

# The Effect of Expected Income on Individual Migration Decisions (Kennan & Walker, 2011)

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- ▶ Most current models “describe patterns in the data”
- ▶ Can we build a model that allows for return migration *and* can be estimated?
- ▶ The main contribution is methodological, no insights about how migration decisions are made.
- ▶ Smart discretization allows for tractable likelihood function...
  - ▶ ... but they still have to assume that some distributions have supports of 3 points, otherwise it blows up.

# Features of the model

- ▶ Many locations: allows for non-trivial reverse migration
- ▶ Migration is search. Workers move to find out the wage in a location (only choice).
- ▶ Partial equilibrium: wages don't adjust to flows

- ▶ Wage  $w_{ij}(a) = \mu_j + v_{ij} + G(X_i, a, t) + \eta_i + \varepsilon_{ij}(a)$ 
  - ▶  $\mu_j$  is average wage in  $j$ , estimated outside from US Census for each age
  - ▶  $v_{ij}$  is the “match quality” that is learned once and forever, does not change. Prospects at unvisited locations are random.
  - ▶  $G(a, x, t)$  is time effect and the effects of observables
  - ▶  $\eta_i$  is individual fixed effect
  - ▶  $\varepsilon_{ij}(a)$  should be simply  $\varepsilon_{ijt}$  (they just emphasize it's a different draw every year).

- ▶  $u_h(x, j) = \alpha_0 w(\ell^0, \omega) + \sum_{k=1}^K \alpha_k Y_k(\ell^0) + \alpha^H \mathbb{1}(\ell^0 = h) + \xi(\ell^0, \omega) - \Delta_\tau(x, j)$ 
  - ▶  $\ell = (\ell^0, \ell^1, \ell^2, \ell^3, \dots, \ell^{M-1})$  reflects all previously visited locations. History is limited to  $M$  periods
  - ▶  $x = (\ell, \omega, a)$  is state vector: visited locations, wage and utility info obtained there, age.
  - ▶  $w(\ell^0, \omega)$  is wage in the current location, given the wage and utility information  $\omega$ .
  - ▶  $\sum_{k=1}^K \alpha_k Y_k$  is the “amenity values” at current location
  - ▶  $\alpha^H$  is the premium for being in “home” location
  - ▶  $\xi(\ell^0)$  is (individual?) location match
  - ▶  $\Delta_\tau(x, j)$  is moving cost

- ▶  $\Delta_\tau(x, j) = (\gamma_{0\tau} + \gamma_1 D(\ell^0, j) - \gamma_2 \mathbb{1}_{j \in \mathbb{A}(\ell^0)} - \gamma_3 \mathbb{1}_{j = \ell^1} + \gamma_4 a - \gamma_5 n_j) \mathbb{1}_{j \neq \ell^0}$ 
  - ▶  $\gamma_{0\tau}$  is heterogenous, indexed by  $\tau$
  - ▶  $\gamma_1$  for distance,  $\gamma_2$  for adjacency,  $\gamma_3$  for returning to prev. location
  - ▶  $\gamma_4$  for age
  - ▶  $\gamma_5$  for the population size in destination

- ▶ Workers choose destination  $j$

$$V(x, \zeta) = \max_j (u(x, j) + \zeta_j + \beta \sum_{x'} p(x'|x, j) E_{\zeta} V(x', \zeta))$$

- ▶ If  $\zeta_j \sim$  Type I EV, then

$$p(x, j) = \exp(\bar{\gamma} + u(x, j) + \beta \sum_{x'} p(x'|x, j) E_{\zeta} V(x', \zeta) - E_{\zeta} V(x, \zeta))$$

# Transition probabilities

Workers can stay, return, or move to new location:

$$p(x'|x, j) = \begin{cases} 1, & \text{if } j = \ell^0, \tilde{x}' = \tilde{x}, a' = a + 1 \\ 1, & \text{if } j = \ell^1, \tilde{x}' = (\ell^1, \ell^0, x_v^1, x_v^0, x_\xi^1, x_\xi^0), a' = a + 1 \\ \frac{1}{n^2}, & \text{if } j \notin \{\ell^0, \ell^1\}, \tilde{x}' = (j, \ell^0, s_v, x_v^0, s_\xi, x_\xi^0) \\ & (1, 1) \leq (s_v, s_\xi) \leq (n_v, n_\xi), a' = a + 1 \\ 0, & \text{otherwise.} \end{cases}$$

Why  $\frac{1}{n^2}$ ? Because the support for  $v_{ij}$  and for  $\xi$  for have  $n_v = n_\xi = n$  points.



- ▶ NLSY79, 432 people, aged 20+, 4274 person-years
- ▶ 124 Interstate moves (only 2.9% of total person years)
- ▶ Locations are States, Home state is state of residence at age 20
- ▶  $\mu_i$  from 1990 US Census (self-reported)

- ▶ They have to assume only one past wage is remembered, otherwise state space is too big
  - ▶ A very disappointing assumption, what kind of learning is it then?
- ▶ Distribution of match effect  $v_{ij}$  is discretized, assumed symmetric and mean-zero
- ▶  $n = 3$ , so  $v_{ij} \in \{-\tau_v; 0, \tau_v\}$ .  $\tau_v$  is estimated
- ▶ “*The location match component of preferences  $[\xi]$  is handled in similar way*”
- ▶  $n = 7$  for the person fixed effect  $\eta_i$
- ▶  $n = 4$  for  $\sigma_\varepsilon(i)$ , individual level variance for the Worker  $i$  wage,  $\varepsilon(i) \sim N(0, \sigma_\varepsilon(i))$

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- ▶  $n_\eta = 7$  for the person fixed effect  $\eta_i$
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# Estimation: likelihood

- ▶ Everything is discretized at quartiles, so all possible realizations of  $\omega^i = \{\omega_\xi^i, \omega_\eta^i, \omega_\varepsilon^i, \omega_v^i(1), \dots, \omega_v^i(N_i)\}$  are equally likely.
- ▶ Ind. likelihood for wage is (recall  $\varepsilon(i) \sim N(0, \sigma_\varepsilon(i))$ ):

$$\psi_{it}(\omega_i, \theta_\tau) = \frac{1}{\sigma_\varepsilon(\omega_\varepsilon^i)} \phi \left( \frac{w_{it} - \mu_\ell^0(i, t) - G(X_i, a_{it}, \theta_\tau) - v(\omega_v^i(\kappa_{it}^0)) - \eta(\omega_\eta^i)}{\sigma_\varepsilon(\omega_\varepsilon^i)} \right)$$

- ▶ Prob of indiv. history is

$$L_i(\theta_\tau) \frac{1}{n_\eta n_\varepsilon n_\xi (n_v)^{N_i}} \times \sum_{\omega^i \in \Omega(N_i)} \left( \prod_{t=1}^{T_i} \psi_{it}(\omega^i, \theta_\tau) \lambda_{it}(\omega^i, \theta_\tau) \right)$$

$$\lambda_{it}(\omega^i, \theta_\tau) = \rho_{h(i)}(\ell(i, t), \omega_v^i(\kappa_{it}^0), \omega_v^i(\kappa_{it}^1), \omega_\xi^i(\kappa_{it}^0), \omega_\xi^i(\kappa_{it}^1), a_{it}, \ell^0(i, t+1), \theta_\tau)$$

# Results

## Utility and cost

Disutility of moving ( $\gamma_0$ )	4.790	0.565	4.514	0.523	4.864	0.601	4.851	0.604
Distance ( $\gamma_1$ ) (1000 miles)	0.265	0.182	0.280	0.178	0.311	0.187	0.270	0.184
Adjacent location ( $\gamma_2$ )	0.808	0.214	0.787	0.211	0.773	0.220	0.804	0.216
Home premium ( $\alpha^H$ )	0.331	0.041	0.267	0.031	0.332	0.047	0.337	0.045
Previous location ( $\gamma_3$ )	2.757	0.356	2.544	0.300	3.082	0.449	2.818	0.416
Age ( $\gamma_4$ )	0.055	0.020	0.062	0.019	0.060	0.020	0.054	0.020
Population ( $\gamma_5$ ) (millions)	0.653	0.179	0.653	0.178	0.635	0.177	0.652	0.179
Stayer probability	0.510	0.078	0.520	0.079	0.495	0.087	0.508	0.082
Cooling ( $\alpha_1$ ) (1000 degree-days)	0.055	0.019	0.036	0.019	0.048	0.018	0.056	0.019
Income ( $\alpha_0$ )	0.312	0.100	–	–	–	–	0.297	0.116
Location match preference ( $\tau_\xi$ )	–	–	–	–	0.168	0.049	0.070	0.099

## Wages

Wage intercept	–5.165	0.244	–5.175	0.246	–5.175	0.246	–5.168	0.244
Time trend	–0.035	0.008	–0.033	0.008	–0.033	0.008	–0.035	0.008
Age effect (linear)	7.865	0.354	7.876	0.356	7.877	0.356	7.870	0.355
Age effect (quadratic)	–2.364	0.129	–2.381	0.130	–2.381	0.130	–2.367	0.129
Ability (AFQT)	0.014	0.065	0.015	0.066	0.014	0.066	0.014	0.065
Interaction (Age, AFQT)	0.147	0.040	0.152	0.040	0.152	0.040	0.147	0.040
Transient s.d. 1	0.217	0.007	0.218	0.007	0.218	0.007	0.217	0.007
Transient s.d. 2	0.375	0.015	0.375	0.015	0.375	0.015	0.375	0.015
Transient s.d. 3	0.546	0.017	0.547	0.017	0.547	0.017	0.546	0.017
Transient s.d. 4	1.306	0.028	1.307	0.028	1.307	0.028	1.306	0.028
Fixed effect 1	0.113	0.035	0.112	0.035	0.112	0.035	0.113	0.035
Fixed effect 2	0.298	0.035	0.296	0.035	0.296	0.035	0.298	0.035
Fixed effect 3	0.936	0.017	0.934	0.017	0.934	0.017	0.936	0.017
Wage match ( $\tau_v$ )	0.384	0.017	0.387	0.018	0.387	0.018	0.384	0.018

## Log likelihood

Exclude income: $\chi^2(1)$	–4,214.880	–4,221.426	–4,218.800	–4,214.834
Exclude match preference: $\chi^2(1)$		13.09	7.93	
	0.09	5.25		

TABLE III  
WAGE PARAMETER ESTIMATES (IN 2010 DOLLARS)

	AFQT Percentile					
	25		50		75	
<i>Average wages</i>						
Age 20 in 1979	25,827		27,522		29,216	
Age 20 in 1989	18,472		20,166		21,861	
Age 30 in 1989	40,360		42,850		45,340	
	Low		Middle		High	
Location match	−8,366		0		8,366	
Fixed effect support	−20,411	−6,498	−2,454	0	2,454	6,498
State means	Low (WV)	Rank 5 (OK)	Median (MO)	Rank 45 (RI)	High (MD)	
	12,698	14,530	16,978	19,276	22,229	