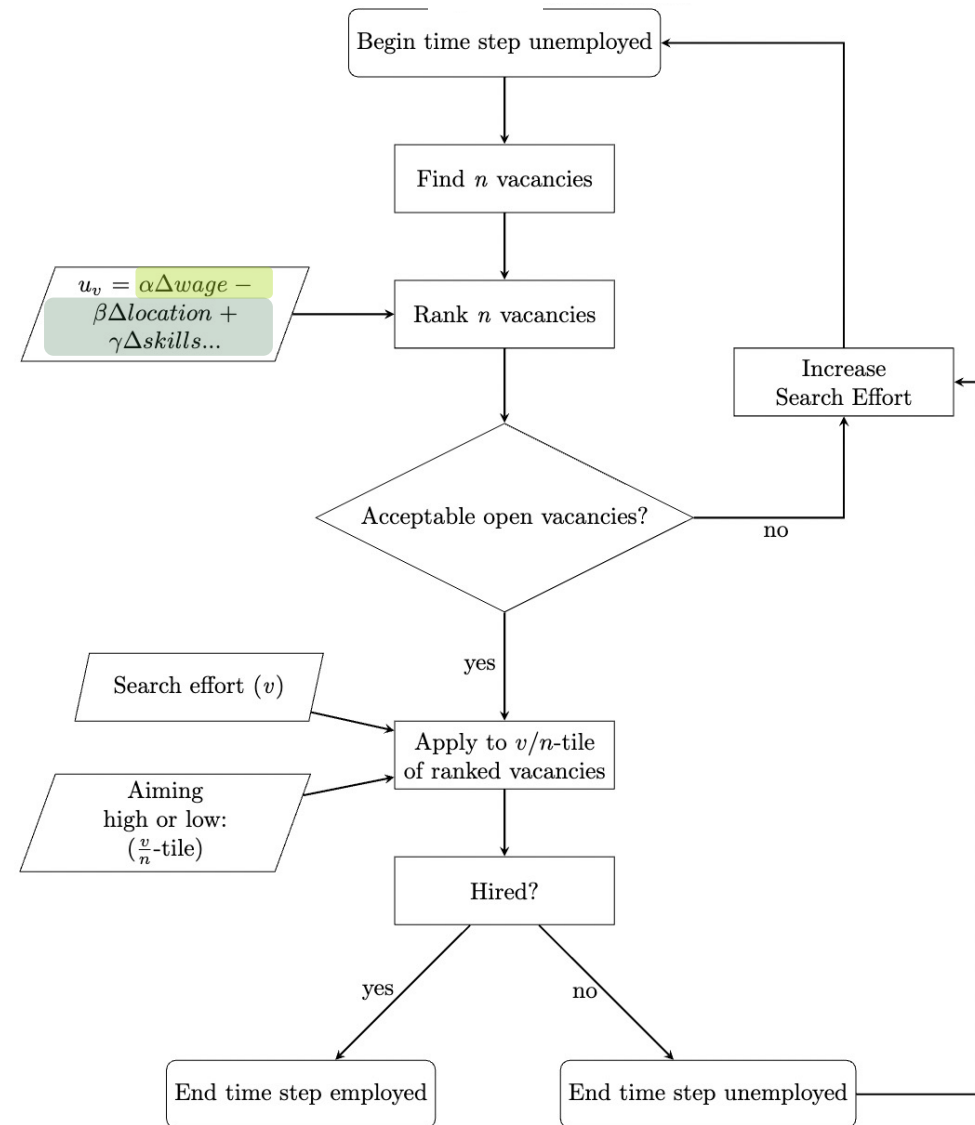


# [1] Preferences

Vacancies:

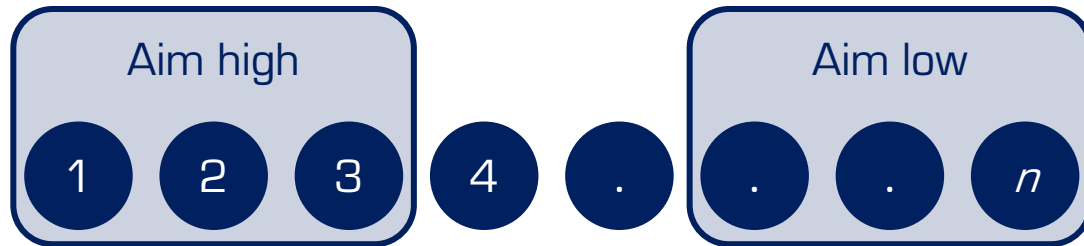


$$u_1 \gg u_n$$

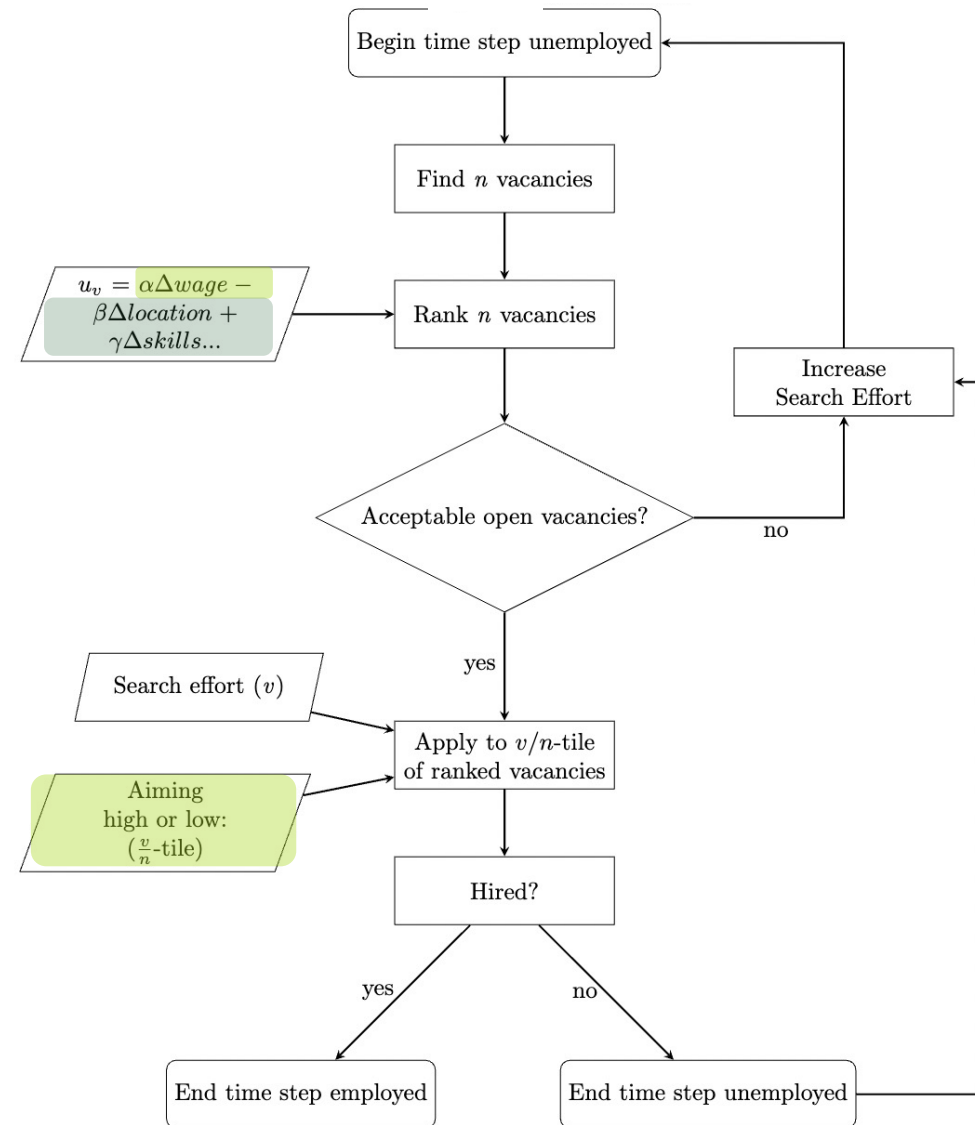


## (2) Aim high or low

Vacancies:

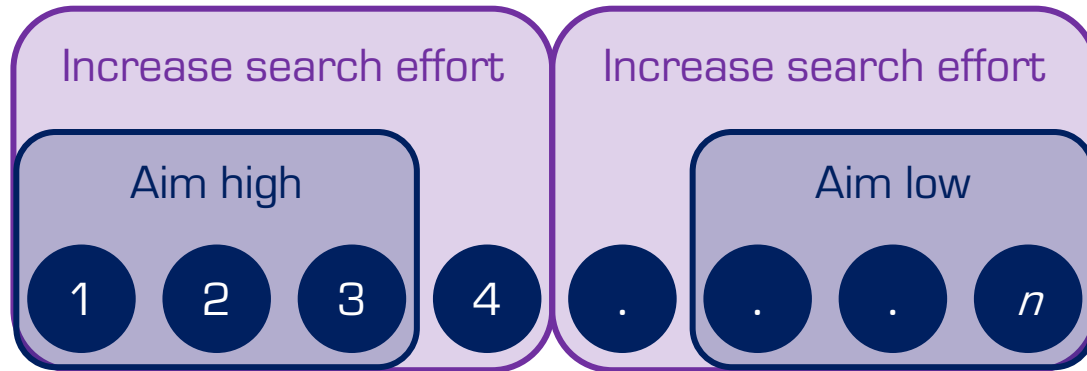


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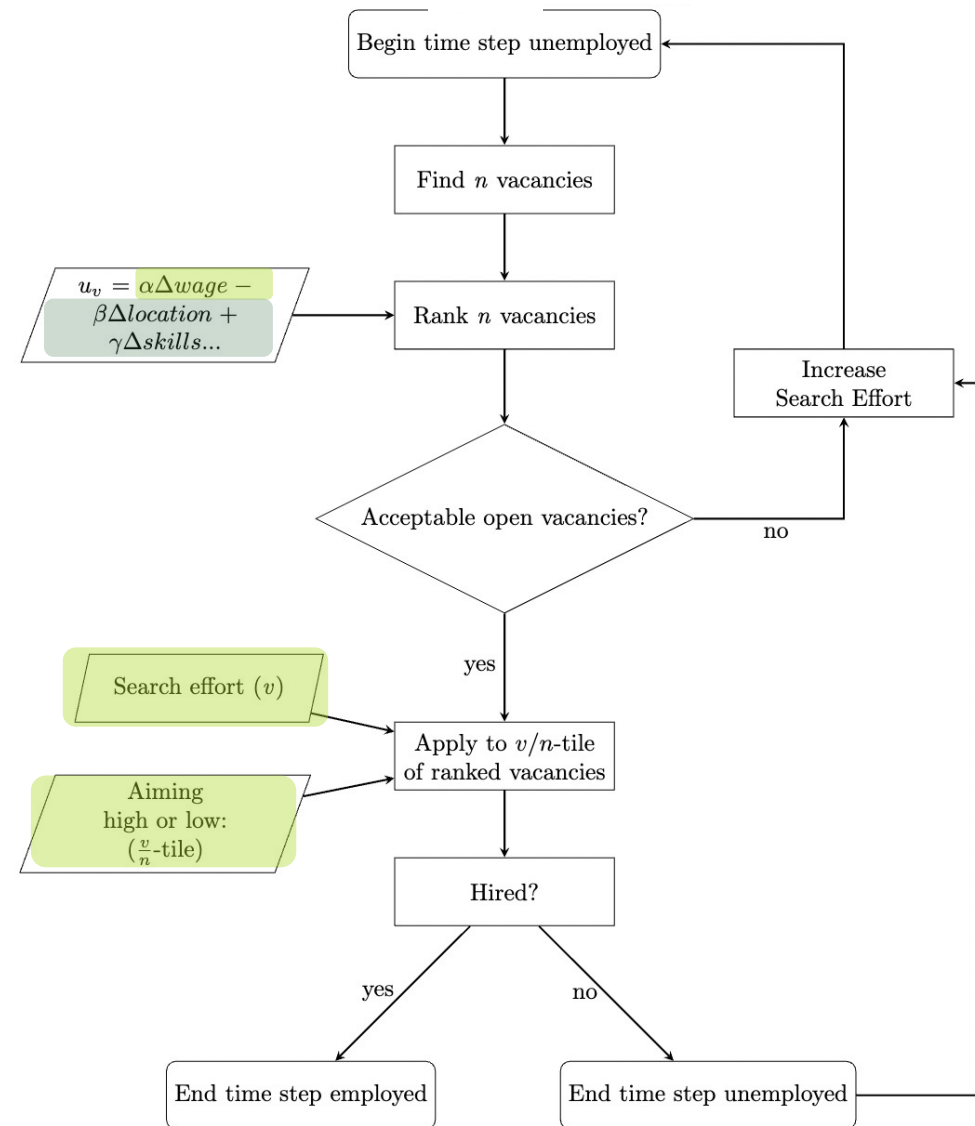


### (3) Determine search effort

**Vacancies:**

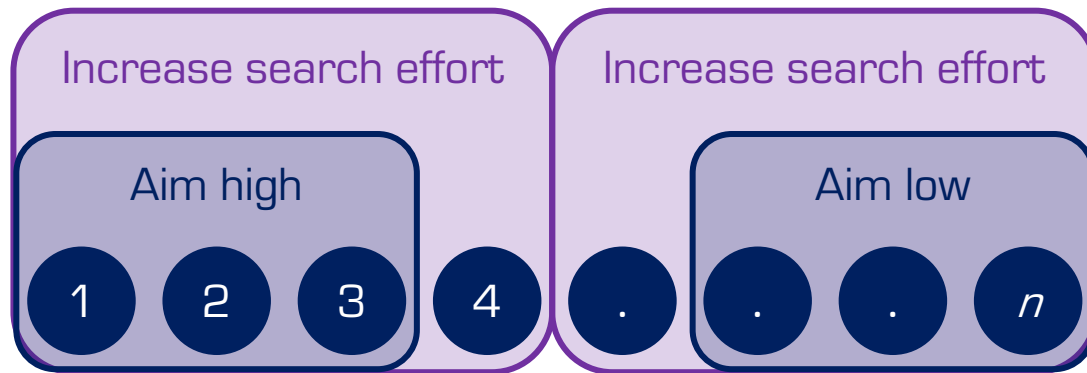


$$u_1 \gg u_n$$

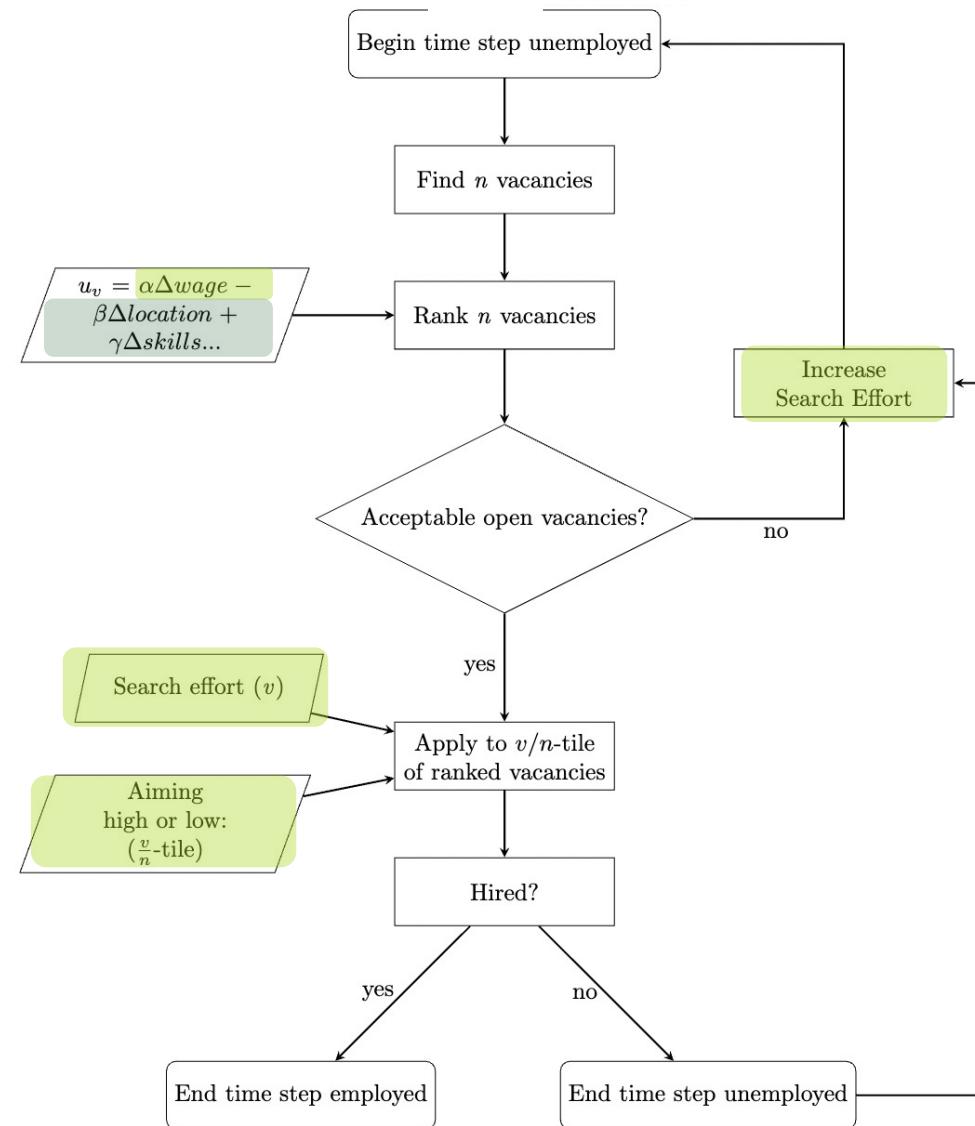


## (4) Non-linear search effort

**Vacancies:**



$$u_1 \gg u_n$$



# Application

*Incorporating behaviour using **sparse data***



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# Validation

Moving toward validation of the model's dynamics  
using three stylized facts about the US labour market:

**Beveridge Curve**

**Gender Wage Gap**

**Long-term  
Unemployment**

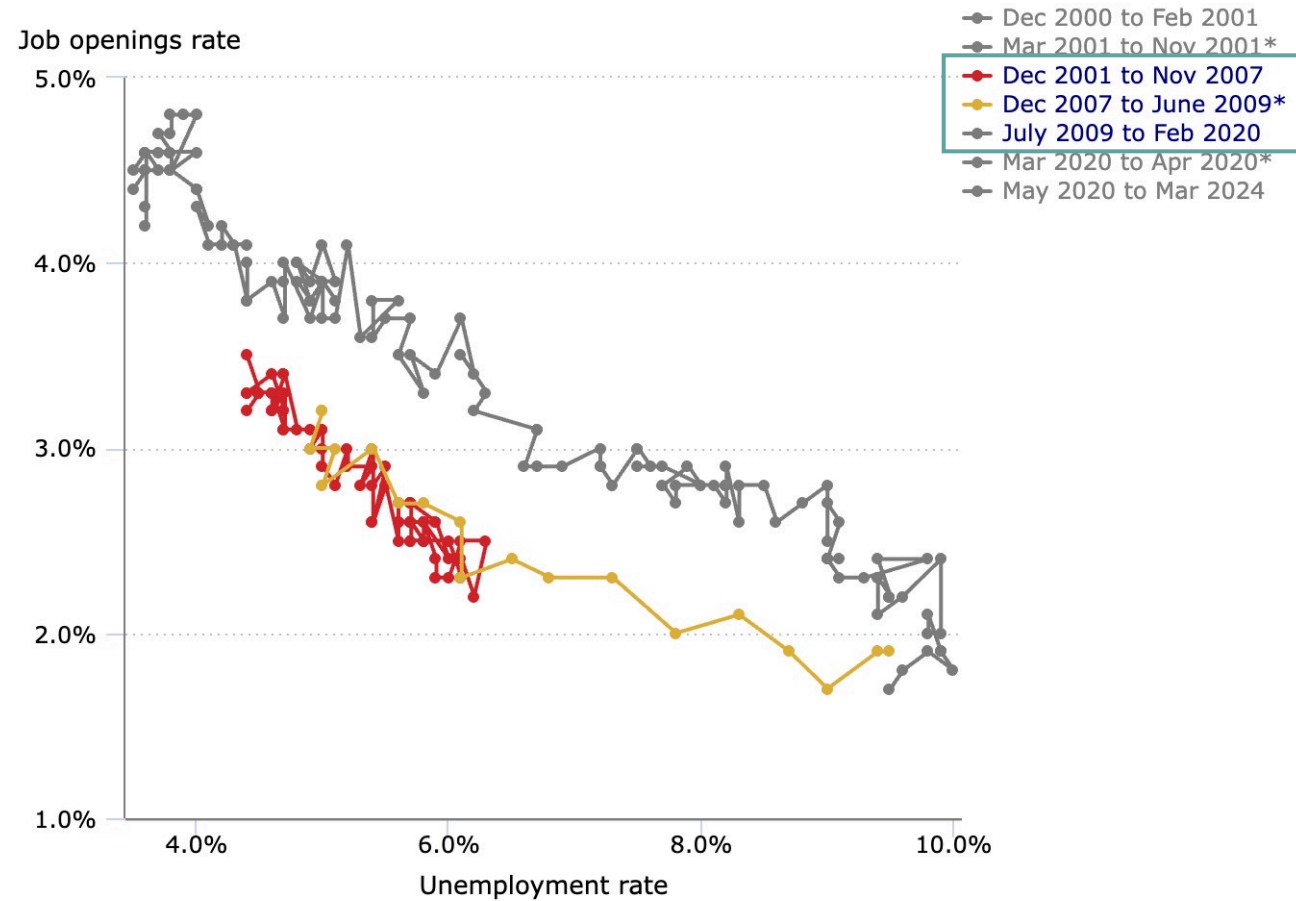
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# Beveridge Curve

# Replicating Beveridge Curve

## The Beveridge Curve (job openings rate vs. unemployment rate), seasonally adjusted

Click and drag within the chart to zoom in on time periods



Source: U.S. Bureau of Labor Statistics.

## Beveridge Curve

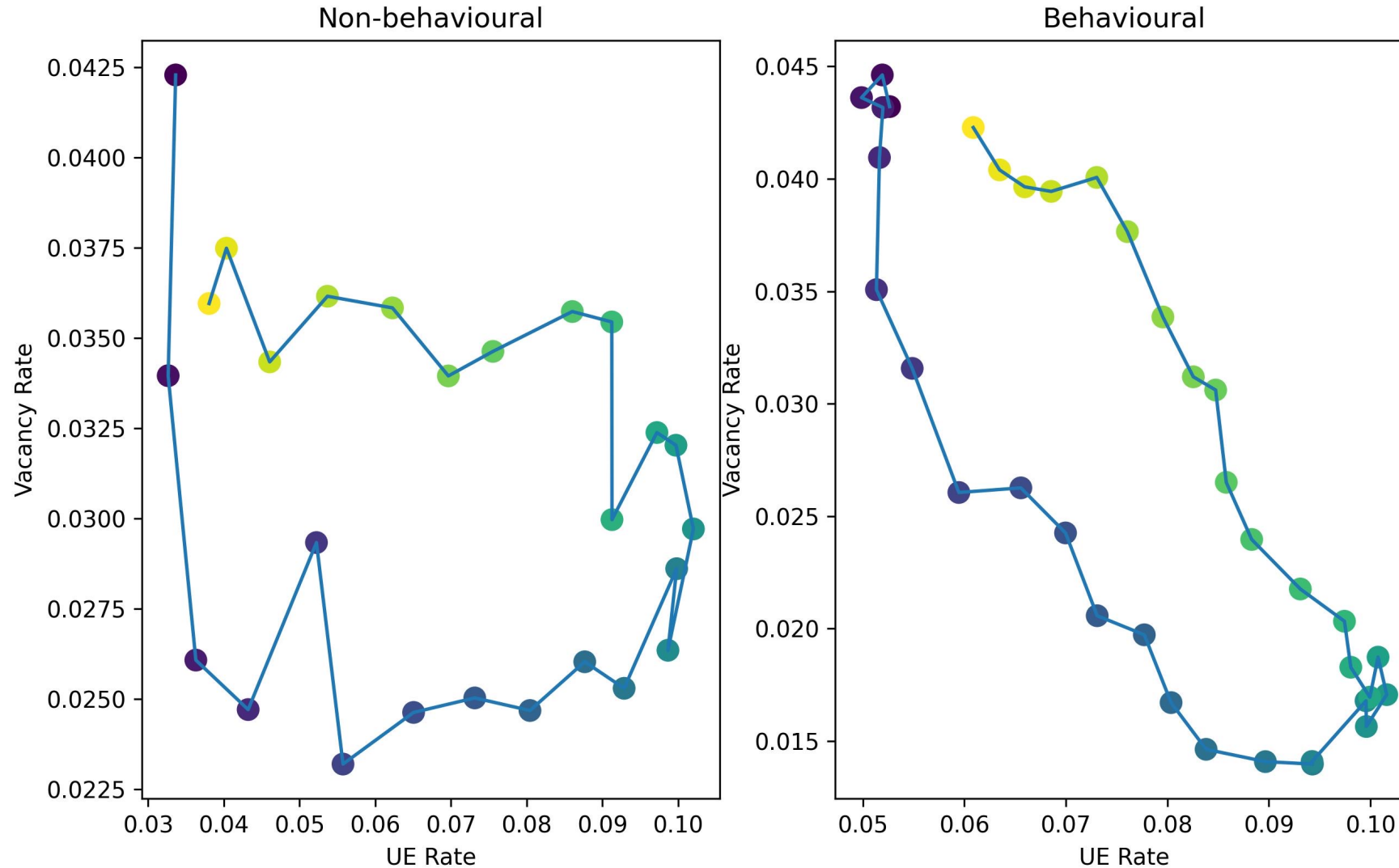
### Data inputs:

Job Openings Rate  
(s.a.)  
JOLTS

Unemployment Rate  
(s.a.)  
BLS

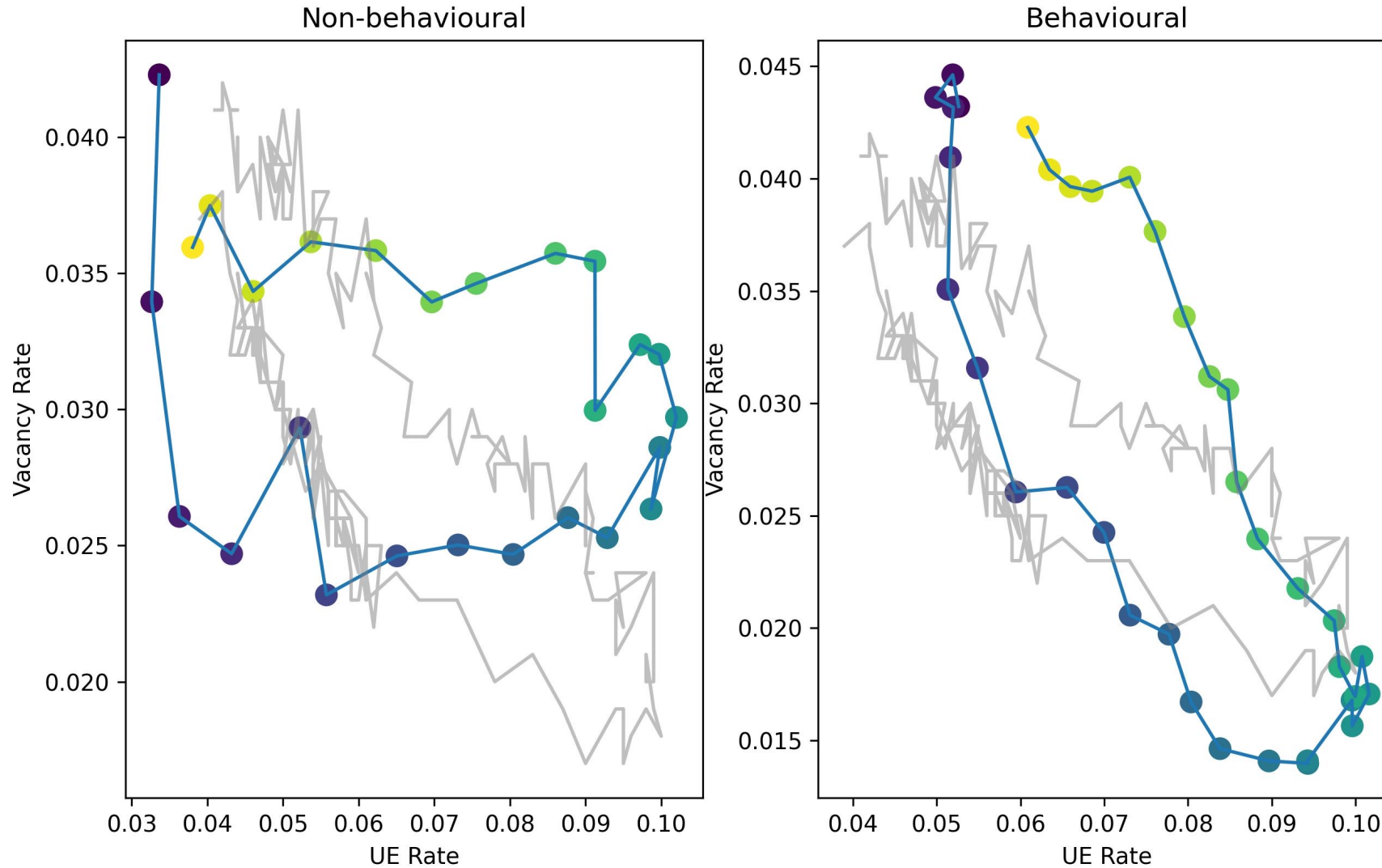
# Replicating Beveridge Curve

**USA Model Beveridge Curve**



# Replicating Beveridge Curve

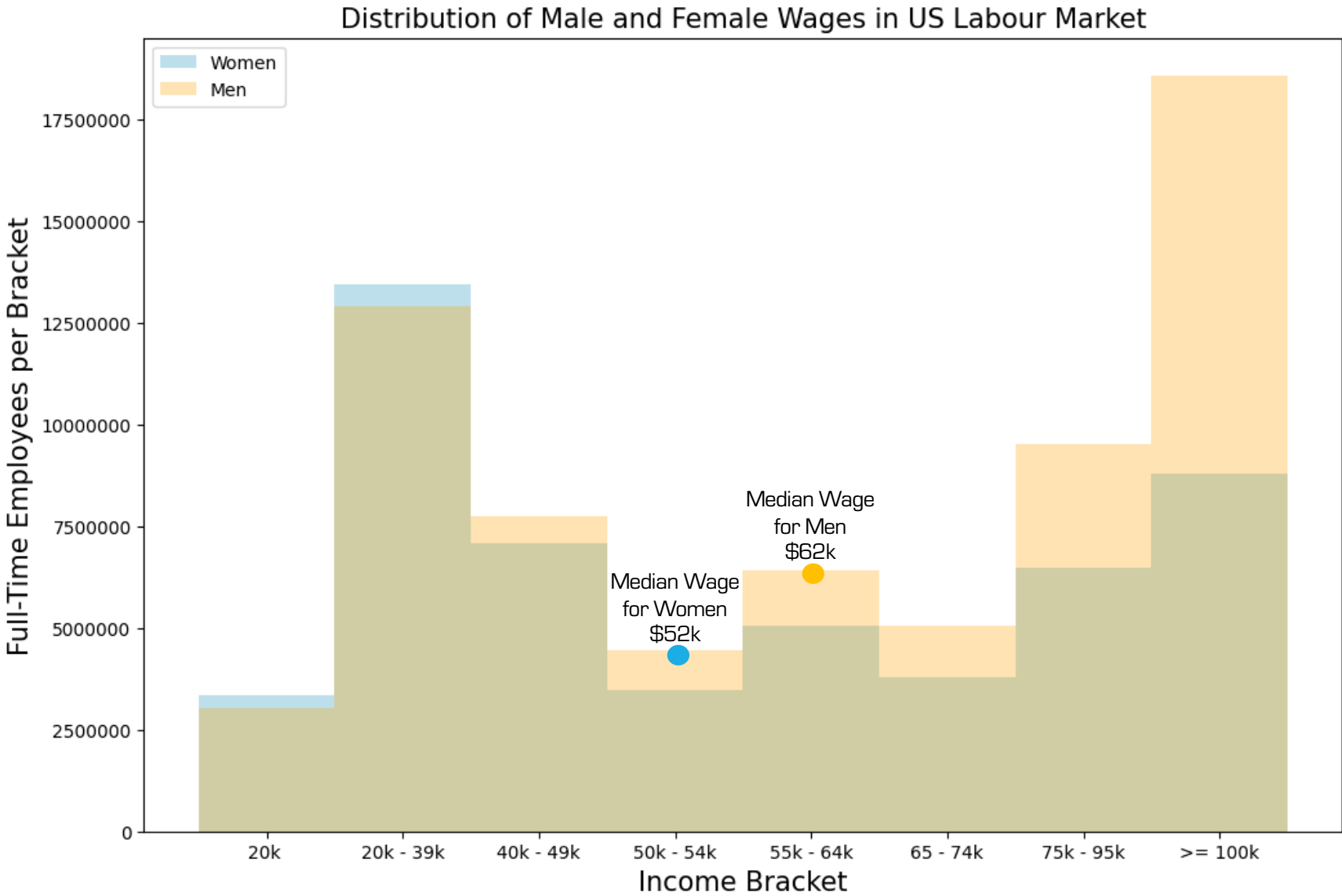
**USA Model Beveridge Curve**



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# Gender Wage Gap

# Nearly all national economies exhibit a gender wage gap

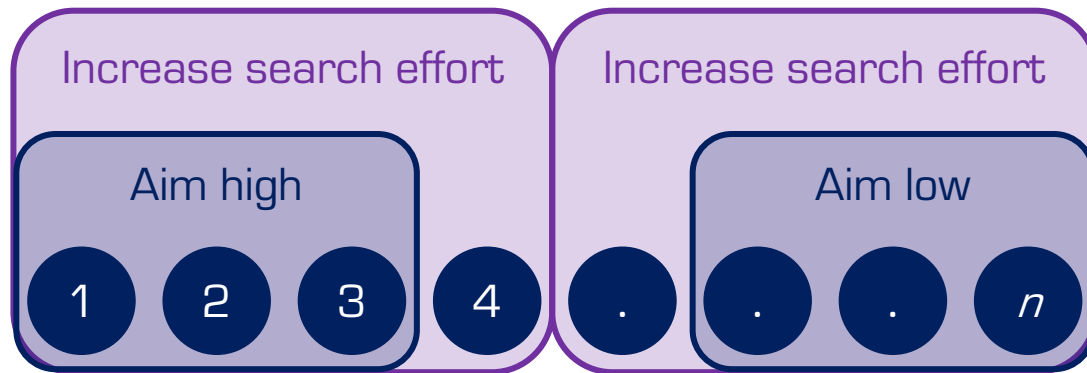


Data from the US Bureau of Labour Statistics

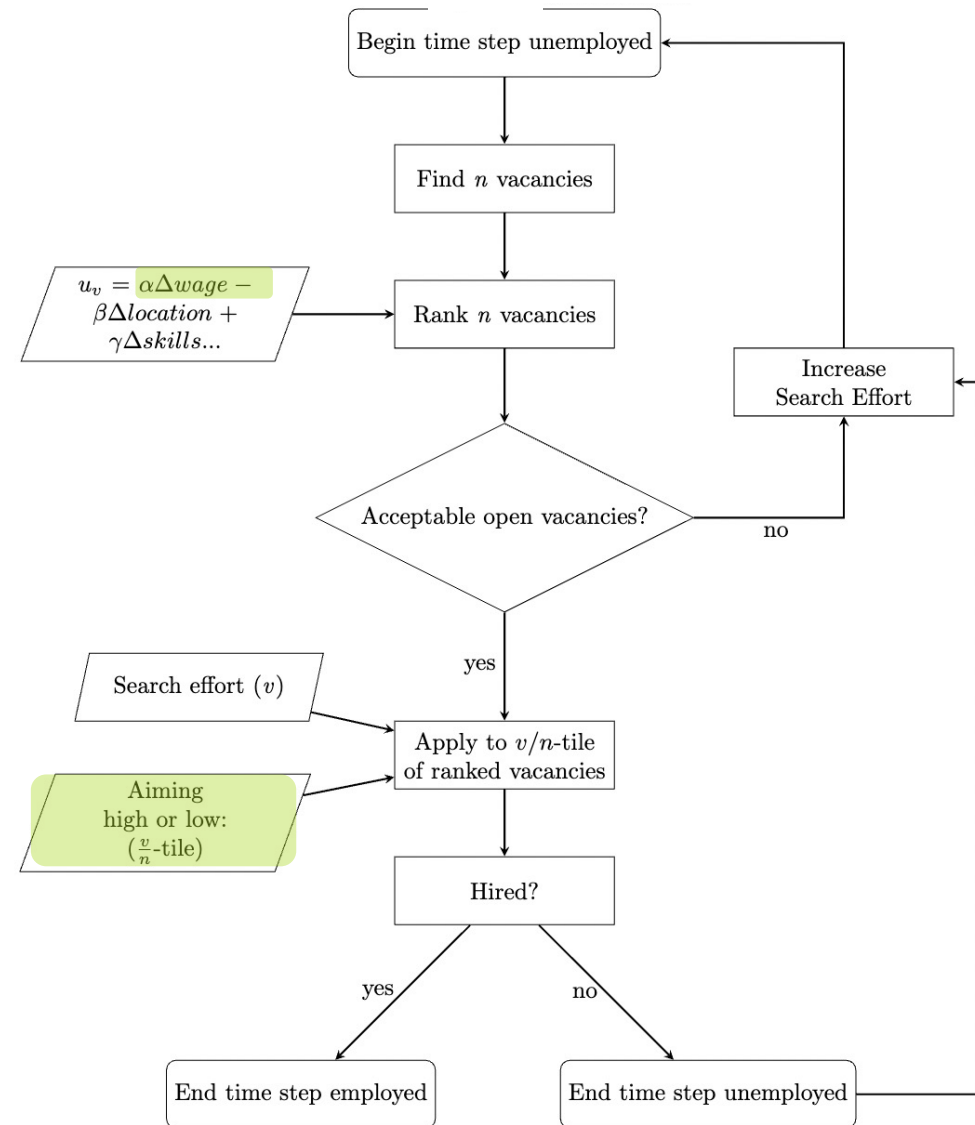


# Incorporating behavioural heterogeneity

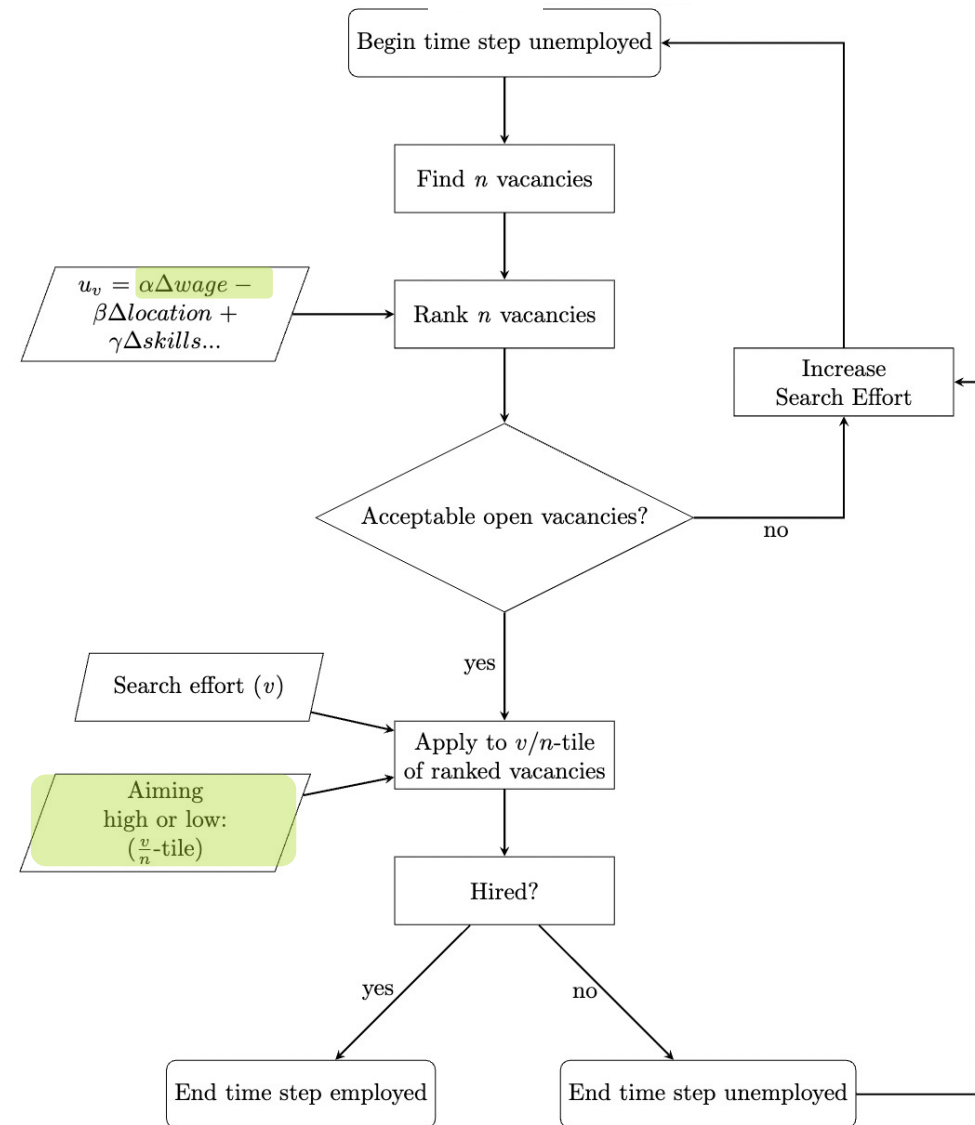
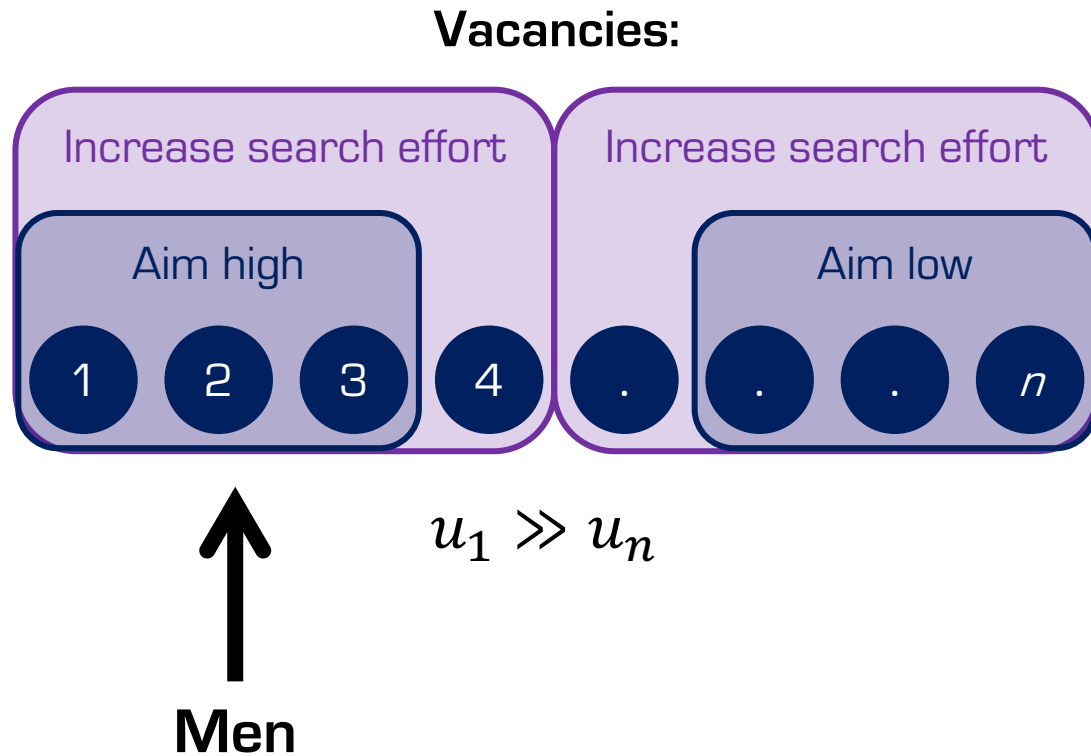
## Vacancies:



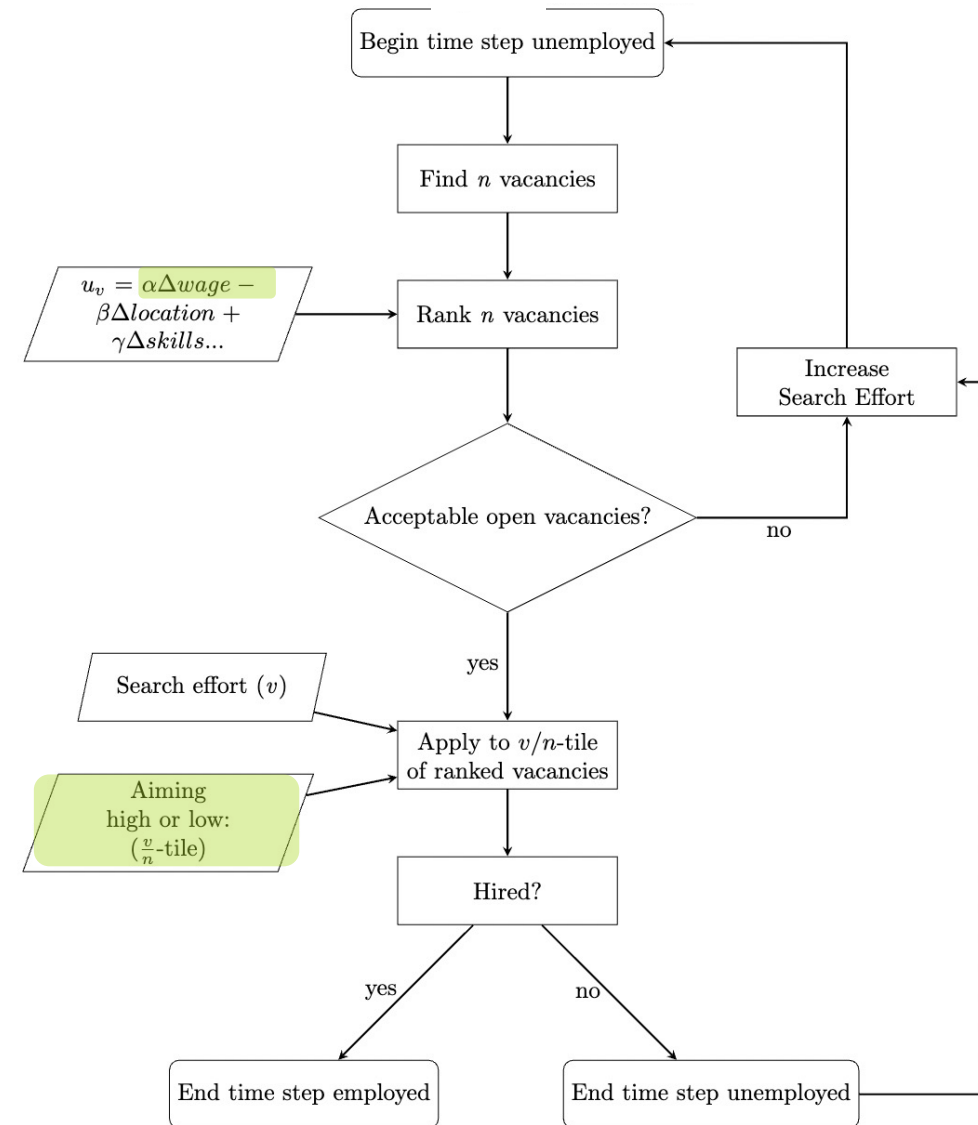
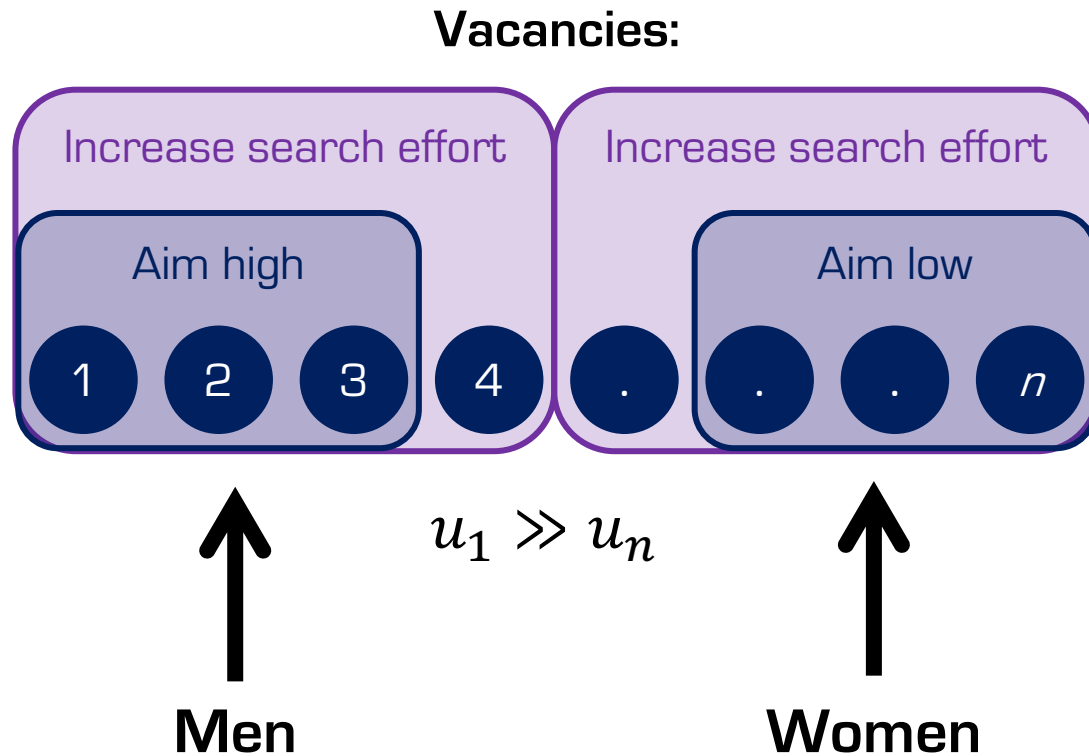
$$u_1 \gg u_n$$



# Incorporating behavioural heterogeneity



# Incorporating behavioural heterogeneity



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## Gender wage gap

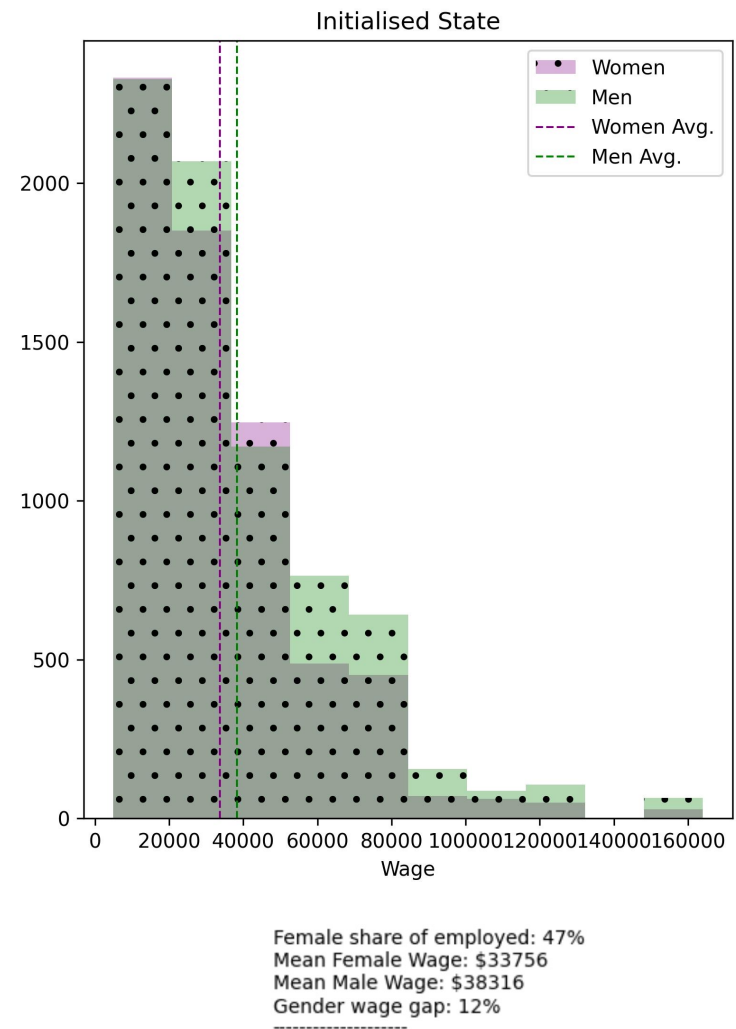
### Data inputs:

Gender share of  
occupations  
CPS – BLS

Median wage of  
occupations  
BLS

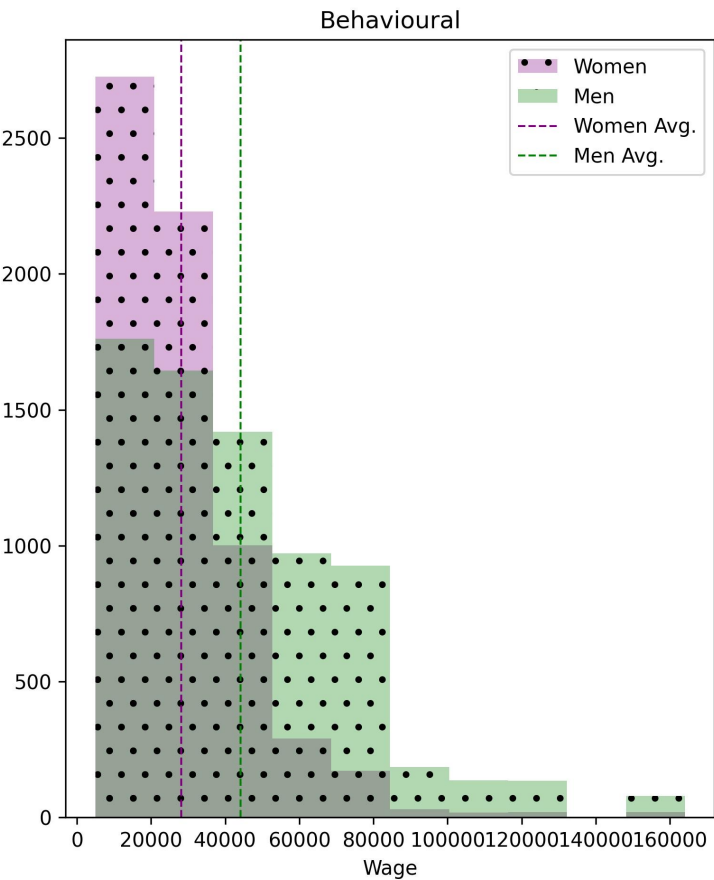
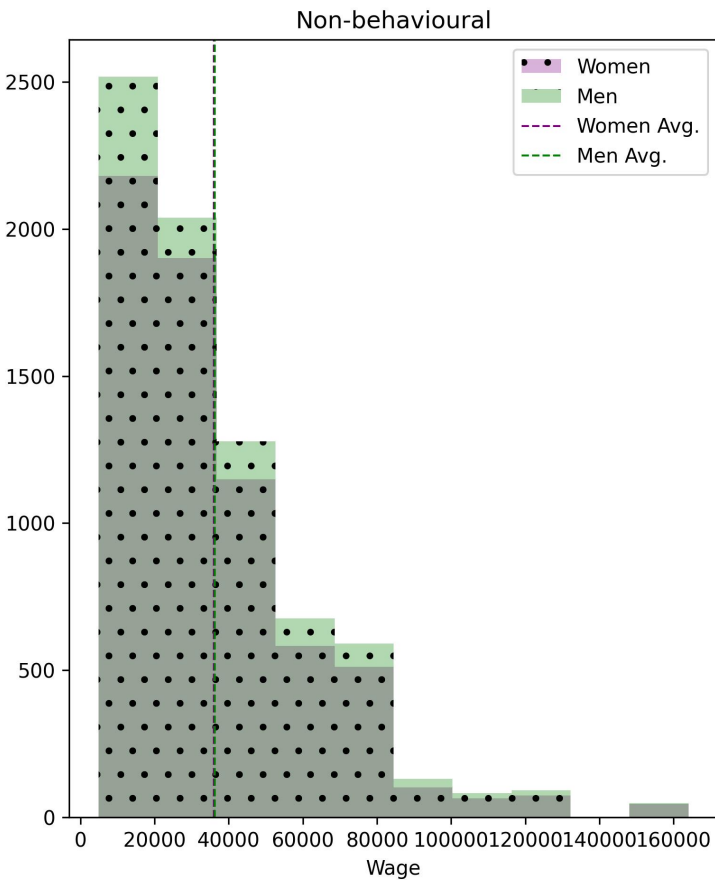
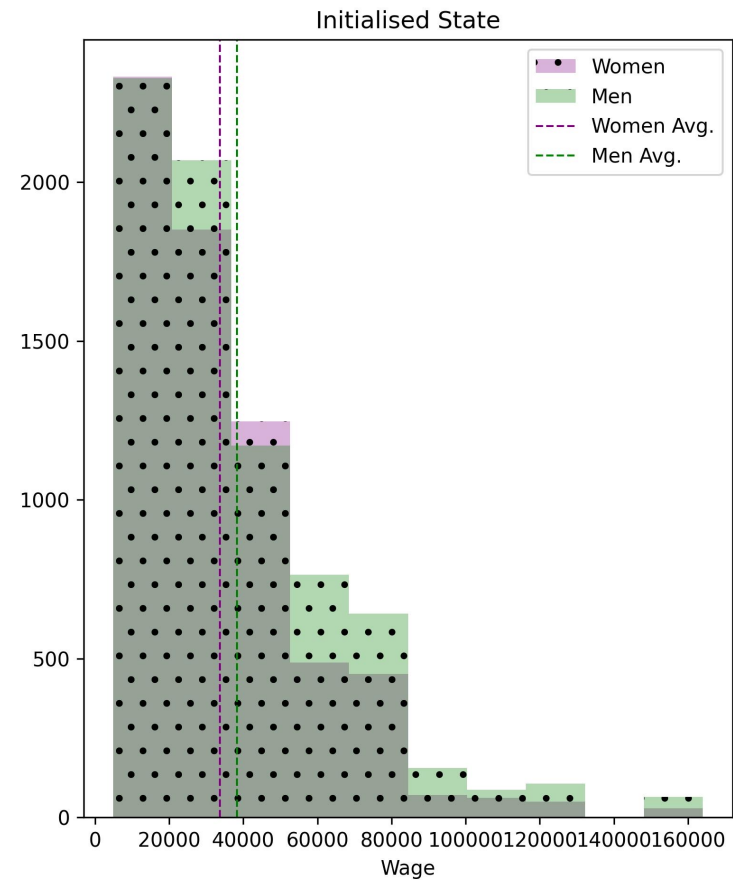
# Replicating gender wage gaps

Distribution of Male and Female Wages in Model



# Replicating gender wage gaps

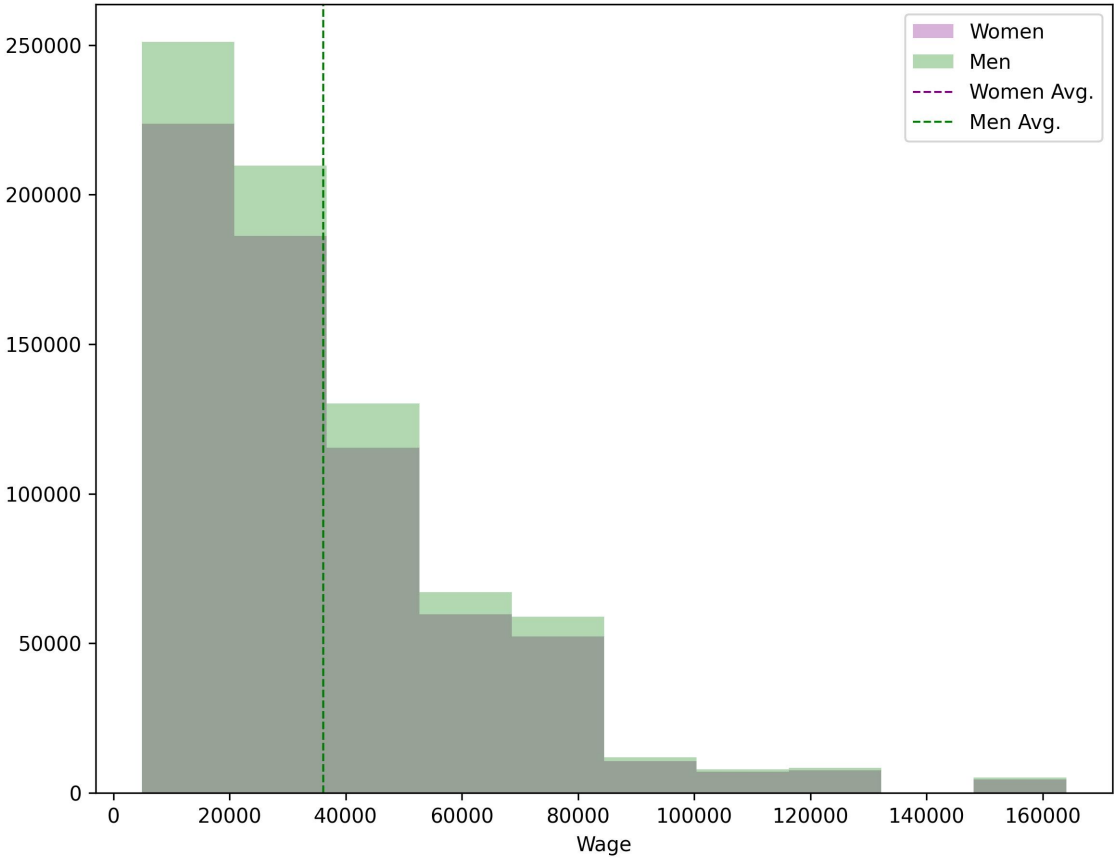
Distribution of Male and Female Wages in Model



# Replicating gender wage gaps

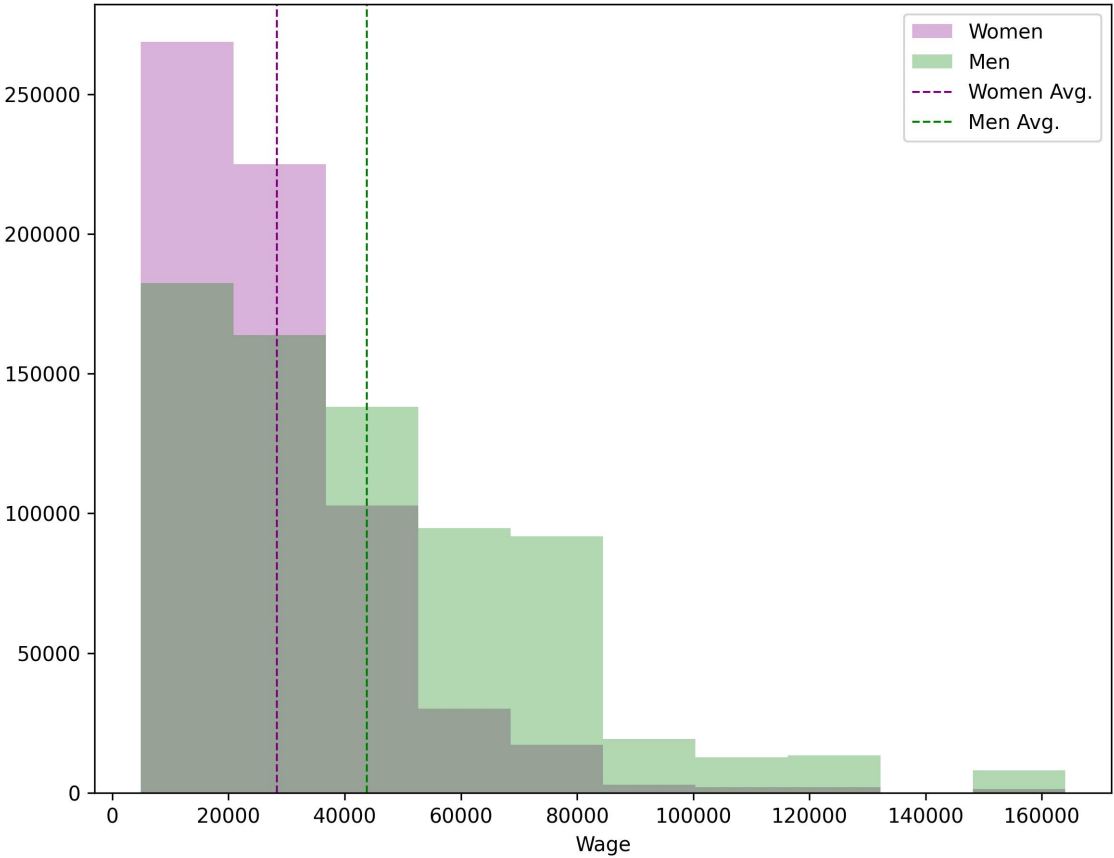
Distribution of Male and Female Wages in Model (100 simulations)

Non-behavioural



Mean Female wage: \$36139  
Mean Male wage: \$36141  
Gender wage gap: 0%

Behavioural



Mean Female wage: \$28327  
Mean Male wage: \$43773  
Gender wage gap: 35%

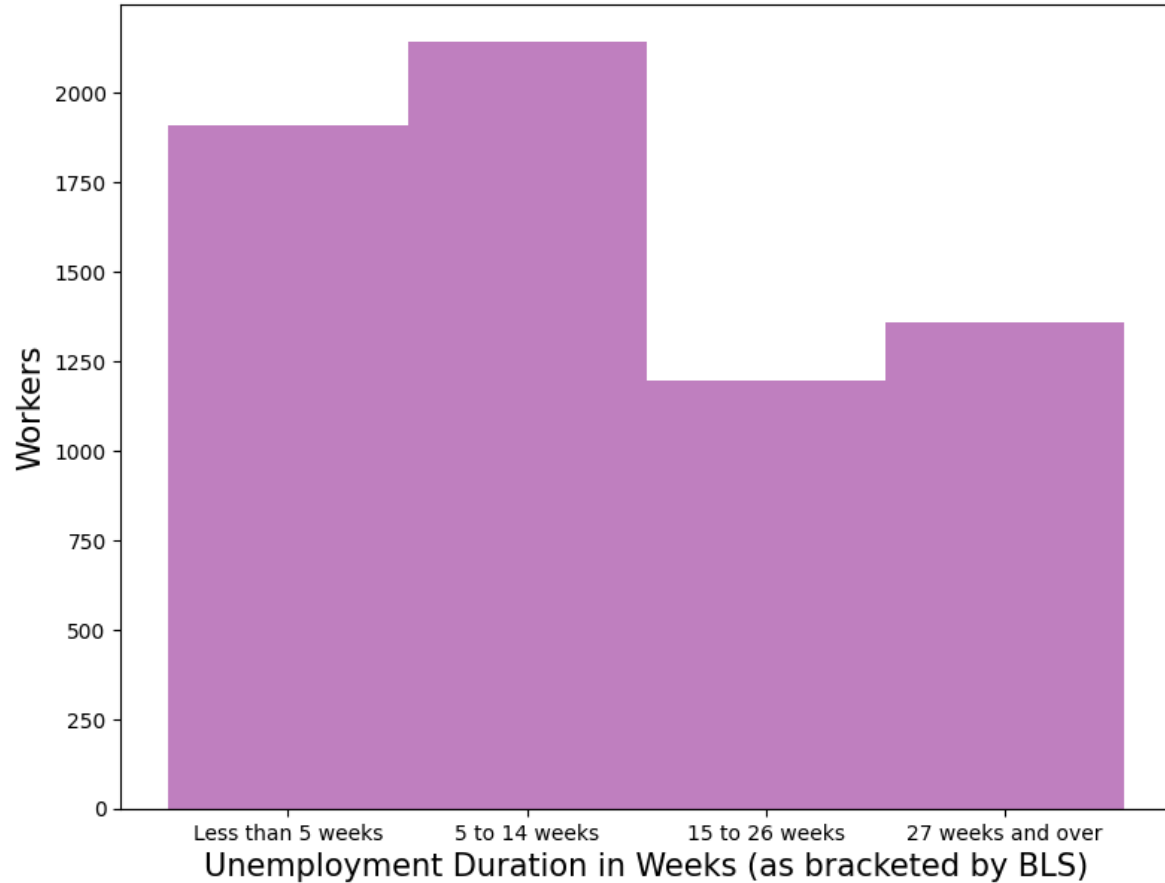
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# Long-term Unemployment



# Distribution of unemployment duration exhibits long tail

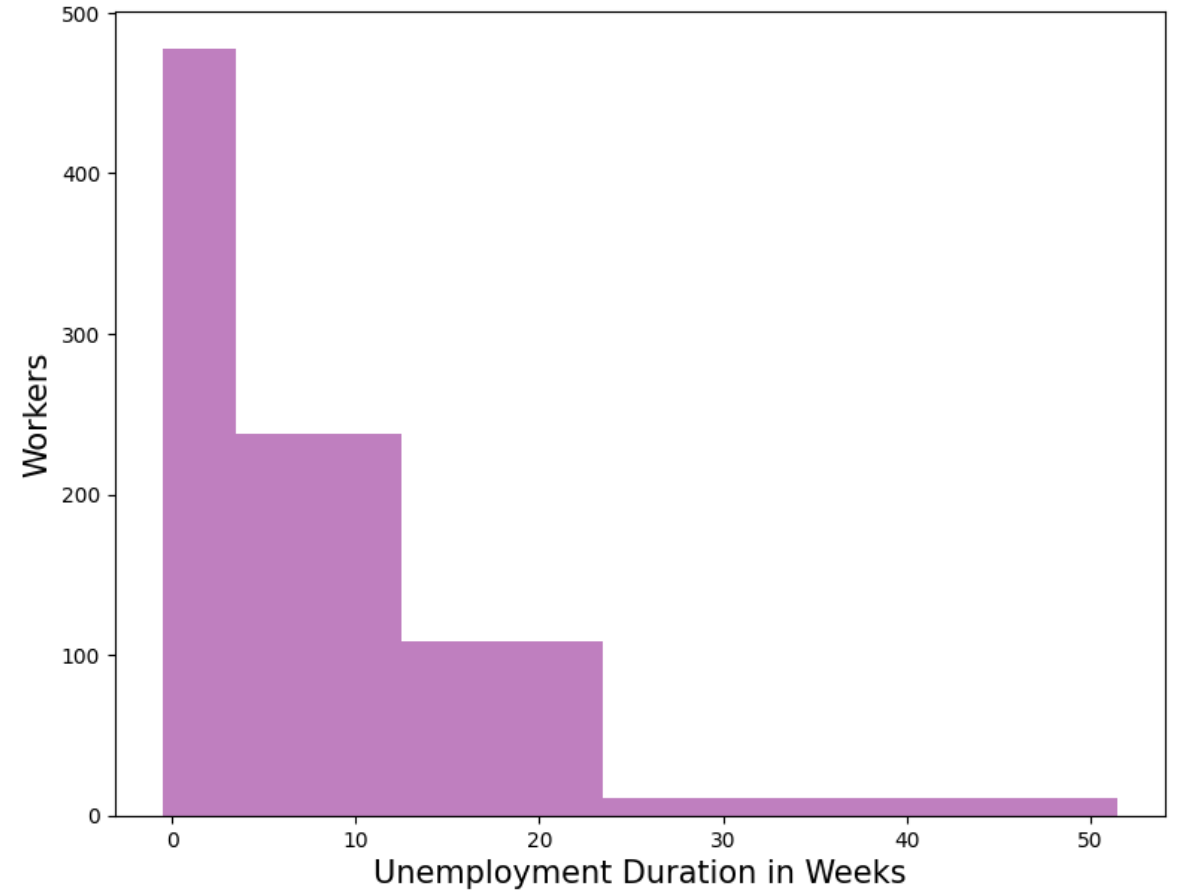
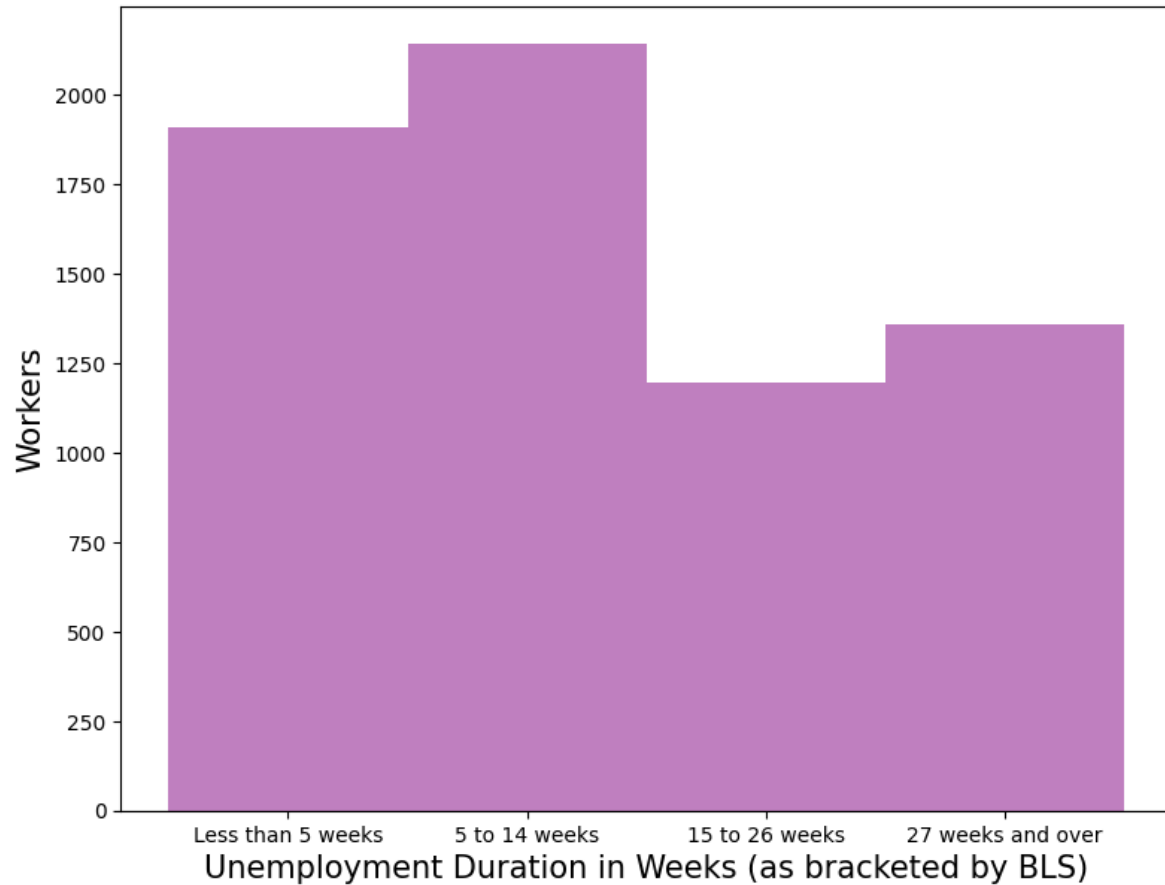
Distribution of Time Spent Unemployed in US Labour Market



*Data from the US Bureau of Labour Statistics*

# Distribution of unemployment duration exhibits long tail

Distribution of Time Spent Unemployed in US Labour Market



*Data from the US Bureau of Labour Statistics*

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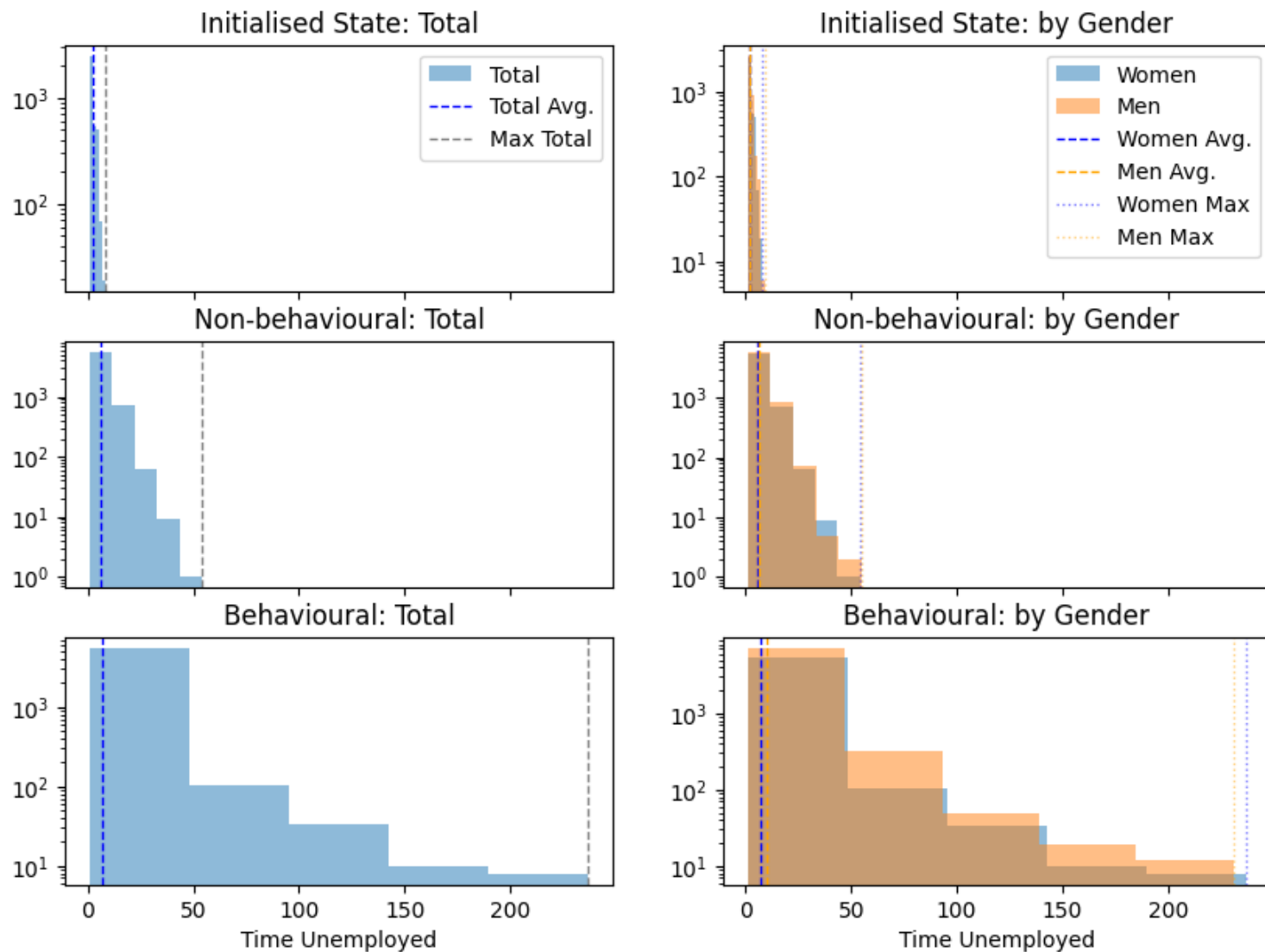
**Long-term  
unemployment**

**Data inputs:**

**TBD – suggestions welcome!**

# Replicating shape of LTUER

Distribution of Time Unemployed in Model



# Replicating shape of LTUER

Distribution of Time Unemployed in Model (100 simulations)

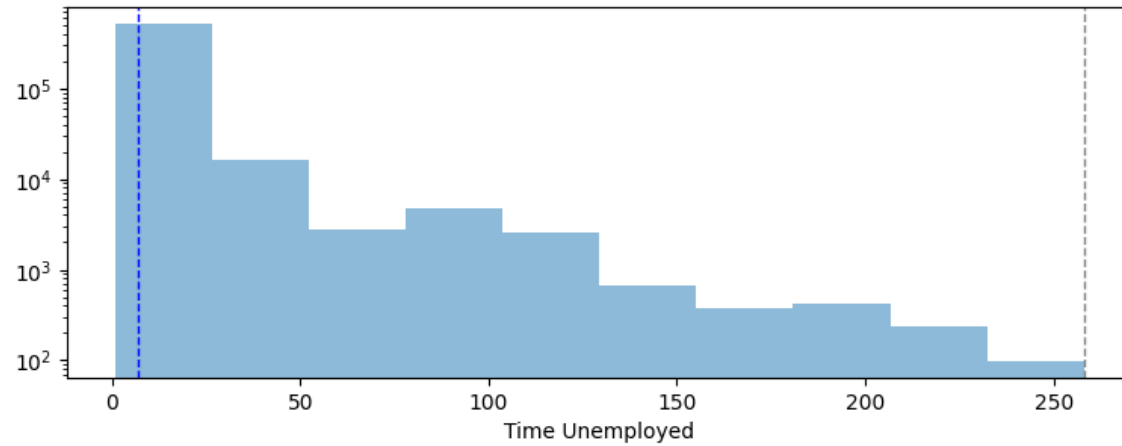
Non-behavioural: Total



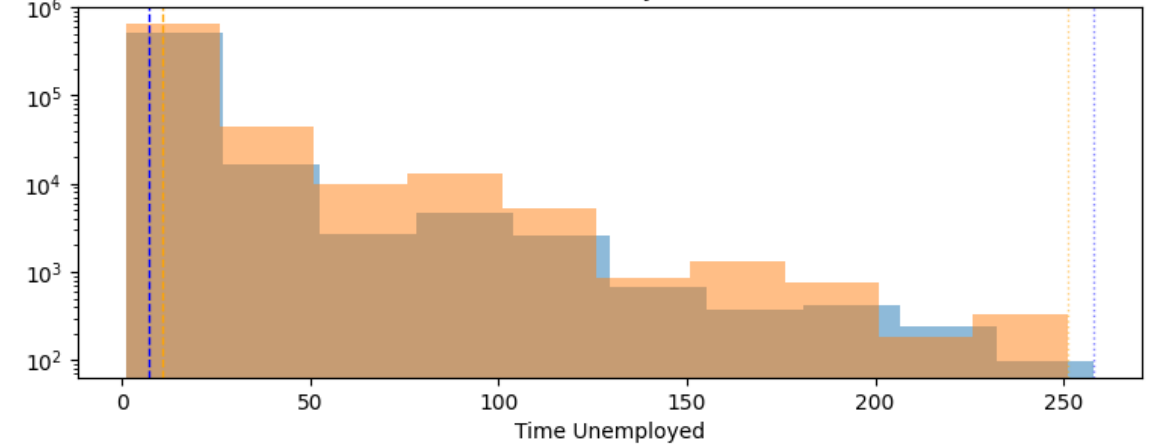
Non-behavioural: by Gender



Behavioural: Total



Behavioural: by Gender



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# Next steps

- **Thoroughly validate against stylized facts presented:**
  - *Calibrate behavioural heuristics:* Incorporate data from additional sources to arrive at key distributions of behavioural parameter values (JOLTS, Time Use Survey, World Value Survey)
- **Replicate co-movements of GDP and unemployment rate** using historical data:
  - More realistic calibration to business cycles
  - Facilitate out-of-sample testing
- **Application:** Examine the interaction of a projected net-zero transformation on existing labour market inequalities.



Thank you!

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# Aspiration

*Incorporating behaviour using **rich micro data***



# Motivation and Research Question

## Motivation

As in the first study, this work will aim to incorporate behaviour into our existing understanding of occupational transitions.

The value added of this latter effort will be:

- The use of rich micro data to understand what determines job to job transitions (ie. skills, wage, gender, location)
- To replace any parameters inferred from aggregate distributions and disparate surveys in the former model with observed variation in the micro-level data
- Ideally, depending on the quality of the data, we will move to deduce the role of identity in job transitions
- This latter point requires significantly more theoretical work at this point

## Research Question(s)

**What determines occupational transitions?**

**Can we deduce information about the role of identity in such transitions?**

# Data Progress

We are closing in on two possible options for our micro-level data:



## Germany

### **SIAB: Sample of Integrated Labour Market Biographies**

2% random sample of all individuals  
in the German Labour Market (whose  
employment is subject to social  
security)



## Brazil

### **Relação Anual de Informações Sociais (RAIS)**

- 67% of total labour force
- 570 occupations and 16 regions
- Contains all active employer- employee contracts  
in each year in the formal labour force in Brazil,  
which accounts for about 67% of the total  
labour force in Brazil.